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Myriam Boueri. Laser-induced plasma on polymeric materials and applications for the discrimination and identification of plastics. Other [cond-mat.other]. Université Claude Bernard - Lyon I, 2010. English. NNT: 2010LYO10207 . tel-00733833

HAL Id: tel-00733833

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N° d'ordre 207 - 2010

Année 2010

THESE DE L'UNIVERSITE DE LYON
Délivrée par
L'UNIVERSITE CLAUDE BERNARD LYON 1
ECOLE DOCTORALE



DIPLOME DE DOCTORAT

(arrêté du 7 août 2006)

soutenue publiquement le

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	M. Jörg	HERMANN	Examineur
	M. Frédéric	PELASCINI	Examineur
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Résumé

La spectrométrie de plasma induit par laser, plus connue sous le nom de LIBS (l'acronyme du terme en anglais Laser-Induced Breakdown Spectroscopy) est une technique analytique qui permet la détection de l'ensemble des éléments du tableau périodique avec des limites de détection de l'ordre du ppm et ceci sur tous types d'échantillons qu'ils soient liquides, solides ou gazeux. Sa simplicité de mise en œuvre, sa rapidité et sa versatilité en font une technique très attractive avec un fort potentiel en termes d'applications que ce soit pour le contrôle en ligne, l'environnement ou l'exploration spatiale. Son point faible reste cependant son manque de fiabilité dans l'analyse quantitative, en particulier lors de l'étude d'échantillons hétérogènes ou de matrices complexes telles que les matrices organiques. Ce travail de thèse propose une étude des propriétés des plasmas induit par laser sur différentes familles de polymères. Une étude du plasma au temps court (\sim ns) par ombroscopie est tout d'abord présentée, ceci pour différents paramètres expérimentaux (énergie laser, durée d'impulsion, longueur d'onde). Un diagnostic complet du plasma par spectrométrie d'émission est ensuite détaillé pour différents délais de détection et montre que la mesure des températures des différentes espèces du plasma (atomique, ionique et moléculaire) permet de vérifier, dans certaines conditions, les hypothèses d'homogénéité et de l'équilibre thermodynamique local. Ceci permet alors la mise en place de procédures quantitatives telles que la méthode dite sans calibration (calibration free LIBS) tout en optimisant le rapport signal sur bruit de la mesure LIBS. Dans nos expériences cette optimisation est mise à profit pour l'identification de différentes familles de polymères en utilisant, pour le traitement des données de la spectroscopie LIBS, la méthode chimiométrique des réseaux de neurones artificiels. Les résultats obtenus, très prometteurs, permettent d'envisager l'utilisation de la LIBS pour l'identification en temps réel des matières plastiques sur chaîne de tri. Par ailleurs et de manière plus générale, ce travail pourrait constituer une base solide pour aller étudier d'autres matériaux organiques plus complexes tels que des tissus biologiques.

Abstract

Laser-Induced Breakdown Spectroscopy (LIBS) is an analytical technique that has the potential to detect all the elements present in the periodic table. The limit of detection can go below a few ppm and this regardless of the physical phase of the analyzed sample (solid, liquid or gas). Its simplicity of use, its rapidity to get results and its versatility provide this technique with attractive features. The technique is currently developed for applications in a large number of domains such as online control, spatial explorations and the environment. However the weakness of the LIBS technique, compared to other more conventional ones, is still its difficulty in providing reliable quantitative results, especially for inhomogeneous and complex matrix such as organic or biological materials. The work presented in this thesis includes a study of the properties of plasma induced from different organic materials. First, a study of the plasma induced on the surface of a Nylon sample at short time delays (\sim ns) was carried out using the time-resolved shadowgraph technique for different experimental parameters (laser energy, pulse duration, wavelength). Then, a complete diagnostics of the plasma was performed using the plasma emission spectroscopy. A detailed analysis of the emission spectra at different detection delays allowed us to determine the evolution of the temperatures of the different species in the plasma (atoms, ions and molecules). The homogeneity and the local thermodynamic equilibrium within the plasma was then experimentally checked and validated. We demonstrated that the optimisation of the signal-to-noise ratio and a quantitative procedure, such as the calibration-free LIBS, can be put in place within a properly chosen detection window. In our experiments, such optimised detection configuration was further employed to record LIBS spectra from different families of polymer in order to identify and classify them. For this purpose, the chemometrics procedure of artificial neural networks (ANN) was used to process the recorded LIBS spectroscopic data. The promising results obtained in this thesis makes LIBS stand out as a potentially useful tool for real time identification of plastic materials. Finally, this work can also be considered as a base for the further studies of more complex materials such as biological tissues with LIBS.

“Logic will get you from A to B. Imagination will take you everywhere.”

Albert Einstein

For Mommsis, Daddie and little Soussie

Acknowledgement

Wow... it's over... and there are so many people that need to be thanked and to be honest, these few pages gave me the biggest writers block possible. How to thank ALLLL the people that made this possible? It goes without saying that, even though this thesis was written by me, it would have never been possible to write without the moral support and scientific enrichment that it bathed in. REALLY! So here goes my attempt to put into words what cannot be expressed in words, so please bear with me (and my humour).

I would first like to thank my thesis supervisor Jin Yu for giving me the chance to surpass my every limit. You gave me the opportunity to create special memories of all the places visited and people encountered all along this thesis. Thank you for also keeping your sanity during the long nights of corrections when I was surely losing mine. In fact, during the whole period, I watched as you sacrificed countless hours of sleep and put your maximal efforts to get the potential of LIBS out of the lab into the world! No potato or carrot or milk and even plastic sample would be safe! It only comes natural to thank Vincent Motto-Ros for his valuable implication during this thesis. Your energy boosts were not only beneficial for me, but to our whole team. Your patience and teaching methods helped me beyond words, especially during those times when I could see no light at the end of the tunnel. Jin and Vincent, you both make an excellent team, and with your motivation and passion, you will attain the heights and merits of your work.

This thesis took place in 3 different laboratories around the world. I thank Christian Bordas, LASIM director for his support and help during difficult situations. It was also a pleasure to meet Rick Russo and Sam Mao during our collaboration with the Berkeley team. Rick, it was a little unreal to have met you in person and I would like to thank you for everything you have made possible for me. Thank you for your constant faith in me. I cannot begin to explain to you how fulfilling it was to work there! Jhanis, Erin, Vassilia, Travis and Dayana: I know I don't stay in touch as much as I would like because I'm just really bad at that. I hope you all know how much I enjoyed working with you and even more how privileged I feel to count you as my friends. Thanks to Xianglei, who took time away from his hectic schedule to help me setup my shadowgraph experiments. Lastly, I would like to thank Heping Zeng for his warm welcome in Shanghai. Working with Bai was a great pleasure and I am sure that her hard work will get her very far in life!

My most sincere thanks go out to Bruno Bousquet and Alexandre Semerok. Every conference was always in highest of spirits thanks to you two. I still laugh when I think about the night we had dinner and made origami swans in Rome! Alexandre, you witnessed the evolution of my thesis from almost

day one, and it was an honour to have you at my jury table. Then there is Jörg! You were there as well from the very beginning. Nothing was too much for you! You were able to juggle science and intellect by day and fun and dancing by night like a pro. Frédéric, you were always such great company and full of lots of useful advice. I hope you can finally see the man in the moon! Thank you Mme. Gilon for allowing us to use your lab for our endless potato experiments. A big thank you to M. Pierre Toulhoat for being the president of my Jury. In fact, it goes without saying, that I had the best jury members one could ever ask for!

I think every PhD student can tell you how long 3 years can be. Before it starts you do whatever you can and work however hard you can to finally get that 'allocation de recherche' to be able to do the thesis. Then when you do get it, you start the customary "why in the world did I want to do a thesis that bad in the first place??" Everyone around you realises you have become obsessed with your thesis but you... well until the last year where you admit this after you work endless hours writing the manuscript up AND even dream about writing it up EVERY night! For the newbies reading this... HA HA good luck!! You'll soon understand what awaits you... I have no idea how you are going to cope but this is how I did.

First, I kept the number to the nearest asylum on speed dial and made sure I was surrounded with people (lots) that enjoy grabbing a pint (or pints depending on the day or week you've had) of beer nearby. Please note that they cannot just be any kind of people! They need to possess the rare gift of being able to listen to you go on and on (and on and on and on) about how insane your week has been and how you are drowning in work. At LASIM I had a number of very patient and magnificent people that filled out that role better than I could have every wished for. The "service informatique" Francisco, Sad and Xavier were my secret asylum. Since day one they dealt with my freak outs, break down, rambling, crazy hysterical laughter and eccentricities so bravely that they should be awarded some kind of Purple Heart for surviving the whole ordeal. I love you guys! I have to tell you that the years 2007-2011 were filled with doctorate students of very high calibre (please insert the cliché joke "not just because I was there" here). We created our own little circle of concentrated madness and delirious moments. I hope the Christmas traditional meal and those runs around the parks to help get rid of extra stress (or love handles, whatever excuse you want to use is ok) we worked so hard is still going on. For the more daring ones all I have to say is: Mardi Gras, Monte Cristo and Poker nights!! Thanks to every student and very extra special thanks to Amandine, Bruno (ah Bruno.... Je resume toute dans une phrase juste pour toi: Je suis passée par Charpennes pour te dire que ma sœur te kiffe (hihi @Cyril!)), Cecilia, Claire, Cyril, Laure, Marion, Maya, Rami, Vincent and Yara. Ok even more special thanks need to go to Cecilia and Cyril how unfortunately had to share an

office with me. Those of you who know me will understand what I'm talking about (ne te vexes pas Cyril, je sais que tu mérites encore plus de remerciements vu que tu as été mon mari de boulot. Mais bon, là je suis à la bourre et cette thèse a besoin d'être imprimée rapidement donc j'ai dû faire court! Je ne peux plus me cacher dans mes obligations... :p). Thanks to Marcus, Framboise, Laurie, Pierre, Julie for the good times especially our lunch special conversations.

You cannot come through LASIM without meeting Aurelie, who always made sure I was OK. We were able to share moments of giggles and old fashion girly gossip in a very mature fashion of course... n'est-ce-pas Aurelie ;) Then one day, out of the blue, came Veronique (a.k.a mon ange gardien) whose presence alone made everything always better. I cannot begin to tell you how important she became to me... That's what was always good with us; we understood things without always having to say them out loud. Je sais que tout le monde remercie toujours les femmes de ménages (car sincèrement il faut voir l'état de certains bureaux de thesard!!). Mais je tenais à vraiment remercier Marie-Louise pour ses petites surprises de bon matins (si vous avez la chance de les recevoir, vous saurez qu'elles ont la magie de garantir que ça va être une très bonne journée!). Thanks to Heather (loved talking to you), Véronique (tes conseils m'ont fait beaucoup de bien), Christophe et Jérôme.

My internship could not have been possible without the help of Estelle (aka Tetelle) and Matthieu. You know when you arrive in a new city and everything just looks the same and you can't find your marks and just don't understand how things work and you just desperately need some guidance. Well, Estelle was that for me. Thank you for explaining things over and over again, teaching me ropes, giving me a heads up on what to do (and many times what not to do!). You helped me survive my introduction to LIBS. Matthieu, who became a brilliant scientist and I'm sure will continue to excel in everything he does, was Jin's first thesis student. It was not always easy coming after someone that was so talented but it did challenge me and allowed me to give the best that I could. Un grand merci également à Marc N., pour les échanges culturelles ainsi que pour ta patience et ton aide pour faire avancer mes manips. Thanks to Wenqi and Ma Qianli for their good humour and for always keeping a smile on their face no matter what was going on! And Josette... wa hen! Your internship with our team was a fresh breath of air! Your hard work and motivation was contagious! You were there until the very last second, coaching me for hours and hours!!!

Now, I really believe that I had some kind of favour left over from my past lives, because the gods (or how Josette prefers to say the God) above blessed me with the perfect (not the best or anything close to the best) but the PERFECT little gifts that life can give you; my friends and family. Sissi, ma bella Sissi! Come posso iniziare a ringraziarti quando sei così lontano!! E 'difficile pensare a tutti quei

momenti che abbiamo trascorso insieme senza avere le lacrime bloccato in gola. What I remember most are how we would meet up with our worlds falling apart (sometimes for real reasons even though most of the time they were just the ones we paranoiacally made up in our head) and were convinced that we could never see sunshine again, and then somehow, god knows how, everything would be just fine. Fine... by fine of course you understand pathological and hysterical laughter combined with martini, kit kat, hamburgers, puzzles (je te l'avais dit qu'il y avait une pièce qui manquait! Peut être que Tom Waits l'a gardée en souvenir) les concerts (NUDO), une danse de pluie un peu spéciale (claro que si (hihi @Cyril!)), berylliums, nymphomanes, la recherche des torros pour retrouver des veaux, les bitch blond girls (pas toujours blonde d'ailleurs) les passages à Charpennes qui sont surement plus rapide.... Bref... ti adoro! Merci à Laurent (tu as été un élément clé dans cette thèse et dieu sait que tu as enduré des moments difficiles aussi avec elle). Un merci du fond de Cœur à des personnes qui suivent mon parcours depuis de longues années (J'espère que vous allez me supporter encore pendant de nombreuses années); Ludi, Maudus, Popo, Thierry, Samia, Catherine et Michel (mes premiers amis à Lyon). Merci à Delphine et Bruno pour votre écoute (des très longues heuressssssss d'écoute et d'analyse!).

And finally, the last but definitely not the least, thank you my family! Please let me start by excusing myself for making you go crazy at times with this thesis. But you know... sharing is caring! Merci Gael pour ta patience qui m'a aidée à surmonter cette fin de thèse et surtout d'avoir eu le courage de rester avec moi pendant cette période baignée de stresse. Thank you my perfect (see message above) parents. I couldn't do this without you. By this I mean my whole life in general. The support you both have given me, the way you've taught me to fight for what I wanted and never give up, the way you were always there to pick up the pieces when things just fell apart, and most of all, thank you for the greatest gift of all, making sure that I never never felt alone no matter where I was in the world and whatever happened. My little Soussie... how to thank such an important part of my life. I always hope that I can live up to the person, not just to the one you see in me, but mostly to the one I see in you. Thanks to my Tata, mon Tonton and Noushie... it's been a crazy ride and I'm so happy to be able to share it with you. Thank you to my Teta, you have always been an inspiration to me. We have a bond that has always been so strong. I love your optimism and the grace you portray despite your khetyara age, you are still one of the most elegant women I know. Thanks to Gramps for always having eyes that shine with pride and having to go through hours of French for me. Thanks to Khalo Elie, Iso and Dido for their constant support. Thanks to Meme, who always makes the moon smile at night and remind that no matter what, everything has a way of working out for the best. I would never be the person I am today without you. I sincerely, from every bit and inch of my heart and soul, love you all...

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GENERAL INTRODUCTION

Since the invention of the laser 50 years ago, one of its primary uses was laser ablation coupled shortly after with optical emission of the ablated matter [1-4]. An intense laser beam is focalised on a sample and the optical emission from the induced plasma is collected by a spectrometer. The analysis of such spectra indicates the elements present in the plasma. This technique is now commonly known as Laser-induced Breakdown Spectroscopy (LIBS). Before 1980, LIBS was more of a laboratory curiosity that fascinated people due to its ease in creating a spark (plasma) in various media (gas [5, 6], liquid, [7-9] and of course solid [10]) providing an intense optical emission from the laser spark. Shortly after, the spectrochemical applications of LIBS, driven by its unique advantages (easy or reduced sample preparation, no contact and remote measurement, instantaneous response and precisely localised (micro) surface analysis), started to develop. The first applications of LIBS were investigated at Los Alamos for the detection of hazardous gases and vapours in air, [11, 12] and small amounts of beryllium in air or on filters [13, 14]. Further development of the technique turned to very practical problems, such as monitoring environmental contamination, [15-18] industrial applications for material processing control, [19-21], and sorting of materials to put them in proper scrap bins. [22-24]

New LIBS applications stimulated by a large range of demands in various domains are still being developed today. A few examples are the analysis of art works and cultural heritages (limited micro destruction) [25-29], homeland security for the investigations on the detection and analysis of bacteria [30-36] and explosives [37-40]. The capability of LIBS to perform remote detection enables its application in nuclear industry [41-43]. A LIBS module has been integrated by the NASA and the ESA for their planetary exploration rovers to analyse the soil on Mars [44-50]. Today LIBS is an elemental analysis technique that provides unique features compared to other analytical techniques.

However, despite these unique features, LIBS remains to this day, still a technique that has not yet found a widespread of applications. This is not the case for other more established analytical techniques such as inductively coupled plasma (ICP). One of the

main reasons behind this comes from the difficulty of LIBS to provide accurate quantitative measurements. The high complexity associated to the laser-matter interaction attracts efforts of some research groups to investigate the fundamental aspects of the laser-induced plasma in order to find a convenient solution to significantly improve the quantitative analysis performance of the LIBS technique. Such efforts include the understanding of different phenomenon involved in the plasma generation and its evolution, as well as the determination of the parameters such as the electronic density and the temperature with time and space resolutions.

In this thesis, we are specifically interested in applying the LIBS technique to organic materials, namely polymers. Compared to the laser ablation of metals, the plasma induced on polymeric materials is much less documented and therefore their properties are still not well established. The difference between the laser ablations of metallic and polymeric materials is due to the differences in absorption and ionisation potentials of these two types of materials. In the case of polymeric materials containing metallic trace elements, the mixture of such different properties can complicate the understanding of the phenomenon. They exhibit discrepancies in their optical, thermal and mechanical properties, which greatly influences the properties of the resulted plasma. Another complication is the chemical reactions within the plasma and those with the background gas, which changes the plasma stoichiometry during its expansion. The purpose of our study is to better understand the laser ablation of organic materials in order to reach a better master of the application of LIBS to such materials.

There is an actual need of improving the performance of LIBS for such materials. In this thesis more specifically, our purpose is to demonstrate the potential of LIBS for the identification and classification of plastic materials. An application of this nature is particularly important for industries that use automated sorting of plastic wastes.

Our approach consisted first of a detailed study of the properties of the laser-induced plasma on the surface of polymeric materials. By studying these properties for a relatively simple and homogenous polymer sample, we are able to infer on the interaction between the laser pulse and the sample, paving the way to understanding such plasma. We used this knowledge to optimise the experimental configuration and spectral data acquisition parameters in order to record the plasma emission in the best conditions. The data

treatment strategy was then developed to get the most significant information for sample classification and identification. Once the ensemble of the experimental procedure was optimised, including spectrum acquisition and data treatment, a range of plastic materials with practical application interests was analysed. The obtained results indicate the actual performance of the method and the further development directions needed to meet the needs of real industrial standards.

The presentation of this thesis is structured in the following way. The first chapter is dedicated to recalling the theoretical basis of laser-induced plasma and that of its diagnostics with the emission spectroscopy. In chapter 2, the early stage expansion of the plasma has been investigated using the technique of time-resolved shadowgraph. This study allowed us to justify our wavelength and laser choices for further analytical application of laser ablation of polymeric materials. Chapter 3 presents the time-resolved spectroscopic measurements that allowed us to observe the evolution of the plasma as a function of the time delay after laser impact on the sample surface. The plasma parameters such as the electron density and temperature are extracted for the assessment of the local thermodynamic equilibrium (LTE) of the plasma. They are also used to define a suitable detection window with an optimal signal-to-noise ratio. Chapter 4 represents an important part of this thesis. A classification and identification procedure was developed for polymeric materials by combining LIBS and data treatment with the artificial neural networks (ANN). Our results put forth the potential of the LIBS technique to provide fast and reliable classification and identification of plastics. The industry of plastics waste sorting and recycling could use this technique to complement and improve the already existing near infra-red spectroscopy technique for such applications. After a general conclusion, we shall finish this manuscript by underlining the perspectives opened by the results obtained during the course of this work.

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CHAPTER I THEORETICAL BASIS

1. INTRODUCTION

Laser ablation of metals has been the subject of a large number of studies leading to a lot of published results when compared with organic samples [1]. The ablation of metals remains easier to understand when compared to organic substances, which add more complications to the mechanism of plasma generation. This is especially true when we consider metallic elements in an organic matrix. Due to different absorption properties, the laser interaction with organic and with metallic samples differs, putting forth even further discrepancies of the plasma properties i.e. non stoichiometric ablation because of the large difference between their ionisation potentials. To improve the understanding of the laser ablation of polymers, we decided to study simple polymer samples such as Nylon or PVC.

In this chapter we wish to recall briefly the basic fundamentals that we used in our understanding of the laser-induced plasma and the spectroscopy of plasmas induced from polymeric materials. Many parameters can influence the plasma properties but we shall first begin with the main differences between the plasma generations with femtosecond or nanosecond laser pulses. A further emphasis will be made between UV and IR nanosecond laser pulses. After briefly discussing the plasma formation, we shall evoke our methods for plasma diagnostics (electronic density and temperature calculations). The spectral emission is used to calculate the electronic density (mainly Stark broadening) and the different plasma temperatures (atomic, ionic and molecular). These measurements will allow us to assess the local thermal equilibrium (LTE) of the plasma. An introduction of calibration free LIBS (CF-LIBS) finishes off this chapter even if we still have not used this method of analysis in this thesis. The advantages associated with this technique have greatly attracted us and we would like to implement it in our further works.

2. PLASMA GENERATION

The mechanisms involved in the ablation depend on a number of variables that effect the interaction between the laser pulse and the sample target. Firstly, the properties of laser pulses can lead to different interaction regimes with regards to the pulse duration, the wavelength and the fluence of the pulses. Secondly, the target (can be in a solid, liquid or gas form) and its properties also influence the ablation due to its specific laser absorption mechanism and matrix effects. Lastly, the surrounding atmosphere during ablation such as the pressure and the ambient gas (inert, air, vacuum etc) will also influence the plasma generation and its expansion. The main differences between nanosecond and femtosecond laser ablation will be evoked here followed by the main mechanisms in which a plasma is formed.

Laser ablation consists of focalising an intense laser emission on a sample and removing matter from the surface. For a LIBS measurement, a focalising lens directs and concentrates the laser pulse energy on the surface of the material of interest. For this reason, we shall often talk in terms of the pertinent factors, which are the fluence (often expressed in J / cm^2) or the irradiance (often expressed in GW / cm^2). They allow us to take into account the size of the section of the focalised beam and the laser pulse duration

respectively. In order to generate a plasma, the irradiance must be greater than what is called the ablation threshold I_{th} (or a fluence greater than F_{th}). As a consequence, the locally removed mass exhibits rapid melting and evaporation to form a vapour. This vaporised mass is exploded out of the material at a supersonic velocity. However, only when the vapour is sufficiently ionised can it be then called a plasma [2]. We can note at this point that our experimental research was conducted so that the laser fluence was always above the threshold value to ensure the plasma generation.

2.1 LASER-MATTER INTERACTION

The practical reason, for which we would like to go into the further detail of the fundamentals of the laser-matter interaction, is because it will enable us to define the appropriate optimisation of our LIBS experimental parameters. For us, this means finding the parameters where we are able to acquire the most significant results. As already discussed, the pulse duration has a significant impact on the ablation process. In the nanosecond regime, there is a significant interaction between the laser emission and the ablated material. In the femtosecond regime, the laser pulse does not directly interact with the generated plasma. This laser-plasma interaction therefore drastically influences the plasma properties. We shall start by discussing the nanosecond regime (in UV and IR) then the femtosecond regime.

2.1.1 Nanosecond Ablation

The most common type of laser used in LIBS is the nanosecond Nd:YAG, which can provide pulse in different wavelengths ranging from UV to IR. Nanosecond lasers emit pulses long enough (around 5 ns FWHM in our case); their “tail” remains present during plasma formation and expansion. This leads to a significant post-ablation interaction between them, i.e. the laser energy continues to be absorbed by the vapour. The laser energy can be absorbed by atoms or molecules leading to their excitation or even their ionisation [3]. The laser energy can also be absorbed by free electrons leading to their heating and the cascade ionisations due to subsequent collisions between electrons and atoms or molecules. The initial temperatures can be very high (6,000 – 20,000 K) with high electron densities ($10^{17} - 10^{19} \text{ cm}^{-3}$) [4]. The energy loss and cooling of the

electrons is mainly due to the thermal propagation into the target [5]. The heat propagation diffuses a few microns into the target causing it to melt in a non-homogenous way creating craters with messy borders.

Depending on the irradiance and the wavelength of the pulse, the ablation mechanisms can be either dominated by thermal (melting, sublimation, and vaporisation) or non-thermal (direct molecular bond breaking) processes. We will focus on laser irradiances greater than 10^8 W/cm^2 to create plasmas with high enough temperatures to ensure the atomic excitation and ionisation of the metallic elements present in certain polymers. The subsequent laser-plasma interaction enables even further heating of the free electrons within the plasma.

The photon absorption in a plasma is mainly due to inverse Bremsstrahlung (IB) and photoionisation (PI) processes since the other absorption mechanisms have smaller cross sections and are ignored at this laser intensity [4,6,7]. The consequence of IB is higher temperature of electrons. These excited electrons in turn collide with neighbouring ions or neutral atoms more violently, causing the liberation of other electrons and in turn ionising more atoms. This phenomenon is known as avalanche ionisation. As long as the electromagnetic field of the laser remains in interaction with the electrons, this process will repeat itself increasing the ionisation degree of the plasma [2]. The cross sections of the collisions between electrons-ions are much greater than those between electron-neutrals.

The inverse Bremsstrahlung (IB) absorption $\alpha_{IB}(\text{cm}^{-1})$ is approximately proportional to λ^3 as can be seen in Equation (1.1). The second part of the right hand side accounts for the simulated emission [2, 3].

$$\alpha_{IB} = \left[Q N_e N_0 + \left(\frac{4 e^6 \lambda^3 N_e Z^2 N_i}{3 h c^4 m_e} \right) \left(\frac{2\pi}{3 m_e k_B T_e} \right)^{1/2} \right] \left[1 - \exp \left(- \frac{h c}{\lambda k_B T_e} \right) \right] \quad (1.1)$$

<i>With</i>	α_{IB}	:	<i>inverse Bremsstrahlung absorption coefficient (cm^{-1})</i>
	Q	:	<i>cross section for electron-photon absorption (cm^2)</i>
	N_e	:	<i>electron density (cm^{-3})</i>
	N_0	:	<i>neutral atom density (cm^{-3})</i>
	λ	:	<i>ablation wavelength (cm)</i>
	Z	:	<i>ion charge (1 or 0)</i>
	N_i	:	<i>ion density (cm^{-3})</i>
	c	:	<i>speed of light (cm/s)</i>
	h	:	<i>Plank's constant ($eV \cdot s$)</i>
	m_e	:	<i>electron mass (MeV/c^2)</i>
	k_B	:	<i>Boltzmann's constant (eV/K)</i>
	T	:	<i>excitation temperature (K)</i>
	e	:	<i>elementary charge</i>

Another photon absorption process that initiates the plasma formation by producing more free electrons is photoionisation. Photoionisation is defined as the ionisation of an atom or a molecule by a single or several photons that have a sufficient amount of energy to directly eject an electron out of the atom or molecule. When the combination of a few photons is required, the process is called multiphoton ionisation (MPI). The energy associated to a photon is defined in Equation (1.2) and is inversely proportional to λ . This indicates that UV radiation will therefore more efficient for ionising atoms or molecules than IR wavelengths.

$$E = h\nu = \frac{hc}{\lambda} \quad (1.2)$$

Laser-plasma interaction is the main ionising mechanism of the vapour, dominating the direct ionisation of particles when vaporised [8], where 10% of the particles are considered to be ionised [9, 10]. The coupling between the laser and the plasma complicates the evolution of the plasma. For example plasma screening (can also be called shielding) can occur in a sufficiently dense plasma, i.e., the electronic density reaches the critical density, $N_c(cm^{-3})$, defined by Equation (1.3) [8, 10]:

$$N_c = \frac{10^{21}}{\lambda^2} \quad (1.3)$$

<i>With</i>	λ	:	<i>wavelength (μm)</i>
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It can be seen that N_c is inversely proportional to the square of the wavelength $\lambda(\mu m)$ i.e. shorter wavelengths have higher critical densities diminishing the probability for shielding. When the plasma reaches this density, the “tail” of the laser pulse is reflected backwards, and its interaction with the plasma (especially the core) is no longer possible.

For laser ablation of polymers, there can be photochemical or photothermal ablation. There is even the possibility of having a mixture of both. Photochemical ablation is defined as the absorption of laser photon (or photons) to dissociate the molecular bonds within the polymer and directly evaporate the matter. Photothermal ablation is when heat is dissipated within the sample to cause melting and etching [11,12], which is more commonly observed in laser ablation of metals.

The ablation depth for polymers per laser pulse L is expressed in Equation (1.4). The logarithmic dependence with the ablation fluence can be applied to most solids even if it is more commonly used for polymers [13].

$$L \cong \alpha^{-1} \ln\left(\frac{F}{F_{th}}\right) \quad (1.4)$$

<i>With</i>	L	:	<i>ablation depth per laser pulse ($\mu m/pulse$)</i>
	α	:	<i>absorption coefficient of the sample (cm^{-1})</i>
	F_{th}	:	<i>ablation threshold fluence ($J\ cm^{-2}$)</i>
	F	:	<i>laser fluence ($J\ cm^{-2}$)</i>

Srinivasan felt the need to introduce another term to Equation (1.4) to put forth the contributions of both the photochemical and photothermal mechanisms [14]:

$$L \cong \alpha^{-1} \ln\left(\frac{F}{F_{th}}\right) + A \exp\left(-\frac{E_1}{RT}\right) \quad (1.5)$$

<i>With</i>	E_1	:	<i>activation energy (cal / mol)</i>
	T	:	<i>temperature of the sample (K)</i>
	A	:	<i>prefactor of the Arrhenius equation.</i>
	R	:	<i>gas constant ($cal / (K\ mol)$)</i>

a. Infrared ablation (1064 nm)

For the ablation of polymeric materials, the photoionisation is very weak in this regime as the infrared photons do not carry a sufficient amount of energy to chemically break the bonds. Due to the insufficient amount of energy of one IR photon to break the chemical bonds present in polymers, it is more likely to see photothermal ablation in this case. The vibrational modes of molecules absorb the laser energy and diffuse it in the form of heat. This leads the sample surface to melt. In order to have ablation, a chemical bond needs to absorb several quanta of laser energy to induce evaporation. During this time however, the material located close to this area will undergo a melting process [11]. This leads to a low efficiency of matter removal by ablation. The post-ablation heating observed in IR ablation is quite efficient which leads to a higher temperature in the plume. In fact, Wang et al. [16] reported the plasma emission to be up to 5 times higher for IR emission for the main carbon line at 248 nm. The observed enhancement in optical emission from the plasma is due to a higher temperature rather an increase in the ablated material due to the high energy of the excited state.

More specifically for the post-ablation interaction between the plasma and the laser, in this regime, the main absorption mechanism is the inverse Bremsstrahlung as indicated in Equation (1.1) causing strong electron heating in the plasma and cascade ionisations. The electron density of the plasma will then increase to the point where it reaches N_c and visible partial screening (also called shielding) of the plasma can be observed. The laser irradiance needs to be greater than 2×10^8 W/cm² for plasma shielding to take place [3]. Chapter 2 presents shadowgraph images of the plasma in expansion to show this phenomenon very clearly [2]. When partial screening commences, the laser pulse “tail” can no longer couple to the target material as its entire energy is absorbed by the plasma or even reflected. Therefore, even when the fluence increases, the ablation depth remains low. The consequence on the plasma expansion is the laser-supported detonation wave (LSD) as it will be shown in Chapter 2.

b. Ultraviolet ablation (266 nm)

Inverse Bremsstrahlung is less efficient for UV ablation than in IR ablation. However, the multi photon ionisation coefficient is much higher leading to a large number of ions [7]. The recombination processes that compete with the electron generation by multiphoton ionisation may reduce the amount of charged particles in the plasma. The critical electron density is higher in this regime, reducing the effect of plasma shielding. This allows a more efficient coupling between the laser pulse and the target. The energy associated to this wavelength is greater than 4.7 eV. This means that the UV wavelength has enough energy to chemically dissociate and break the bonds, through multi photon process, within the matrix for most organic materials.

When bonds are broken in a non-thermal manner i.e. not heat induced but more chemically broken, the ablation is said to be photochemical [17]. When a polymer absorbs a UV photon, it can either excite the kinetic energy of its internal vibrational transitions (photothermal) or directly break a chemical bond within the target [11]. For UV nanosecond ablation, there is usually a mixture between both photothermal and photochemical ablation. The photochemical processes that intervene in UV ablation results in higher ablation rates (ablation efficiency, which is the ablated mass per laser pulse per unit area [2,18]). Furthermore, the rotational, translational and vibrational temperatures of ablated species are not in equilibrium. The shockwaves have higher initial velocities, the surface swelling is faster and a lower ablation threshold value is found when compared to IR ablation [2,19].

To conclude about the differences between the nanosecond UV and IR ablation regimes, regimes, we present a model by Garrison et al [11] in Figure 1, which illustrates the photochemical ablation (no melting can be seen around the crater just a clean expulsion of matter) and the photothermal ablation (melting and evaporation).

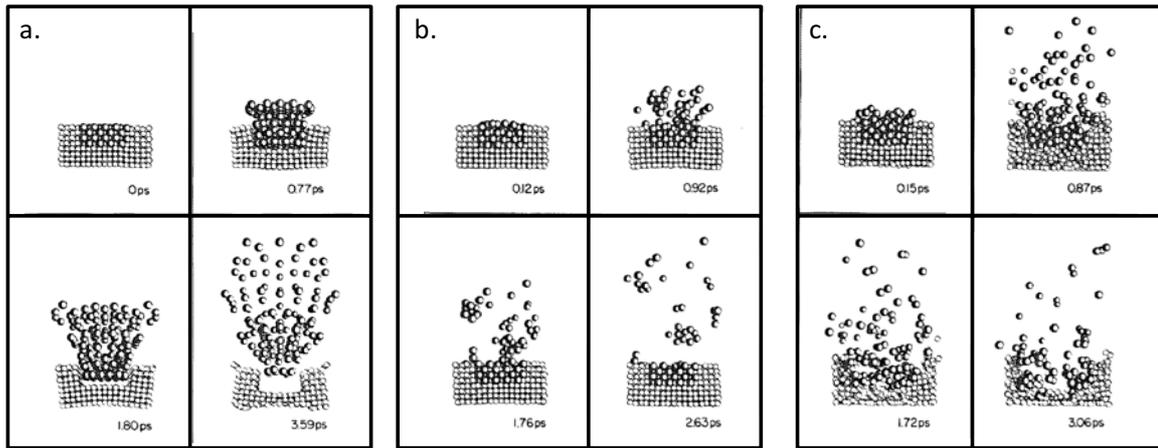


Figure 1 Nanosecond laser ablation of polymers in air [11] at different time delays after laser impact. a) UV ablation (1 photon per monomer) b) IR ablation (1 photon per monomer) c) IR ablation (4 photons of energy delivered to each monomer). There is a slight perspective in the drawings.

2.1.2 Femtosecond Ablation

In the case of femtosecond ablation, the phenomenon responsible for the ejection of matter from the surface of a target is the Coulomb explosion, otherwise known as non-thermal melting. When the electronic excitation energy due to the intense laser field is transferred to the lattice phonons in the form of kinetic energy, the atomic motion can be so violent as to break the bonds that hold solids together. The release of the electrons of the solid [20] leaves behind a locally positively charged lattice. Since the electrons are negatively charged, this mutual repulsion (Coulomb repulsion) creates an explosion where a small cluster of charged and energetic ions creates a plasma.

The energy absorbed by the electrons increases instantaneously but the time it takes for them to transfer their energy to the lattice phonons depends on the thermal conductivity of the sample [15] and cannot be neglected for femtosecond ablation. For femtosecond ablation, the heat diffused to the target represents only a tiny part of the laser energy. The rest of the very intense beam is used to create photothermal changes and multi-photon ionisation.

The craters produced in this regime have smoother edges and can be processed more precisely compared to the craters induced by nanosecond pulses. This can be due to the absence of the liquid phase in femtosecond ablation. The velocity of the ejected mass is

considerably faster than the thermal velocity of the particles [18]. Photons from visible and IR femtosecond pulses do not have enough energy for ionisation. The absorption of several photons (multi-photon absorption MPA) makes the ionisation possible, but makes the process more sensitive to the laser intensity [21] $MPA = \sigma I^m$ where m satisfies the equation $m\hbar\omega > E_g$. Due to the higher peak intensities of ultrashort laser pulses, the interband transitions are significantly increased. Ablation efficiency is best for femtosecond pulses and it is improved for ns pulses when the laser wavelength decreases on metal targets [18]. Our shadowgraph experiments will demonstrate that this is also the case for polymeric targets.

In femtosecond ablation, the laser pulse is absent during the plasma expansion meaning that there is no interaction between the laser pulse and the expanding plasma. The pulse shielding effects is totally absent. Due to high intensity of a femtosecond pulse it can however create an ionised channel by breaking down the air molecules during its passage. This can be observed in the shadowgraphs in Chapter 2.

2.2 PLASMA FORMATION

Figure 2 shows the different steps observed during the nanosecond ablation of a solid sample. The laser pulse strikes the sample surface (a) where it is first absorbed. The energy absorbed is rapidly converted into heat and causes the sample to locally melt. Due to the focalisation that increases the irradiance, the laser beam contains enough energy to induce vaporisation and fragmentation (b). The boiling point is reached when the pressure of the vapour exceeds that of the ambient environment. As seen before, for intensities higher than 10^8 W/cm², a large number of electrons, ions and excited atoms are present in the vapour and absorb part of the laser energy leading to a breakdown in the vapour and its ionization (c) [6,7]. At this moment, the plasma is dominated by the continuum emission that overbears any other atomic emissions. The emission detected in this phase is mainly due to the Bremsstrahlung and the recombination processes

When the laser pulse is over, the plasma continues to expand very rapidly attaining supersonic velocities [11] ($\sim 10^6$ cm/s) perpendicular to the surface. The ions formed at the beginning start to capture the free electrons leading to neutrals. De-excitation of ions

and neutrals lead to the emission of characteristic photons, a very important point that allows us to determine the concentration of the elements in the plasma with the LIBS technique (d). This phase is followed by the formation of aggregates and clusters due to the decrease in temperatures (e), which are deposited around the crater along with the molten material created by the laser melting (f).

Different time stages can be studied during the expansion of the plasma. Figure 1 shows the time evolution of these stages (A, B, C, and D) with regards to a femtosecond and nanosecond laser pulse duration. The first phase A represents the time that the laser pulse transfers its energy to the sample. The laser photons are immediately absorbed by the surface electrons ($\sim 10^{-15}$ s). It then takes ~ 100 fs for them to thermalise amongst themselves and transfer their energy in turn to the lattice phonons (a few ps). Followed by phase B where the plasma vapour, containing ions, atoms and molecular fragments, initiates its expansion into the surrounding environment. Due to the supersonic velocity of this expansion, a shockwave engulfs the plasma. This expansion continues for several microseconds. From the formation of the plasma and until its dissipation, it emits radiation. The continuum radiation is followed by the ionic, atomic and molecular emission lines that can be used to determine the elements present within the plasma. In phase C, the excited ions, atoms, and molecules return to a more stable state by emitting photons. The plasma further relaxes in phase D, where its temperature falls below the fusion temperature and the condensation starts to form nanoparticles [1]. In this thesis we will mainly focus on parts B and C where the expansion and the spectral emission of the plasma will be studied.

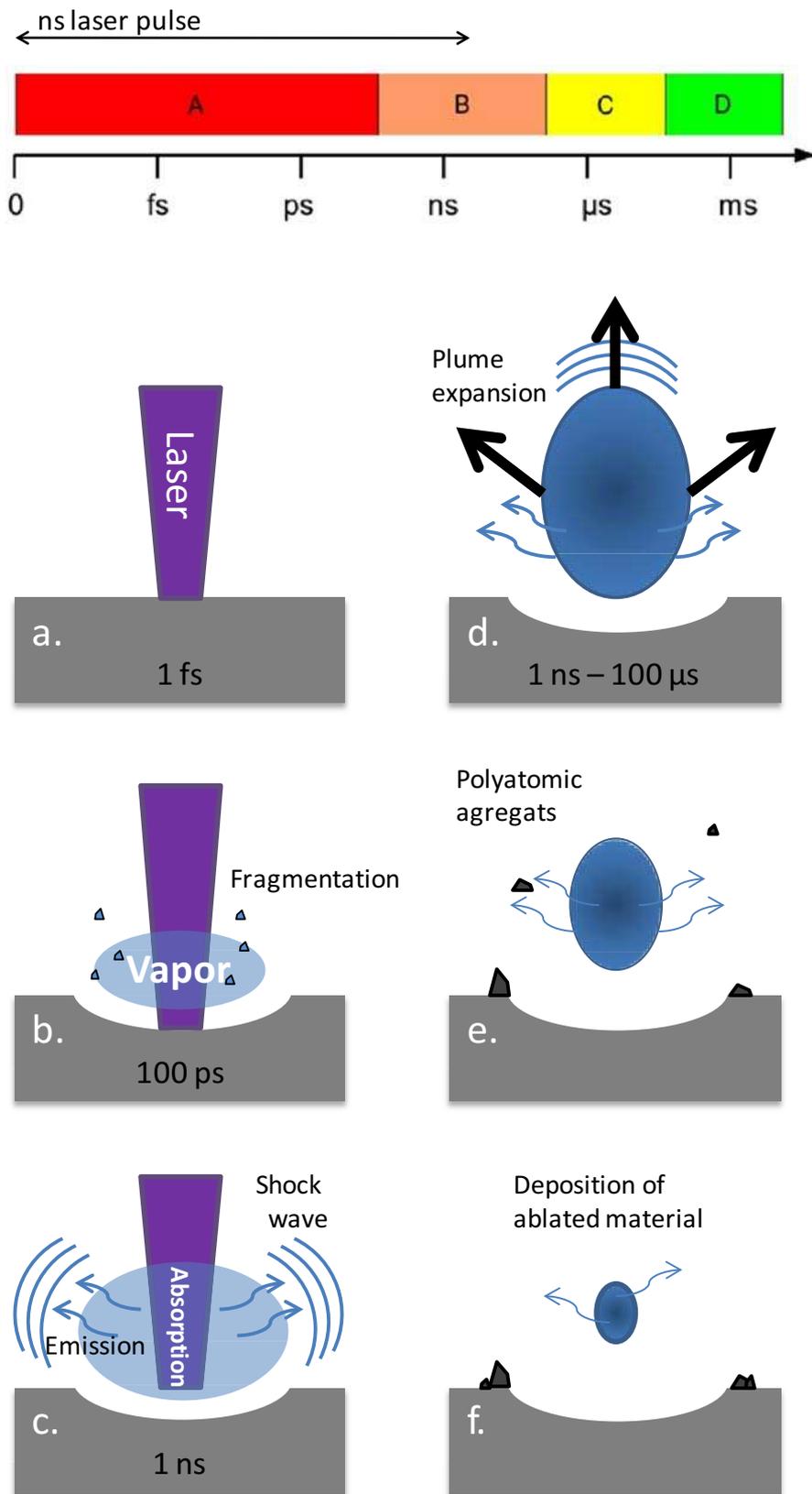


Figure 2. Plasma formation for nanosecond pulses with sufficiently high fluence [22,23].

3. LIP EMISSION SPECTROSCOPY

3.1 LIP EMISSION

Laser-induced breakdown spectroscopy (LIBS) uses the spectral emission from the laser-induced plasma (LIP) for elemental analysis of the target. However, from the moment the plasma is formed until the moment it is dissipated, it goes through different stages. The spectra collected at different delays and different gate widths are therefore different. Figure 3 shows the evolution of the optical emission intensity with regards to the time after the laser pulse strikes the sample. It can be divided into two parts: a very intense signal that comes from the continuum and the atomic or molecular signals used to characterise the sample. At early delays (especially when the laser pulse is still present) the optical signal is dominated by a strong continuum emission. During laser ablation, ions and neutral atoms are liberated and are further ionised by the laser pulse. Therefore, ionic lines are first seen, followed by atomic and finally molecular spectral emission lines. The delay after laser impact chosen and the gate width are important criteria depending on what is to be analysed. Time-resolved studies are the best way to find the most adequate time delay and gate width according to the application.

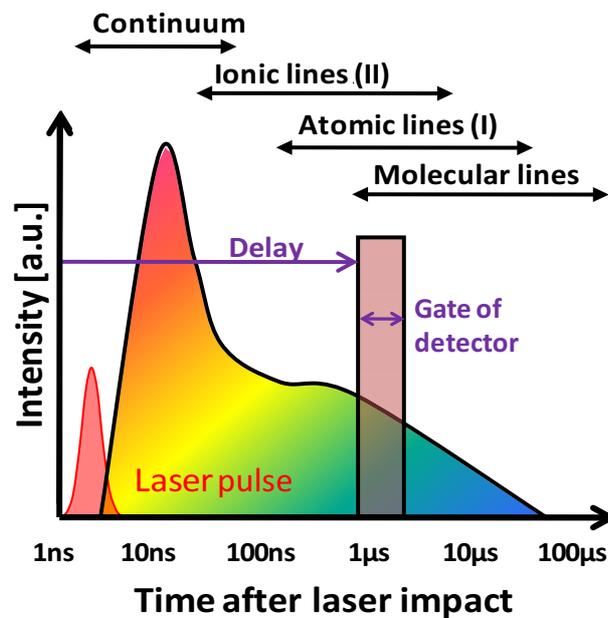


Figure 3. The evolution of the intensity emission from a laser induced plasma [24].

Since these excited states are not stable, the electrons within the atoms and molecules will return to more stable energy levels by releasing their excess energy in the form of photons (Figure 4). This can be done either by spontaneous emission or by intense enough collisions. Contrary to the continuum radiation, the photons emitted ($h\nu_{ij}$) are specific to their transitions and therefore represent different atomic or molecular information. This information is the essential for the diagnostics of the plasma i.e. calculate the electron density and the temperature.

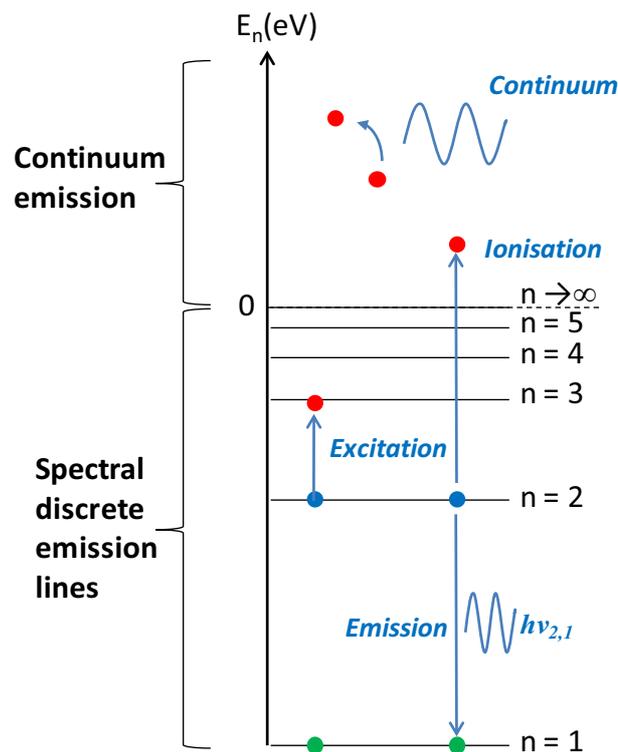


Figure 4. Continuum and Spectral line emission

Bremsstrahlung Radiation

In the first picoseconds the electrons and ions start colliding and interacting. When a free-electron is decelerated in the presence of a strong electromagnetic field, such as the one of an atomic nucleus, the loss of kinetic energy is transformed into radiation shown in Figure 5. This process is called Bremsstrahlung, which is the German term meaning “braking radiation”. Since the electron started being free and after the interaction remained free, it is known as a free-free transition. However, when a free electron is captured by an ion instead of being deviated, this leads a free-bound transition. The

excess kinetic energy is given off in the form of a photon resulting in recombination radiation. These two radiation mechanisms are responsible for the continuum emission. Since this emission is mainly due to the free electrons present in the plasma, it does not represent any spectral lines.

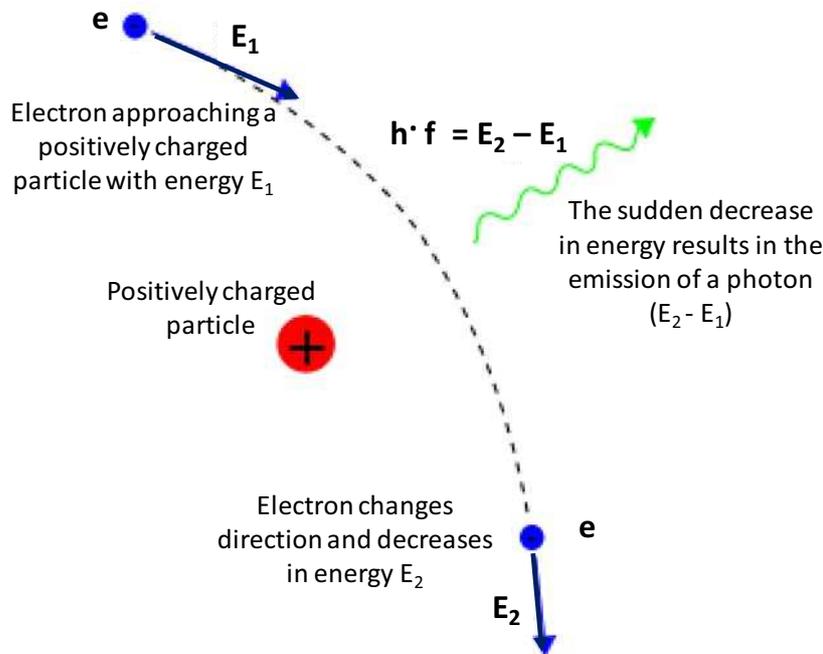


Figure 5. Bremsstrahlung radiation.

Spectral line emission

The LIBS technique is based on the study of the light emission from a laser-induced plasma. After an enough long delay, a discrete spectrum is recorded with peaks produced at specific wavelengths. This means that every peak corresponds to an element and this is how we use the plasma to characterise the ablated matter. For molecular lines, it often comes in a group of lines which can be observed in the results in Chapter 3. Between the time the photon is emitted and the time it is detected, it has the possibility to get reabsorbed by other ions or atoms. An optically thick plasma means most of the radiation cannot escape the plasma due to high absorption. When the photons are directly detected, the plasma is said to be optically thin. This is the ideal situation for LIBS but unfortunately this is not always the case. Corrections have to be made and some spectral lines have to altogether be ignored to avoid erroneous interpretations.

Self-absorption and Self-reversal

The corrections are not the same depending on how the photon is reabsorbed. If the photon emitted from one atom is absorbed by another atom of the same element close by, considered to be at the same temperature and surrounded by the same electron density, the spectral line is said to be self-absorbed. Transitions which involve the fundamental level are more affected by such perturbations. When a photon is emitted from the hot core of the plasma and absorbed by an atom of the same element in the cooler exterior where the temperatures and the electronic density differ, the phenomenon of self-reversal is observed. Both phenomena give rise to distorted spectral lines, which can be seen in Figure 6. Self-reversal results in broadening but most importantly; it is marked by a dip in the centre of the peak. Self-absorption broadens the spectral line, making it difficult to distinguish from other non-distorted lines.

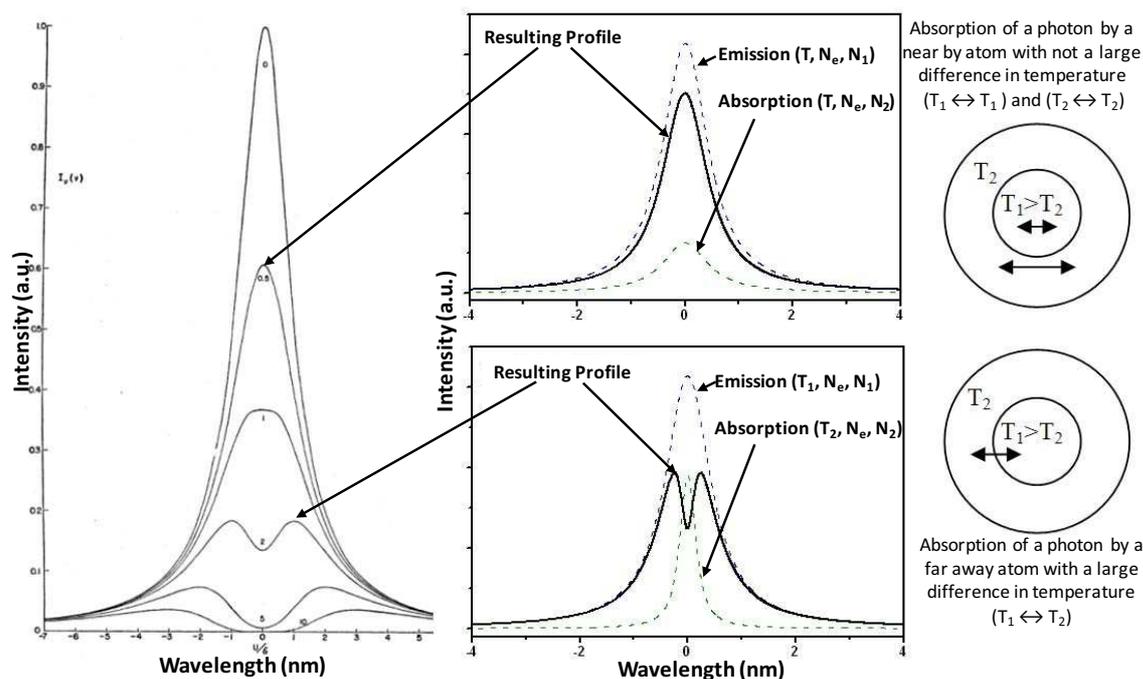


Figure 6. . Self-absorption (upper) and Self-reversal (lower) spectral distortions [25].

A lot of emphasis is given to spectral line analysis because they are the basis of plasma diagnostics and the technique of LIBS. For example, the electronic density can be found by studying the spectral line width and the temperature can be calculated using the line intensities. Due to the self-absorption or self-reversal, it is crucial to carefully select the

spectral lines. Self-absorbed lines are not always easy to identify because of having the same allure as non absorbed lines. The spectral lines chosen should ideally have [26]:

- strong intensity in the working spectral range
- minimum overlapping with other lines
- minimum self-absorption
- without saturation for the detector.

In practice, these ideal lines are very difficult to attain altogether. There are a few compromises that need to be done. For example, when a line is known to be self-absorbed, their intensities should be corrected before being used or otherwise excluded from the calculations. Lines that include the fundamental level should also be avoided [27] from a certain value of the electron density.

3.2 PLASMA DIAGNOSTICS

The plasma dynamics and time evolution have interested many researchers for a few decades [13, 28, 29]. Temporal and spatial evolutions have been extensively studied to understand the different mechanisms that occur after laser impact. A certain emphasis is taken on electron density and temperature calculations from spectral data. Some of these calculations, particularly the those for estimating the plasma temperature can only be used when the plasma is considered to be at a certain equilibrium known as the local thermodynamics equilibrium (LTE).

Before going into more detail regarding LTE, explanations on the electron density and temperature calculations will be first explained. For now, very simply, LTE first requires that the plasma has a certain electron density, larger than a critical value (equation (1.42)) and then that all the species present (atomic, ionic and molecular) be defined by the same and unique temperature. In other words, the excitation temperatures of different species (atoms, ions, and molecules) must have the same value as temperature of the electrons. For the LIP plasma, a transient and expanding plasma, other criteria which takes into account the time evolution and the inhomogeneity of the plasma, need to be considered

[30, 31]. Only when these criteria are fulfilled, the LTE can be considered. In addition, for a typical LIBS plasma, only neutral and singly charged ions are taken into account for the following calculations [24].

3.2.1 Electron Density Calculation

The different spectral line broadening mechanisms need to be understood in order to correctly extract the electronic density. Each of them has a particular influence on the spectral line. Some are represented by the Gaussian profile and others the Lorentzian one. This will be explained in more detail below followed by stating how the electron density can be calculated.

a. Broadening Mechanisms

Natural line broadening

The natural line broadening is due to the finite time that an atom spends in its excited state E_i and results in a Lorentzian profile [32]. If the mean time at E_i is Δt , then the Heisenberg uncertainty principle states that:

$$\Delta E_i \Delta t_i \geq \frac{h}{2\pi} \quad (1.6)$$

The frequency ν_{ij} (s^{-1}) considered for a transition from E_i to E_j is:

$$\nu_{ij} = \frac{E_i - E_j}{h} \quad (1.7)$$

Considering the finite line width $\Delta\nu^{Natural}$ (s^{-1}), the natural broadening can be expressed as:

$$\Delta\nu^{Natural} = \frac{\left(\frac{1}{\Delta t_i} + \frac{1}{\Delta t_j}\right)}{2\pi} \quad (1.8)$$

The value at the spectral range of 300 nm is of the order of 0.000005 nm which means that it will be neglected for atomic spectra from a laser induced plasma.

Doppler broadening

The Doppler broadening [32] is the broadening of spectral lines due to the Doppler Effect resulting from the thermal motions of atoms and molecules in the plasma (some moving towards and others away from the detector). The higher the temperature or pressure within the plasma, the faster the particles move increasing the chances of collisions. The particles in motion are considered to have a Maxwellian velocity distribution resulting in a Gaussian spectral broadening profile. The FWHM of the profile depends only on the frequency of the spectral line, the mass of the emitting particles, and their temperature, as can be stated in (1.10):

$$\Delta\lambda_{1/2}^{Doppler} = (7.16 \times 10^{-17}) \lambda \left(\frac{T}{M}\right)^{\frac{1}{2}} \quad (1.9)$$

With T : emitters' temperature (K)
 M : atomic weight (a.m.u)
 λ : wavelength (\AA)

For the spectral range frequently used in LIBS (UV to IR), Doppler broadening is not the main line broadening mechanism. If an extremely high plasma temperature is considered (10^5 K), the corresponding Doppler broadening for the visible range is only 0.01 nm. This is negligible compared to actual spectral line widths.

Stark broadening

Collision-induced broadening, sometimes called pressure broadening, is a result of the deformation of atoms or molecules when they bounce off each other. For example, they may not be as symmetrical after a collision as they were before. These deformations perturb the quantum mechanical energy levels of the atom or molecule, slightly shifting the frequencies of the emission or absorption lines. Just like Doppler broadening, the spectral lines are therefore broadened. This effect depends on both the pressure and temperature of the gas.

Stark broadening is a particular case of pressure broadening where atoms interact with the neighbouring charged particles, such as ions and electrons. According to quantum mechanics, the interaction will slightly modify the energy levels considered in the transition inducing the spectral line to broaden. The broadening is homogeneous and corresponds to the Lorentzian profile. It follows from the Lorentz Force Law that the magnitude of the perturbing electric field (F) created by fast moving electrons and relatively slow ions (e) at a certain distance (r) is given by:

$$F = \frac{1}{4\pi\epsilon_0} \frac{e}{r^2} \quad (1.10)$$

During the plasma expansion, where the electron density of the plasma is quite high, this is the main broadening mechanism, dominating the natural and Doppler mechanisms [8].

Instrumental broadening

Measured line profiles also experience influence from the instrument resolution. It is dependent on the slit width, the grating dispersion and the detector. Diffraction from a slit gives us a cardinal sinusoidal pattern, which can be approximated using a Gaussian profile.

b. Electron Density

Finally with all these mechanisms taken into consideration, the spectral lines can be fitted with a Voigt profile, the result of the convolution of Lorentz (Stark broadening) and Gaussian (Instrumental broadening). Natural and Doppler broadening are considered as negligible for atomic spectra of a laser-induced plasma [26].

$$V(\lambda (nm)) = G(\lambda, \Delta\lambda_{Doppler}, \Delta\lambda_{Inst}) \otimes L(\lambda, \Delta\lambda_{Stark}) \quad (1.11)$$

$$\Delta\lambda_{Gauss}(nm) = \sqrt{\Delta\lambda_{Doppler}^2 + \Delta\lambda_{Inst}^2} \quad (1.12)$$

Linear Stark Effect

The line width and the electronic density of the plasma are closely linked with each other. There exists a direct relationship between Stark broadening and the electronic density, especially at short delays where the plasma is quite dense [32]. For hydrogen, this relationship is linear because of its linear dependence with the Lorentz force, which results in a symmetrical broadening. The electron density can be calculated using the following equation [32]:

$$N_e (cm^{-3}) = 8.02 \times 10^{12} \left(\frac{\Delta\lambda_s}{\alpha_{1/2}(Ne, T)} \right)^{3/2} \quad (1.13)$$

<i>With</i>	$\Delta\lambda_s$:	<i>FWHM of the line (\AA)</i>
	$\alpha_{1/2}$:	<i>half width of the reduced Stark profile(\AA)</i>

The parameter $\alpha_{1/2}$ depends slightly on the excitation temperature T_{exc} , and the electron density N_e . Values of this parameter can be found in ref [32] with precision in the range of 10%.

Quadratic Stark Effect

For other non-hydrogen-like atoms, stark broadening is proportional to F^2 and is thus called the quadratic Stark effect. Their density can be calculated using the following equation:

$$N_e(cm^{-3}) = N_{ref} \frac{\Delta\lambda_{Stark}}{2\omega_{ref}} \quad (1.14)$$

With N_{ref} : reference N_e ($\sim 10^{16} - 10^{17} cm^{-3}$)
 ω_{ref} : impact parameter at this reference density.

Hydrogen Balmer H_α Emission

Even though both linear and quadratic Stark effects are observed in the LIBS spectra, more focus is taken on extracting information from hydrogen or hydrogen-like atoms. Their linear dependence allows more accuracy and ease when calculating the half width of line profiles providing more reliable results [8, 26]. The hydrogen Balmer line H_α (656 nm), emission line width in function of the detection delays is often used to study the temporal evolution of the electron density in the plasma during its expansion and decay [33].

In organic materials, the presence of hydrogen in the matrix allows for a well isolated and significantly intense peak. In addition, this line is strongly affected by the linear Stark effect [32]. Its use for electron density determination has been extensively reported for the diagnostics of laser ablation plasma [34, 35]. The surrounding electrons are mainly taken into consideration rather than the ions, which have a much weaker impact on the total line broadening [36]. Even though hydrogen is part of the polymer matrix H_α the auto-absorption can be neglected as long as the electronic density is less than $10^{18} cm^{-3}$ [35].

Figure 7 shows H_α spectral line fitted with Voigt profiles for a potato. Measurement were taken at different delays (100, 300, 600 and 1000 ns) after laser impact to put forth its evolution with time. A narrowing of the line along with a shift towards smaller wavelengths can be seen.

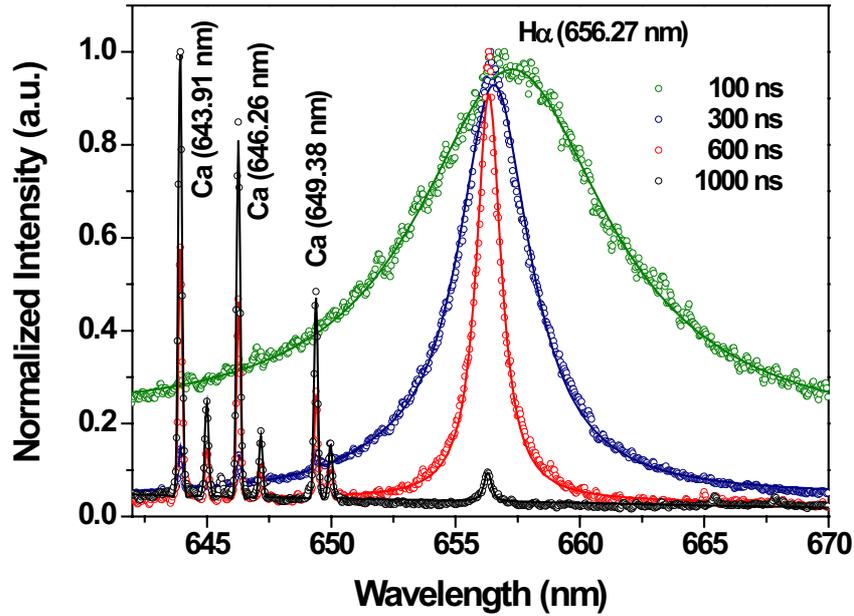


Figure 7. Stark Broadening The H_{α} fitted with Voigt profiles for a potato sample. Different laser impact delays were taken (100, 300, 600, and 1000 ns after laser impact). [37]

3.2.2 Temperature Calculation

For temperature calculations, the differences between the line intensities are taken into account rather than the broadening. There are different methods, Boltzmann (one element, one species (neutral or ions)) plot, Saha-Boltzmann plot (one element, different species (neutral and ions)), ionic temperature (one element in its neutral and ionic state) and molecular temperature (vibrational). To estimate the temperatures, it must be assumed that the plasma is in LTE.

a. Boltzmann method (Excitation temperature)

There is a correlation between the transition emission (observed by a spectral line in the spectrum) and the Boltzmann equation. At 0 K, all the atoms are in their fundamental electronic configuration. In addition, at this temperature, they are considered to be motionless with absolutely no kinetic energy. As the temperature increases, they begin to move and gain more energy to access different excited configurations as can be seen in Figure 8. The distribution of this population follows the exponential nature of the Boltzmann law.

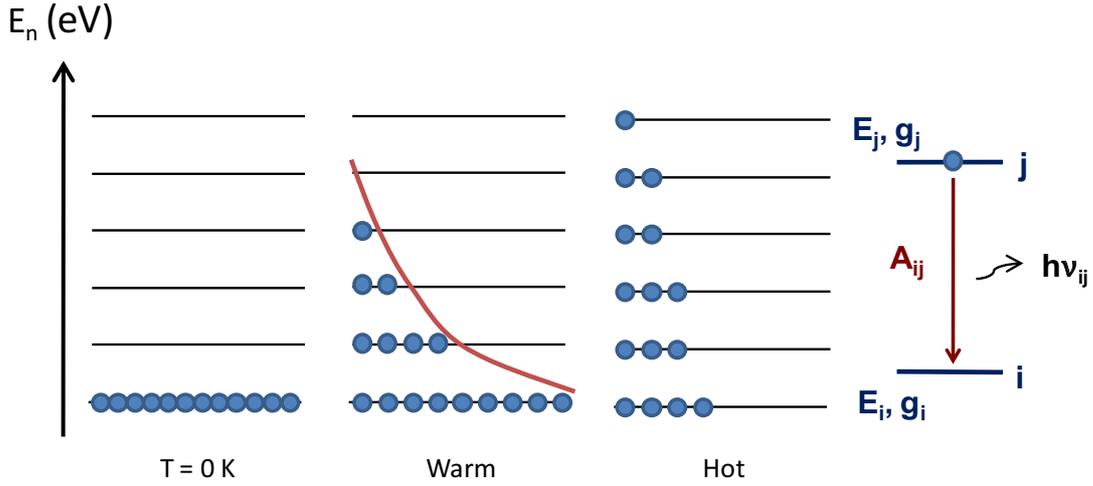


Figure 8. Boltzmann electron population distribution at different energy levels.

At a sufficiently high temperature, collisions between electrons and atoms can become efficient to excite atoms. The population distribution among the different energy levels obeys the exponential law present in the Maxwell-Boltzmann distribution [34]:

$$N_i^Z = N^Z \frac{g_i e^{-\left(\frac{E_i^Z}{k_B T}\right)}}{U^Z(T)} \quad (1.15)$$

With N_i^Z : number of particles in state i (cm^{-3})
 E_i^Z : energy of the i -th state (eV)
 g_i : degeneracy of energy level i ,
 T : absolute temperature(K)
 N : total number of particles (cm^{-3})
 Z : $Z=0$ (neutral) and $Z=1$ (singly-charged ions)

The denominator $U^Z(T)$ is known as the partition function and can be calculated by equation (1.22). Due to the difficulties associated with these calculations, we only state the equation required but we do not calculate it ourselves.

$$U^Z(T) = \sum_i g_i e^{-\left(\frac{E_i^Z}{k_B T}\right)} \quad (1.16)$$

With $U^Z(T)$: partition function

The excited atoms return to their lower energy configurations by emitting photons. Spectral lines corresponding to the transitions between these different levels can be observed in emission spectra. Each photon has a specific wavelength that is unique for every transition enabling us to detect the elements present. The line intensity can be calculated by the following formula:

$$I_{ij}^Z = \frac{1}{4\pi} h\nu N_i^Z A_{ij} = \frac{1}{4\pi} \frac{hc}{\lambda_{ij}} N_i^Z A_{ij} \quad (1.17)$$

$$\begin{array}{lll} \text{With} & I_{ij}^Z & : \quad \text{line intensity} \\ & \lambda_{ij} & : \quad \text{wavelength of the transition (nm)} \\ & A_{ij} & : \quad \text{transition probability} \end{array}$$

If the Boltzmann equation is used in the equation (1.26) the intensity can be written as:

$$I_{ij}^Z = \frac{1}{4\pi} \frac{hc}{\lambda_{ij}} \frac{N^Z g_i A_{ij}}{U^Z(T)} \exp\left(-\frac{E_i^Z}{k_B T}\right) \Rightarrow \frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} = \frac{1}{4\pi} \frac{hc N^Z}{U^Z(T)} \exp\left(-\frac{E_i^Z}{k_B T}\right) \quad (1.18)$$

In a more simplified version, the linearization of the spectral intensity line represents the wavelength intensity with regards to the different energy transitions for the same species (neutral or ionic).

$$\ln\left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}}\right) = \left(-\frac{1}{k_B T} E_i^Z\right) + \ln\left(\frac{1}{4\pi} \frac{hc N^Z}{U^Z(T)}\right) \quad (1.19)$$

If the upper excited energy levels of a given species are known, along with the statistical weights, transition probabilities and wavelength for some transitions, then the excited temperature can be deduced by measuring the emission radiances [38]. The use of transitions between largely separated levels leads to more precise results because the line fit becomes more reliable. This method is known as the Boltzmann plot (BP) method [26]. Assuming the LTE is established within the plasma, the population in different levels is governed by the Boltzmann distribution.

The value of A_{ji} is only known with a precision in the range of 10 - 30% depending on the transitions used. Some authors believe that in practice, the linearity of the plot is an

indication for the satisfaction of LTE [39, 40], while others think that a straight line can also be obtained even when the level population is out of LTE [41]. Both agree that the excitation temperature of a species within a plasma in LTE, is inversely proportional to the slope of the Boltzmann plot $\left(-\frac{1}{kT_{exc}}\right)$ shown in equation (1.29). The measurement of the line intensities is therefore a crucial factor to obtain the optimal temperature calculations using the BP method. For this, an intensity calibration of the detection system must be performed especially when the selected lines extend over a large spectral range. It goes without saying that self-absorbed or self-reversed lines will lead to inconsistent temperatures.

For more reliable results, the spectral lines should be well isolated without interference from other transitions. These conditions can be satisfied for transition metals, such as iron or titanium, or alkaline earth metals, such as calcium, for which a large number of transitions are available in the visible spectral range. The elemental metallic trace lines present in the polymer spectra are the result of contaminations that come from fillers and additives added to enhance physical properties. Their presence facilitates the temperature calculations, especially titanium, because the organic elements (C, H, O, and N) do not have enough lines for such measurements.

b. Saha-Boltzmann Plot Method

The Boltzmann plot method has the convenience of being relatively simple, but it is not very accurate. The energy gap between the upper levels of the two lines is usually small for the same ionisation degree [42]. One way to make it more precise is to significantly extend the range of E_i . This can be done by using both ionic (II) and atomic (I) elements on the same plot by combining the Boltzmann distribution with the following Saha equation:

$$\frac{N_e N^Z}{N^{Z-1}} = \frac{2U^Z(T)}{U^{Z-1}(T)} \left(\frac{2\pi m_e k_B T}{h^2} \right)^{3/2} \exp \left(- \frac{E_\infty^{Z-1} - \Delta E_\infty^{Z-1}}{k_B T} \right) \quad (1.20)$$

With ΔE_∞^{Z-1} : *correction of the ionization energy E_∞^{Z-1}*
 E_∞^{Z-1} : *ionisation energy of the lower ionisation stage*

This method, named the Saha-Boltzmann plot method, allows a significant extension of the range of E_i distributing the data from the spectra over a larger range of energy allowing for a more accurate linear plot.

The following calculations are considered for a plasma in LTE, therefore we can consider that the excitation temperature, T_{exc} , and the ionisation temperature, T_{ion} , are equal. The Boltzmann and Saha equation are then defined by a single temperature value. The result is an equation similar to that of the BP method but with a few corrections:

$$\ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right)^* = \left(-\frac{1}{k_B T} E_i^{Z*} \right) + \ln \left(\frac{1}{4\pi} \frac{hc N^0}{U^0(T)} \right) \quad (1.21)$$

With N^0 : number of neutral atoms ($Z=0$)
 $U^0(T)$: partition function for neutral atoms ($Z=0$)
 $*$: ions and electron values included

The equation is valid for atoms that are both neutral and ionised but in order to plot them on a same graph, there must be some adjustments (designated by *) made to the abscise (E_i^Z) and the ordinate ($\ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right)$) values in equation (1.28). There is a likelihood law to find atoms ionised more than once in a LIBS spectra. It is important to note that for ionised marked species ($Z \geq 1$) the quantities with the superscript * must be replaced by the following expressions.

For the abscise values:

$$E_i^{Z*} = E_i^Z + E_{ion} \quad (1.22)$$

With E_{ion} : ionisation energy

For the ordinates:

$$\ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right)^* = \ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right) - z \ln \left(2 \frac{(2\pi m_e k_B)^{3/2} T^{3/2}}{h^3 N_e} \right) \quad (1.23)$$

:

$E_{ion} E_i^Z$ Obviously for $Z = 0$ (neutral atoms), the Saha-Boltzmann equation becomes identical to the Boltzmann equation since the last term becomes equal to 0. For ionised species, this term depends on the temperature, T , therefore, an iterative procedure needs to be used to determine the temperature and plug it into the equation. In practice, only a few iterations are needed to converge.

c. Ionic Temperature

The ionic temperature also uses the combination of the Saha and the Boltzmann equation previously mentioned. In this case, the temperature can be found for an ionised species of a given element. The use of LTE conditions which states that the ionic and excitation temperatures are equal along with the transition information of two different ionisation degrees ($z = 0$ and $z = 1$) gives the following intensity ratio between ions and neutral atoms [42, 43]:

$$\frac{I_{ion}}{I_{atom}} = \frac{4.83 \times 10^{15}}{N_e} \left(\frac{g_i A_{ij}}{\lambda_{ij}} \right)_{ion} \left(\frac{\lambda_{ij}}{g_i A_{ij}} \right)_{atom} (T^{3/2})_{ion} \exp\left(-\frac{E_{ion}^Z - E_{atom}^Z}{k_B T_{ion}}\right) \quad (1.24)$$

With E_{ion}^Z : energy of the upper level of transition for ions
 E_{atom}^Z : energy of the upper level of transition for atoms

With a known ionisation energy E_{ion} and electronic density N_e (measured by Stark broadening described previously) along with the spectroscopic parameters of both these transitions, the ionic temperature can be calculated.

d. Molecular Temperature

The plasma is expected to have both ionic and atomic spectra, but of polymers (detailed in Chapter 2), molecular emission is also expected. Mostly we see spectral lines emitted by CN and C₂ which can be used to calculate the molecular temperature. The theoretical considerations in Figure 4 show that molecules are created after the atoms and ions, i.e. their spectral emission should be observed after the ionic and atomic ones, as shown by our experimental results in Chapter 3.

The vibrational temperature is determined with CN molecular band head emission present in organic materials. The vibrational temperature is extracted using the method described in ref [42]. Diatomic molecules possess many vibrational levels where molecules in the upper vibrational state v' can decay into lower electronic states v'' . They put forth the correlation between the sum of the band strengths and with the number of molecules in each respective state as can be seen in Equations (1.32) and (1.33).

$$\sum_{v''} \left(\frac{I(v', v'')}{\nu^4} \right) \propto N_{v'} \quad (1.25)$$

$$\sum_{v'} \left(\frac{I(v', v'')}{\nu^4} \right) \propto N_{v''} \quad (1.26)$$

With	ν'	:	vibrational upper electronic state (cm^{-1})
	ν''	:	vibrational lower electronic state (cm^{-1})
	ν	:	frequency (cm^{-1})
	$N_{\nu'}$:	numbers of molecules present in ν' (cm^{-3})
	$N_{\nu''}$:	numbers of molecules present in ν'' (cm^{-3})

If we consider that the LIBS plasma is in local thermodynamic equilibrium, it can be possible to calculate the number densities of an excited molecule by using Equation (1.34).

$$v_{ij} = \left(\frac{E_i - E_j}{h} \right) \ln \sum_{v''} (\lambda^4 I_{v', v''}) = C_1 - G(\nu') \left[\frac{hc}{k_B T_{vib}} \right] \quad (1.27)$$

With	λ	:	wavelength of each transition ν' and ν'' (nm)
	$G(\nu')$:	vibrational term value to ν'
	C_1	:	constant
	T_{vib}	:	vibrational temperature (K)
	$I_{\nu', \nu''}$:	measured intensity

By plotting $\ln \sum_{v''} (\lambda^4 I_{\nu', \nu''})$ with regards to $G(\nu')$, the slope provides the determination of the vibrational temperature. The results have been further confirmed by using the temperature calculation by the spectrum simulation with the LIFBASE software [37] for temperature ranging from 5000 to 20000 K. The comparison showed good agreements with a typical accuracy of 5-10%.

4. LOCAL THERMODYNAMIC EQUILIBRIUM

The first optical thinness has already been discussed. According to the talk presented by Omenetto at EMSLIBS [45], there are two essential criteria that need to be filled in order to consider the LTE. The criteria are an electronic density minimal requirement (McWhirter criterion) and the definition of a unique plasma temperature for all species in the plasma.

4.1 ELECTRON DENSITY CRITERION

For the population of excited states for the elements in the plasma to be predominately caused by collisions with electrons and not by radiative processes, it requires a minimal electron density to ensure these collisions. The radiative processes explained before such as photon absorption, recombination radiation and spontaneous emission, are in this case, negligible. This minimal electron density is known as the McWhirter criterion and is defined as [50]:

$$n_e \geq 1.6 \times 10^{12} T^{1/2} (\Delta E)^3 \quad (1.28)$$

With ΔE (eV) : largest gap between 2 adjacent energy levels
 T (K) : plasma temperature

This condition is deduced for hydrogen and hydrogen-like atoms in an optically thin, stationary and homogenous plasma [45]. The largest gap for hydrogen is indeed between the ground state and the first excited energy state and corresponds to 4 eV. This is not always the case for other elements. Figure 9 shows that the largest energy gap for oxygen does not include the ground state. In this case, the largest gap must be used.

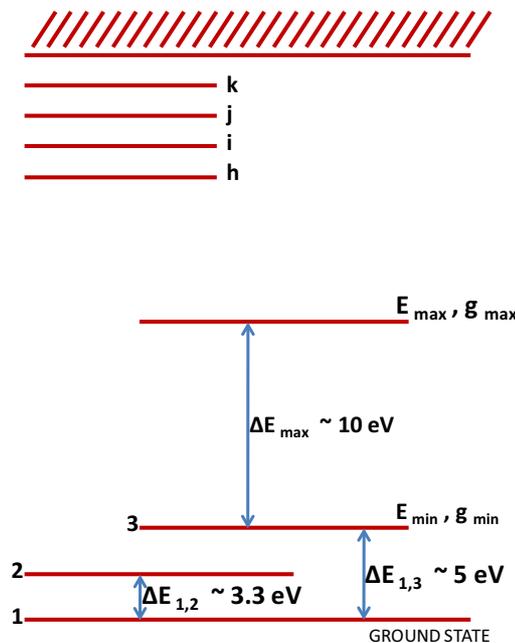


Figure 9. Oxygen energy levels [45].

Considering a plasma at 7000 K, the required electronic density comes out to:

$$N_e = 8.6 \times 10^{15} \text{ cm}^{-3} \quad (1.29)$$

This is the necessary but not sufficient requirement for the LTE that is typically fulfilled during the first stages of the plasma lifetime. Due to the large ΔE often associated with the lower lying energy states, this condition can be difficult to satisfy. A solution would be to use higher excitation levels where ΔE are close enough to allow the equation to hold, the plasma would be in partial LTE.

4.2 TEMPERATURE CRITERION

As explained previously, the plasma consists of different particles (atoms, ions, molecules, ...) in excited or fundamental states. In addition, different energy processes take place within the plasma and can each be characterised by their own temperature. Therefore, there is not clear definition of the plasma temperature unless the different temperatures converge and one can describe the entire plasma with one and unique temperature. Otherwise, the notion of plasma temperature would be confusing as it would not be clear as to 'what' temperature was in question. Four different temperatures can be used to characterise the plasma and are listed below:

➤ **Kinetic energy (Maxwell-Boltzmann distribution law) : electron temperature**

This energy is associated the motion of the electrons. In the case of plasma, it defines the energy of the electrons that obey the Maxwell distribution law i.e. the Maxwellian velocity distribution. The interactions involving electrons and other particles are not taken into consideration here, just their free motion between subsequent collisions. The Maxwell-Boltzmann distribution law is characterized by the electron temperature.

➤ **Internal energy (Boltzmann law): excitation temperature**

Elements (atoms, ions or molecules) in the plasma have the ability to access higher energy configurations as the temperature increases. The population amongst these excited configurations obey the Boltzmann law characterised by the excitation temperature.

➤ **Ionisation energy (Saha law): ionic temperature**

Saha's law is used to describe the population distribution among successive ionisation degrees within the plasma i.e. the distribution of the same element in different ionisation stages. In a typical LIBS spectrum and mainly for what concerns us, this is limited to species ionised only once.

➤ **Radiation energy (Plank's law): radiation temperature**

Plank's law describes the spectral radiance distribution from electromagnetic radiation emitted from a black body, i.e. optically thick plasma.

Each one of these processes can attain their own individual equilibrium without being in equilibrium with the rest. However, for complete thermal equilibrium (CTE), all the temperatures must be simultaneously equal, defining one and unique plasma temperature.

$$T_{e^-} \equiv T_{exc} \equiv T_{ion} \equiv T_{rad} \quad (1.30)$$

Complete LTE

For the LIBS plasma, the complete thermodynamic equilibrium cannot be attained, because one expects an optically thin plasma. As stated, the Plank law requires the plasma to be optically thick at all frequencies [24]. Optically thick plasma means the line intensities are distorted and misleading due to re-absorption of emitted photons within the plasma. The radiative life of the emitting particles and the duration of the laser pulse are much smaller than the duration of the radiative life of the plasma. So, the emission is not due to the photo-excitation mechanism but rather the thermal electron impact excitation. In a typical LIBS plasma in the typical time frame, emission from electron collisions are said to be the rate determining mechanisms dominating the radiative emission. These reasons explain why the radiation temperature is often decoupled from the others allowing the opportunity to find a temperature that satisfies Maxwell, Boltzmann and Saha laws. In this case, we cannot talk about CTE but instead we talk about LTE.

When considering a plasma in LTE, $T_{e^-} \equiv T_{exc} \equiv T_{ion}$, meaning the temperature that depicts the population of atomic and ionic energy levels is the same for the atoms of the same element in different ionisation stages. Light emission collection depends on the angle aperture of the collection apparatus. Even if a plasma is in LTE, it takes into consideration only the mean values of T and Ne, but not the spatial decoupling where the plasma has different temperatures at different spatial positions. Panne et al [46] noticed that the geometrical observation in LIBS setup generated different temperatures and electronic densities. The influence of the collection system must be taken into consideration when comparing spectral data obtained by different methods.

The use of BP method to evaluate the LTE leads to erroneous results [47] due to the lack of sufficient spectral lines used. The ionisation temperature obtained by the Saha law provides a better estimate for the real plasma properties [48]. Milan et al [49] found through spatially resolved measurements of ablation of a silicon sample, that the cooler regions near the exterior of the plasma affected the excitation temperature more than the ionic temperature. Hence, since most temperature evaluations are based on a space averaging technique, the ionisation temperature seems to provide more accurate results. They concluded that $T_{exc} = T_{ion}$ on silicon sample after 2 μ s. Others have taken another step forward by calculating the vibrational temperature when molecular spectra are detected. Hermann et al [51] deduced that the electronic temperature converged with the vibrational temperature associated to graphite samples only after 1 μ s. At this point there was a thermalisation between the electrons and heavier particles. Rusak and al [52] added the temperature associated with the rotational energy of molecular bands only to notice no LTE during the first μ s of the plasma for power densities lower than 1 (GW) / cm^2 . They did deduce the equilibrium between the excitation and vibration temperature however.

Whilst some authors like Capitelli et al [4] doubt that there can be a true existence of LTE in laser induced plasma due to the high imprecision of the partition function values. Simeonsson and al [48] estimated that kinetic equilibrium between electrons and heavier ions and neutrals was reached only 10 ns after the laser pulse impact, while the excited states needed 50 ns to reach a Boltzmann-like distribution for H, C, O and Cl. Amoruso and al [6, 7] support this hypothesis stating that the electrons could effectively transfer

their absorbed energy on to other ions during the laser pulse. The time needed for the transfer is between $10^{-10} - 10^{-11}$ s according to [53], much shorter than most laser pulse durations. They conclude that the electron temperature can be assumed to be nearly the same as the ion temperatures.

Finally, the state of LTE can be considered as a good approximation for the plasma used for LIBS measurements, generally true for delays greater than 1 μ s and over an interval in the order of 1 μ s. For polymers, the signal intensity is less intense and does not last as long as that of metals. By choosing an inadequate acquisition delay, the McWhirter criterion might be violated as there might not be enough electrons due to the weak emission signal. This hypothesis needs to be backed by proper plasma modelling or time and spatial resolved plasma detection as it depends very strongly on the properties that influence the plasma formation i.e. laser pulse energy, pulse duration, ambient gas, acquisition gate time etc. These parameters need to be accurately detailed to ensure the reproducibility of the experimental results.

4.3 INTRODUCTION TO CF-LIBS

LIBS has the ability to detect the elements present in a sample sometimes at a ppm level with little or no sample preparation required. Other techniques such as ICP-MS have this ability as well but need time for sample preparation often required to be dissolved into a liquid state. Their main advantage over the LIBS technique is their precision at quantifying the element concentrations. The task of rendering LIBS capable of somewhat similar quantitative measurements is undergoing.

Quantitative analysis can be done in several ways [54]. One method to determine the concentration of each element independently is through the emission lines of accurately known transition probabilities followed by absolute measurements of the integrated intensities. Another way would be to compare the elemental emission lines relative to the most abundant element in the sample. This method only requires the ratios of the transition probability of the emission lines. Of course, as with all information attained

from the spectra, a greater number of lines provide better results. Therefore, to increase the accuracy of these methods, a well calibrated broad band spectrum is required.

An approach widely used compares the spectra intensity lines with known calibration standards i.e. calibration curves. As long as the properties of the sample being analysed and the standard sample matrices are similar, this method can be very practical. The result of the matrix effect can be observed when comparing two known identical concentrations of an element. It would be expected to see identical emission intensities but this can be distorted due to the matrix effect. The sensitivity of the measurements are sensitively dependant on the laser-matter interaction, this includes the matrix effects.

The use of calibration curves when possible provides promising results when dealing with homogenous samples i.e. constant matrix effect [54]. However, as the sample gets more complicated, for example organic samples, the calibration curves are not always easy to create. The notion of a reference sample, where all the element concentrations are already known, for polymers can be accepted. A problem arises when dealing with more complicated samples such as vegetables or soil. How can one create a reference sample for an inhomogeneous sample or a sample with unknown matrix? Add to this the complications of varying matrix effects associated to the local in homogeneity of such samples. In this case, the reproducibility of LIBS is therefore reduced and the errors associated with the measurement increase drastically. Construction of a calibration curve is highly complicated and nearly impossible.

The notion of Calibration-Free LIBS (CF-LIBS) was proposed by Ciucci et al [55]. Such procedure can in the ideal case, provide solution for inhomogeneous sample or those with unknown matrix. The idea behind this method was to go around these matrix effects by making each measurement self consistent i.e. avoid the need to compare it with calibration curves. For this method works, these are three requirements for the ablation and the resulted plasma [54]:

- Stoichiometric ablation
- Plasma must be in LTE
- Optically thin plasma

The last two requirements have already been discussed previously. Stoichiometric ablation means that the composition of the ablated mass is representative of that of the sample. This is the basis of LIBS and other laser ablation techniques in general. This allows to characterize the sample analyzed using the light emitted from the plasma. Russo and al. mentioned in their paper in 1991 [56] that stoichiometric ablation existed for laser power densities greater than 10^9 W/cm^2 , a condition generally satisfied for nanosecond or shorter duration pulse ablation. Borisov and al [57] stated that the degree of stoichiometric ablation was reduced when the crater depth increased. He put forth two hypotheses for this statement. The first imagined that the geometric factors reduced the power density and the second consisted of shielding and thermal variations within the crater.

Elemental fractionation of the sample was observed with longer laser pulses [2]. Femtosecond ablation reduced this problem and showed an increase in the ablation efficiency. Russo et al. 2002 [58] performed nanosecond and femtosecond ablation on glass samples and did not come to the same conclusion. They believed that the pulse duration alone was not enough to account for fractionation. The generated plasma and the shockwave induced with the plasma could influence it as well. The mass removal depends on the heat conduction of the laser pulse within the sample as well as the recoil pressure, which is a consequence of the shockwaves. Shockwaves present in femtosecond laser ablation are due to the high velocity of the ejected mass.

For the CF-LIBS calculations, the use of the Boltzmann plot method was first used (includes the total plasma number density and volume and the experimental geometry) [54]:

$$\bar{I}_{ij} = F C^s A_{ij} \frac{g_i \exp\left(-\frac{E_i}{k_B T}\right)}{U^s(T)} \quad (1.31)$$

With \bar{I}_{ij} : measured integral line intensity (counts)
 C^s : concentration of the emitting species S
 $U^s(T)$: partition function of the emitting species
 F : optical efficiency of the collection system.

Equation (1.32) requires information either protruding from spectral databases or directly from the experimental category.

$$\begin{array}{ll} A_{ij}, g_i, E_{ij}, U^S(T) & \rightarrow \text{spectral database} \\ F, C^S, T & \rightarrow \text{experimental data} \end{array}$$

If y, x, m and q^S are introduced as the following:

$$y = \ln\left(\frac{\bar{I}_{ij}}{g_i A_{ij}}\right) \quad (1.32)$$

$$x = E_i \quad (1.33)$$

$$m = -\frac{1}{k_B T} \quad (1.34)$$

$$q^S = \ln\left(\frac{C^S F}{U^S(T)}\right) \quad (1.35)$$

The Boltzmann equation can be written simply in the form:

$$y = mx + q^S \quad (1.36)$$

Where m is associated with the temperature of the plasma and q^S is associated with the concentration of the emitting species S . One spectral line emission produces one point on the Boltzmann plot diagram. This way, all the emitting species can be represented in one diagram. If the condition of LTE is fulfilled and the lines are carefully selected to avoid self-absorption, each emitting species should align on parallel lines. The same temperature means the same slope m for all elements. These calculations require that the spectral data contain:

- At least 2 emission lines from the same species to determine the electronic temperature
- At least 1 emission line with known spectroscopic parameters to calculate the electronic density.

The y-intercept given by q^S of each line is proportional to the natural logarithm of its concentration. Once the electronic temperature is calculated, the concentration of species of the corresponding element, which sums up the concentrations of both the neutral and single ionised atoms, can be determined. The F parameter creates a bit of a problem but is solved by using the normalisation relation if an internal standard is not known.

$$\sum_e C^e = 1 \quad (1.37)$$

where the sum of the total element concentration should be 1. Even if in theory, this method can provide very accurate results, the LIBS measurements with CF-LIBS are only at best 3-5% experimentally reproducible [59] and the errors associated with the probability transitions can go from 10-30% [54]. If more spectral lines were used, this could average out the uncertainties linked with the probability transitions and provide more accurate line plots.

Initially, CF-LIBS introduced by Ciucci [54] excluded self-absorbed lines. However, it is possible to include these lines by using a recursive method. This means the use of more points (more spectral lines) for even more accurate information. This method predicts the intensity of the self-absorbed line. The scattering around the best fit line can be reduced, putting forth evidence that it is mainly due to self-absorption effects and not experimental errors, fitting errors or even uncertainties related to transition probabilities.

5. CONCLUSION

The emission line intensities depend on the plasma properties that are influenced by the ablation laser pulse (wavelength, pulse duration, pulse energy), the sample (matrix effects, inhomogeneities) and the ambient gas (vacuum, inert, air and the pressure). It can also be related back to the concentration of the elements present within the plasma. A detailed understanding of the laser-induced plasma is the basis of high quality quantitative measurements with LIBS. A thorough knowledge on evaporation, ejection of particles, hydrodynamic explosion, shock waves, plasma initiation and expansion not to forget the laser-plasma interaction is needed. All these parameters can influence the amount and the composition of the ablated mass.

In the following chapters of this thesis we will present the study of such parameters for the ablation of different polymer samples. First, we study the plasma formation and expansion by shadowgraphic measurements under different ablation regimes. This is followed by time-resolved analysis of the spectroscopic plasma emission for the optimisation of experimental acquisition parameters (local thermodynamic equilibrium, signal-to-noise ratio, acquisition gate width etc). We are interested in CF-LIBS because it is the only method that allows a direct analysis without the use of calibration curves. Before applying this method however, certain notions and understanding of the physics behind the induced plasma, especially for organic substances, is needed. Therefore, the last chapter has for aim to classify and identify various polymers through different multivariate analysis techniques. It emphasises that the work in this thesis is more qualitative than quantitative. The aim behind this reasoning, is to be able find an application for the industrial sorting of plastics with what we already know today. In the future, we hope that the additional information presented here of plasmas induced on polymers will allow us the ability to use calibration free methods to provide more precise quantitative measurements.

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CHAPTER II EARLY STAGE EXPANSION OF A LASER-INDUCED PLASMA FROM A POLYMERIC MATERIAL BY TIME-RESOLVED SHADOWGRAPHY

1. OPTICAL SHADOWGRAPHY TECHNIQUE

The main aim of this chapter is to study the laser interaction with both the target, in our case a polymer, and the generated plasma. The shadowgraphy technique can put forth a different amount of information regarding the expansion of the generated plasma. However, here we shall essentially focus on how the laser interacts with the plasma in order to help us choose the appropriate laser ablation regime for our future study regarding emission spectroscopy of the induced plasma. The work reported in this chapter was done with the collaboration of the research group “Laser Spectroscopy and Applied Materials Group” led by Dr. Richard Russo in the Environmental Energy Technologies Division at the Lawrence Berkeley National Laboratory (LBNL) in Berkeley, California, U.S.A. The experimental setups used to obtain shadowgraph images are presented along with their experimental procedure.

We studied the expansion of the induced plasma into the ambient air at atmospheric pressure in three different laser ablation regimes. The used target is a polyamide (nylon sample):

- Ultraviolet nanosecond (wavelength: 266 nm, pulse duration: 4 ns FWHM)
- Infrared nanosecond (wavelength: 1064 nm, pulse duration: 4 ns FWHM)
- Infrared femtosecond (wavelength: 800 nm, pulse duration: 100 fs FWHM)

Shadowgraph images of the induced plasma for each of these regimes were taken for pulse energy ranged from 1 to 15 mJ for nanosecond pulses, and from 1 to 4 mJ for femtosecond ones. They will be compared in terms of shockwave expansion velocity, plasma screening and morphology. With this information we shall explain the reasons behind our choice in ablation regime for our time-resolved spectroscopic measurements in the next chapter.

During the plume expansion, the resulting shockwave has consequences on the surrounding environment by creating disturbances that are known to refract light rays, being observed in the form of shadows [8]. The disturbances change the refraction index causing the light rays to be deviated within a transparent media (in our case air) [11]. A simple demonstration of how the shadowgraph technique works is shown in Figure 1. In the absence of a dense but not opaque object (sometimes called a schliere that bend light in any direction other than the normal producing gradient disturbances of light), the light beam shown on the screen is uniform. With a dense schliere placed between the light source and the screen, otherwise regular light rays are now bent, refracted and cross amongst each other. The refractive angle ϵ puts forth these irregularities and the image on the screen shows concentrations of light in certain areas while a total absence of illuminance in others [8]. A simple everyday example of such a phenomenon is also shown in Figure 1. A frying pan illuminated from above puts for the difference in gradients between the oil drops and water present. As the light refracts, it creates a shadowgram.

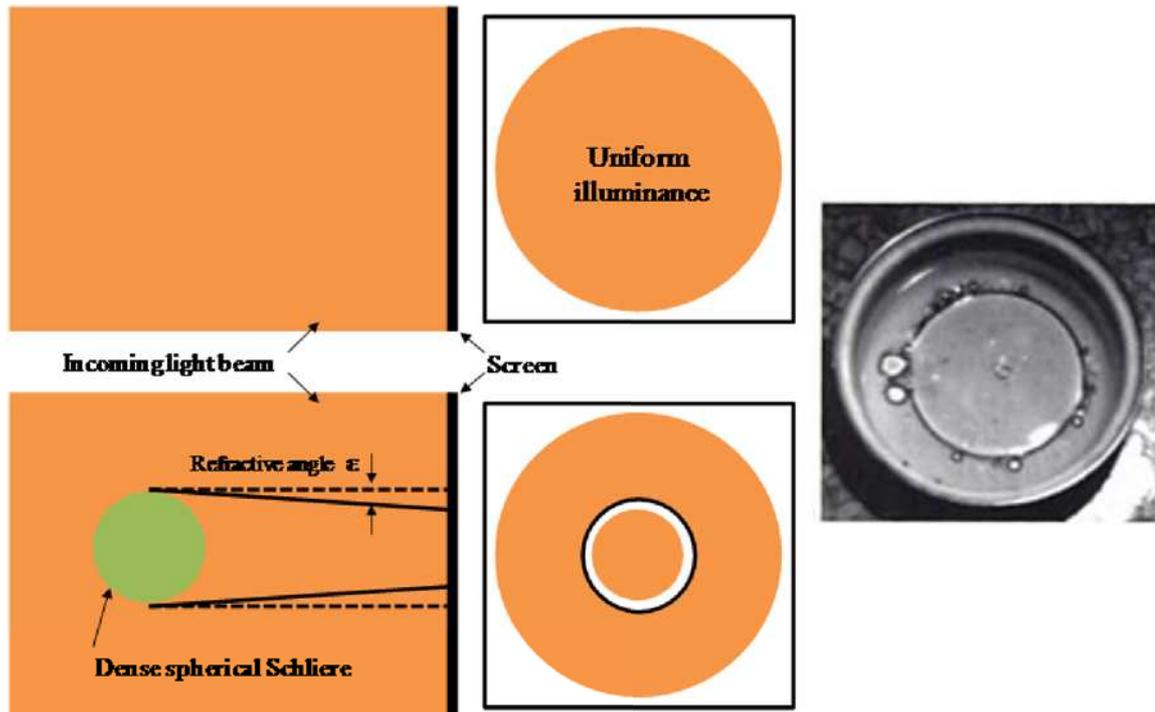


Figure 1. Shadowgraph principle puts forth the consequence of light refraction due to a dense media on a screen.

For our optical shadowgraphy method, we used two laser beams: one to ablate the target and the other to obtain time-resolved shadowgraph images. The nylon sample (target) was placed so that the ablation beam hits the surface at a perpendicular angle intersecting the probe beam that was parallel to the surface, as can be seen in Figure 2. As the imaging beam goes through the plasma, certain rays are reflected due to the difference in gradient caused by the presence of the plasma. A CCD camera coupled to a computer is used to take images (pictures) as the screen shown in Figure 1 to present the results in the form of shadowgram pictures. They provide a visual aspect of the laser-target and laser-plasma interaction. In certain cases, we can even see the direct interaction between the laser and the surrounding air. An X-Y-Z micro-displacement stage was used to accurately move the sample horizontally between two different ablation shots. It is also used to place the sample at the same position for all duration of the experiment by its vertical displacement to minimise the experimental fluctuations.

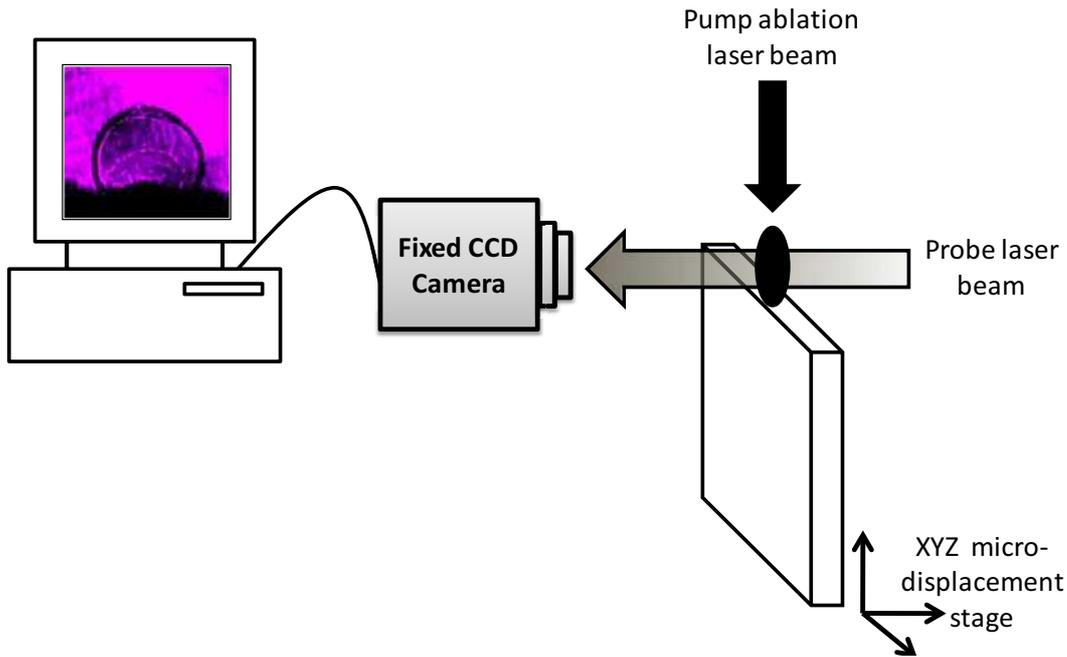


Figure 2. Position of the nylon sample with regards to the incoming laser beams to produce shadowgrams. Image is not to scale [1].

The nylon samples were cut from a large nylon block of dimension $30 \times 15 \times 2 \text{ mm}^3$ to make sure that they have all the same chemical composition. Thinner samples produce higher quality images due to less light diffraction on the surface. This is why the ablation beam was focused on the thin-edged surface of the sample (Figure 2). Craters were separated with a gap of 0.3 mm to avoid any overlapping. The nylon samples were cleaned with a methanol solution to remove any impurities from their surface.

Upon laser impact, the pulse energy is transferred to the sample, which heats the target to the point where a vaporised cloud of excited atoms and ions is created on the surface. The vaporised plasma starts off with extremely high temperatures and electronic density pushing the ambient gas away from the target surface with a high velocity [2]. This forms a densely packed area called a shockwave, which propagates at a speed that exceeds the sound velocity (supersonic velocity). The compression due to the shockwave can be observed as an exterior dark ring in the shadowgraph image in Figure 2 with the plasma confined within it. With time, the plasma reaches the shockwave front. An internal shockwave can also be observed moving back to the target surface. When it reaches the target surface, it is reflected back and undergoes oscillations between the external shockwave and the target surface. [3]

1.1. EXPERIMENTAL SETUP

1.1.1. PROBE AND CAMERA

Femtosecond laser pulses are known to interact only with the target sample, while nanosecond ones with longer pulse duration interact with the created plasma. In order to observe the evolution of the plume, time-resolved measurements are recorded by taking images at different delays after the laser impact on the target. The arrival times of probe and pump pulses on the sample are shifted experimentally determining the delay time at which a shadowgram was taken. Different ablation regimes (nanosecond or femtosecond) requires a specific method to delay the pulses. For more accurate time-resolved measurements, both methods used a femtosecond laser as the probe. A Ti: Sapphire femtosecond laser (Spectra-Physics, TSA) with a pulse duration of 100 fs at full width at half maximum (FWHM) and with a fundamental wavelength of 800 nm, provided the probe beam.

The used CCD camera (Nikon D200) is sensitive to visible light so a KDP (Potassium Dihydrogen Phosphate) crystal was used to double the femtosecond beam to attain a wavelength of 400 nm. A narrowband filter around this wavelength was placed at the front of the camera to minimise the optical noise by selecting the light protruded by the passage through the plasma. It allows the formation of images with higher contrast.

1.1.2. NANOSECOND ABLATION SETUP [10]

Here, a Nd:YAG laser (New Wave Research, Minilase II) was used to provide the ablation pulse of the nylon surface. We used the fundamental wavelength at 1064 nm (infrared) and 4th harmonic at 266 nm (ultraviolet). The pulse duration at FWHM was 4 ns and its energy that can be varied between 1 and 15 mJ. A beam splitter allowed a part of the beam to be deflected into a power meter to monitor the pulse-to-pulse energy fluctuations. The main laser beam was focused on the sample using a quartz lens as shown in Figure 3. When the laser parameters were changed (energy, wavelength), the position of the sample's surface was adjusted to get the same laser spot diameter,

ensuring the same irradiance for same energy. The measurement of the focused laser spot was performed using a white-light interferometric microscope (New View 200, Zygo). The focused beam was positioned to make craters that were about 100 μm in diameter large.

The femtosecond laser provides the probe beam. The delay between the probe pulse and the ablation pulse is adjusted using a delay generator. Two photodiodes coupled to an oscilloscope monitored the two pulses, which allowed a precise measurement of the actual delay between the two pulses. The CCD camera takes images corresponding to different time delays for time-resolved studies of the plasma expansion in ambient air.

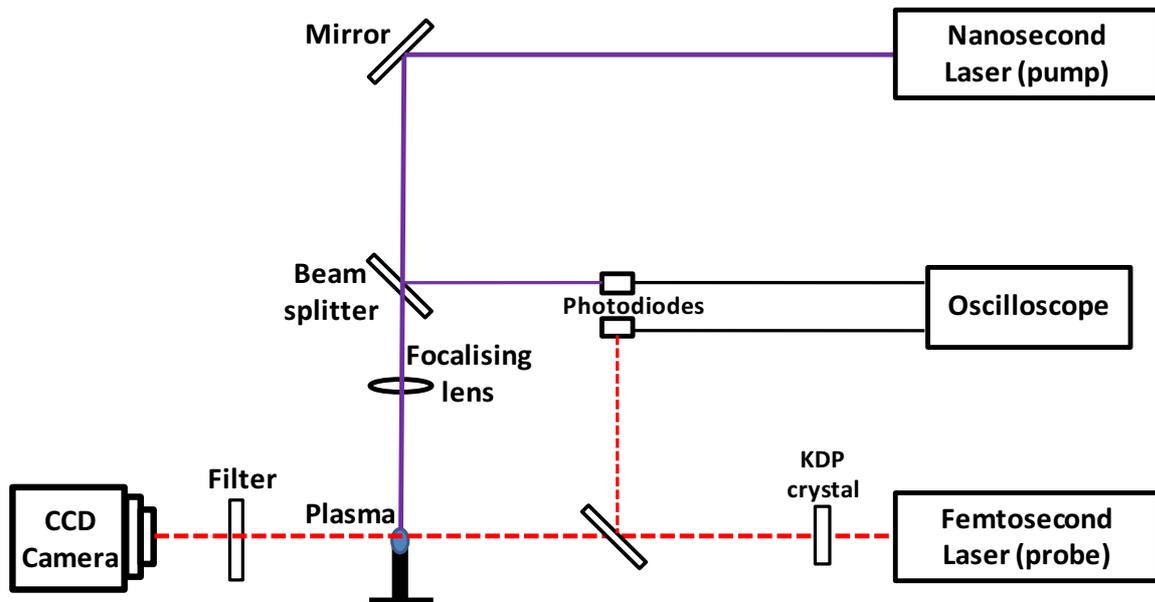


Figure 3. Optical shadowgraph setup for nanosecond ablation.

Determination the delay between the ablation and the imaging pulses

A delay generator (Stanford Research, DG 535) was used to adjust the time delay between the ablation and the imaging laser pulses by synchronising the Q-switches of the nanosecond laser and that of the amplifier of the femtosecond laser. However, since the oscillator of the femtosecond chain is not synchronised with its Q-switched amplifier, the ejection of an amplified femtosecond pulse can not be more precise than the interval

between two successive pulses. The femtosecond oscillator operated at 80 MHz, which corresponds to an interval of 12.5 ns between the successive pulses. The actual delay between the two laser pulses is the difference in time when they actually arrive on the sample or the plasma. For precise determination of this delay, two photodiodes were used to detect the light scattered from the two respective pulses after emission. Since there can be a difference between the time fixed on the Stanford delay generator and the real emission times of the two pulses, an oscilloscope (Tektronix DSA 602A) recorded the signals from the two pulses simultaneously. This allowed us to determine the exact delay between the two pulses every time for our time-resolved measurements as can be seen in Figure 4. For our delay measurements, we calculated $T_{fs} - T_{ns}$ and since the femtosecond pulse is much narrower compared to the nanosecond one, we can observe certain negative values in our shadowgraph images for the time delay.

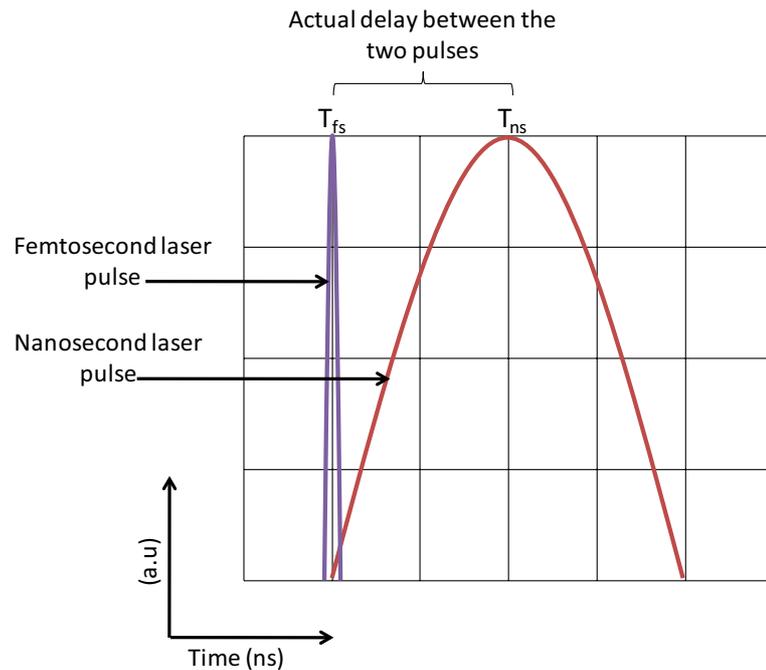


Figure 4. Oscilloscope representation of the delay between the two laser pulses.

However, slight corrections were added to this measurement. The first is due to the 12 cm difference between the cable lengths linking the photodiodes to the oscilloscope. We calculated the propagation speed of the electric pulse at 2×10^8 m/s. The delay generated by this difference in length is therefore 0.6 ns.

There was a second correction related to the different laser beam optical trajectories. The Stanford generator only controls the delay between the emissions of the pulses (as mentioned before with an uncertainty of 12.5 ns). However, if the two laser paths do not follow the same optical path, they will not reach each other at the same time. In fact, this is what we play with to have time-resolved images but we need to control all the parameters to be sure of our measurements. This difference in optical trajectory was 222 cm generating a delay between the two of an additional 7.4 ns (considering the speed of light 3×10^8 m/s).

Therefore, by combining the delay determined by the photodiodes coupled to the oscilloscope and laser to target or to plasma trajectories, an additional delay of 8 ns needs to be added to the pulses observed on the oscilloscope in Figure 4. For example, if the delay generator is fixed for a 10 ns delay between the femtosecond and nanosecond laser pulses, we can expect to see a delay of $((10 \pm 12.5) + 8)$ ns between the two pulses. The fluctuations of this delay can be directly observed with the oscilloscope providing us with the “actual delay”.

1.1.3. FEMTOSECOND ABLATION SETUP

In this case, the femtosecond laser beam is split into two: ablation and imaging. One part remains at the fundamental wavelength and is used as the ablation beam to induce the plasma on the polymer surface. The other is used at 400 nm as the imaging beam in the same manner as the nanosecond ablation setup previously explained. The femtosecond pulse used at its fundamental wavelength of 800 nm and always hits the sample at a perpendicular angle (Figure 4) with an energy variation between 1 and 4 mJ. The laser was focused on the sample, in the same manner as before, using a quartz lens leading to a spot of diameter of about 100 μ m. A beam splitter allowed part of the beam to be deflected onto a power meter to monitor the pulse-to-pulse energy variations. The CCD camera is used again to take shadowgrams at different time delays.

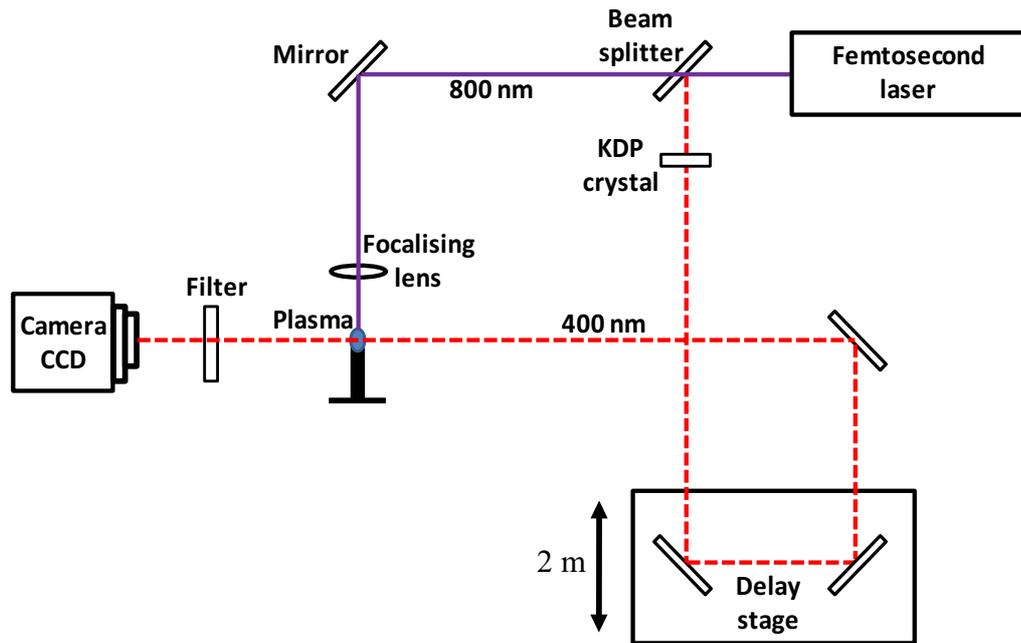


Figure 5. Optical shadowgraph setup with femtosecond pulse ablation.

Delay determination

Previously, for the nanosecond ablation regime, the ablation and the imaging beams were emitted from two different lasers and a delay generator was used to control the delay. This method is not used here as only the same laser is used for both the ablation and imaging pulses. The imaging pulse is deferred by the imaging pulse through means of an optical delay stage. The delay stage consists of movable stage with two mirrors carefully aligned. By moving this delay stage, the distance travelled by the imaging pulse changes allowing us to obtain time-resolved measurements. The additional distance travelled by this pulse is calculated, providing the time delay information required.

Obviously, the delay that can be generated is limited by the total additional distance travelled by the imaging pulse, 2 m in the case of our experiments (further fine alignment adjustments can be required when displacing the stage). With this constraint, the imaging pulse could only be deferred by an additional distance of 4 m so that we could only study the plasma expansion between 0 and 11.5 ns after laser impact on the target in the femtosecond ablation regime.

2. EARLY STAGE PLASMA EXPANSION: OBSERVATION AND INTERPRETATION

As we shall see in this section, there are obvious differences between the different ablation regimes. Due to the difference in the pulse durations, the laser-plasma interaction is more obvious in the nanosecond ablation regime. We shall first start by presenting the different shadowgrams in the nanosecond regimes for infrared and ultraviolet ablations followed by the femtosecond ablation shadowgrams.

2.1. NANOSECOND ABLATION REGIME

In this regime, the laser pulse is long enough to have a significant interaction with the generated plasma. At first, the laser hits the target surface and excites its electrons until a certain temperature, where irradiated matter undergoes phase transition and is ejected violently in the direction of the laser beam [9]. A shockwave is initiated creating a thin layer of very condensed air around the plasma. If the plasma critical density value is attained, then we should observe plasma shielding reflecting the laser beam back. The critical density value is much lower for infrared than ultraviolet.

2.1.1. INFRARED REGIME

Shadowgraph Images

Shadowgraph images have been recorded for various energies of 1, 2, 3, 4, 5, 7, 10, 15, and 20 mJ for different delays after the laser pulse impact, but the three energies presented (1, 7, and 15 mJ) are enough to depict the general plasma characteristics. In our case, this corresponds to a fluence of 12.7 J/cm^2 , 88.9 J/cm^2 , and 190.5 J/cm^2 in figures 4, 5 and 6 respectively.

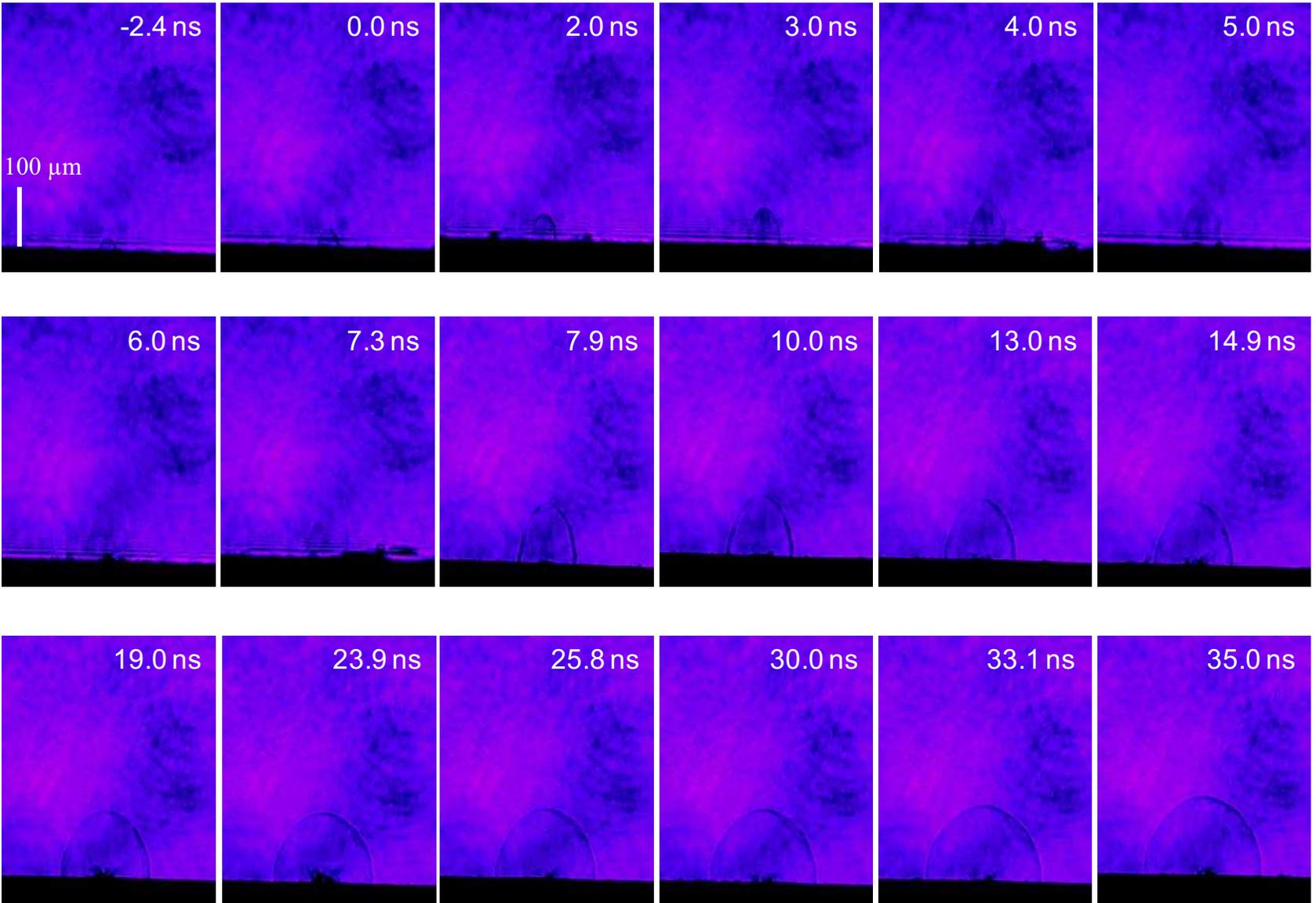


Figure 6. Shadowgrams for ablation with 1064 nm nanosecond pulse at $12.7 J/cm^2$

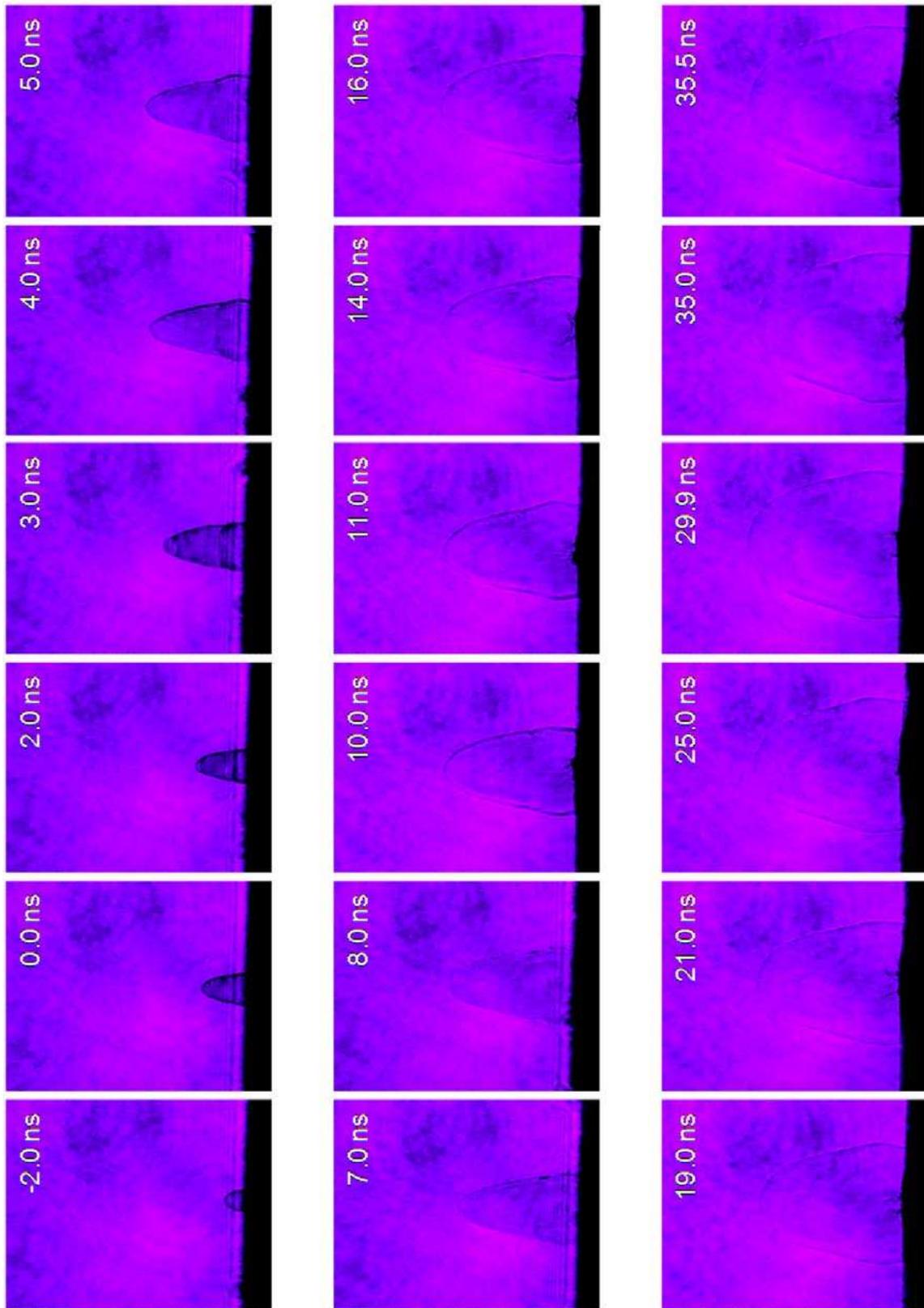


Figure 7. Shadowgrams for 1064 nm nanosecond ablation at $88.9 J/cm^2$

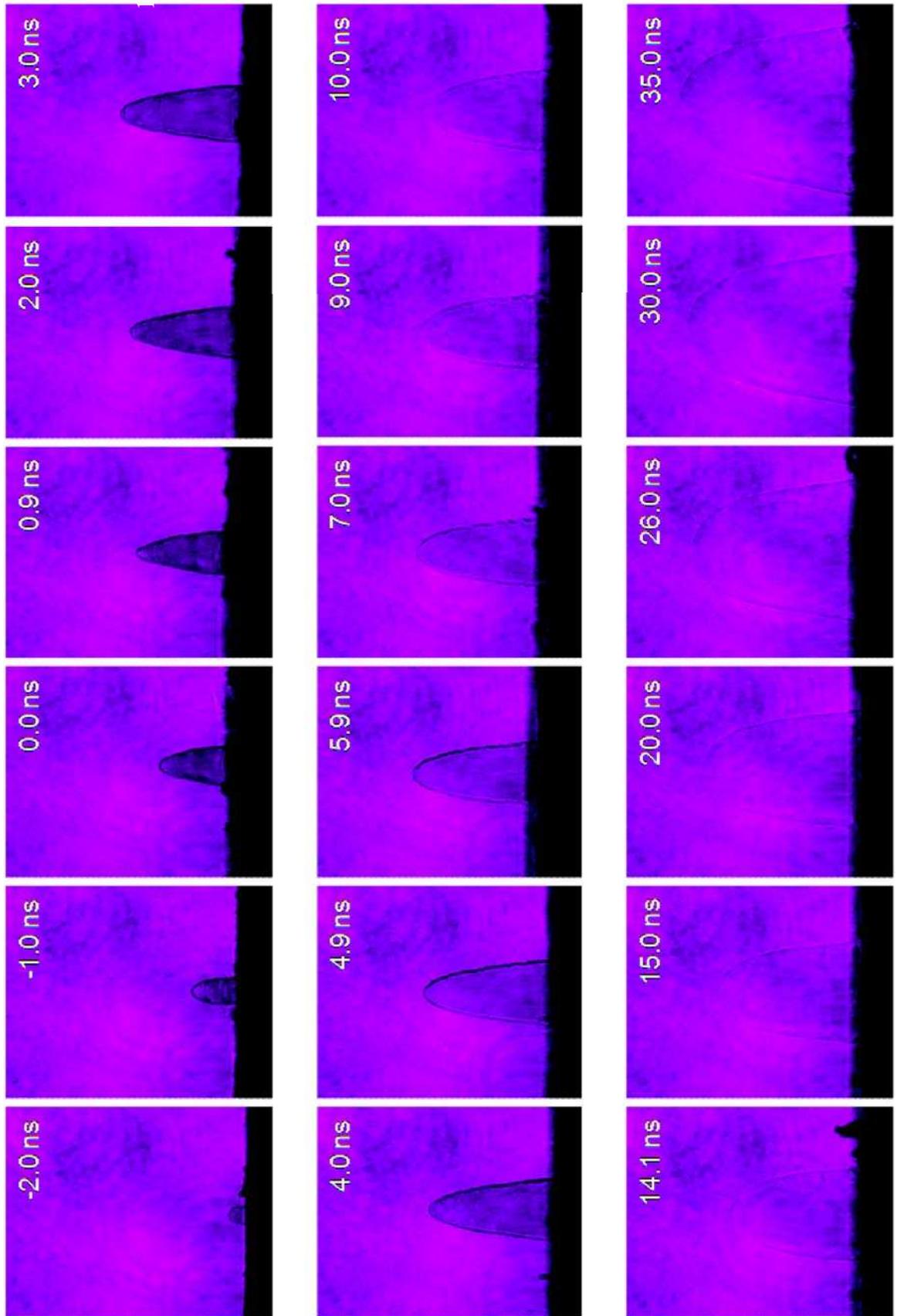


Figure 8. Shadowgrams for ablation with 1064 nm nanosecondpulse at $190.5 J/cm^2$

As expected, the infrared regime presents a strong absorption of the laser pulse at its interface even at low fluence. This phenomenon is observed by the presence of two components in the expanding shockwave, which can be seen in Figure 9. The main component, spherical in shape, is best observed at short delays (represented by the dotted line). The second component, known as a laser-supported detonation wave (LSDW), is observed on top of the first one (represented by the plain line). This wave is the consequence of the absorption of the laser energy leading to a second breakdown process on the tip of the shockwave. The result is an elongated anisotropic shockwave expanding towards the axis of the incoming laser pulses. This anisotropy becomes more pronounced as the laser pulse energy increases due to greater plasma absorption of the laser pulse tail [1, 2]. The interface of the external shockwave has the highest electron density due to the compressing the surrounding gas by the ablated vapour. This particular point can reach values that are considerably higher than the average plasma density [2, 3]. This explains the observed laser supported detonation wave on the tip of the shockwave even at early delays.

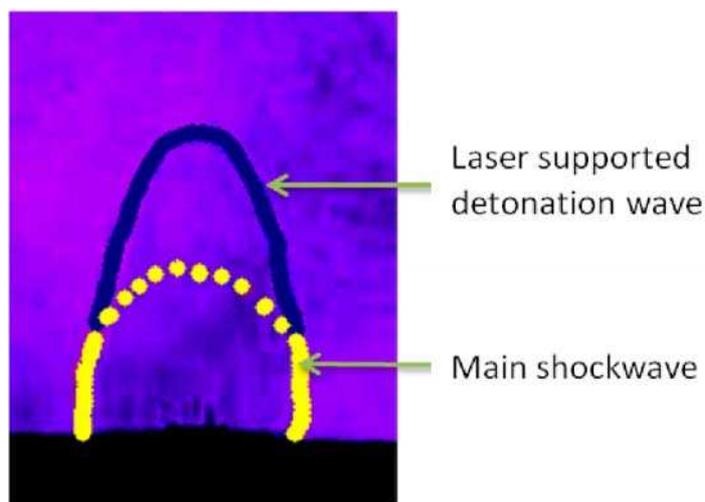


Figure 9. Laser Supported Detonation Wave of a plasma induced from a nylon sample in air for infrared nanosecond ablation. [1]

The coupling between the incoming laser pulse and the sample is diminished as the plasma front absorbs the incoming laser pulse, preventing it from reaching the target surface. A part or all of its energy, depending on the extent of the absorption, is

transferred to the plume. Total shielding only occurs when the plasma has exceeded the critical electron density and starts to act like a mirror reflecting the laser pulse energy instead of absorbing it. This critical density has already been defined in Chapter I [12, 13] and reminded in Equation (1.2). We do not observe total plasma shielding as even at our highest fluences we do not attain such values.

$$N_c = \frac{10^{21}}{\lambda^2} \quad (2.1)$$

$$\begin{array}{lcl} \text{With} & \lambda_{1064 \text{ nm}} & \Rightarrow N_c \approx 9.0 \times 10^{20} \text{ cm}^{-3} \\ & \lambda_{266 \text{ nm}} & \Rightarrow N_c \approx 1.4 \times 10^{22} \text{ cm}^{-3} \end{array}$$

Chapter I also mentioned that for longer wavelengths, the main laser absorption mechanism is inverse Bremsstrahlung. For this mechanism, only one photon is absorbed. It means that it has a linear relationship with the laser intensity as opposed to a multi-photon ionisation that depends on the number of m photon absorbed (I^m). We believe therefore that the laser pulse is absorbed by a greater surface area instead of being concentrated in the centre.

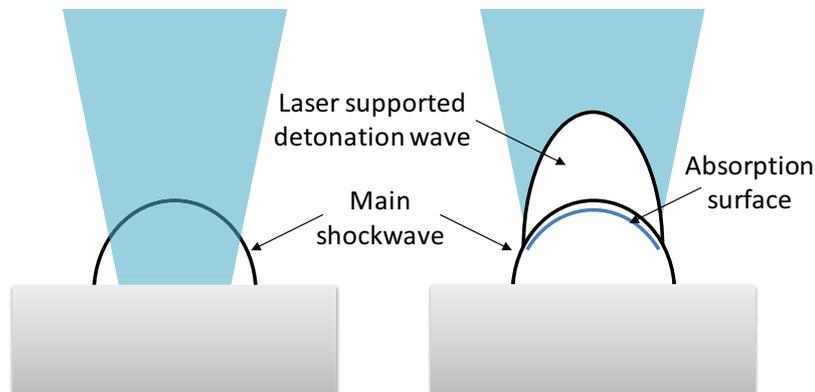


Figure 10. Absorption of the income infrared laser beam by the plasma induced on a nylon sample.

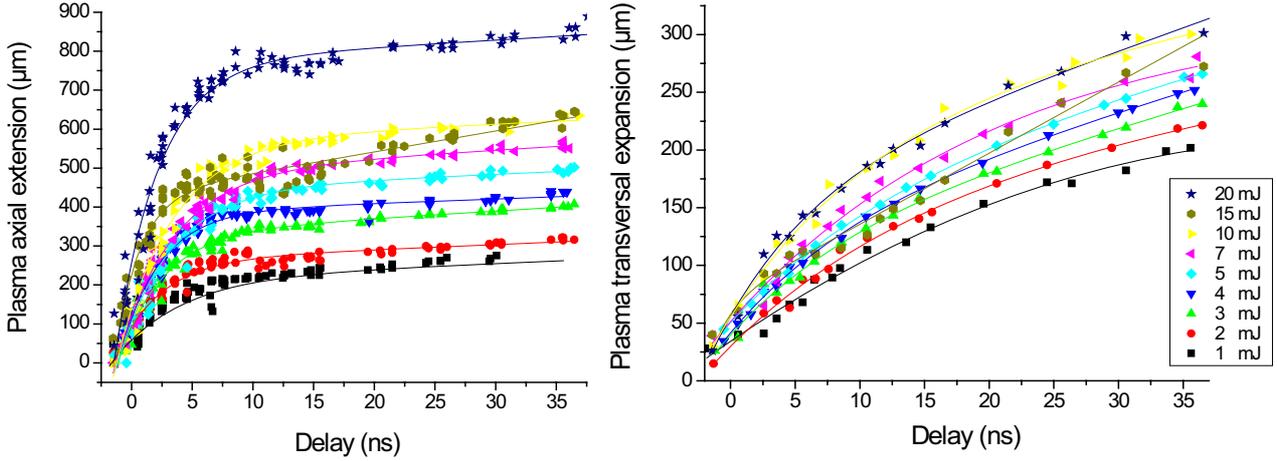


Figure 11. Axial and transversal expansion of the plume.

For a quantitative study of the expansion, the height and the width of the shockwave expressed in μm are presented in Figure 11. The radial expansion of the shockwave is the total width of the observed plasmas in figures 6, 7 and 8 divided by 2. The predominant axial expansion is obvious when comparing the two graphs. The axial expansion starts with a very high growth spurt and then levels out. The transversal expansion is more uniform with smaller expansion velocities. The difference in the two graphs put forth the dominant axial expansion associated with the anisotropic plasma.

The curves for different energies follow the same tendencies except for the curve at 15 mJ where a few abnormalities can be noted. The axial expansion graph starts off with an exponential growth followed by a linear expansion. The great expansion speeds observed at short delays (< 5 ns) correspond to the laser supported detonation wave effect on the plasma. From the data presented in figure 11, we can extract the axial expansion velocity as a function of the delay and for different laser energies by fitting the graphs with the equation below:

$$y(t) = \left(1 - \exp\left(\frac{t-t_0}{\tau}\right)\right)(ct + d) \quad (2.2)$$

The parameters t_0, τ, c, d are adjusted to fit the experimental data. The derivative of equation (2.2) was calculated and the expansion velocities are represented in figure 12 in function of the delay after laser pulse impact on the sample. It is important to state here

that this method remains preliminary and a more detailed model is necessary to understand the physics behind this phenomenon. Equation (2.2) simply allows us to have an idea of the expansion velocity.

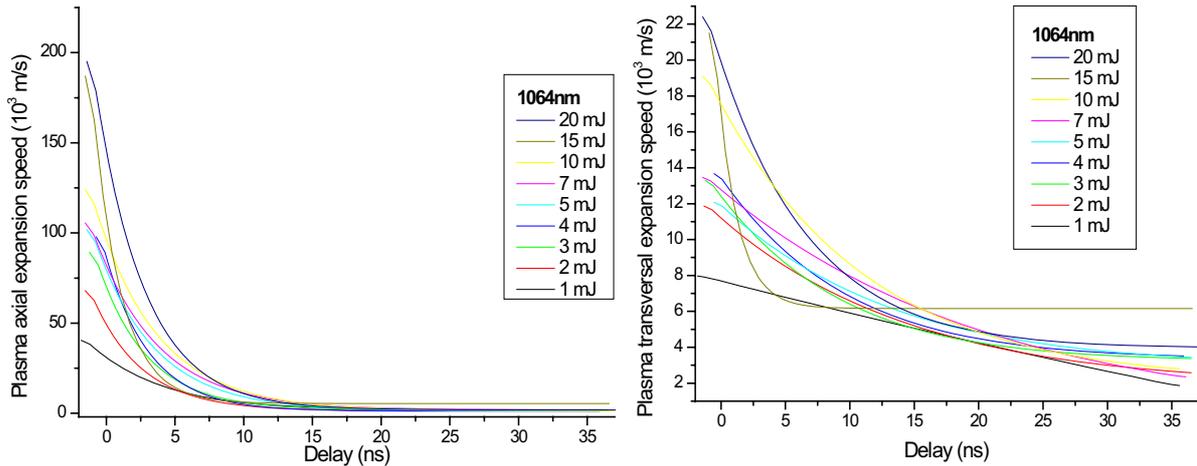


Figure 12. Calculated axial and transversal expansion speeds.

The axial velocity starts out extremely fast (between 5 to 20×10^4 m/s) and drastically slows down after $5 - 10$ ns. After 3 times the value of τ , a uniform expansion can be observed. Considering the velocity is the derivative of the equation (2.2), the graphs will obviously tend towards c (around 2×10^3 m/s except for 15 mJ where c tends towards 5.5×10^3 m/s). We note that the velocities are much higher than the speed of sound (344 m/s in the atmosphere at 20 °C), proving the agreement with the shockwave definition. The transverse velocity follows a different trend. Apart from the graph at 15 mJ that tends towards a certain c value, the other curves do not follow this behaviour. The velocities continue to decrease even after 10 ns. [2]

2.1.2. ULTRAVIOLET REGIME

Shadowgraph Images

The following images were obtained with the shadowgraph procedure presented above. The wavelength of the ablation pulse was 266 nm provided by a quadrupled Nd:YAG nanosecond laser. We obtained images for various energies (1, 2, 3, 4, 5, 7, 10, and 15 mJ) in the nanosecond ultraviolet regime, but again the following energies, 1, 7, 15 mJ, are enough to depict the general plasma characteristics. They correspond to fluences of 12.7 J/cm^2 , 88.9 J/cm^2 , and 190.5 J/cm^2 in figures 12, 13 and 14 respectively.

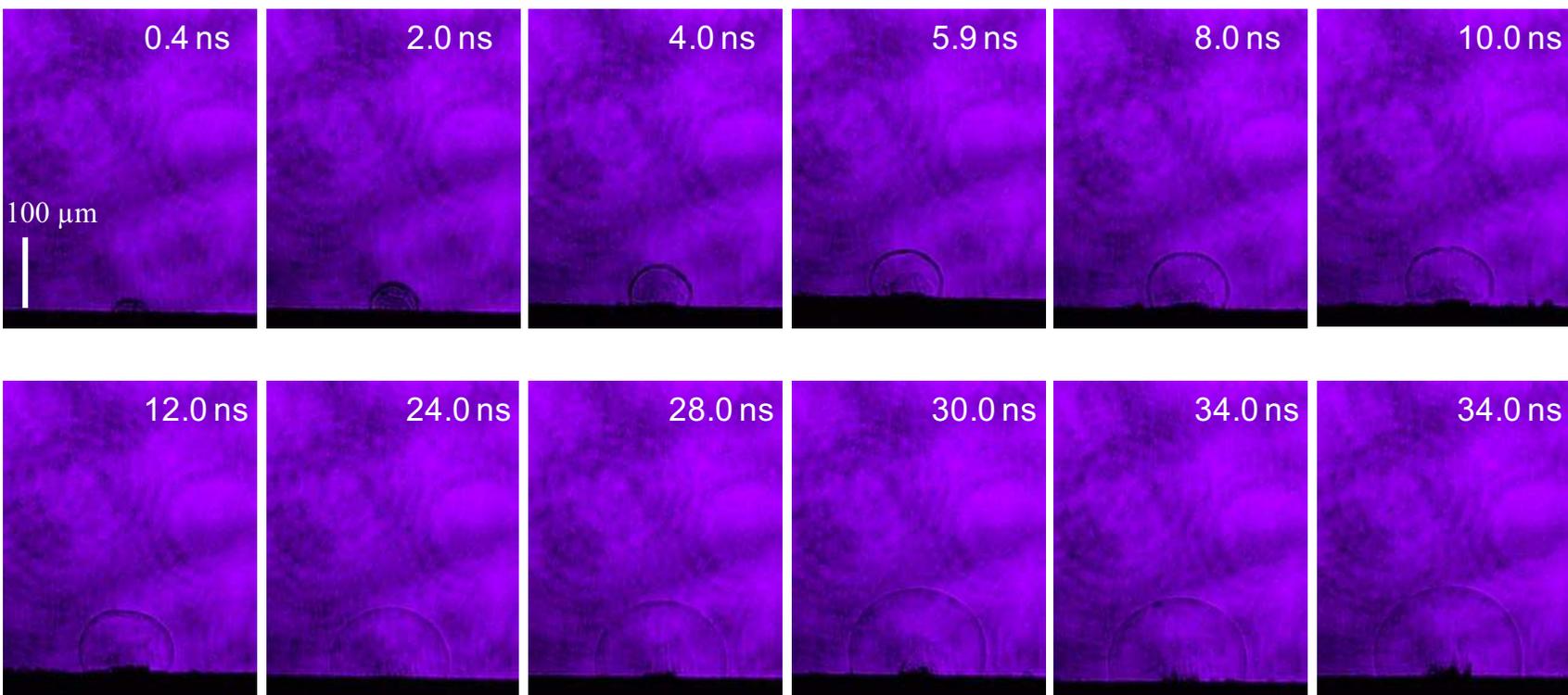


Figure 13. Shadowgrams for 266 nm nanosecond ablation at 12.7 J/cm^2

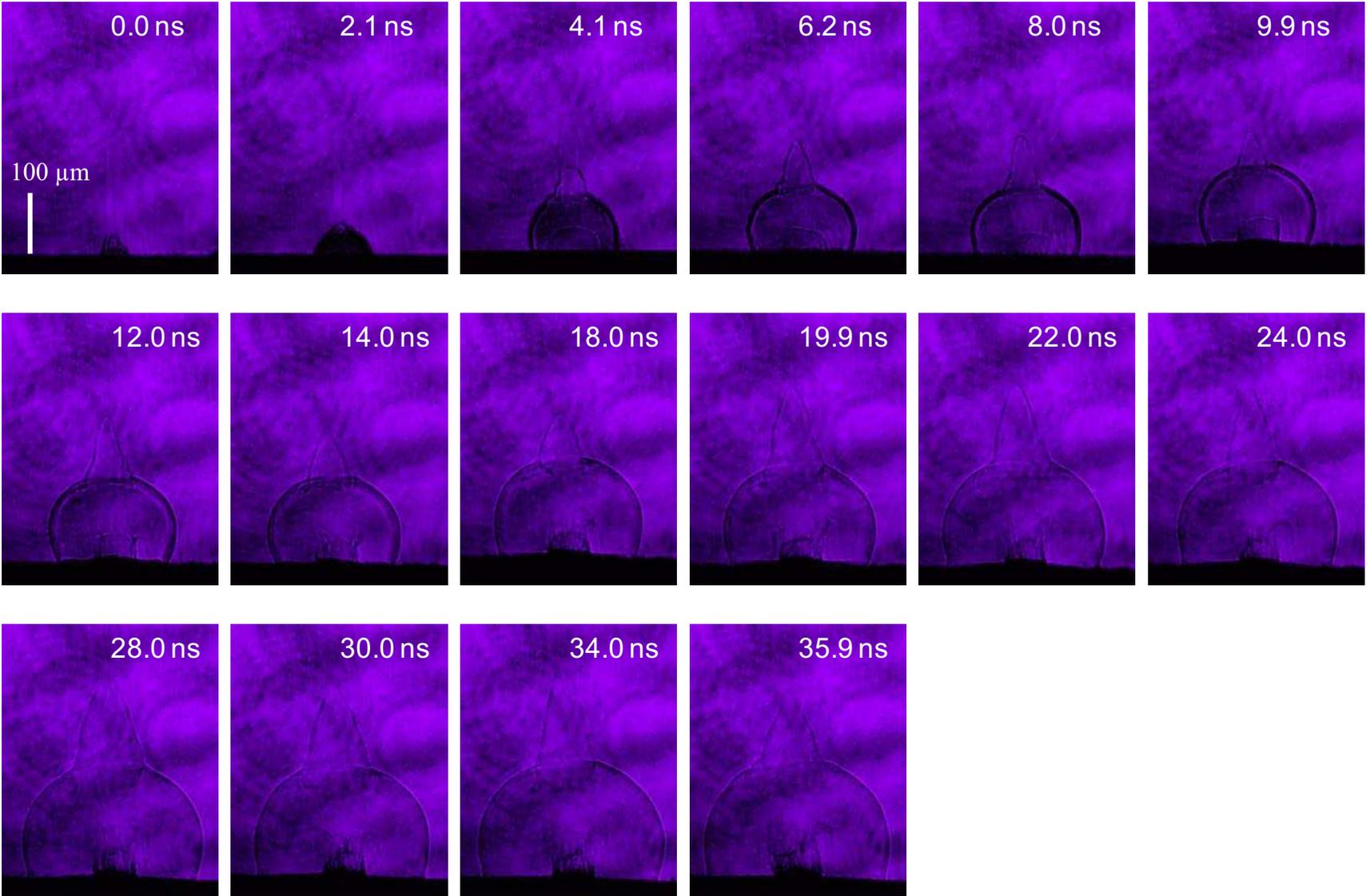


Figure 14. Shadowgrams for 266 nm nanosecond ablation at 88.9 J/cm^2 .

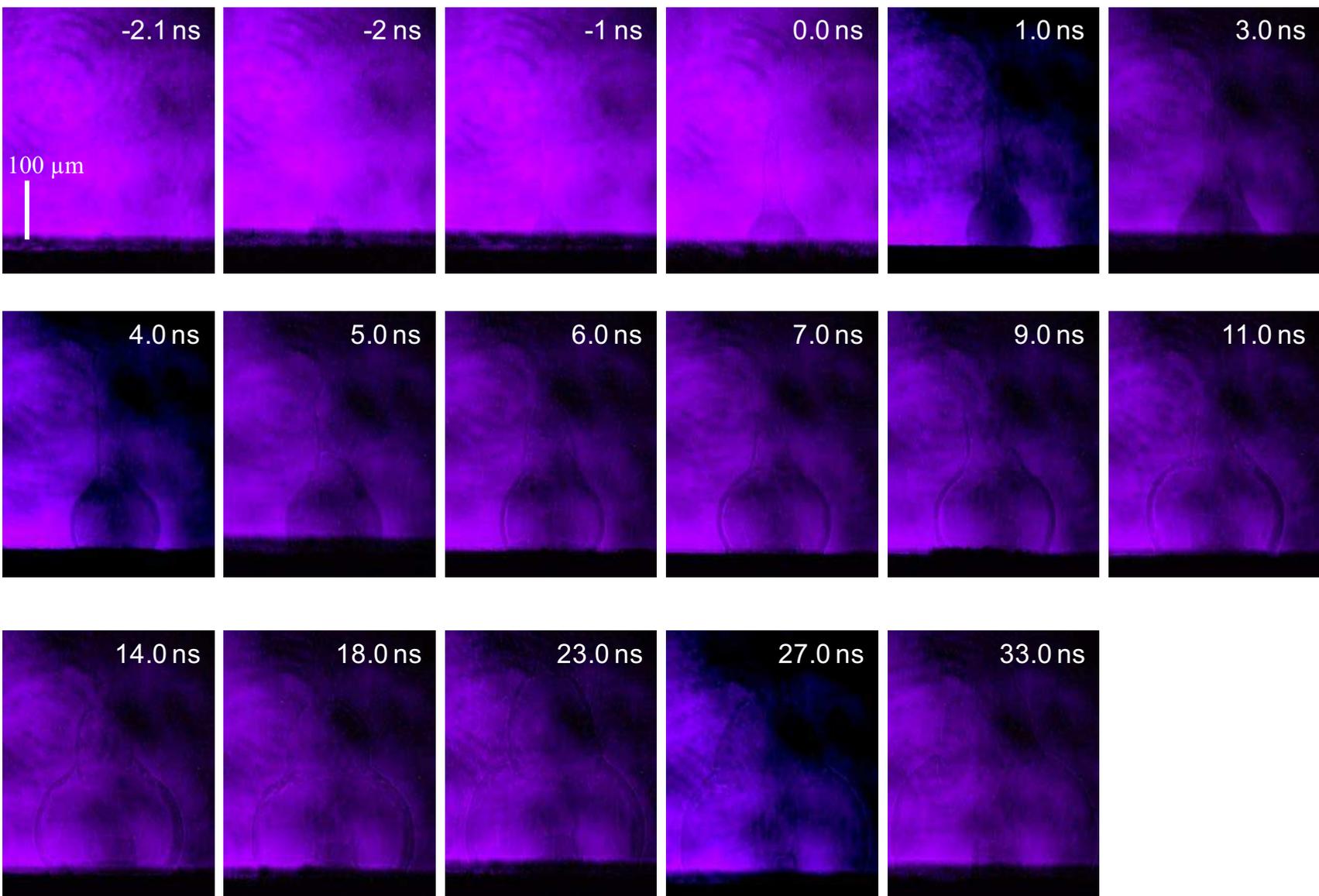


Figure 15. Shadowgrams for 266 nm nanosecond ablation at 190.5 J/cm^2

Our results show that the ultraviolet regime exhibit shockwaves that have a predominated semi-spherical form. However, for energies above 3 mJ, we observed small structures that grow on the tip of the shockwave, seen in figure 15. The appearance of these peaks is not regular from pulse-to-pulse, mainly due to the laser intensity variances that may occur and the difference in surface conditions at each ablation spot. This structure affects only part of the shockwave compared to figure 8 where the laser supported detonation wave protruded from a more global absorption.

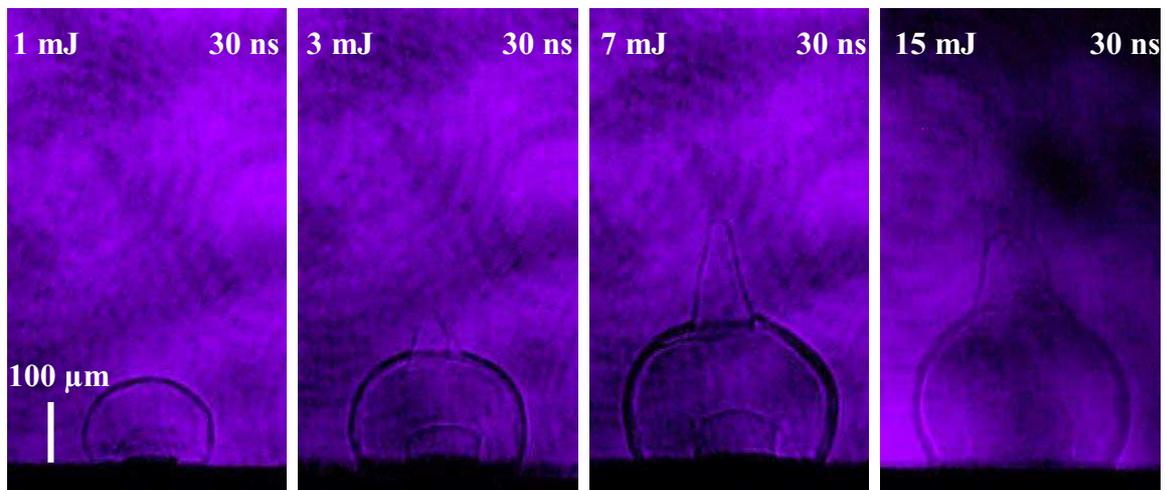


Figure 16. Shadowgraph images for 1 mJ, 3 mJ, 7mJ and 15 mJ at 30ns.

The critical electron density is calculated at $n_{e_{crit}} = 1.4 \times 10^{22} \text{ cm}^{-3}$, which is 16 times greater than that for infrared ablation. Therefore, the electron density needs to be 16 times greater for shielding to occur, which cannot be obtained in our experiments. However, as the irradiance gets higher, the density of the interface increases and we can observe laser absorption.

Ultraviolet ablation has a higher dissociation probability when compared to infrared ablation, due to its more powerful multi-photon absorption by air molecules (O_2 and N_2). The dissociation energies of the oxygen and nitrogen molecules are 5.2 eV and 9.8 eV respectively. O_2 dissociation needs to only absorb 2 photons while N_2 needs only 3 if we consider that a 266 nm photon has the energy of 4.66 eV. The irradiance required to form an air plasma is approximately $9.0 \times 10^{11} \text{ W/cm}^2$ [5]. Due to the nonlinear

dependence of the photo dissociation rate with respect to the laser intensity, only the central part of the laser beam can induce a significant dissociation. This may explain the reason why the apparition of the second component is only located at the very tip and does not cover the whole upper part of the shockwave. The newly emitted free electrons continue to absorb the laser energy, which in turn, heat the incoming laser pathway. A laser ionised channel is therefore created locally, further facilitating the expansion at this particular point. The combination of both the non linear partial shielding and the laser ionised channel induces a sharp peak to emerge.

Figure 17 shows the axial expansion of the main spherical shockwave with respect to the delay time. The different graph lines follow the same tendencies. Due to a higher slope below 10 ns, we can see that the plasma grew at a much faster rate at the beginning. We remark that in the ultraviolet regime, shockwaves expand spherically, which demonstrates the absence of a laser supported detonation wave in the ultraviolet ablation regime. We can also note that higher energies have a smoother fit.

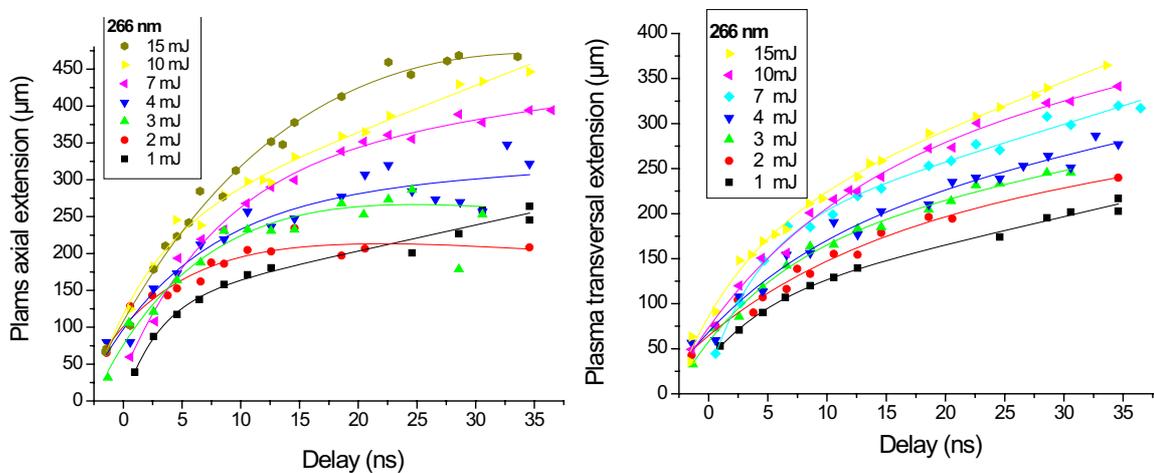


Figure 17. Axial and transversal expansion.

Again, equation (2.2) was used to fit the different curves in Figure 17. The derivative is calculated and plotted onto the graph below in Figure 18. Again, the graphs only give an estimate of the expansion velocities. The expansion speed starts out at a maximum and gently slows down to give us, in most cases, a constant value c , which varies more for these graphs than in the precedent Figure 10. The velocities reach a certain stability for energies 1, 2, 4 and 10 mJ, while the others are still decreasing.

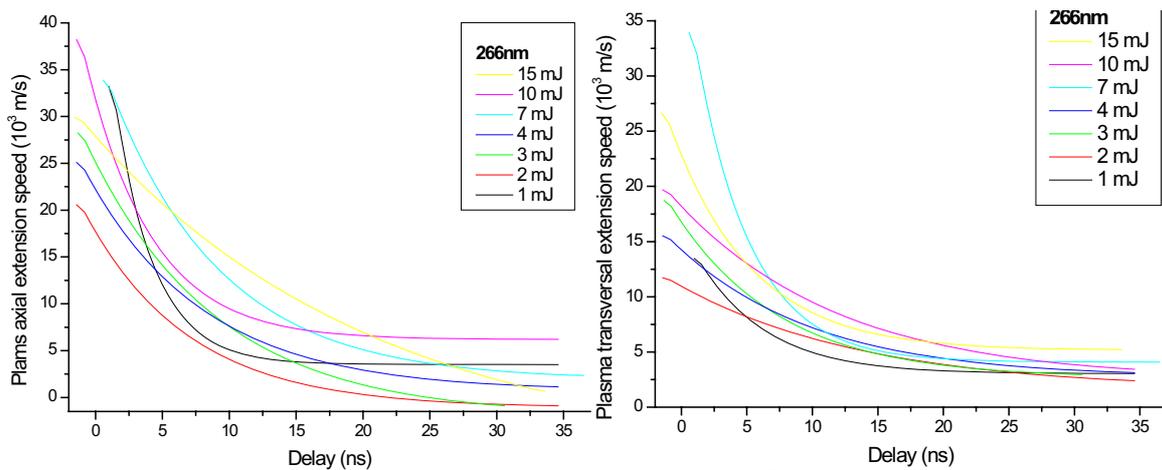


Figure 18. Axial and transversal velocity expansion.

As the energy increases, as shown in the shadowgraphs, other components can be observed at different times. Unlike the infrared regime, where only two components could be observed, the ultraviolet regime exhibits different components that become more complicated at high laser energies. This can be seen in Figure 19 where shadowgrams at 15 mJ portray different components that appear at different time delays on the tip of the shockwave. Inside the plasma, we can see the presence of an interior shockwave and material fragment ejection [2, 3].

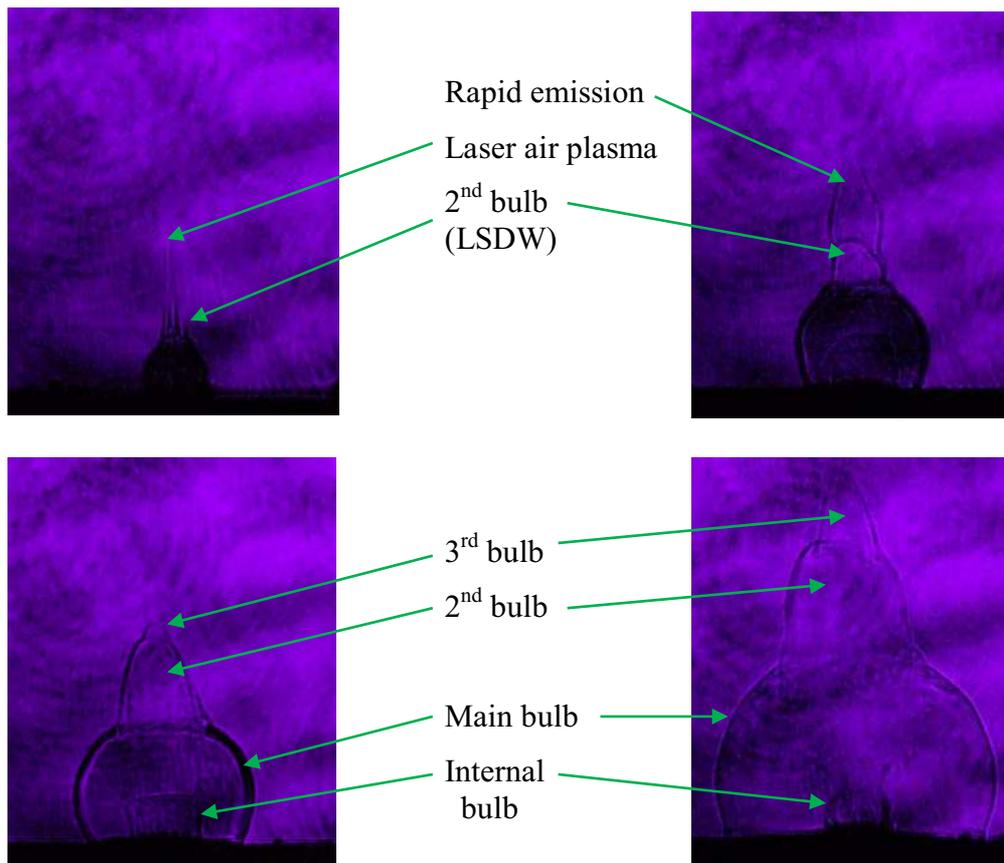


Figure 19. Shadowgraph UV images at 10mJ. We can see the different components of the plasma clearly in these images.

The thin peak that we see in the image (a) is due to the dissociation of N_2 caused by the passage of the laser. In figure (b) we see the laser supported detonation wave (LSDW) creating a second bulb. In (c) we see the apparition of a third bulb and in (d) the evolution of these three bulbs. Inside the main bulb, we see the presence of an interior bulb, which is due to the internal shockwave. The graph in Figure 19 below shows the specific expansion velocities of the components: the main spherical shockwave component with its two superimposed bulbs, and the two faster expanding components due to multiphoton absorption of the air molecules (green and blue). At higher energies, the ultraviolet shockwave expansion takes on different forms as the time delay varies. The earlier components (blue and green) are due to the laser passage in the air. The breakdown of N_2 and O_2 create a sensitive zone only locally. Due to which, the high interface density of the external shockwave explodes into the air where different components are interposed.

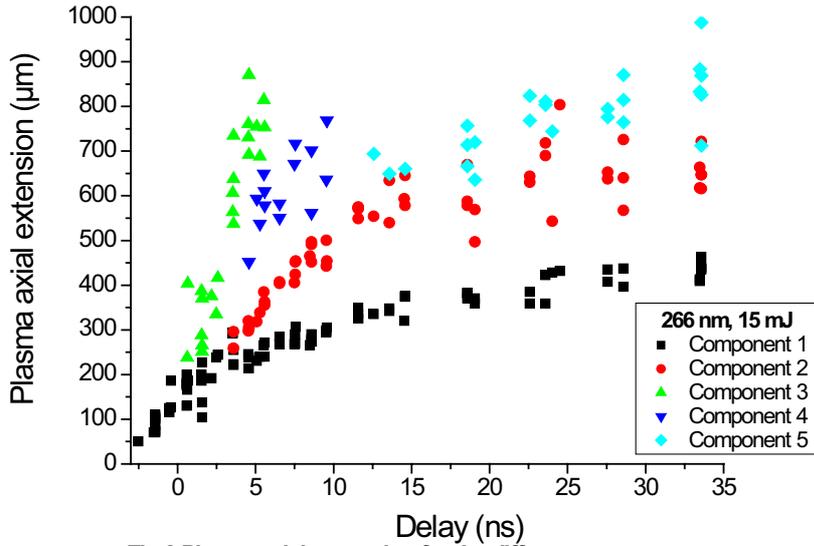


Figure 20. Axial extension of the different components of the choc wave.

At very short delays, a very sharp peak was noticed but could no longer be observed after a few nanoseconds. This peak corresponds to the plasma induced in the air directly by laser pulse which can be clearly seen in figure 16a. An ultraviolet photon at 266 nm is equivalent to 4.67 eV and can dissociate oxygen molecules first. Another peak starts growing at the top of the main spherical shockwave (main bulb). As the plasma gets wider, so does this peak. However, there is a very erratic behaviour with its height. This may be due to the laser shot-to-shot energy fluctuation and the slight changes in the homogeneity of the target surface. At higher energies, there is even a third peak that appears. For larger delays ($> 18 \text{ ns}$) these three areas form a pear shape choc wave that gets larger as shown in Figure 16.

2.2. FEMTOSECOND ABLATION REGIME

Figure 18, 19 and 20 show shadowgraphs of plasmas induced on the surface of a nylon sample by femtosecond laser pulses (800 nm) with pulse energies of 1, 3 and 4 mJ. The observed shockwave expansions are essentially spherical putting forth the lack of interaction between the laser pulse and the induced plasma. However, the high intensity of a femtosecond laser pulse can induce an ionised channel in air before the pulse hits the sample, which is clearly observed in the shadowgrams for pulse energies at 3 and 4 mJ. This ambient air conditions remain changed due to the filament even after the laser pulse is over, which creates a guided structure at the tip of the shockwave.

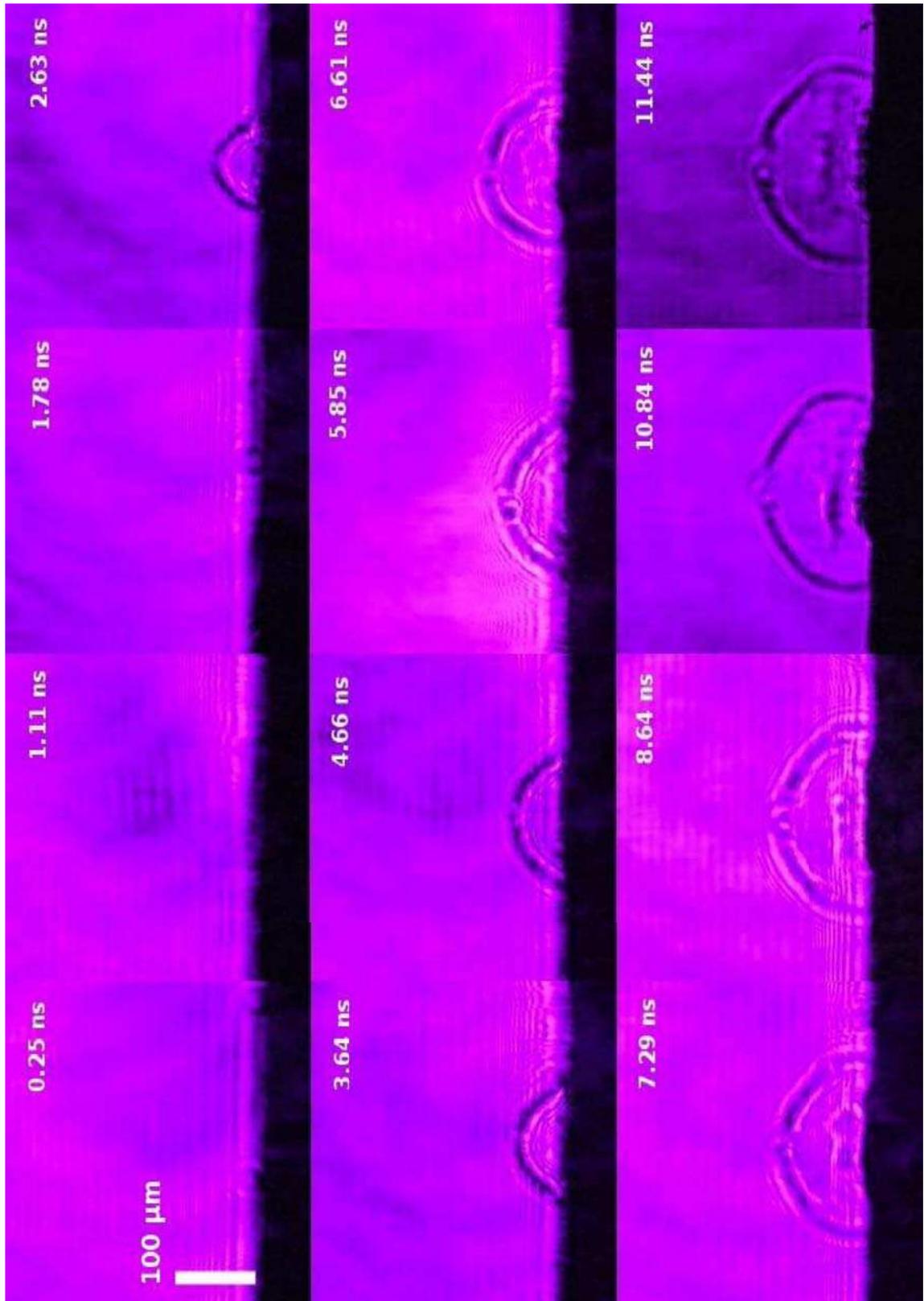


Figure 21. Shadowgrams for 800 nm femtosecond ablation at 1 mJ

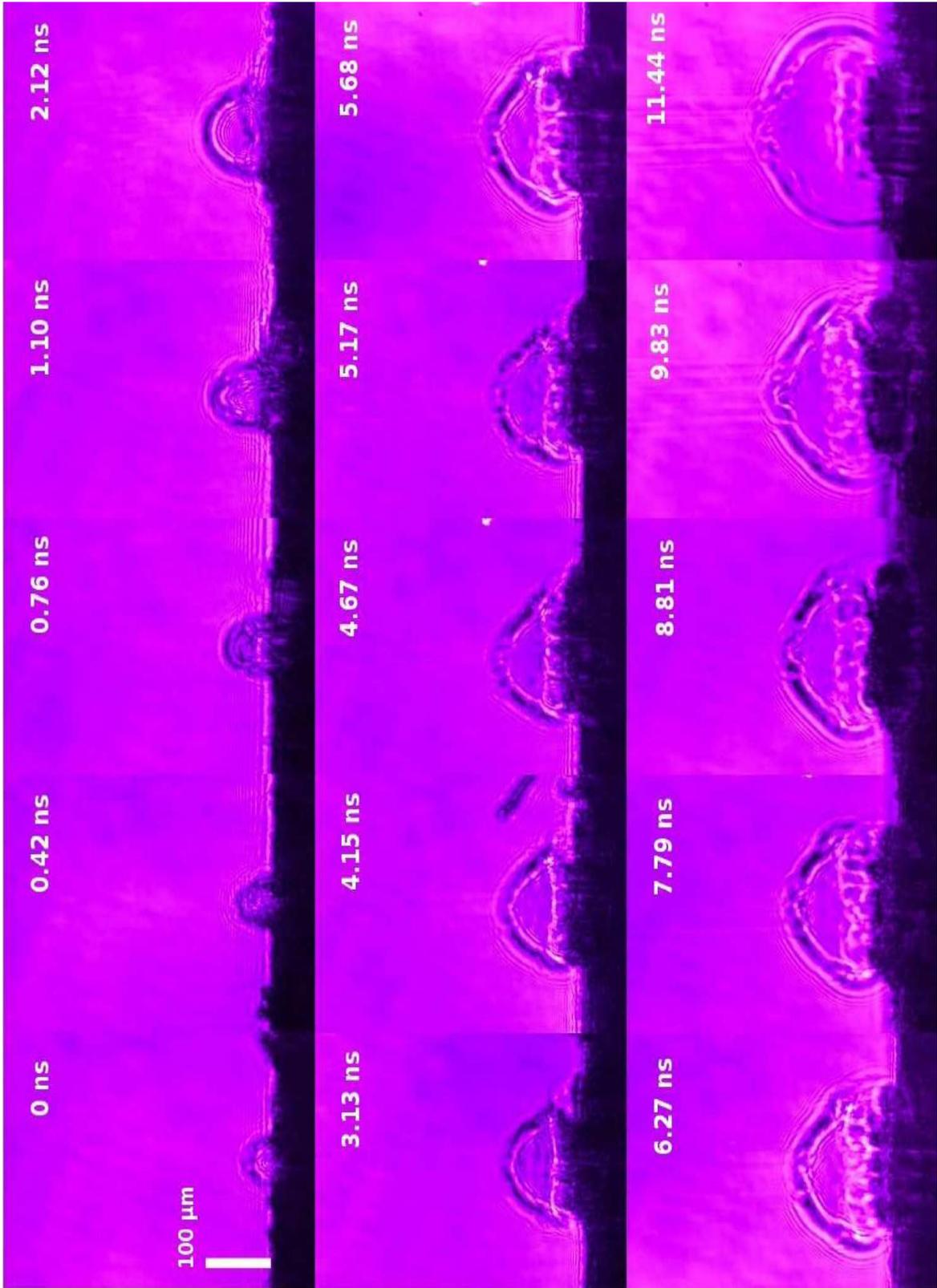


Figure 22. Shadowgrams for 800 nm femtosecond ablation at 3 mJ



Figure 23. Shadowgrams for 800 nm femtosecond ablation at 4 mJ

Figure 21 shows the axial expansion of the shockwave as a function of the delay. We can explain this observation by the fact that the femtosecond ablation regime allows a more efficient generation of nanoparticles creating a plasma with a lower degree of atomization [6].

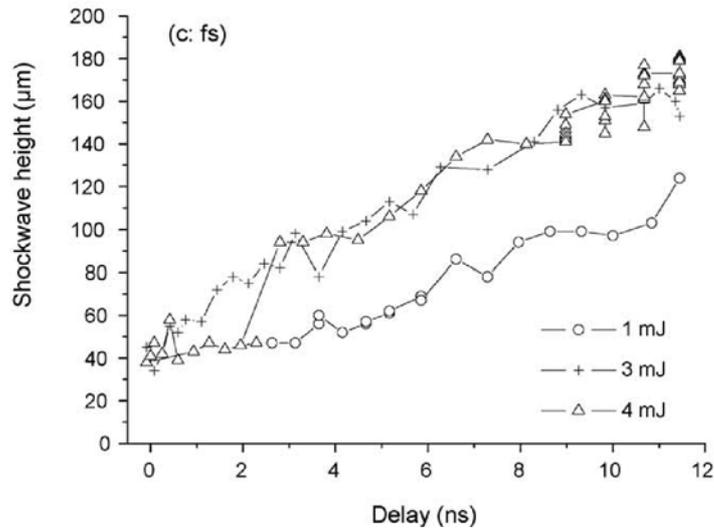


Figure 24. Femtosecond linear axial expansion [7].

3. DISCUSSIONS

The primary melting and evaporation of the sample occurs during the first couple of nanoseconds. When the plasma plume is in contact with the incoming laser beam, its temperature and electron number density continue to increase. If the electron number density becomes higher than the critical number density, the laser pulse is blocked by the plasma front and can no longer attain the target surface. The entire laser energy interacts solely with the plasma in this case, subsequently heating it even more and causing its expansion speed to increase. The result is the formation of a second component called laser-supported detonation wave (LSDW), clearly observed in the infrared nanosecond shadowgraph images at higher energies.

In the nanosecond ultraviolet regime, the critical electron density required for shielding is much higher than in the infrared regime, making it more difficult to reach. The sharp peak could be the consequence of a local breakdown of the surrounding atmosphere due to the multi-photon ionisation caused by the high laser intensity. This dissociation becomes more evident in the femtosecond regime where the available higher laser intensity favours the air molecule dissociation followed by electronic detachment and ionisation.

In the femtosecond regime, the laser pulse is already over by the time the plasma is formed and therefore cannot interact with it. Due to its ultrashort duration however, the intensity is significantly higher in this regime. The shadowgraph images (Figures 18 through 20) show the consequence on the ambient air of the passage of the laser pulse, which creates an ionised channel.

The differences mechanisms involved in the three ablation regimes lead to for each regime, different shockwave morphologies. The infrared regime portrays an axial expansion which is significantly faster than the radial one due to a larger absorption and screening by the plasma of the incoming laser pulse. In the ultraviolet regime, the shockwave expands mainly in an isotropic way with partial screening leading to a second component that has a sharp peak. In the femtosecond regime, the expansion is spherical except for the very tip of the wave front that is influenced by this laser ionised channel. More detailed theoretical modelling is certainly necessary to understand the interplay of all these mechanisms which are responsible for the observed shockwave forms in the different ablation regimes. However, such modelling is out of the scope of the present thesis.

We do know from the previous results that the shockwave front expands supersonically. To balance the velocity and the high backpressure generated by the external shockwave, an internal shockwave is produced, observed in our experiments only for the 266 nm ablation regime. The Berkeley team has a model that confirms the existence of the internal shockwave and how it moves back and forth between the sample surface and the external shockwave. They even show how some liquids are ejected as it strikes the sample surface due to the large amount of momentum transferred. They believe that the

interaction between the ejected liquid and the shockwave is the reason behind the rim around the crater after ablation. Our results also show a decrease in the shockwave expansion velocity as the ablation wavelength decreases. In the femtosecond shadowgraphs we remark a higher contrast compared to the nanosecond ablation regimes, which corresponds to a plume with higher density.

For our choice in ablation regime, we eliminated nanosecond infrared ablation due to the strong screening presented. Despite that the plume had a higher temperature, we preferred to have the greatest amount of the laser energy used to ablate the surface of the sample and not heat the plume. Our choice was therefore between the nanosecond ultraviolet ablation and femtosecond regime. Before even starting our experiments we knew the delicateness and the cost of the femtosecond laser could be a problem for industrial applications. However, if we could prove that the results obtained by this regime were really more interesting than the ultraviolet ablation then it should be worth to use it.

Since our aim is to find industrial applications from the results of this work, we opted for the nanosecond regime. The femtosecond regime remains very sensitive to the operation environment and demands more technical skills for its use. Nd:YAG nanosecond lasers cost less money and are less fragile, making them the more suitable choice for industrial application. In addition, published articles confirm that ultraviolet ablation is more efficient for organic samples.

4. CONCLUSION

In this chapter, we have studied the early stage expansion of the plasma induced on the surface of a nylon sample in different laser ablation regimes with regards to pulse duration, wavelength and fluence. The aim was to select an optimised ablation regime for our LIBS analysis of organic materials in the atmospheric pressure. This regime can be provided by either nanosecond ultraviolet pulses or femtosecond pulses. Nanosecond infrared laser is less suitable for organic material analysis, because of smaller absorption coefficient of organic materials for infrared radiations. In addition, in this regime on top of the generated plasma, at high fluence, the absorption of laser energy by the plasma leads to shielding of the plasma. Such shielding decreases the energy coupled into the sample, reduces therefore the ablation efficiency. The correlation between the laser supported detonation wave and the high rate of recombination with ambient air for this regime was observed [1].

For nanosecond ultraviolet laser pulses at high fluence, energy deposited on the sample increases because the absence of pulse shielding by the plasma. The significant energy deposited on the sample increases the initial energy of the plasma which leads to a higher ionisation degree and a reduced molecule production. For application purposes, the use of a nanosecond ultraviolet laser with a much more compact and reliable system is obviously advantaged compared to a femtosecond laser who actually remains still more complicate to operate.

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CHAPTER III. TIME-RESOLVED LIBS FOR CHARACTERISATION OF THE PLASMA INDUCED FROM POLYMERIC MATERIALS

1. INTRODUCTION

The basis of LIBS is to detect and study the light emission from the induced plasma, which provides information of the analyzed sample. Previously, we made a choice regarding the wavelength that most suitable for polymeric samples and found that a nanosecond ablation wavelength at 266 nm was the most appropriate. The shadowgraph images also showed the supersonic expansion of the plasma, confirming that the characteristics of the plasma (temperature and electronic density) rapidly changes with regards to the delay after the laser impact on the target.

This chapter shall focus on the choice of other experimental parameters, most importantly the delay after the laser pulse impact at which the spectrum acquisition should start (t_d)

and for how long it should record the spectrum (t_b). For this, a presentation of our LIBS experimental setup will be given along with preliminary choices of our experimental procedure for spectra acquisition. Before going forth with any analysis, instrumental line broadening measurements were performed. This is followed by our spectrum acquisition procedure, explaining our choice for the different delays and gate widths for time-resolved LIBS. The number of spectrum chosen, along with the elements used for the plasma diagnostics are introduced as well.

Once these criteria are chosen, time-resolved LIBS on polymeric materials can be properly investigated. We present our results regarding the electron density and different temperatures (mainly through Boltzmann plots and Saha-Boltzmann plots). The notion of local thermodynamic equilibrium is evoked for the use in calibration-free LIBS. For this thesis however, we are more interested in the signal-to-noise ratio that allows us to extract the most significant spectral information. Once these studies are finished, the optimal delay and gate width can be selected for the most reliable spectral line emission range (used in Chapter 4 for further LIBS applications).

2. EXPERIMENTAL SETUP

In the following experiments, the fourth-harmonic generated (266 nm) by a Q-switched Nd:YAG laser (Thompson B. M. Industry 5000 series) was used at a repetition rate of 10 Hz. The laser pulse duration at FWHM was about 5 ns. The energy of this laser was varied between 1 and 20 mJ. The harmonic generation was first optimised for generating 20 mJ output at 266 nm. To obtain smaller energies, the phase-matching was detuned. The diameter of the laser beam was 7 mm. The laser beam was focalised onto the sample with the means of a fused silica lens. The focal length of the lens was 50 mm. An ablation crater with a 20 mJ laser pulse on an aluminium sample is estimated to be 100 μm in diameter. Therefore, the estimated fluence was $150 \text{ J}/\text{cm}^2$. The emission from the laser-induced plasma was collected by two aluminium-coated parabolic mirrors. The second one focalises the light emission directly into an optical fibre with a core diameter of 50 μm . A schematic representation is presented in Figure 1.

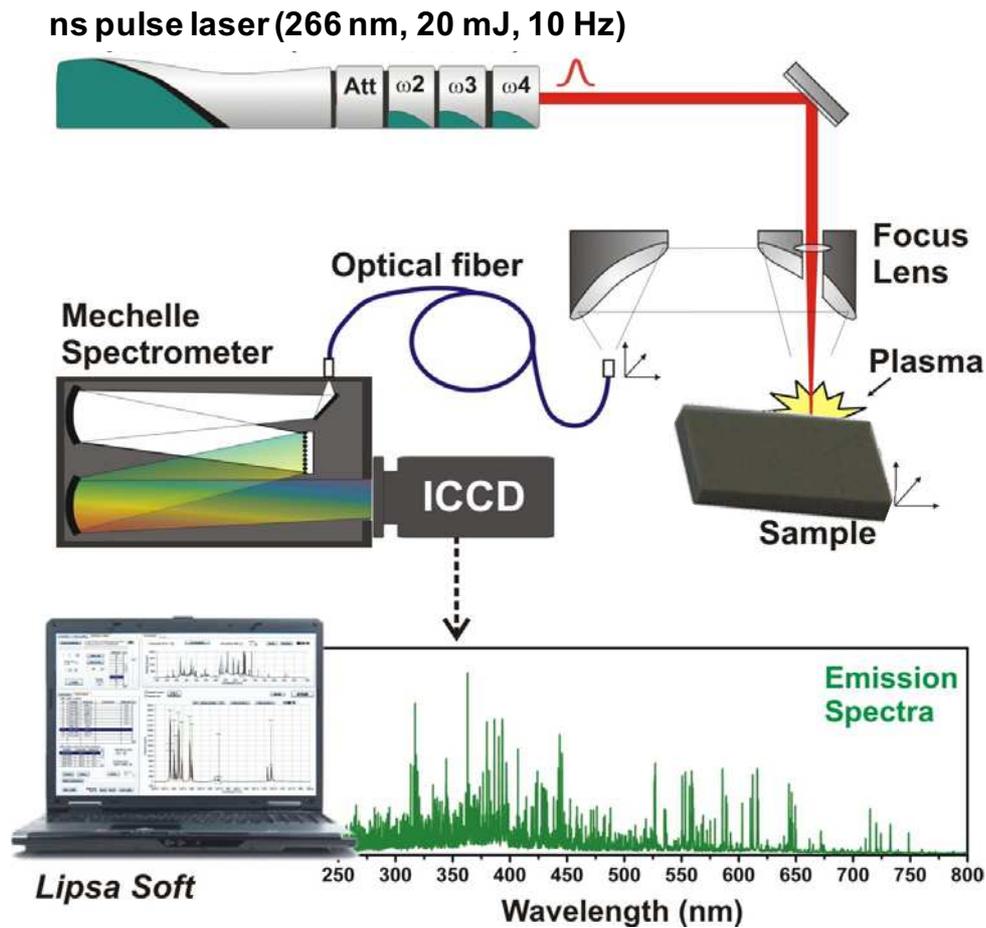


Figure 1. LIBS experimental setup. The emission spectra are analysed with the LIPSA software developed by our research group.

To ensure the repeatability of the experiments, it is obvious that a rigorous experimental protocol has to be put in place. The first thing to do is to optimise the amount of light collected. Before any alignment is done, the distance between the focalising lens and the sample has to be fixed. We have chosen to place the sample right before the waist point of the laser beam to make sure that the ablation takes place inside the sample without direct air breakdown. The optical collection system is then aligned for this particular position. The parabolic mirrors are fixed to attain the smallest light spot. To finish the signal optimisation, the optical fibre entrance is positioned to attain the highest output signal.

To clearly understand how this system works let us go back to the basics of light dispersion. The grating equation is stated in Equation (3.1).

$$\sin \alpha \pm \sin \beta = kn\lambda = R\lambda \quad (3.1)$$

With	α	:	<i>incident angle</i>
	β	:	<i>diffracted angle</i>
	k	:	<i>diffraction order</i>
	n	:	<i>grating density</i>
	λ	:	<i>wavelength of diffraction</i>
	R	:	<i>resolution of diffraction ($R = kn$)</i>

An echelle spectrometer has the capability of separating both the diffraction orders k and the wavelength of the emission light received. This is why it is said to have a two-dimensional wavelength dispersion. For high spectral resolution, a high dispersion element is needed such as echelle grating (Figure 2). When an incident ray is diffracted by grating, many spectra can occur. A low dispersion system like a prism is used to separate the overlapping of the light emitted by the echelle grating to produce high resolution and order separated spectra. This unique two-dimensional dispersion allows an atomic spectral line to produce an image in the form of a spot.

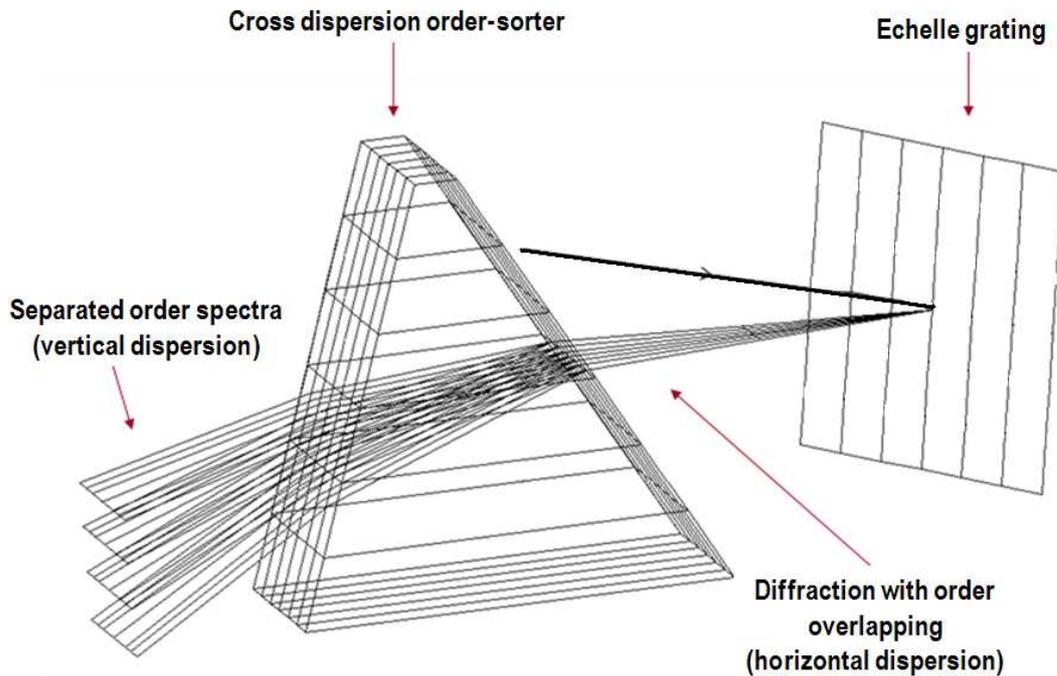


Figure 2. Echelle Grating [1].

The Andor Mechelle spectrometer is a patented optical design that produces high resolution, broadband spectral ranges with low cross talk, based on the echelle grating

illustrated in Figure 3. The collected plasma light delivered by the optical fibre enters through the pinhole. The two correction lenses are used to decrease the loss of resolution that may occur. The collimating mirror reproduces an image of the source onto the prism. The downside to this is the overlapping of all the different spectral orders. The prism is used as a refractive element to separate these different orders.

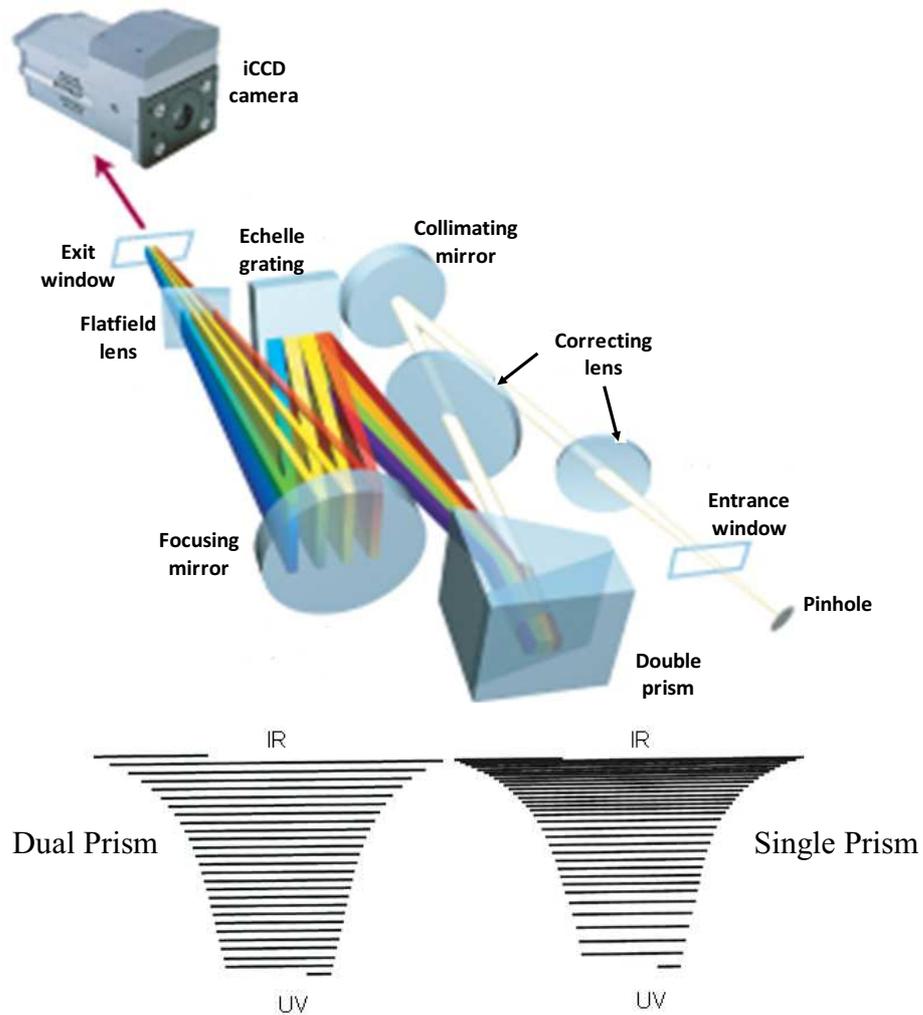


Figure 3. Mechele Andor Spectrometer and iCCD [1].

To make the dimensional display even better, the use of two prisms results decreases the cross talk that can occur between the orders in the visible and NIR region, where the low dispersion forces them to bunch up. The resulting image is shown below in Figure 3, where a regular spacing between the spectral orders is seen with a dual prism. The advantage of such a system is the low cross-talk and equally spaced order separation for an optimal resolution with a wavelength ranging from UV to NIR.

The 2D map is focused onto the detector, which is an iCCD camera (iStar DH734 Gen II, Andor) in our case. The iCCD camera has the advantage of offering high sensitivity in conditions with very low light available thanks to the intensifier. Its gating properties also provide very good temporal resolution for time -resolved plasma evolution studies. By receiving the two dimensional image, the camera can produce spectral images with high resolution. At the entrance of the CCD camera, there is an array of pixels. In our case we have 1024×1024 pixels. The time it takes for the information to be gathered and sent out is a couple of microseconds. When the aim of our experiments is to look at what happens at very early and small time delays, the data readout time becomes a problem.

To fix this problem, an intensifier is placed in front of the CCD. It acts as an electronic shutter that permits or prohibits light to enter. In other words, if the light beam lasts for one second, it will open the slit to CCD only for the desired amount of time, allowing only a fraction of the light to enter. It then closes, blocking the rest of the light. It can be done very fast and can be opened and closed within a matter of a few ns. This is the principle of gating shown in Figure 4.

The intensifier presented in Figure 4 consists of a photocathode, a multichannel plate (MCP) and a phosphorus screen. The photons hit the photocathode, which due to the photoelectric effect, emits a certain amount of electrons. These negatively charged electrons with the help of a gate voltage are drawn to the MCP, which acts like an electron multiplier. The MCP has a lot of channels and when the electrons hit one of them, they get multiplied. The generated secondary electrons are accelerated towards the phosphorus screen due to the applied voltage. Once they hit the screen, their energy is transferred into the emission of a number of photons on the other side. The higher the voltage, the faster and stronger the electrons hit the screen, producing more photons. The photons are then projected onto the CCD camera. The advantage is that a lot more information is gained using this technique but because it has more optical components, we lose some resolution.

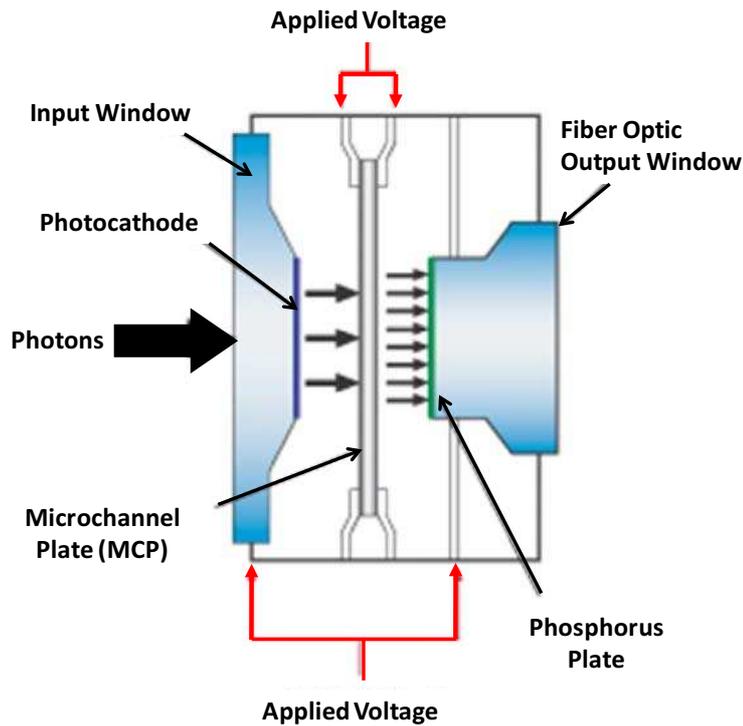


Figure 4. The intensifier [1].

3. PRE-SPECTRA ACQUISITION

Before starting our time-resolved experiments, a few necessary details need to be adjusted. Notion of instrumental broadening will allow us to refine our calculations regarding the temperatures and electronic density of the plasma. Preliminary spectra were recorded to provide us with information regarding the delay and gate width to choose at different moments. Finally, the choice of how many spectra is sufficiently enough for our time-resolved analysis.

3.1. INSTRUMENTAL BROADENING

As explained in Chapter 1, an instrumental broadening must be accounted for spectral analysis. Our best line fit for such broadening is the Gaussian fit with $R^2 = 0.99$ as compared to the Lorentzian fit with $R^2 = 0.96$ as can be seen in the Figure 5 representing a spectral line emitted by a mercury lamp.

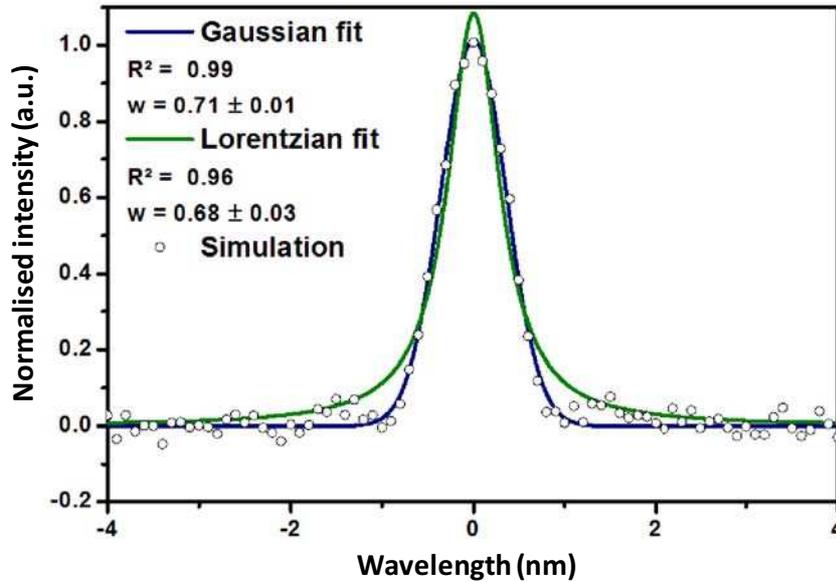


Figure 5. Simulation of a mercury lamp emission fitted with a Gaussian and a Lorentzian graph

The instrumental broadening may depend on the wavelength. Different wavelengths from a low-pressure mercury lamp were used to measure this broadening as a function of the wavelength. Figure 6 puts forth our results and the instrumental resolution power determined experimentally was $\lambda/\Delta\lambda_{1/2} = 9010 \pm 150$ (the constructor had announced 10 000). There is less fluctuation at shorter wavelengths (~ 0.03 nm) than at longer wavelengths (~ 0.07 nm).

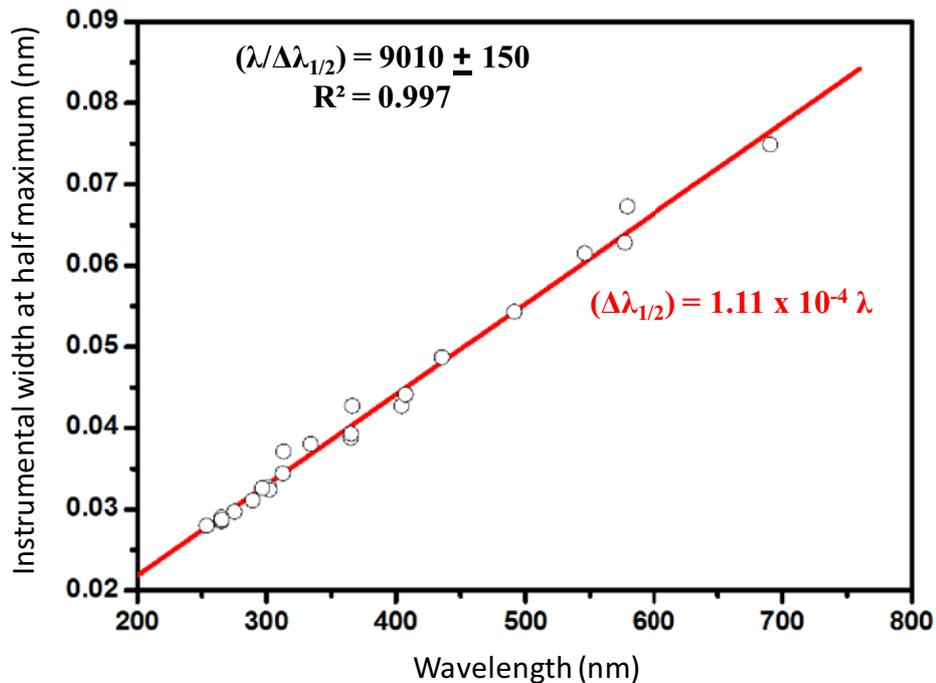


Figure 6. Instrumental broadband broadening.

3.2. DELAY AND GATE WIDTH

Our first aim was to study the time-resolved spectral emission of the plasma by calculating different temperatures and electronic densities as functions of the delay after the laser impact. For this study, we fixed the following experimental parameters accordingly. The camera exposure was always fixed at 1s, which means that each spectrum is the accumulation of 10 laser pulses since the laser is working at 10 Hz. A reason behind this choice was to accumulate the spectra to increase the signal intensity and reduce some of the noise. The delay t_d is the moment after laser impact upon the sample where the acquisition starts and the width t_b corresponds to the duration that the camera is gated (Figure 7).

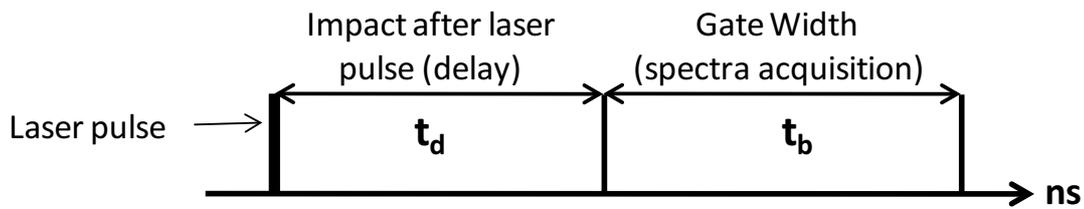


Figure 7. Chronogram of spectra acquisition.

Longer delays allow the plasma to reach equilibrium but also give less intense signals. At earlier delays, there is a rapid change in the electronic density and temperatures due to the extremely fast expansion of the plasma as explained in Chapter 1. For earlier delays, in order to avoid large differences between the spectral data acquired at the beginning and at the end of the acquisition, smaller gate widths should be chosen. At longer delays, the fluctuations dampen out i.e. these effects are less consequential, allowing us to choose larger gate widths to acquire more signal. In other words, we have segmented the plasma acquisition times into different categories as can be seen in Table 1. We assume that between 70 ns and 130 ns (gate width of 30 ns), the plasma is almost stationary. The same reasoning goes for 1980 ns and 3980 ns.

Delay (ns)	Gate width (ns)
70 – 100	30
100 – 130	30
130 – 180	50
180 – 230	50
230 – 280	50
280 – 380	100
380 – 480	100
480 – 580	100
580 – 780	200
780 – 1280	500
1280 – 1780	500
1780 - 2280	500
1980-2980	1000
1980-3980	2000

Table 1. Spectrum acquisition parameters.

4. EXPERIMENTAL RESULTS

4.1. STATISTIC SPECTRAL ACCUMULATION

The question on how many spectra should a statistical series contain was evoked. When gathering information for a statistical spectra batch, we kept all the parameters fixed while we scanned the surface of our sample. Every spectrum was issued from a plasma that was induced on a fresh surface and not the accumulation at the same ablation spot. For example, for a fixed delay of 380 ns and gate of 100 ns, all the spectra were acquired under the same conditions except for the experimental fluctuations and the local sample inhomogeneities. We did a quick study comparing the results regarding the temperature and electronic density of the statistical batch containing information from 10 and 100 spectra.

100 ACCUMULATIONS		10 ACCUMULATIONS	
Delay (ns)	Gate width (ns)	Delay (ns)	Gate width (ns)
170	200	170	100
370	200	270	100
570	400	370	100
970	400	470	200
1370	800	670	300
		970	300
		1070	500

Table 2. Spectra acquisition parameters for quick measurements

Figure 8 shows the estimations for both the electronic density (with regards to the H_{α} line) and the temperature calculations with regards to calcium elements [4, 5]. With a 100 spectra statistical batch, we are able to scan a relatively large surface area of the sample. Since a perfectly homogenous sample is only really theoretically possible, this scanning can dampen out any local differences on the sample's surface. The experimental conditions are also prone to slight fluctuations (energy of the laser, the position of the sample with respect to the beam focusing point, inhomogeneity of the sample etc.)

Figure 8 shows that the accumulation of a 100 spectra provided more accurate results, especially when calculating the electronic density. The dispersion of the points was far too great with regards to the statistical batch containing 10 spectra to provide us with measurements that are good enough. Figure 8 also shows dots that are present for one statistical series and not the other. The time required for the 100 spectra acquisitions are 10 times more time costly. Therefore, the graphs were composed of the data corresponding to the parameters in Table 2.

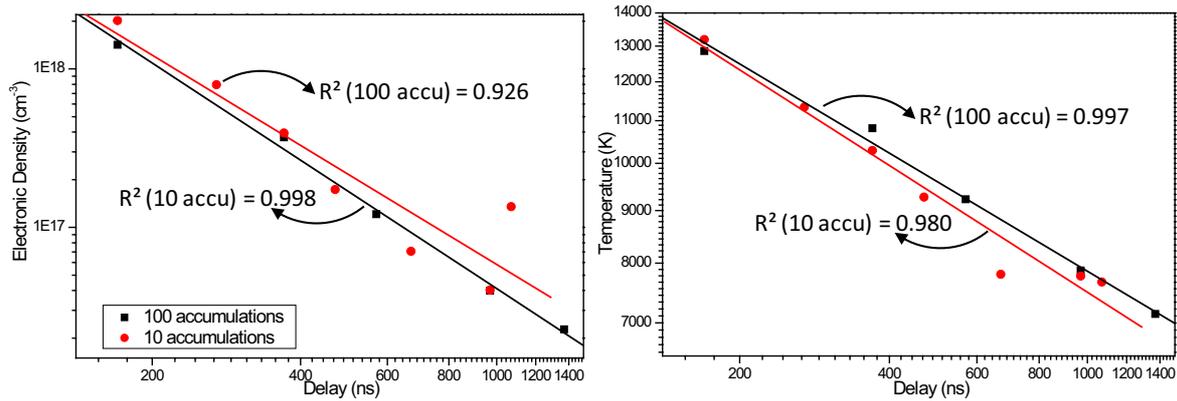


Figure 8. Comparison between the accumulation of 10 and 100 spectra taken on a polypropylene sample ablated in air.

4.2. TEMPERATURE COMPARISON

The atomic temperature can be calculated with regards to 3 elements in our case: magnesium (Mg), titanium (Ti), and calcium (Ca). Their temperatures are presented in Figure 9 showing the different values produced by Saha-Boltzmann plots previously explained in Chapter 1. Magnesium exhibits the highest temperature values and calcium the lowest ones. Titanium contains much more spectral lines compared to the other two elements that only have 3 lines for each (either ionic and neutral lines). For the rest of our temperatures calculations, it seemed that the titanium results provided a more accurate estimation and shall be used.

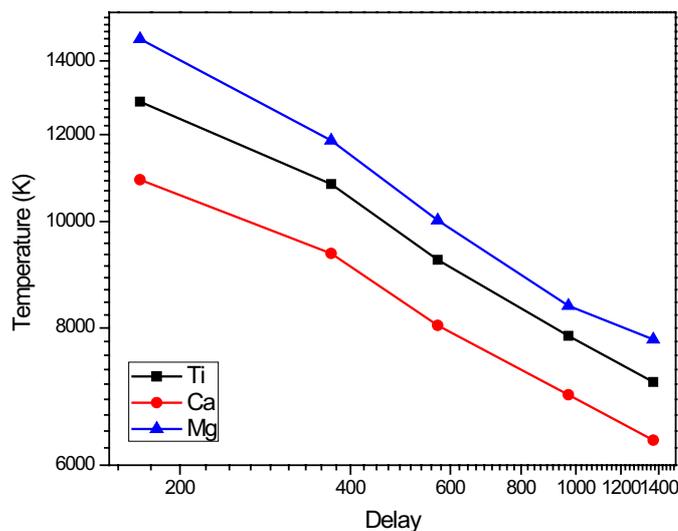


Figure 9. Temperature calculations on a polypropylene sample.

Figure 10 puts forth the average of a 100 spectra for a polymer sample at different laser pulse impact delays. For shorter delays, the total emission spectrum experiences a large influence of the continuum emission since the baseline of the spectrum is high. As the delay increases, the radiative recombination between electrons and ions and Bremsstrahlung emission drop, decreasing their influences. After about 500 ns, our spectrum is dominated by spectral line emissions associated to optical transitions between discrete states of different species, ions, atoms and molecules. As the delay increases, the electronic density decreases resulting in spectrum narrowing.

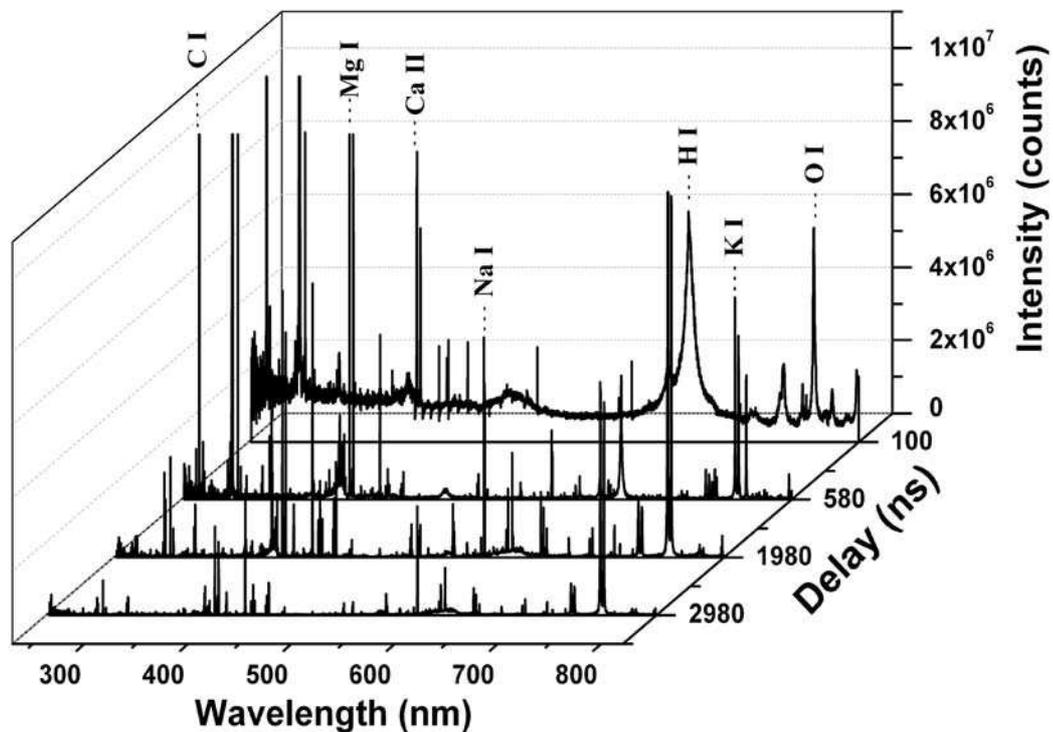


Figure 10. Evolution of the spectrum for a polymer sample.

Carbon and hydrogen are the elements most commonly present in all polymers. Oxygen and nitrogen can sometimes be present as well. However, if the ablation process is done in the presence of air, hydrogen, oxygen and nitrogen can also be due to the contributions from the ambient air. Certain additives and fillers are added to polymers to improve their properties or simply to reduce the cost of fabrication. The most commonly found are

TiO₂, CaCO₃, and Mg, with variable concentrations [2]. This is explained in more detail in the next chapter. In our experiments, we use these elements to calculate electron density and different temperatures. To find the best compromise between maximal intensity and maximal signal, we also studied the signal-to-noise ratio (SNR). It will be shown that this compromise coincides with the conditions for the local thermodynamic equilibrium as well.

4.3. ELECTRON DENSITY

Time-resolved LIBS was used to characterise the plasma induced on a polypropylene sample. Temporal evolution of the electron density was determined for the ablation energies of 18 mJ. The Stark broadening of the H_{α} line was used to calculate the electron density [6]. As shown in Figure 11, it starts out higher at about $1 \times 10^{18} \text{ cm}^{-3}$ at short delays then decreases to about $1 \times 10^{16} \text{ cm}^{-3}$ for longer delays. The self-absorption of the H_{α} line only starts to appear for electron densities greater than $3 \times 10^{18} \text{ cm}^{-3}$ [3]. Therefore, despite hydrogen being part of the matrix, we can ignore the self-absorption of the H_{α} in our calculations. After 1500 ns the H_{α} line became undistinguishable amongst the noise, rendering the evaluation of the electronic density impossible.

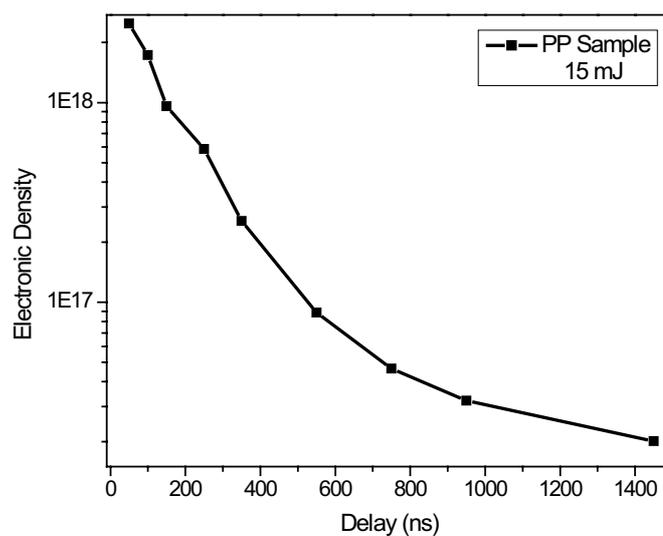


Figure 11. Electron density for polypropylene sample at 18 mJ.

As mentioned in Chapter 1, the McWhirter criterion provides the lowest value for the electron density N_e necessary (but not sufficient) for the plasma to reach local thermodynamic equilibrium. For a 7000 K plasma, it can be estimated at $N_e = 8.6 \times 10^{15} \text{ cm}^{-3}$. In Figure 11, the measured electron densities are larger than this value. This means that the plasma has enough electrons to reach local thermodynamic equilibrium through collisions.

4.4. TEMPERATURE CALCULATIONS

Given these electron density, the following temperature calculations were done using plasmas induced by 18 mJ laser pulses upon a polypropylene sample. For the evaluation of the line intensities, a question on whether it was better to use the total surface area under the spectral line or just the peak maximum value was asked. Figure 12 shows the results from the both methods taken simultaneously. We can see that for delays less than 300 ns there is a difference between the two methods. However after this delay, both provide comparable results. Therefore to simplify the calculations, we use the peak maximum since it is less time consuming.

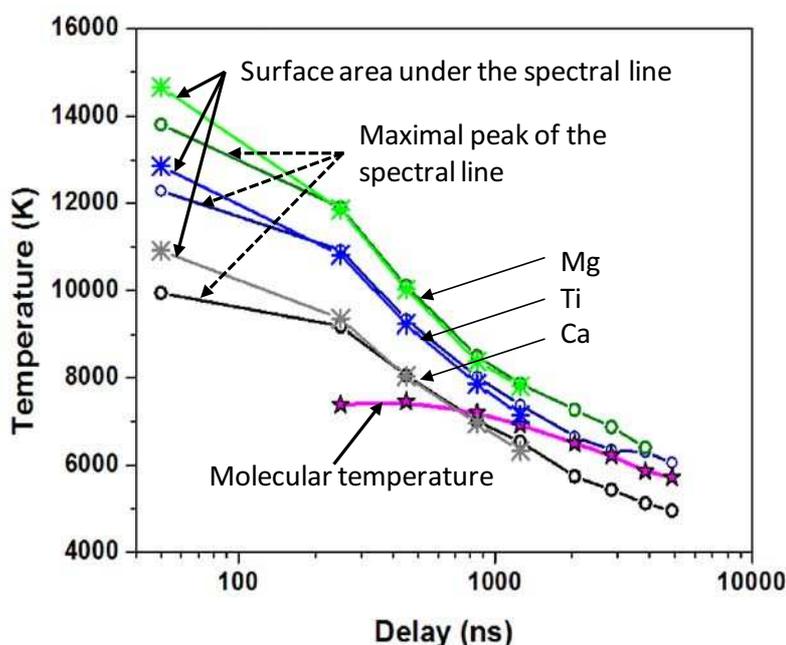


Figure 12. Temperature calculations comparing the spectral peak maximum (open dots) or total surface (asterisk) and the molecular CN temperature (stars)

Boltzmann plot

The aim is to plot information taken from different spectral lines of the same species (neutral or ionic) and draw the best fitting line. According to the Boltzmann distribution law, the slope of this line is related to the temperature of that particular species. As can be seen by Equation (3.2), the intensity of a spectral line is inversely proportional to the temperature [6].

$$\ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right) \propto \left(- \frac{E_i^Z}{k_B T} \right) \quad (3.2)$$

The temperature of the plasma can be found by calculating the slope of the line, as shown in Figure 13. The atomic neutrals are on the upper left side of the graph while the ions are in the bottom right side. It can be seen that it is not a very accurate means due to the large vertical dispersion of the points and a narrow energy scale concerned, especially for the atomic neutrals. We will see later that the Saha-Boltzmann plot provides more accurate results in for temperature calculation.

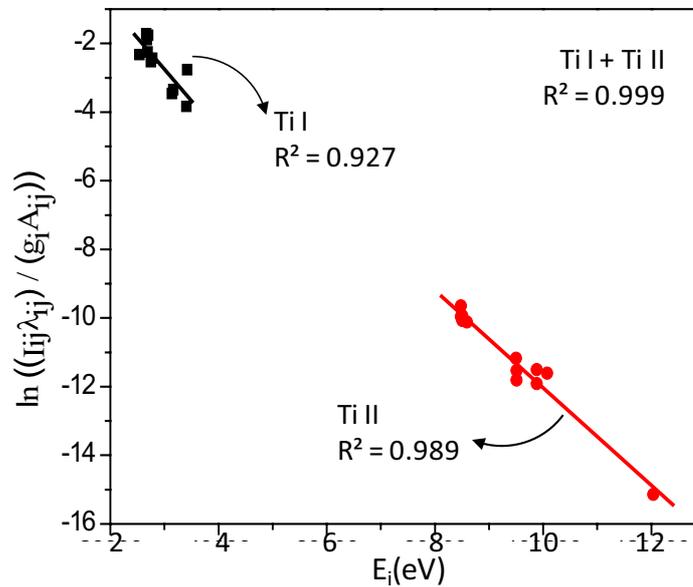


Figure 13. Boltzmann plot for titanium species at a delay of 1370 ns and gate width of 800 ns

Saha-Boltzmann Plot

If we assume that the ions and the neutral atoms of the same element are said to be at the same temperature, we lift the problem encountered with the Boltzmann plot. We can now trace a line that goes through both species simultaneously. The errors associated to linear fits are now improved as the horizontal energy range E_j has increased considerably compare to previous Boltzmann Plot. We recall the Saha-Boltzmann ordinate correction in Equation (3.3). The approach remains very similar to the Boltzmann plot where the aim is to find the best fitting line and calculate its slope to find the plasma temperature [4]. Typical Saha-Boltzmann plots are shown in Figure 14 for different delays starting off from 170 until 1370 ns to show the temporal evolution of the temperature. These plots exhibit a good linearity with statistical errors associated to the retrieved temperature at less than 100 K. Good agreements are especially observed between the atomic (left side) and ionic (right side) transitions. This agreement indicates that equilibrium is reached between neutral and ionised titanium atoms [6].

$$\ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right)^* = \ln \left(\frac{I_{ij}^Z \lambda_{ij}}{g_i A_{ij}} \right) - z \ln \left(2 \frac{(2\pi m_e k_B)^{3/2} T^{3/2}}{h^3 N_e} \right) \quad (3.3)$$

The wider energy range E_j allows for more precise linear plots by combining both species of the same element. Different delays are presented on the same figure to put forth the difference in their slopes. A greater slope indicates a lower temperature.

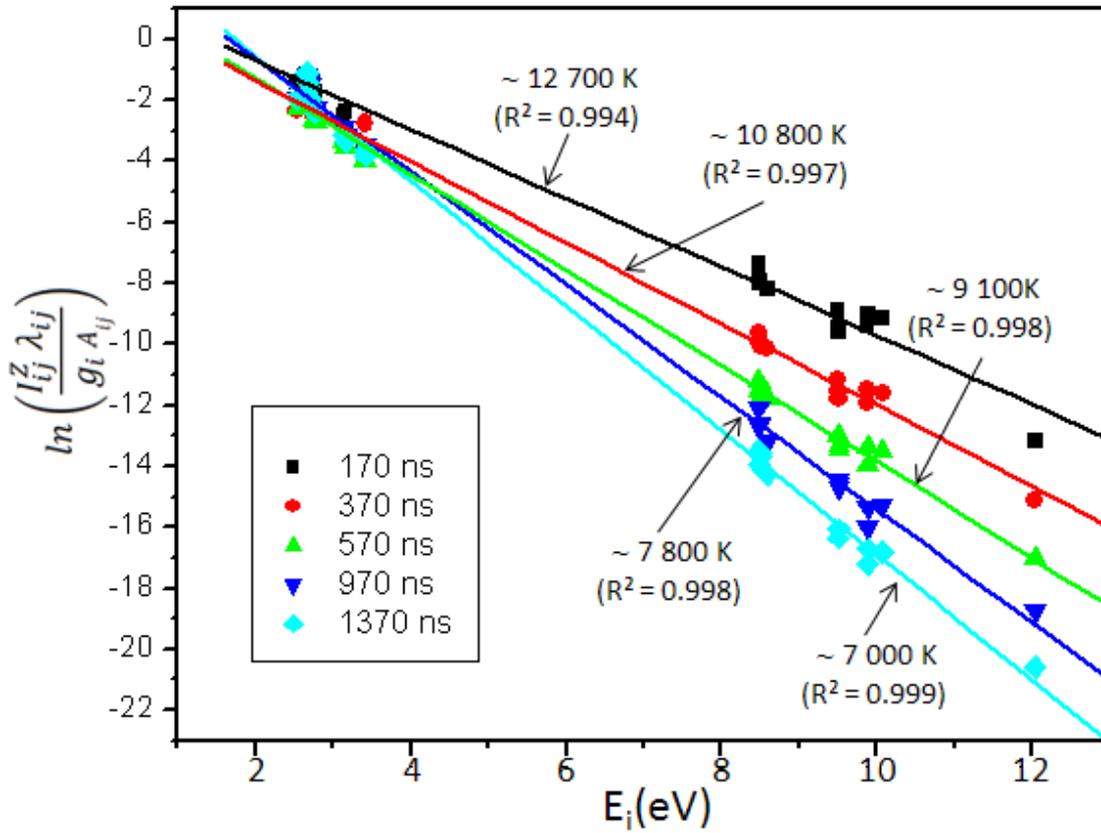


Figure 14. Saha-Boltzmann plots with lines from Ti.

4.5. LOCAL THERMODYNAMIC EQUILIBRIUM

Considering the McWhirter criterion, the electron density must be high enough for the local thermodynamic equilibrium to be considered [4]. In Chapter I we calculated this density for a plasma at 7000 K using the H_{α} line is $N_e = 8.6 \times 10^{15} \text{ cm}^{-3}$. The calculations show that the electron density ensures this density until a delay lower than 1 μs . The second criterion for local thermodynamic equilibrium states that the plasma needs to be defined by a single temperature to describe the different species present. This means that the neutral and ionic species of all the elements must be at the same temperature along with the molecular temperature. For this, we perform a study of the evolution of the temperatures in the plasma to verify this condition.

The different plots are represented in Figure 15. In this figure, the excitation temperature of Ti ions is calculated for different delays using the Saha-Boltzmann plot. According to

Equation (3.3) the electron density N_e is needed. The electronic density data shown in Figure 15, which corresponds to 18 mJ ablation laser energy, is fed into Equation (3.3). Calculations are only available until a delay of 1 μ s due to the disappearance of the H_α line. The electron densities beyond this point are obtained by extrapolating the experimental data using a power law fitting. The different temperatures merge together after 600 ns, with a maximal value for the molecular temperature between 700 and 800 ns. The density criterion is satisfied only until about 1000 ns. In order to consider ourselves in local thermodynamic equilibrium, it would be better to place ourselves with a delay of 600 ns. The gate width of 400 ns was chosen in order to get the maximal amount of intensity out of the spectral information. The fluctuations between the temperatures are still seen to fluctuate for this time frame, however it was the best compromise to satisfy the local thermodynamic conditions.

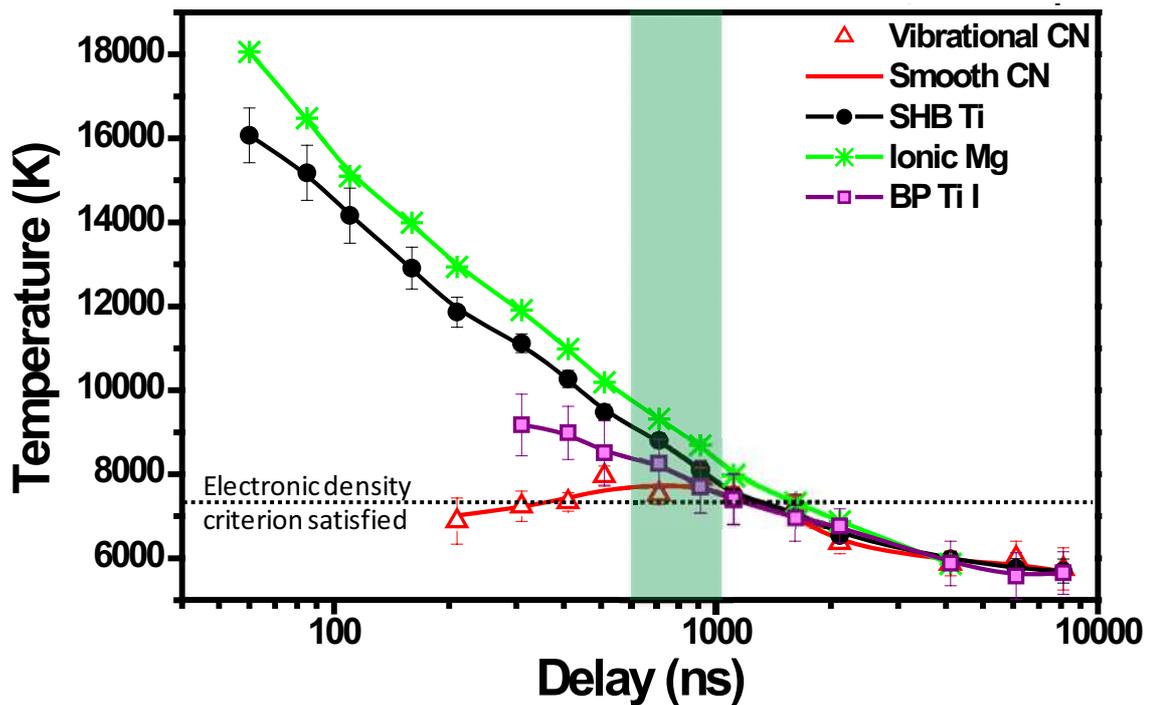


Figure 15. Temperature temporal evolution.

4.6. SIGNAL – TO – NOISE RATIO

To estimate the signal-to-noise ratio, the background intensity I_{bg} is subtracted from the maximal peak intensity of I_{max} of a given spectral line. The result was then divided by the noise, calculated by the standard deviation σ_{bg} of the background emission close to the considered line. The signal to noise ratio is thus given by Equation (3.4) [4]:

$$SNR = \frac{I_{max} - I_{bg}}{\sigma_{bg}} \quad (3.4)$$

Figure 16 shows the temporal evolutions of the SNR of some elements detected in a polymer sample. We can see that the maximum values of SNR are achieved between 300 and 1000 ns if we take the carbon line and the CN line maximums for our limits. The local thermodynamic equilibrium range found before was satisfied between 600 and 1000 ns. The best compromise between these two values regarding our measurements, would be to take start the acquisition after a delay of 600 ns after laser impact for a duration of 400 ns.

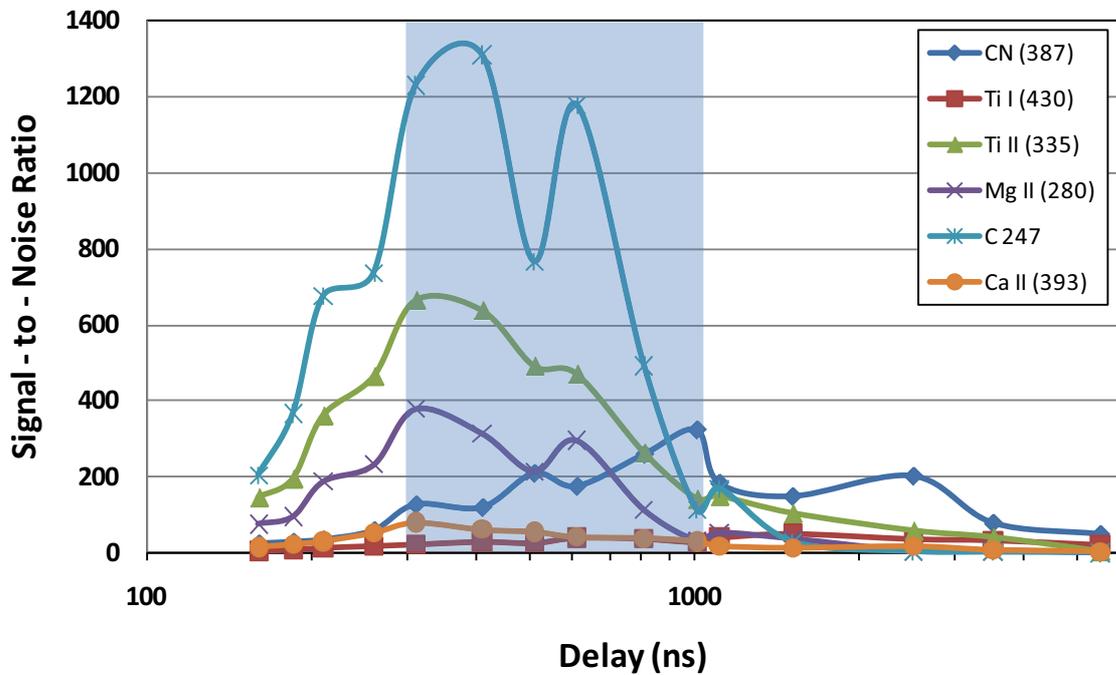


Figure 16. Signal-to-noise ratio for a polypropylene sample at 18 mJ

5. CONCLUSION

Through time-resolved LIBS we observed the evolution of the electron density of the plasma H_α . This allowed us to measure the different temperatures of the plasma using Boltzmann or in a more advantage way, Saha-Boltzmann plots that provided us with more accurate temperatures due to the larger energy range by using emission from both ions and neutral atoms. We were also able to estimate the molecular temperature for CN and C_2 molecules. The convergence of the different types of temperatures and the sufficient electron density ensure the condition for the local thermodynamic equilibrium.

For the following works concerning the coupling of LIBS with artificial neural networks (ANN), local thermodynamic equilibrium conditions are not what are important for us. For LIBS to find its place in the industry it needs to be efficient i.e. provide accurate results in a minimal interval of time. For this, it needs to identify polymer samples using their essential and optimal information, which means at the best signal-to-noise ratio. The fact that this temporal range coincided with the local thermodynamic equilibrium range is reassuring as local thermodynamic equilibrium provides the best conditions for the extraction of spectral information. Therefore, our time-resolved LIBS analysis presented in this chapter has led us into choosing our ideal experimental configurations for the next step. The LIBS spectrum will be recorded 600 ns after the laser pulse impact upon the sample during 400 ns. ($t_d = 600 \text{ ns}$ and $t_b = 400 \text{ ns}$).

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CHAPTER IV. DISCRIMINATION AND IDENTIFICATION OF POLYMERS USING LASER-INDUCED BREAKDOWN SPECTROSCOPY

1. INTRODUCTION

1.1 POLYMERS

Polymeric materials are used in our everyday life in many forms, from disposable containers for household products to construction materials used industrially. The main building blocks, except for a few exceptions, are carbon and hydrogen, but other elements such as oxygen and nitrogen are very commonly found as well. These main elements form monomers, which are essentially organic based (C, H, O and N). The specific ways in which these chains are aligned and twisted together define the material's physical properties. If the alignment of the monomers is random it gives rise to amorphous plastics, which are known for their impact strength and toughness. Examples include polyvinylchloride (PVC), polycarbonate (PC), and polystyrene (PS). If the monomers are

orderly and densely packed it leads to crystalline polymers, which are less flexible and have lower elongation properties than amorphous ones. Polyamide (PA), polyethylene (PE), and polypropylene are some examples. However, even though the physical properties are important for their diverse uses, it is also essential to understand what kind of bonds make up the plastic to help understand how to recycle them.

1.1.1 Thermoplastics and Thermoset Plastics

Apart from how the monomers can be arranged, the nature of the polymer bonds also has an impact on its physical properties. Depending on the nature of these bonds, polymers can be either thermoplastics or thermoset. The bonds are weaker for thermoset plastics giving them the property of being heat sensitive, i.e. they can be melted given the right temperature. This property is extremely useful as it facilitates the recycling purposes. Thermoplastics suffer only chemical degradation when a significant amount of heat is applied due to their stronger bonds. These plastics require more complicated recycling processes that are not yet very attractive to the recycling industries. Even though technically all plastics have the potential to be recycled, for the time being, the focus is on thermoset plastics (especially polypropylene (PP) and polyethylene (PET)). The ones that cannot be recycled are dispatched into landfills or sometimes burned [1].

1.1.2 Plastic Consumption and Production

One can logically assume that if the production of plastics increases the amount of plastic waste increases accordingly. Figure 1 shows the increase in European and world plastic production between 1950 and 2005. Figure 2 shows the evolution of how landfilled and recycled plastics between 1995 and 2005. Even though recycling tendencies have risen, the efforts remain less significant compared to the total waste increase. Improving the recycling process might be one way of reducing this huge difference between the plastic production and waste recycling.

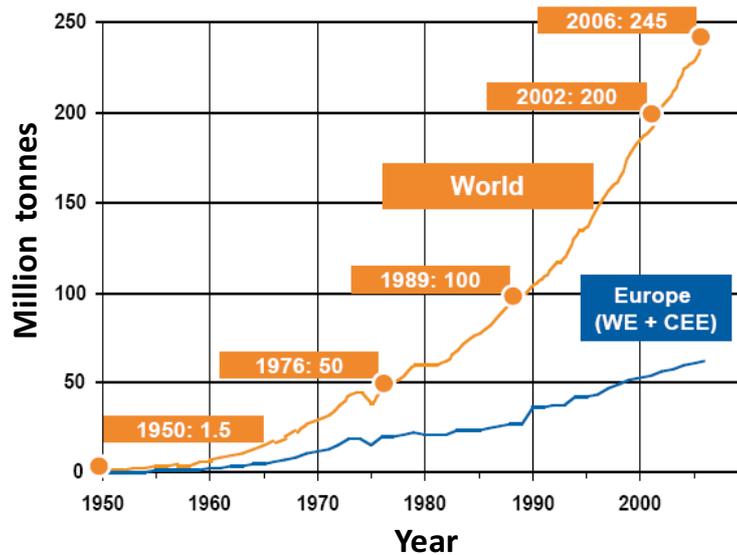


Figure 1. Increasing European and world plastic production. Includes thermoplastics, polyurethanes, thermosets, elastomers, adhesives, coatings, sealants, and polypropylene-fibers but not polyethylene-, polyamide- and polyacryl-fibers [2].

For the time being, virgin plastics remain relatively cheap to produce compared to recycled plastics. Recycled plastics are becoming more attractive from an ecological and soon a financial point of view. Plastics have an extremely long lifetime adding to the landfill overpopulation problem. Burning them leads to the possibility of transmitting potentially dangerous and harmful elements in the atmosphere. The production of plastics consumes a lot of petrol, where it is said that 4% of worldwide petrol consumption is used for the production of plastics and another 3-4% is used in their manufacture. Petrol, being natural resource, is running out and will cause the prices to increase drastically over time.

This tendency is already beginning to show in the past years already, as shown in Figure 2. As petrol prices keep increasing, so will the price of virgin plastics (a). Weekly Gasoline prices in the USA that show an exponential increase since 2002 (b). Therefore, the 1.8 tonnes of oil saved per tonne of PET bottles recycled might not be worth much now, but with future sky rocketing oil prices, the save will become more valuable. Other solutions other than recycling may exist, nonetheless, it has a positive and non-negligible participation to the global efforts for an ecological solution for the future. Figure 2c shows the increase in recycling of PET bottles between 1995-2005 but also the increase in the amount that are landfilled. [3]

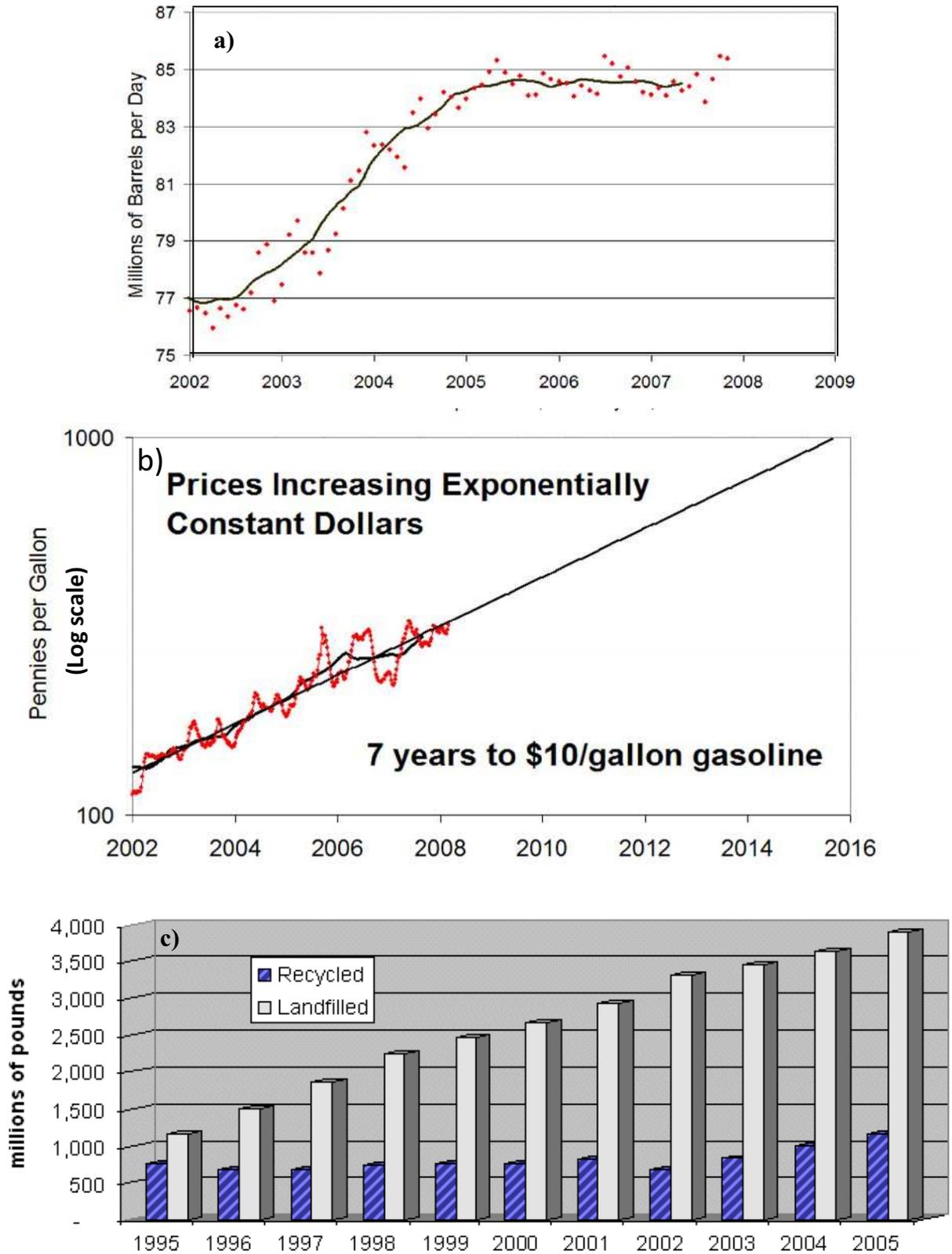


Figure 2. a) Total world oil production [4]. b) Oil prices since 2002 [5]. c) Evolution of PET bottles [4].

1.2 RESIN IDENTIFICATION CODE

Therefore, industrial and domestic waste recycling offers an essential contribution towards environmental protection through sustainable development all the while providing economical benefits. The first step of recycling starts with sorting, whether it is at a domestic level or an industrial level. The American Society of the Plastic Industry (SPI) introduced a resin identification code system in 1988 to identify the main polymers encountered in the residential waste stream [6, 7]. Their aim was to help people sort their domestic waste to initiate recycling in homes. The code brings together the six main polymers that are mostly used in the household. Each number is associated to the type of material used and are listed in the Table 1.

<i>Code System</i>	<i>Associated Plastic</i>	<i>Examples</i>
	PETE (polyethylene terephthalate)	Water and soft drink bottles
	HDPE (high density polyethylene)	Household bottles (cosmetics, food, detergent), industrial wrapping
	PVC (polyvinyl chloride)	Medical bottles, water pipes, window panes, credit cards, flooring
	LDPE (low density polyethylene)	Garbage bags, food wrap
	PP (polypropylene)	Yoghurt containers, margarine containers, milk and beer crates
	PS (polystyrene)	Disposable cups and plates, egg cartons
	Other (the rest of all plastics)	Plastic mixtures

Table 1. Resin identification coding system implemented by the Society of Plastic Industry (SPI)

Due to blending incompatibilities, different plastics cannot be mixed together when recycled. When different types of plastics are melted together they tend to phase-separate, like oil and water, and produce layers. The phase boundaries cause structural weaknesses in the resulting material, meaning there exists only limited useful applications for polymer blends. A main concern is the emission of toxic gasses due to certain combinations such as PVC and PET. Therefore, to avoid dangerous outcomes to both the employees and the environment, they must be identified and sorted beforehand. Another barrier to recycling is the widespread use of dyes, fillers, and other additives in plastics. If the separation of fillers is not carefully optimised the plastic can be damaged in the process. Additives are less widely used in beverage containers, allowing them to be recycled more frequently [8].

1.3 SORTING TECHNIQUES

Even though sorting can start in the household, large scale and more refined methods are required at an industrial scale. This leaves sorting industries with a choice between two methods: manual or automated sorting.

1.3.1 Manuel sorting

Manual sorting consists of employees manually looking for the Society of Plastic Industry resin code identification on arriving plastic on waste stream belt as shown in Figure 3. (a) Employees first sort out the whole waste stream taking out things that should not be present. (b) Once plastics have been separated from the basic waste stream, a team does extra sorting to separate the different kinds of plastics. Once the code is found, the plastic is disposed in the specific batch with other like materials. This method has not proved to be very efficient for many reasons. Firstly it is time consuming as the resin coding system is not always easy to identify, especially with crushed and damaged containers. Secondly, employees are in direct contact with the waste stream, where appropriate protection is needed due to the heavy bacteria filled environment. This task is physically enduring due to the long hours and repetitive movements. There are also possible toxic gas releases if PVC is wrongly mixed with PET bottles [9, 10, 11]. The non-negligible human errors are urging industries to find automated and cost-effective methods to perform reliable sorting of plastic wastes.



Figure 3. a) Pre-manual sorting to remove big contaminants. b) Manual plastic sorting.

1.3.2 Automated Systems

Automated systems that provide fast identification are the best answer to these industrial demands [12]. Pre-sorting solutions can be found within the optical sorter technologies. Their ability to detect plastic materials from paper, metals and glass rapidly is very useful [13]. They are most commonly used in the food sector to separate plastic bottles from other materials and may also provide colour detection. The second step, which is polymer sorting, is done with more sophisticated spectroscopic methods. There are a few techniques that stand out for the moment [14].

a. Near infrared (NIR) spectroscopy

One of the most commercially available technologies, operating in a spectral range of 700 – 2500 nm [15-18]. It identifies the molecular structure of polymers by detecting their reflection or transmission spectrum. The identification is done quickly with a response time sometimes in the order of 100 ms. Despite its accuracy and rapidity, it cannot analyse black or very dark samples. Most of the radiation gets absorbed leaving hardly anything reflected back to allow the analysis to take place i.e. featureless spectra [19, 20]. This represents a problem, for example for the automobile industry where black or dark plastics are largely used. This technique also needs to be in contact with the sample to identify it.

b. X-ray fluorescence spectroscopy

Mainly used in the industry for its ability to detect a single element: the chlorine atom. When a plastic stream is predominantly filled with PET bottles it can quickly detect the PVC bottles that can be scattered by making them stand out [19]. There are two kinds of X-ray detectors in the industry [7,20] :

- XRF (x-ray fluorescence): the x-rays are detected once they bounce off the surface of the bottles. PVC bottles that are shielded by other PET bottles are not detected with this method.
- XRT (x-ray transmission): the detection is done once the x-rays pass through the bottles. More efficient due to its ability to detect PVC bottles even if they are hidden behind other PET bottles.

The disadvantages of this technique are its ability to only separate PVC bottles from the plastic stream and the security risks associated with x-rays. X-rays are a form of radiation and special precautions are required to protect employee exposure. X-ray detection systems need to be cleared and registered by the Nuclear Regulatory Commission (NRC).

c. Raman spectrometry

A spectroscopic method that uses monochromatic laser light (mostly visible or NIR) that is scattered once it is reflected upon the Raman sensitive samples such as a plastic surface. The re-radiated scattering of light causes differences with the original monochromatic light that puts forth the different molecular structures of each polymer [21, 22]. The detection time remains relatively too long to use for industrial uses.

d. Laser-induced breakdown spectroscopy (LIBS)

As already recalled throughout this thesis, LIBS has strong potential that should meet the requirements of the industry. It is not limited by the colour of the plastic like NIR spectroscopy, i.e. dark and black samples. It can also take things a step further by detecting metallic trace elements, fillers, reinforcements and additives [12]. With its intrinsic advantages, LIBS can become an attractive technique, particularly when combined with other existing ones for online and real-time identification of plastics [1, 23]. In this chapter we will especially demonstrate that a large LIBS spectral range combined with the artificial neural networks (ANN) [13, 15] provide reliable and fast identification of different types of plastics.

2. SAMPLE PRESENTATION AND SPECTRAL RESULTS

The same experimental setup as demonstrated previously in Chapter 3 was used here. It was found that a delay of 600 ns and a detection gate width of 400 ns provided the highest signal-to-noise ratio (SNR). In this detection window, the plasma can also be considered homogeneous and in local thermodynamic equilibrium (LTE). Statistical series were then performed with these detection parameters on the polymer samples. This means that for

each sample, two sets of 100 spectra were recorded in the same conditions with the same experimental parameters at an integration time of 1 second per spectrum. To account for the possible environmental and experimental changes that can exist in the industry and affect a LIBS setup, the first set was taken a few weeks before the second one. This was the only difference between the two. In order to compare our experimental conditions to those used during routine analyses, each spectrum was recorded with the short integration time of 1 second, which corresponds to the accumulation of 10 laser shots. Samples were cleaned with methanol before measurements. Finally, the data analysis was done with the help of the LIPSA software developed in our research team.

2.1 SAMPLE PRESENTATION

LIBS measurements were first done on the eight following reference samples (Figure 4), which cover different varieties of polymers used domestically and industrially. A combination of dark, white and transparent samples is used in our studies. The common element being carbon and hydrogen (except for PTFE which has fluorine instead of hydrogen). This set contains polymers within the resin identification code (PP, PVC, PE500) along with a thermoset plastic (PTFE), a polyamide (PA 6,6), and other plastics commonly used in the industry. They are homogeneous sheets of plastic and their colour ranged from transparent to dark and opaque. Prior to our experiments all the samples were initially smooth and cleaned with methanol.

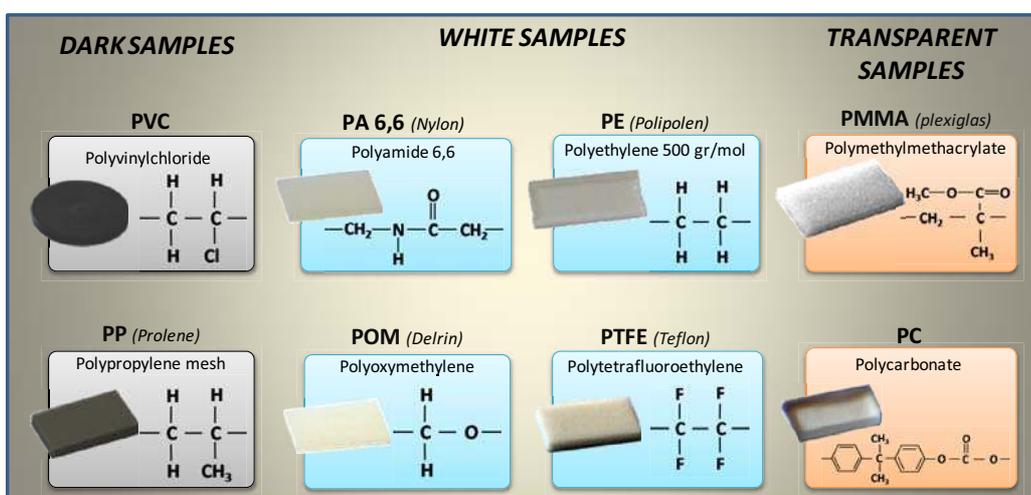


Figure 4. Polymer reference samples

2.2 EXPERIMENTAL SPECTRA

Typical LIBS spectra are presented for PVC, PTFE, PP and PC in Figure 5. A broad spectral range between 240 nm and 820 nm permits the detection of spectral lines used to identify different atomic elements and molecular bands. Carbon and calcium are saturated in the graphs in order to accentuate the intensity of weaker spectral lines. The molecular elements, such as CN and C₂ bands, are more likely to be seen in the ultraviolet region. Most of organic elements, such as hydrogen, nitrogen, and oxygen, can be observed in the near infrared range. We can also notice the presence of trace elements such as Ti, Mg and Ca, visible in the ultraviolet range for PP and PVC, while none were observed for PTFE and PC. These metallic traces are often used to reduce production costs and/or to improve physical and mechanical properties [24]. The atomic elements, such as chlorine and fluorine, are observed in the visible and infrared regions. The absence of hydrogen in PTFE can also be noticed. PC and PMMA, which are both transparent samples, did not contain any trace elements or charges. This absence can also be noted for the white PTFE sample.

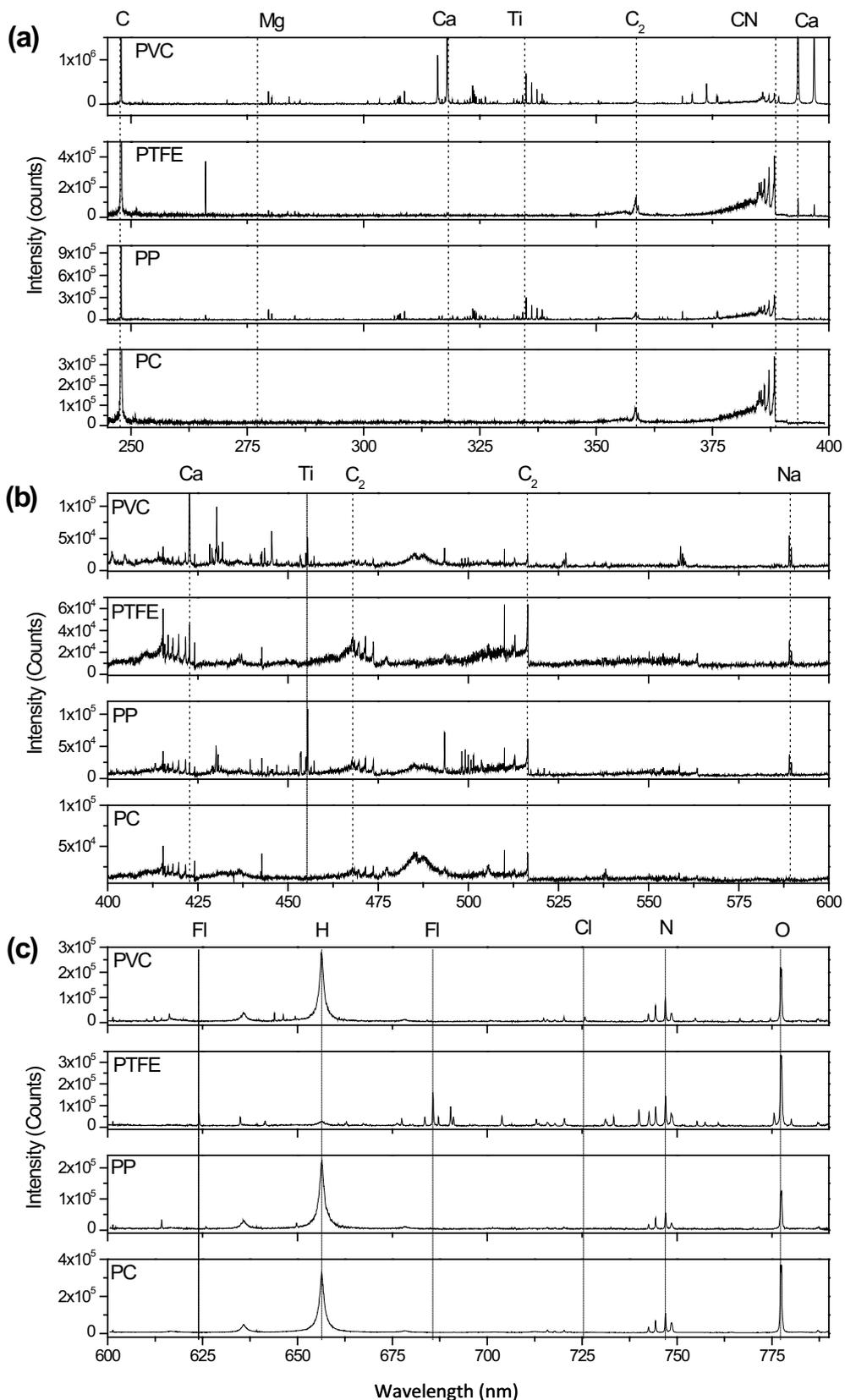


Figure 5. Spectra of PVC, PTFE, PP and PC in the a) Ultraviolet, b) Visible and c) Visible and near infrared ranges.

3. SPECTRA ANALYSIS

3.1 STOICHIOMETRIC RATIO

At first, we were curious to see the difference between the C/H experimental and theoretical ratios. The stoichiometric ratios from the molecular formula of the matrices were calculated for each sample as shown in Table 2. The theoretical variations of C/H go from 1/2 to 2/3 with the exception of PTFE, which contains no hydrogen. The experimental ratios measured using the H_{α} spectral line at 656 nm and the C line located at 247 nm were calculated and compared with the theoretical ratios in Table 2 and displayed in Figure 6. Other elements present in the matrix could be used for this ratio, but given that carbon and hydrogen are the main chemical components, it would make more sense to use them.

Sample	H/C theoretical ratio
PVC	0.132
PP	0.163
POM	0.175
PTFE	0.000
PE	0.163
PA 6,6	0.156
PMMA	0.133
PC	0.066

Table 2. Theoretical ratios of carbon atoms with hydrogen atoms in the different polymer matrices.

In Figure 6 we traced the best fit line and notice a large dispersion with regards to this line. The dispersion ($R^2 = 0.75$) of the different polymer samples indicate clearly that there no linear correlation between the two. This means that we could not use calibration curves to identify samples according to their organic elements. One of the reasons for this dispersion could be due to the strong matrix effects. Another reason could lie behind the fact that the experimental carbon line includes the carbon concentration not only from the molecular matrix but from the additives and fillers as well. This spectral line was also very intense and possible auto-absorption could have been present. These limitations confirmed our doubts regarding the difficulties associated between organic material and calibration curves and encouraged us to find solutions with the use of multivariate chemometric methods.

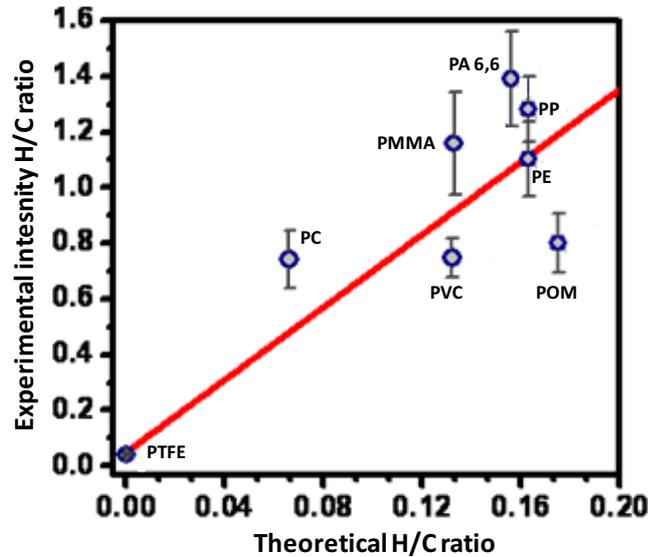


Figure 6. Theoretical H/C intensity ratios compared with the experimental data.

3.2 PRINCIPLE COMPONENT ANALYSIS DATA PROCESSING (PCA)

Multivariate chemometric methods are particularly helpful when dealing with data that have a lot of variables. A simple form of multivariate data is a matrix with N rows and M columns. The columns correspond to the objects, which in our case are the samples, and the rows correspond to certain features or variables of the objects such as the spectral line intensities. It is the easiest form to stock up a lot of data with a lot of features. The aim is to focus on the most important features and disregard all the information that is useless to the analysis. One of the common chemometric methods is the Principal Component Analysis (PCA). PCA is a very useful statistical technique that finds patterns in data that may contain a large number of variables. The aim is to re-express the data in the form of linear combinations of its basis vectors and then reducing the amount of variables by keeping only the most important. The procedure contains the following main steps. This method has provided appropriate results when used to discriminate between bacterial spores, molds and pollens [31, 32], characterisation of Iron Age pottery [27], and used in considerable success for the treatment of LIBS data obtained from rock [28, 29]. Godoi et al. [30] used PCA combined with LIBS to determine the levels of toxic elements in children's toys. Lasheras et al. [14] used this combination for the identification of polymers using an Nd: YAG nanosecond laser at 1064 nm. Their experiment was not meant for real-time identification.

3.2.1 PCA procedure

a. Gathering the data

The 100 polymer spectra were averaged together to obtain 10 spectra per sample. The data from the spectra are gathered up to form a $N \times M$ matrix. The intensities at different wavelengths of a spectrum of a given sample are placed in a column m ($m = 1$ to M). The same is done for the different spectra then for different samples. The data are thus stacked together in M columns, where each column contains information of a particular spectrum of a given sample. Therefore, the 8 samples provided 80 columns ($M= 1$ to 80), each with 14 different spectral intensities to identify corresponding to N rows ($N = 1$ to 14) to represent the features of the spectra as can be seen in Table 3.

N=14 Spectral Intensities	M=80			
	Sample 1 (10 spectra)		...	Sample 8 (10 spectra)
λ_1	N_1M_1	N_1M_{10}	...	N_1M_{80}
\vdots	\vdots	...	\vdots	\vdots
λ_{14}	$N_{14}M_1$	$N_{14}M_{10}$...	$N_{14}M_{80}$

Table 3. PCA input matrix. This matrix represents the values of the averaged 10 spectra per sample (8 samples in total so $M=80$). There are a total of 14 spectral lines (N) chosen to identify the samples.

b. Subtraction of the mean and normalisation

This step allows for different line intensities to be analysed together by rendering them dimensionless. For example in our case, the chlorine line is very hard to detect but has great importance in the identification of PVC. To proceed, we need to centre and re-dimension the data at a certain wavelength for all elements. The mean \bar{X} is calculated for each column m ($m = 1$ to M) by

$$\bar{X}_n = \frac{1}{M} \sum_{m=1}^M X_{nm} \quad (4.1)$$

and is then subtracted from all the variables of the column to produce a set of data with a zero average value. The result is further normalised by dividing by the standard deviation of the column:

$$\sigma_m = \sqrt{\frac{1}{N} \sum_{n=1}^N (X_{nm} - \bar{X}_m)^2} \quad (4.2)$$

giving the centred and normalised scores:

$$X'_{mn} = \frac{(X_{nm} - \bar{X}_m)}{\sigma_m} \quad (4.3)$$

c. Construction of the covariance matrix

Covariance is defined as a means to present the way two multi-dimensional variables covary with respect to each other. It is calculated using the following formula:

$$cov(X, Y) = \frac{1}{N-1} \sum_{n=1}^N (X_n - \bar{X})(Y_n - \bar{Y}) \quad (4.4)$$

where X and Y represent two specific samples, \bar{X} and \bar{Y} the means values of the normalised intensities for the samples X and Y respectively (for example X =PVC and Y =POM then for a next study X =PVC and Y =PE). The covariance calculation between the 8 different samples generates a matrix of $N \times N$ (with symmetrical elements with respect to its diagonal). The covariance matrix is used to characterise the scatter of data. The procedure of PCA consists in diagonalising this matrix with the aim of putting forth the most important features in the data.

d. Diagonalising the covariance matrix

The aim is to diagonalise the covariance matrix by using its eigenvectors to change the vector basis, accentuating the most important features. The eigenvalues and eigenvectors are real and positive values because the covariance matrix is square and symmetrical. The eigenvectors are orthogonal and linearly independent therefore uncorrelated. This removes the redundancy previously encountered and in order to put forth in the form of a graph the differences and similarities amongst the data. This should allow us to determine which spectral lines have a real contribution or if some can be eliminated to speed up the process.

e. Choosing the different components

The data compression takes place in this part. The eigenvectors, previously calculated, are associated to the different components. The eigenvector with the highest eigenvalue is also called the principle component. This component represents the feature that varies the most and therefore has the greatest impact. The aim is to choose a number of key components that represent at least 95% of the data so not to lose valuable information. This is called data compression. The main spectral lines are put forth at this moment.

3.2.2 PCA Polymer Results

The results are best seen in the form of a graph in Figure 8, where each score corresponds to a different variable. This was a basic test done with only one iteration. We used the eight different polymer samples (Figure 5) and added 7 different kinds of PVC samples shown in Figure 7. For the first batch of polymer samples, 100 spectra were taken from each one and 60 from the second batch that only included the PVC samples. In Figure 7, each dot is the average information accumulated from 10 spectra (this explains why there are 10 dots for the first batch and 6 for the PVC batch). The results show that a possible discrimination is possible amongst the polymers. Certain are a little more difficult to separate for example PP and PVC. Further PCA analysis is able to separate these samples more clearly.

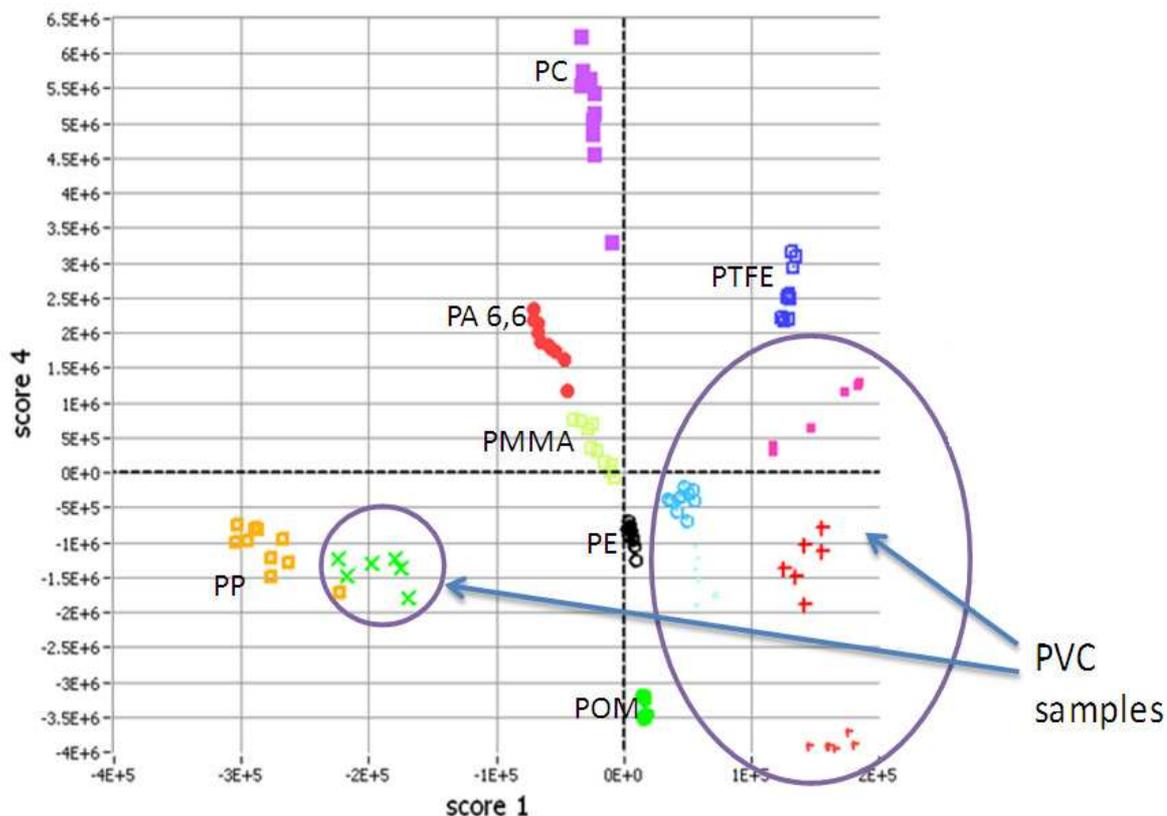


Figure 7. A comparison of two scores using the PCA technique.

Figure 7 gave us the best discriminating results (score 4 vs. score 1) even though we would have thought that it was score 1 vs. score 2 that would be better. Averaging the spectra in our case (equivalent to 10 s integration time) to provide clean results takes time. We decided to exploit artificial neural networks (ANN) that have the possibility to provide better suited results for single spectra real-time identification when correctly optimised [31], especially when the calibration curves are nonlinear [28, 32]. Therefore, even though these results are satisfactory for the discrimination of polymers, due to the non-linear nature of the calibration curves often seen with organic samples, we decided to examine the results provided by artificial neural networks with the same polymer spectra.

3.3 ANN DATA PROCESSING

Artificial Neural Networks (ANN) are aspired to work like biological neural networks that make up the nervous system. The brain remains a much more complex tool than any machine for certain computations, such as motor control, pattern recognition and perception. In other words, it can only take a human brain between 100-200 ms to recognise a person in disguise or in an unfamiliar scenario, whereas it can take the most powerful computer longer to deal with even simpler tasks. ANN have found useful applications in domains where pattern recognition and classification are important, such as face recognition, animation technologies, artificial “intelligence” or autopilot mode in aerospace engineering. ANN combined with LIBS have been already used for soil analysis [32, 33]. Results proved better than PCA especially when the calibration curves had a non-linear nature. Sattmann et al. [18] have used this method with the same industrial application aim as in this thesis but with different ablation parameters (wavelengths of 1064 nm and double pulse ablation).

The aim of the ANN procedure developed in this thesis is to model the way a human brain recognises patterns and therefore apply this model for the identification of polymers. Applications can find their way into the plastic recycling industry where a need to improve their sorting process is required.

3.3.1 Evolution and Principle of the ANN

a. Biological Neuron

The nervous system is made up of billions of specialised cells called neurons, connected in a vast network, creating what is known as a neural network [21]. An example of a simple neuron is shown in Figure 8. The dendrites receive input signals from the ears, nose, skin, eyes and so on. This signal is processed in the nucleus and an activation signal is delivered through the axon towards the axon terminals. The axon terminals can either be connected to another neuron or a muscle tissue, for example, to create a physical reaction. If someone were to pick up something that is extremely hot, their brain would send a signal to their

hand muscles to let go of this object as quickly as possible. The input data is the cup temperature, the hand receptors are the transmitters of this data to the brain, which processes the information and sends back a signal through the neurons to the hand muscles. The reaction (letting go of the cup really quickly) would be the output result. The brain then learns and remembers that it should remain cautious in the future when dealing with this particular kind of object. This kind of life experience trains humans to survive in their natural habitat.

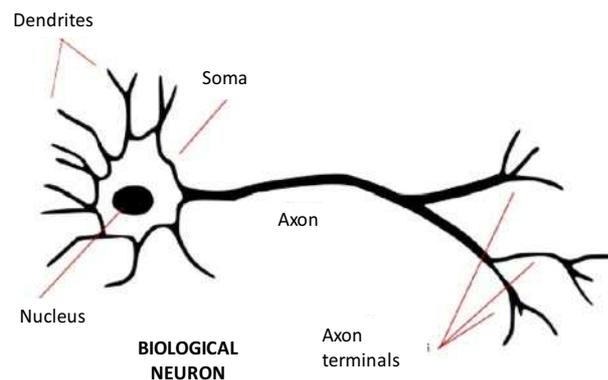


Figure 8. Biological neuron cell.

b. Artificial Neurons

In 1943, Warren McCulloch and Walter Pitts published a thesis, which showed that it was possible to construct neural networks using only mathematics and algorithms, to help solve certain computable tasks [21, 24]. However, this model had many limitations as it could not learn or be trained like the human brain. Information in this case only went forward, so when an error was produced, it had no way of going back to eliminate the cause of error and improve. In 1962, Frank Rosenblatt implemented the notion of synaptic strengths, called *weights* in ANN. Their function is to increase or decrease a certain neural influence. If we take the example of the extremely hot object mentioned before, the person would not have the same reaction if it was only mildly warm. Even though the reaction would not be the same, the neurons that transmit the information are not different, they are still the same. So how can this reaction change so drastically depending on the stimuli? The answer lies in the synaptic weights that transmit this kind of information. Figure 9 represents how Rosenblatt included this weight in the artificial neural unit. Each input value was associated a specific weight that Rosenblatt had to change by hand, one at a time, to

produce the appropriate result. It is important to note here that one neuron unit can receive a number of input values but can only produce one single output value. Rosenblatt called this neural unit a perceptron.

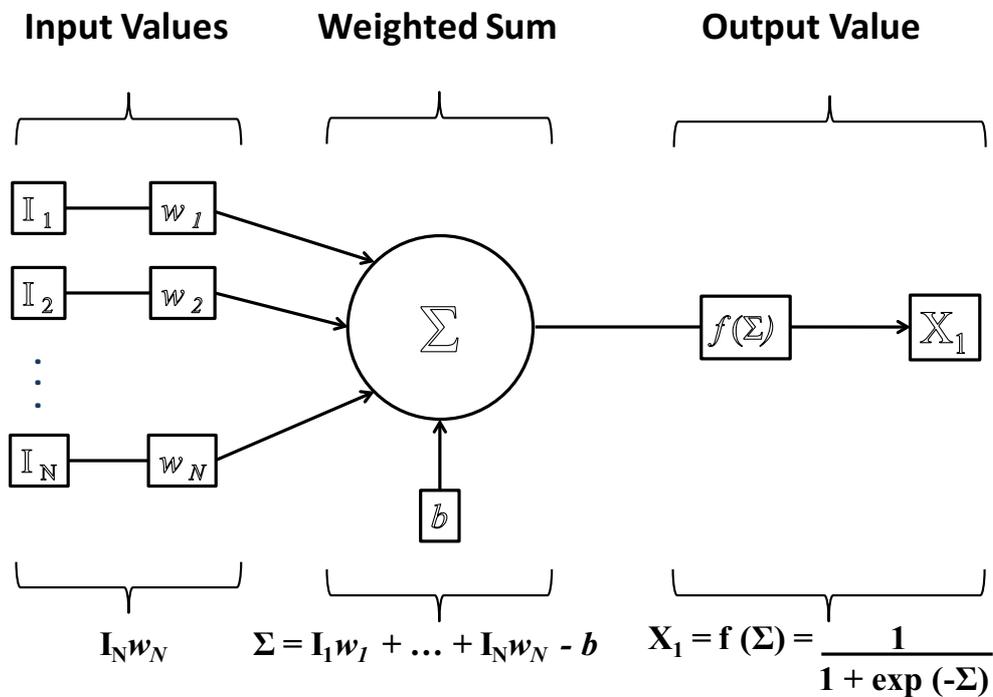


Figure 9. ANN single perceptron neuron unit. $X = f(\sum_i w_i I_i - b)$, where w_i are the weights, I_i the input data, and b the bias. The transfer function $f(u)$ is a non-linear transfer function.

So how does the neural network work? The input values are each multiplied by their specific weight values. Their total sum is then calculated at the core of the unit and a bias is subtracted. This weighted sum is computed into what is called a *transfer function* $f(u)$ to produce the output value X defined by:

$$X = f\left(\sum_i w_i I_i - b\right) \tag{4.5}$$

The bias b increases or decreases the net input value of the transfer function, which can either be linear or non-linear. The most commonly used is a sigmoid function which is a compromise between perfectly linear and a perfectly non-linear threshold function. The different transfer functions can be seen in Figure 10.

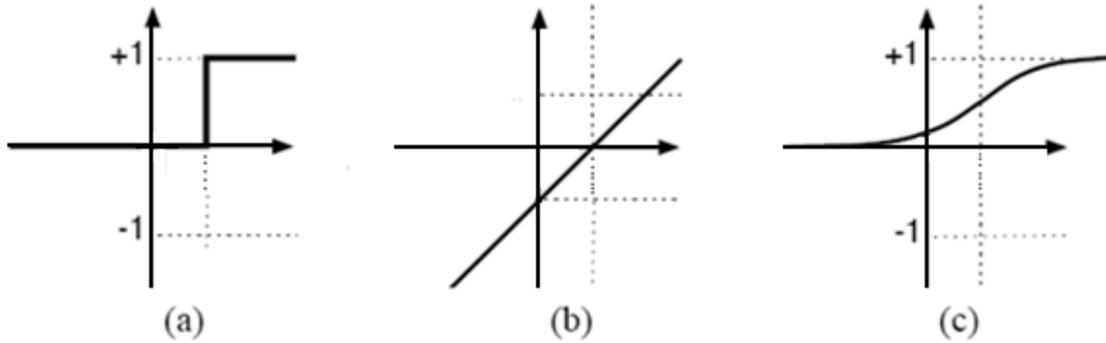


Figure 10. Transfer functions. They take the sum of the weighted inputs to provide a single output answer. There are three main functions used in ANN:

(a) Perfectly non-linear function threshold/ step function.

(b) Perfectly linear function

(c) Non-linear sigmoid function $f(\Sigma) = \frac{1}{1+\exp(-\Sigma)}$

c. Feed-Forward Network

Limitations related to this artificial neuron were put forth in a thesis written by Marvin Minsky and Seymour Papert in 1969 [21]. A great weakness was the incapability of this neuron to solve exclusive OR (XOR, which means one or the other but not both) problems as simple perceptrons can only solve problems that are linearly separable. This problem found a solution by adding consecutive layers of neurons together, known as multi-layer perceptrons. This perceptron can also solve problems that are not linearly separable. A 3 layer network is shown in Figure 11, which consists of one input layer, one hidden layer and one output layer.

Neuron units are placed in parallel layers. The output of one neuron is an input value for the next neurons. Information is fed in only one direction with the exception of errors, which are propagated backwards. Figure 11 shows an example of 13 input neurons and 8 output neurons because this is the multilayer feed-forward network that we are going to use in this thesis. The input neurons and output neurons can fluctuate depending on the requirements of the ANN program.

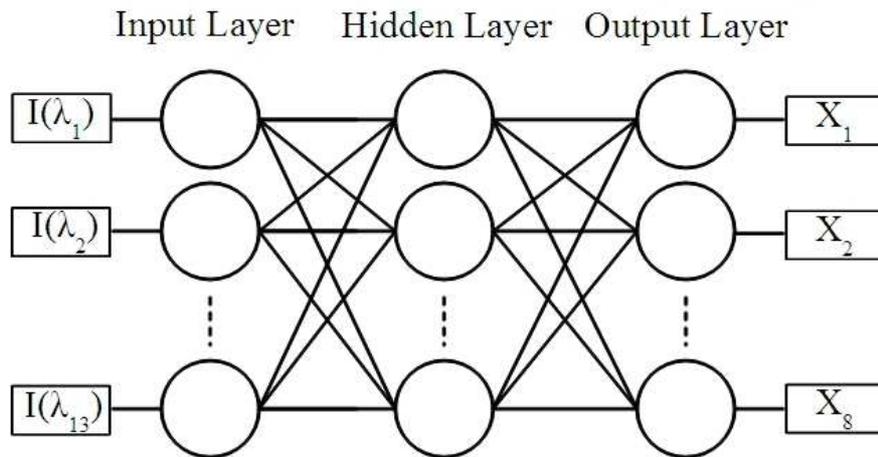


Figure 11. Multilayer feed-forward Perceptron.

This network is called a feed-forward network where information propagates from the input layer and works its way to the output layer without going backwards. Each layer has its specific number of neurons, i.e. if the input layer consists of 10 neurons, the other layers do not necessarily contain 10 neurons. The manner in how these neurons are chosen will be discussed later on in more detail. In an effort not to overcrowd this figure, the weights have been omitted from the drawing. The single output value produced by each neuron is distributed to each neuron in the next layer. The last layer (Output Layer) provides the overall computational values. Again, since one neuron can only provide one answer, the number of output values depends solely on the number of neurons in the Outer Layer.

d. Back-propagation Algorithm

Another great weakness put forth by Minsky and Papert, was the difficulty in training such a neuron. This consisted of changing the weights by hand to provide the expected output results. This meant that the programmer needed to change each value one at a time to get the expected results. One cannot tell which one should be changed and by how much in advance. Just imagine the example of the 3 layer network in Figure 11 with all the different weights it encloses. The time it would take to firstly identify the weight or weights that needed to be modified and then find out the correct variation needed in order to produce the proper result would be horrendous! They demonstrated that this training method became very difficult to handle if the number of input values was high. In fact, the

training time would increase exponentially with every input value added, greatly handicapping the efficiency of the network i.e. it would take more time training the perceptron than actually doing the calculations oneself. The ANN had great potential at this point but could not be properly exploited just yet.

The invested time would not be worthwhile unless it was performed by an algorithm (for certain computations, a computer remains much faster than the human mind, given it is properly used). It was only in 1974 that Paul Werbos introduced the back-propagation algorithm used to train the multilayer feed- forward network. The errors generated by the ANN were estimated and were allowed to travel backwards to previous layers to make the appropriate changes automatically to the weights in question. This is the only time where information is fed backwards. This back-propagation algorithm removed a tremendous burden off the programmer. Iterations and tedious calculations were executed automatically by the computer and training was done much more quickly. After this point, ANN started to gain a significant amount of interest from the industries and their applications exploded. These newly enhanced ANN became efficient i.e. they now had the ability to mimic the way biological neurons work mainly in two ways:

- gain knowledge from its surroundings through a learning/training process
- interneuron connection strengths (weights) have the ability to store this gained knowledge

The ANN once created can be considered as a child. It has the ability to understand but needs to be taught and fed information. A normal child has the possibility to learn any kind of language. It learns from its environment to pronounce the words and how to associate them together. Without this environment, it would not be able to learn and integrate himself. A child cannot start speaking German all of a sudden if it has heard only English all of its life. ANN gains knowledge from its environment through a learning process, and stores this learned information in its weights for future use. Now, depending on what is expected of the ANN, different learning processes can be used. The aim in this thesis is pattern recognition so I will focus on the back-propagation algorithm used in supervised training and not go into all the other learning processes.

e. Pattern Recognition

Pattern recognition is formally defined as “the process whereby a received pattern/signal is assigned to one of a prescribed number of classes” [21]. To state a little differently, the input data is analysed according to its traits (feature extraction) and then is correctly placed in one category or another. The ANN needs to learn and memorise the traits related to each category through a learning/training phase. In the different families of polymers stated in the experimental section (Figure 5), certain present traits at an elemental level that distinguish them almost immediately (chlorine, fluorine, oxygen, and nitrogen). Others present molecular specificities such as aromatic cycles, aside from the basic carbon and hydrogen bonds. Air is mainly composed of molecules of nitrogen (78 %) and oxygen (20%). Since the ablation is done in the presence of air, all samples will contain traces of these basic ingredients. It also contains humidity (vapour of H₂O) that varies from day-to-day depending on the climate. Without this training process, the ANN cannot acquire the right experience required to identify different polymers efficiently.

f. Supervised Training

The most suitable training method for our case is called supervised training. A teacher or optimiser, transfers his knowledge to the ANN. Like a child, it is otherwise incapable of providing accurate results without the proper teachings. The ANN still does not know what results it needs produce. However, it does know how to produce the results asked of it by changing its weights. The back propagation algorithm estimates the error between the expected values (information provided by the teacher) and the calculated value (results provided by the ANN). This estimation ε is then sent backwards to allow the appropriate changes to be made to the weights where necessary. The mean square error (MSE) is a risk function, which is used to quantify the error between the two values. Figure 12 puts forth a simple model explaining each iteration. The input data is multiplied by weights initially chosen randomly. The ANN processes the information and provides the calculated value, which is the training value. This value is compared to the expected value to provide the training error ε that is propagated backwards. This error is taken in consideration to make the necessary change to the weights in question and the cycle continues until an acceptable training error is reached.

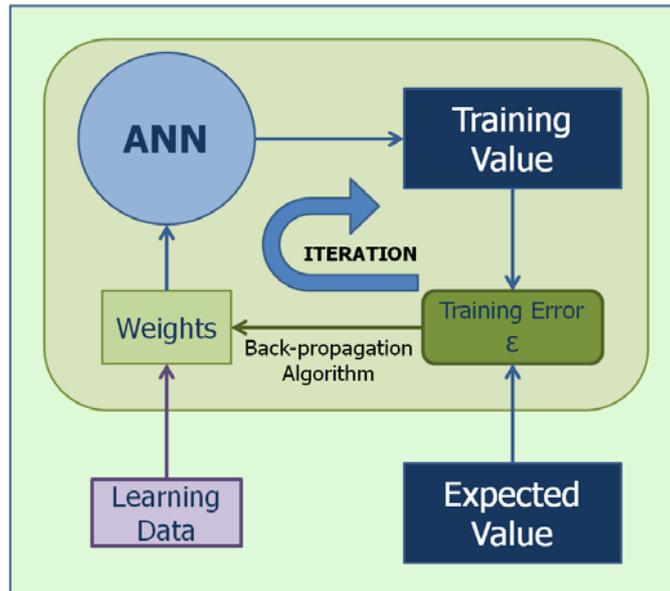


Figure 12. Supervised Training.

The aim of the back-propagation algorithm is to tend the training error ϵ towards a minimal value as can be seen in Figure 13. These iterations will converge until it either finds a minimum (sometimes impossible) or until a fixed limit previously defined is reached. The teacher fixes the limit by either it telling it to stop converging after a certain number of iterations or by choosing a satisfactory MSE. Most commonly a MSE limit is chosen to ensure that the error is small enough to provide good results. A certain reasoning must be made when choosing this limit that will be explained later. The weights are fixed and will no longer be touched once the learning phase is over. The mean square error quantifies the training error between the expected value and the calculated value. After each iteration, the ANN provides answers that are closer to the expected value. A back-propagation limit stops the convergence when it has attained the iteration or MSE limit.

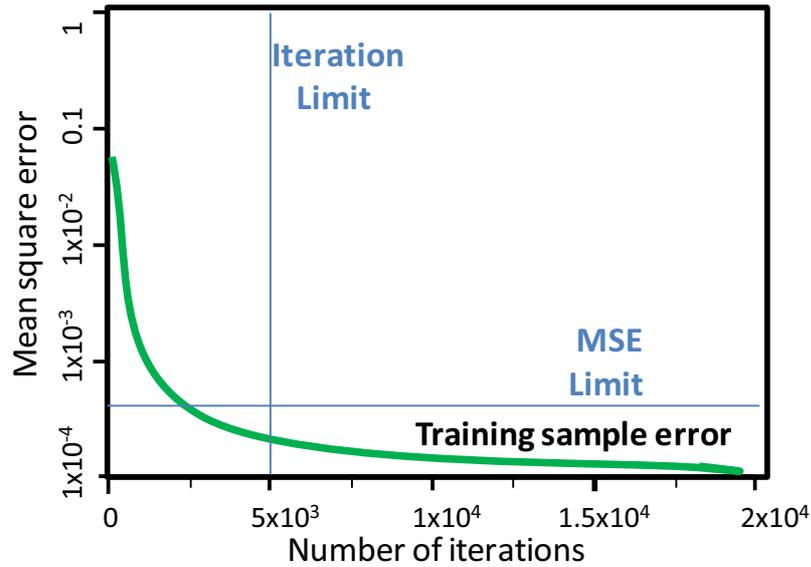


Figure 13. Training Error ϵ .

g. Validation Phase

When the learning phase is over, it is followed by what is called a validation phase. The validation phase uses the same basis as the data used during the learning phase but taken at different moments. The ANN processes this data and provides results in the same way as indicated in Figure 12 with a main difference. This time the weights are defined according to the optimisation done by the teacher during the learning phase as the ANN internal parameters, the data from the learning phase (training value) is compared to the validation results. Again, the MSE is used to quantify the errors between the two, providing us with a validation error ϵ' .

This method of optimisation is called the cross-validation method and an illustration is shown in Figure 14a. This method compares the training output with the validation output and generates a validation error ϵ' . For this validation phase, the aim is to optimise the internal parameters of the ANN i.e. the number of neurons in each layer, the back-propagation MSE limit (Figure 13), and a threshold value required for real-time identification. The latter will be explained in the optimisation phase. The error ϵ' needs to be at its minimal value for the right optimisation as can be seen in Figure 14b. Changes are made to the ANN parameters and the cycle continues. The aim is to use this error to identify the moment where the internal parameters are optimised.

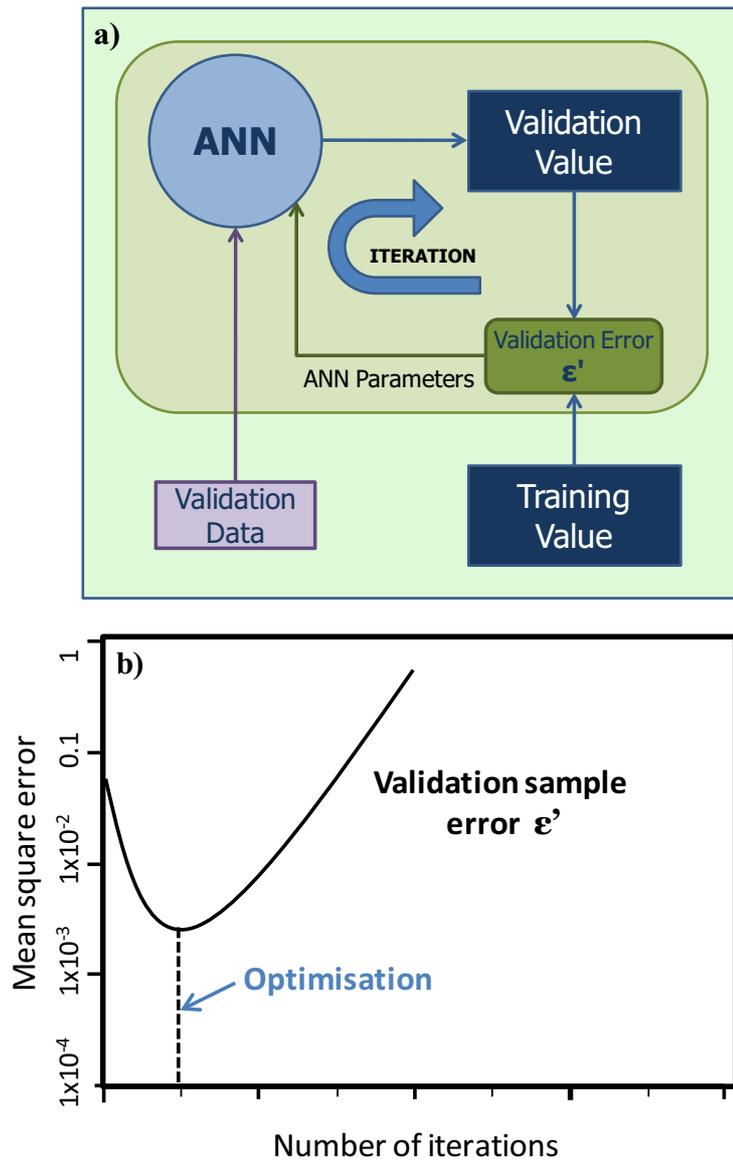


Figure 14. a) Cross-validation method. b) Validation sample error ϵ' .

h. Over-learning and Under-learning

The back-propagation MSE limit is calculated by finding the minimal value of the mean square error ϵ' associated with the cross-validation method (Figure 14b). Figure 15 plots both MSE values, ϵ and ϵ' , at the same time. The learning error, ϵ , continues to decrease with the number of iterations as the validation error, ϵ' , shows a very clear minimal value. This minimal value links both graphs because it allows us to find the back-propagation

limit where the most efficient results are provided. After and before this limit we can see that the validation process starts generating more errors despite the fact that ϵ continues to decrease. We need to avoid these two scenarios for the most efficient processing.

When an ANN starts to learn, it begins by memorising the data and associates patterns to each sample. As ϵ decreases, the memorisation becomes more and more acute. When the ANN is allowed to converge after this limit, it will start memorising random noise lines as potential patterns when they should just be completely random [21]. This is why errors start to generate and we experience over-learning of the ANN. Before this limit, the algorithm was not given enough time to converge and memorise the useful information, lacking the essential tools needed to analyse the samples. This is otherwise known as under-learning. Just looking at Figure 13 is not enough to determine both limits, while ϵ' indicates the right MSE value for the back-propagation limit.

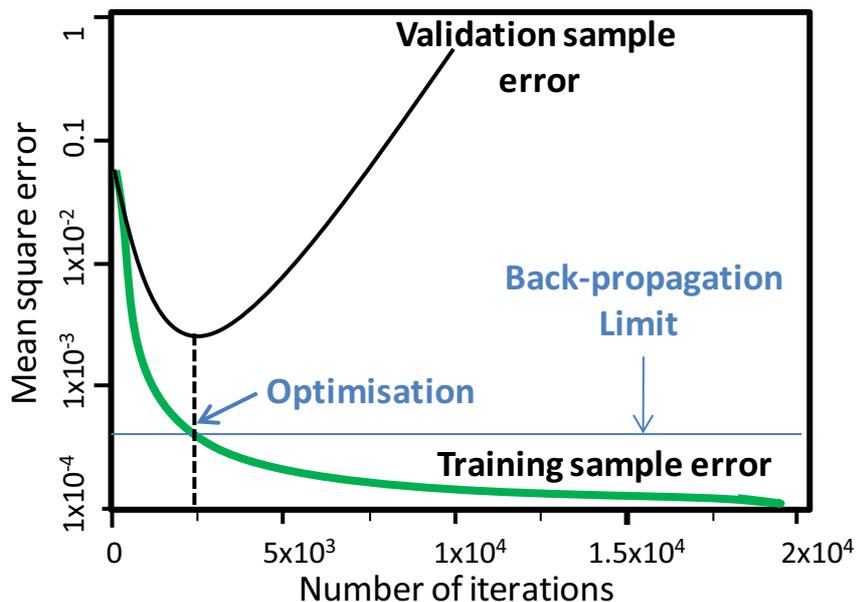


Figure 15. Cross-validation error and back-propagation error by plotting both MSE values plotted.

Another way of understanding over or under-learning can be seen in Figure 16. Ideally, what we would be looking for is the best-fit to the line in blue. If the ANN was under-

learning, the fit would probably just show an average of the blue lines by tracing the green line. If the ANN was over-learning, the fit would find a complex polynomial to pass through each point shown by the red line in the Figure 16. Therefore, it is important to note that the training and optimising of an ANN revolves around find the minimal validation error. If not, the ANN will either under-learn or over-learn, both generating inadequate results. The broken line represents the best fit line that is ideally sought after. The horizontal line represents the under-learning fit (an average of all points). The line that passes through every point represents over-learning (polynomial fit). The back-propagation limit is used to avoid these two scenarios. The other ANN parameters are fixed using the same cross-validation method. They will now be explained in more detail in the next part.

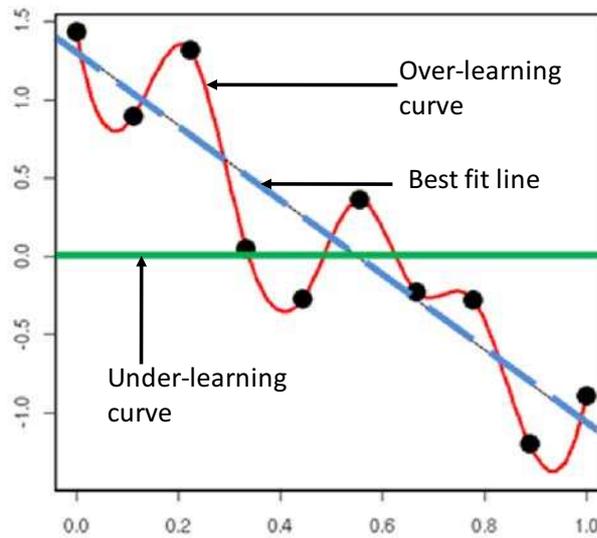


Figure 16. Over-learning and under-learning.

3.3.2 Optimisation of ANN parameters

Thanks to optimisation methods, ANN have many purposes and applications in various domains i.e. for material identification [18] or quantitative analysis [33]. The multilayer perceptron allowed them to provide solutions for non-linear separable problems especially since the back-propagation algorithm that revolutionised the training process. The purpose of training and optimising ANN, is to allow it to use past experiences to correctly identify

future never-seen-before data. The dangers of over-training and under-training must be avoided as they can lead to parameters that are inadequately chosen, rendering the results useless and at best misleading. The optimisation of the ANN has to go through two steps. The first requires the optimisation of the weight values (learning/training phase) and the second step is the optimisation of the different ANN parameters (validation phase).

The core of the algorithm used in this thesis was initially developed by researchers at the Canadian Space Agency (CSA) [34]. Due to the difference in nature of what we need to analyse, the ANN has been modified and adapted to our needs. The supervised training and optimisations previously done by the Canadian Space Agency cannot be applied to our case and needs to be modified. This means we need to re-optimize the weights, the number of neuron units in each layer, the back-propagation limit, and threshold values. Once our training and validation phases are completed, our next goal will be to test the ANN with data from PVC samples not used during the training or validation process, otherwise completely unknown to the ANN. ANN are used to help us but they need our guidance and validation to accomplish their tasks correctly since they are only mathematical tools.

In the previous section, I explained how a 3 layer feed-forward network is trained. In this section I will give detailed explanations on how our ANN was trained and in turn optimised. Firstly, our input data comes directly from the spectra of each polymer. A selection of lines are needed initially for the input layer otherwise the ANN has no data to start with. As explained before, the time it takes to train an ANN (optimize its weights) is greatly influenced by the number of data input. Each spectral line is given its own neuron in the input layer. Considering the 25,000 points provided by our Mechelle spectrometer to its broad range and resolution, one can only imagine the amount of neurons and the considerable time it would take to train. The polymer spectra, shown previously in Figure 6, show that samples studied do not contain lines that are present and significantly important throughout this range. Therefore our first priority, before any other ANN parameter, was to choose characteristic lines that represent all the polymers. An insufficient amount of lines or randomly chosen ones will not provide the ANN with the necessary information required to fulfil its task efficiently (under-learning). If too many lines are fed into the ANN, it might lead to over-learning, in addition to wasting time.

a. Input Neurons

We started to test the program with 24 lines shown in Table 4 as input value, which correspond to 24 neurons in the input layer. Other than the obvious C, H, O, N lines, this group included lines from non metallic elements such as chlorine and fluorine, molecular band lines such as CN and C₂, along with lines from trace metallic elements, titanium and magnesium. One by one, the spectral lines were removed to see their impact on the results. The ones with little or no impact were permanently removed from the group leaving behind only the most significant lines (for example we noticed that Na had a greater impact on our results than Aluminium and C₂ that did not change much). The cross-validation method is applied in order to study the impact of each line. At the end of this study, we are left 14 spectral lines to needed to differentiate and identify the polymer samples shown in Figure 5.

<i>Elements</i>	<i>Wavelength (nm)</i>
C I	247.869
Mg II	279.553
Mg I	285.213
Al I	308.215
Al I	309.271
Ti II	326.160
Ti II	334.941
Ti II	334.945
CN	388.308
CN	421.600
Ca I	422.689
C ₂	473.710
C ₂	516.520
Na I	589.048
H I	656.411
F I	685.752
F I	703.889
F I	720.236
Cl I	725.662
N I	746.952
Cl I	754.671
K I	766.527
Cl I	774.497
O I	777.285



<i>Elements Chosen</i>	<i>Wavelength (nm)</i>
C I	247.869
Mg II	279.553
Ti II	334.945
CN	388.308
Ca I	422.689
Na I	589.048
H I	656.411
F I	685.752
F I	703.889
Cl I	725.662
N I	746.952
Cl I	754.671
K I	766.527
O I	777.285

Table 4. ANN input intensities. The initial ANN input values included information of the intensities of 24 spectral lines. After optimisation, a total of 14 lines were kept for our needs.

The impact on the ANN results was not the only criterion when choosing the 24 initial spectral lines. When possible, the lines were chosen to be the most intense with minimal overlapping and self-absorption [35]. This was not always the case for lines like chlorine, which are hard to detect due to their low intensity. Chlorine and fluorine, due to their low intensity were chosen to be represented by 2 wavelengths each even though lines that accentuate differences and similarities amongst polymers are given greater weights, which means greater importance like PCA. Lines that are self-absorbed or self-reversed tend to lose part of their intensity making them unreliable, especially since we use their intensities to differentiate the polymers. If for the same polymer, the line intensity observed tends to fluctuate more than the expected shot-to-shot variations, it will only end up confusing the ANN. The data were normalised using the carbon line, located at 247.869 nm, given that it is the one common element amongst the polymers. The chosen wavelength values are placed in Table 5 along with their corresponding polymer line intensities that represent the normalised data. Since we normalise with regards to the carbon line at 247.87 nm, instead of having 14 wavelengths, we are left with only 13.

#	Name	λ (nm)	PVC	PA 6,6	PC	PE	PMMA	POM	PTFE	PP
1	C I	247.869	100	100	100	100	100	100	100	100
2	Mg II	279.56	3.4387351	0.1783783	0.1018691	0.2263056	0.1517826	0.3018691	0.4692874	8.9800443
3	Ti II	334.945	8.8537549	0.1675675	0.1130841	0.1715667	0.1850594	0.1602803	0.1326781	16.895787
4	CN	388.308	5.0197628	4.6486486	3.2056074	5.5319148	6.2648556	7.4299065	5.5405405	7.3170731
5	Ca I	422.689	9.8814229	0.2702702	0.1046729	0.5938104	0.1833616	1.9859813	0.3488943	0.4611973
6	Na I	589.048	1.4011857	0.6283783	0.0666355	0.2998065	0.4210526	0.5000000	0.3501228	0.7605321
7	H I	656.411	3.1422924	5.0270270	3.0747663	4.3713733	4.1595925	3.5560747	0.2309582	4.5232816
8	F I	685.752	0.0590909	0.0894594	0.0557943	0.0669245	0.1005093	0.0630841	1.0429975	0.0844789
9	F I	703.889	0.0693675	0.1083783	0.0642056	0.0814313	0.1157894	0.0635514	0.3427518	0.1119733
10	C II	725.5	0.2905138	0.0935135	0.0571028	0.0717601	0.0988115	0.0612149	0.0584766	0.0949002
11	N I	746.952	1.0988142	1.8108108	1.0186915	0.7234042	1.5908319	0.9813084	1.2899262	1.1707317
12	C II	754.671	0.2312253	0.0950000	0.0570093	0.0744680	0.1042444	0.0640186	0.0587223	0.0946784
13	K I	766.527	0.3418972	0.1295945	0.0448598	0.1212766	0.0845500	0.1813084	0.1047911	0.0787139
14	O I	777.285	2.4308300	4.6216216	3.4579439	1.5280464	4.8217317	3.5700934	2.8869778	2.7272727

Table 5. Matrix normalised line intensities with regards to the carbon 247.869 line. The intensities shown in the tables correspond to the averaged value of 100 spectra.

This table represents the form in which information is fed into the ANN. The spectra series consisted of 2 statistical data sets of 100 spectra, each spectra contained information from the accumulation of 10 laser shots. The reason we used a set of 100 spectra was simply to ensure that the possible local inhomogeneous concentrations are levelled out when a sufficiently large area is scanned. Our samples measured about 3 x 2 cm, and each sample provided one data set. The 1st spectra series were used to train the ANN where the 100 spectra were averaged and their peak intensities are recorded and represented by each column in Table 5. The back-propagation algorithm is used to optimise all the weights associated to each wavelength. This is the learning/training data.

Once the optimisation of the weights is finished, the 2nd spectra series was fed directly into the ANN without undergoing the learning process. If compared to Table 5, this spectrum series is very similar. It will have the same number of rows (it uses the same wavelengths) but will consist of a 100 columns per polymer. Each spectrum is treated as if though it were a new sample each time unlike the 100 columns that were averaged into one column in Table 5. These spectra were not averaged because we needed to gather statistical output results for each sample. In other words, it will test each spectrum, one by one, and then provide a result indicating the efficiency of the ANN. Once the total 100 spectra are individually tested, we are given report that includes how many are correctly identified (ANN provided the right prediction), wrongly identified (ANN predicted the wrong answer for the plastic) or non-identified (ANN came up with no answer). However, as discussed before, we can train the ANN but without the back-propagation limit, we might be faced with an over-trained or under-trained network. Our next priority is therefore the detection of the limit that separates the two regimes, otherwise defined as the back-propagation limit.

b. Back-Propagation Limit

Once the characteristic spectral lines have been chosen, the learning/training can begin. There is a problem associated with over or under-learning that cannot be detected during the training phase. It would make no sense for us to just feed the spectra into the ANN and expect it to learn correctly by stopping at the right time. The limit that stops the convergence of the back-propagation algorithm is needed. To find it, we start by randomly

selecting MSE training values to limit the iterations. Directly after the convergence is attained, we feed the ANN with the validation data to test the performance of this randomly chosen ϵ (training error) value. This is what was previously explained as the cross-validation method represented in Figure 17.

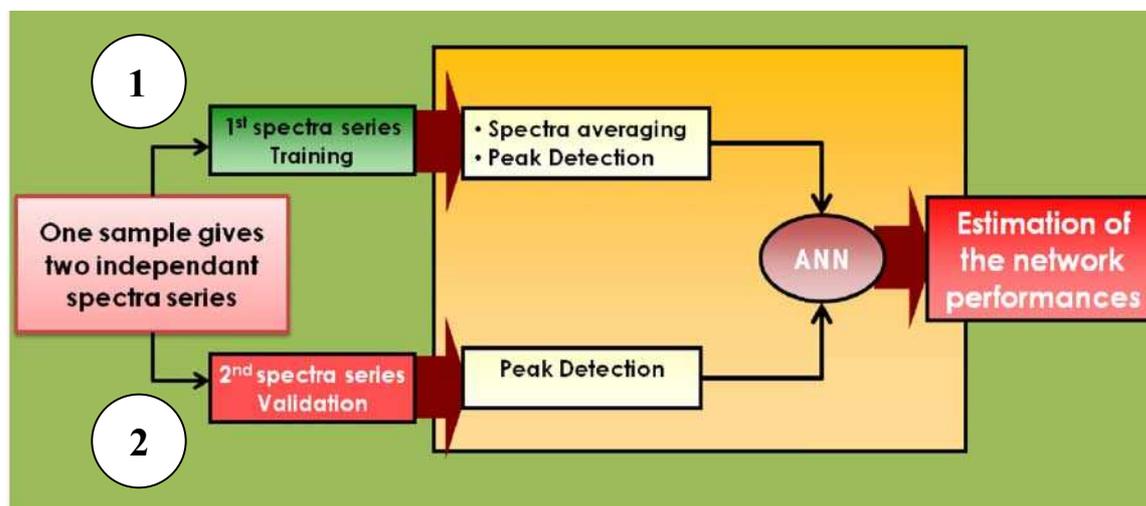


Figure 17. Cross-validation method.

For proper optimisation, the training and validation phases are done hand in hand. One by one, the internal parameters are studied. First the weights are optimised through training (1), and the validation phase (2) allows us to calculate the different errors associated to different training errors ϵ . Two independent spectrum series are required for each. This cross-validation technique was used to study the back-propagation limit value on a nylon sample. PA (nylon) was used because it is the polymer with the most complete matrix composition, containing C, H, O, and N. An MSE training value ϵ was fixed according to one of the values in Figure 18 and the weights were optimised for this value. Then we fed the validation data to the ANN and the results noted. 10 different tests were done per MSE value chosen to provide statistical data. Once this was over, we went back to change the ϵ value and redo the whole process providing the results in Table 6.

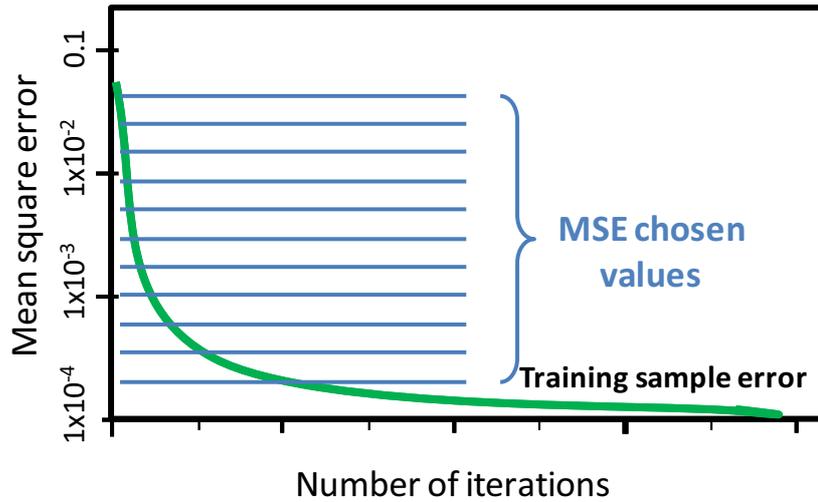


Figure 18. MSE training values.

Table 6 shows the results given by the cross-validation method and the complete study of the MSE results is plotted in Figure 19, where its values varied between 0.1 until 1×10^{-3} . It is important to understand that each set of answers (each column) has its own specific weights. Since the training is stopped at different moments, this means that each training process learned something different. The average per series was calculated along with its relative error produced. This table represents how many of the 100 spectra used were correctly identified. The best results were at $\varepsilon = 7 \times 10^{-3}$.

MSE (ε)	1 E-01	5 E-02	2 E-02	1 E-02	7 E-03	5 E-03	4 E-03	3 E-03	2 E-03	1 E-03
ID 1	100	87	92	98	93	98	99	96	99	98
ID 2	98	84	97	100	96	100	98	100	100	95
ID 3	53	92	87	100	98	95	88	100	97	100
ID 4	51	74	86	92	87	91	92	94	98	100
ID 5	84	86	87	87	100	99	94	100	98	42
ID 6	94	87	90	90	88	75	100	100	64	99
ID 7	32	97	98	98	100	100	87	77	98	62
ID 8	43	87	90	95	100	97	77	99	90	61
ID 9	91	40	84	87	100	92	95	90	77	37
ID 10	89	95	66	61	93	100	100	95	63	99
Average	73.5	82.9	87.7	90.8	95.5	94.7	93	95.1	88.4	79.3
Sigma	25.7	16.4	8.9	11.6	5.0	7.7	7.3	7.2	14.8	25.9

Table 6. ANN validation results. The test was done with a PA 6,6 sample and 10 were done per ε value.

A simple polynomial fit characterises the results. However, instead of using the results in Table 6 directly, we subtracted them from a 100 to calculate the number of wrong identifications, equivalent to the MSE validation error. The minimum depicts the ideal MSE validation value. On the left side of this value is the over-learning area, and on the right side is the under-learning area. As can be expected, the error bars are larger in these zones compared with the minimal zone, i.e. the ideal zone produces better results even statistically. This study is crucial for the minimisation of wrong identifications. For our studies, the MSE limit was fixed at 0.007 throughout the rest of our ANN analysis.

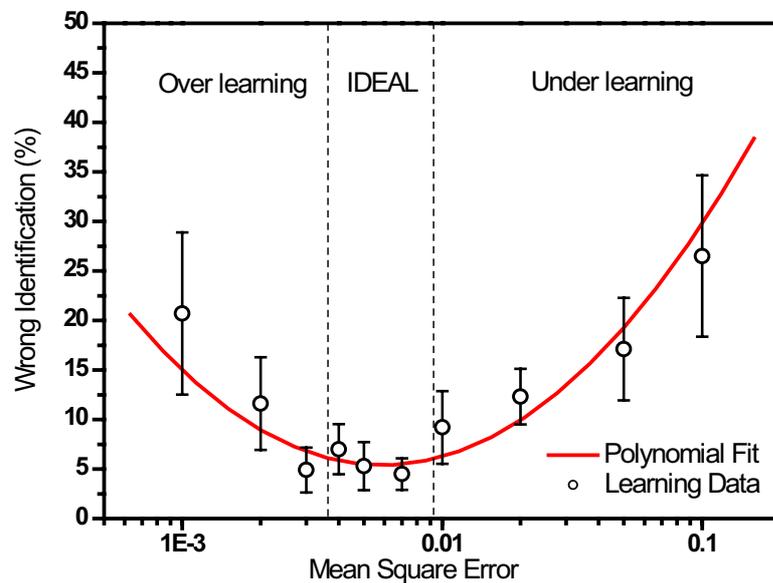


Figure 19. Validation error.

c. Hidden Neurons

The hidden layer is the next parameter that needs to be optimised. Optimally, the number of neurons in this layer should be between those in the input (13 selected lines) and the output (8 samples) layers. The main function of the hidden layer is to extract the salient traits in the input data, which results in a preliminary pattern recognition. The study of this layer was done in the same fashion as before. We trained the ANN with a certain number of neurons in the hidden layer and the validation followed directly to estimate the error

generated by these hidden neurons. The final number of hidden neurons was determined by the minimal error. Finally 12 neurons provided the most accurate results.

The ANN is optimised after the accomplishment of the above steps. This means that the layers have their specific neurons with their specific weights associated to each input value. The training phase was over at this point and the last ANN parameter to be studied was the threshold value. It still needed to be defined and optimised for identification purposes. In the next part, the weights will not be touched (training is over) and the number of neurons is each layer along with the characteristic wavelengths have been fixed. The spectra will be fed into the ANN and the threshold value will be the only parameter to be adjusted.

d. Threshold Value

The cross-validation method is not used here. The previous optimisations have now been fixed, and this parameter can be changed during the validation phase. The threshold value is essential for real-time unsupervised correct identification of polymers. A neuron, in our network, only gives answers that range between 0 and 1 due to the sigmoid function properties previously shown in Figure 11. The output neurons will provide us 8 values in this range. 0 represents no identification possible and 1 meaning a perfect match found. If all output values were identically equal to 0, the ANN is not able to identify the sample in any regard. If they are all equal to 1, the ANN believes that each plastic sample is a perfect match! In practice, we are usually between these two extremes. The maximal output value indicates the sample with the greatest chance of resembling the input data.

To allow the ANN to find this maximum value we introduce what is called a confidence level CL_X . It is defined by Koujelev et al. [34] as

$$CL_X = \frac{P_X}{\sum_j P_j} \quad (4.6)$$

where P_X is the prediction of one output neuron, and $\sum_j P_j$ is the sum of all the 8 output values. The aim is to ease the identification by calculating the relative output values of the network.

In order to have real-time identification with no direct supervision, we define a threshold value (T_{value}). The threshold value allows an automatic response of the network. For identification, CL_X must be above this value. Figure 20 shows the CL_X values and the domain that the threshold value can fluctuate, T_{value} , all the while providing the correct answers. As can be seen, if T_{value} is too high, above all the CL_X values ($T_{value} = 0.7$), no kind of identification is possible. A value too low, where more than one CL_X values rises above it ($T_{value} = 0.1$), this would result in multi-identification. The threshold value must be optimised to avoid these two scenarios.

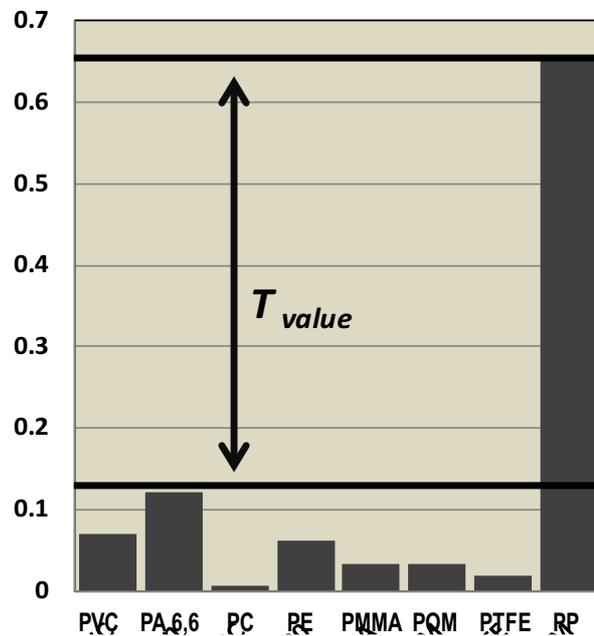


Figure 20 ANN prediction values for polypropylene (PP).

Figure 20 shows results for a PP sample, where the threshold value can range from about ~ 0.13 until ~ 0.65 and still provide correct results. Each polymer has a probability of being identified. The PP value is clearly above the others but we need to find a method to confirm this. This is where the threshold value T_{value} comes in. This study needs to be performed for all samples to find the best T_{value} for the global samples as it is not the same for all polymers. A common T_{value} must be found that can satisfy the requirement of the all plastic samples. For this, we study the lower limit and upper limit of the threshold for

all the samples. It evolves differently for different polymers and a few examples are shown in the graphs in Figure 21. Each graph is composed of curves that correspond to:

- correct identification
- incorrect identification
- unidentified

With a threshold value decreasing and becoming smaller than the lower limit the zone T_{value} , the probability of wrong classification increases. An increasing threshold value above the upper limit of the zone T_{value} reduced the probability of correct identification. The ideal choice would be a value that is common to all polymers but with the smallest possible value to maximise the correct identification. In practice, with certain samples, such as PVC, it is more important to not identify it rather than wrongly classify it, which may provoke a dangerous mixture with other polymers. The choice of the threshold therefore also depends on the applications and the consequences of wrong identifications. Certain samples like POM have a quite small optimal range compared to that of PA. In addition, the minimal limit of the T_{value} zone for POM is higher compared to others. The minimal limits in Figure 21 are around 0.4 (PA and PE), 0.5 (PMMA) and 0.55 (POM). Taking this into consideration, the ideal value is the smallest value common to all.

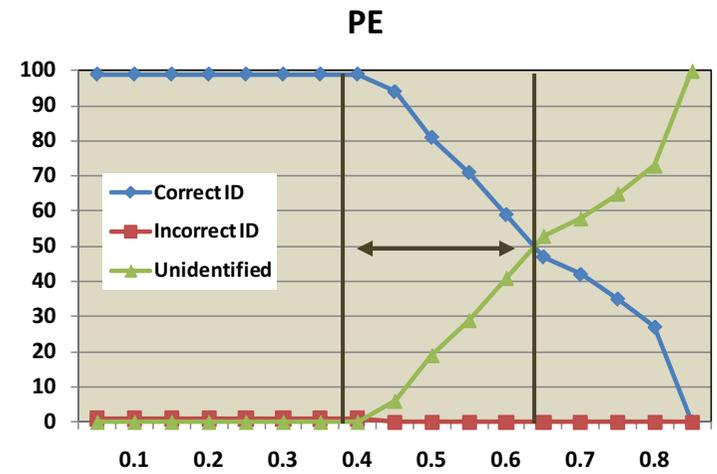
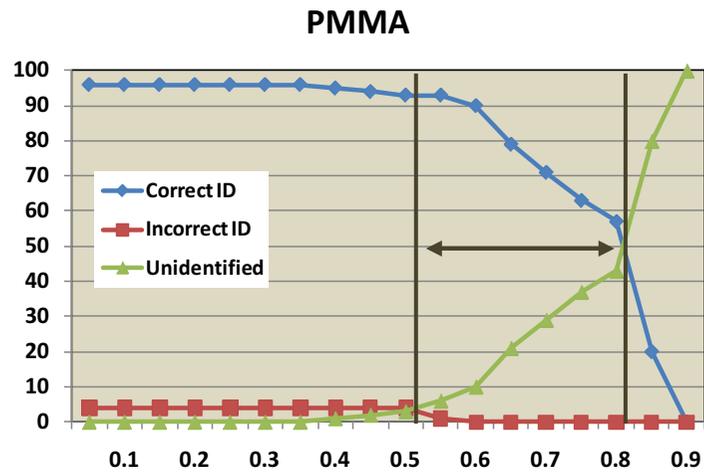
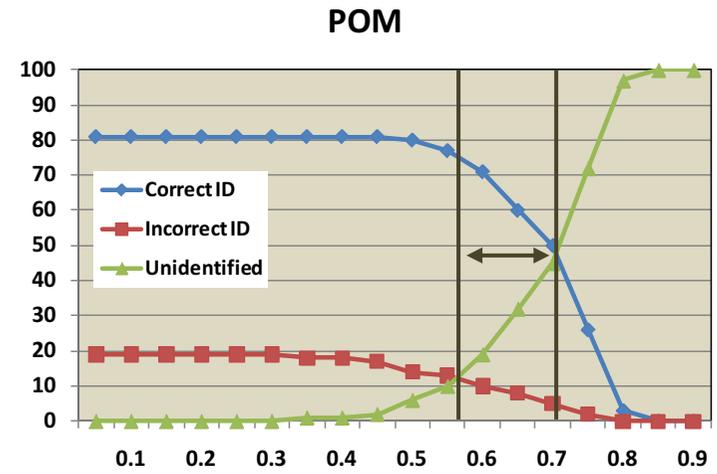
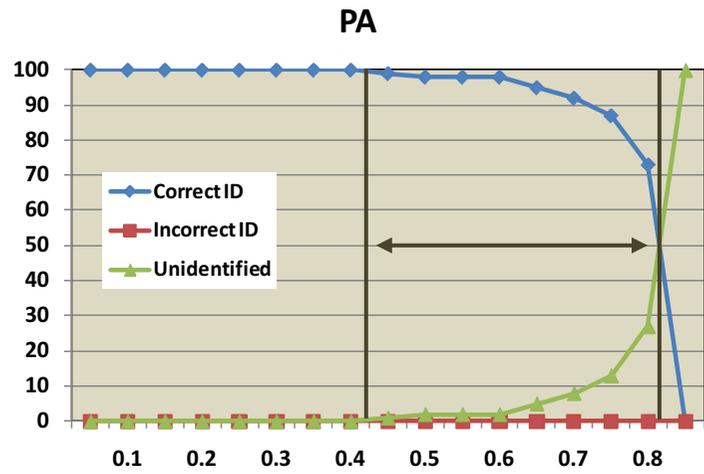


Figure 21. ANN predicted results for PA, POM, PMMA, PE.

For single spectrum identification, we decided to evaluate two different threshold values. The first one is the minimal limit of POM and the second one smaller than this limit. It is expected to have an increase in wrong results for POM but we want to see whether it would be globally more interesting to work with this compromise. We averaged the errors on all samples and plotted a graph accordingly shown in Figure 22. The correct results decrease very rapidly for a threshold larger than 0.4. As a compromise of all these values we decided to choose $T_{value} = 0.45$. As can be expected, the lower the threshold, the higher the correct identification becomes. However, this graph does not show us that the lower the threshold, the higher the wrong identification becomes. It does indicate a sharp decrease that starts after $T_{value} = 0.45$. We use this graph to find the best upper limit for our optimisation.

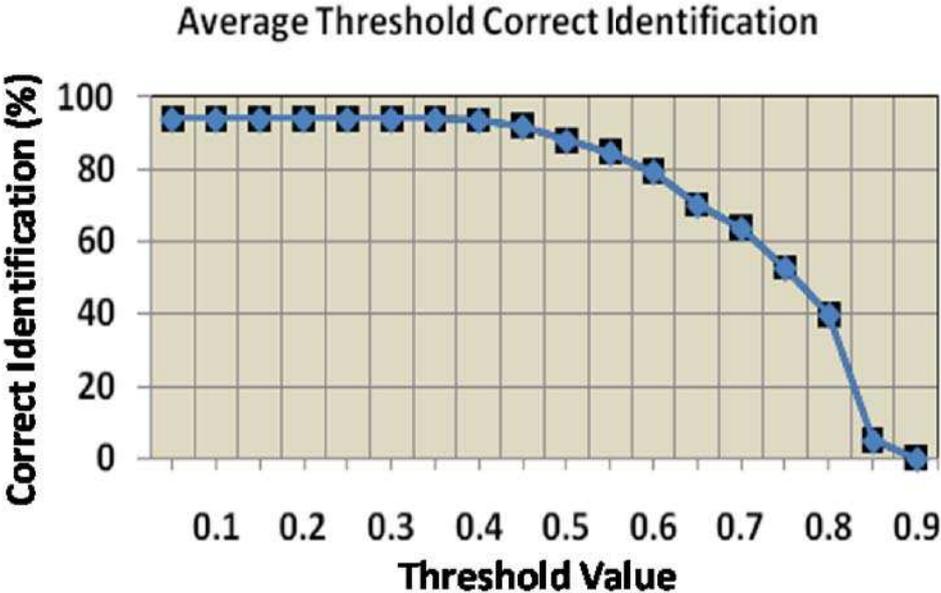


Figure 22. Threshold global identification.

3.3.3 Summary

In summary the parameters of the ANN after the optimisation are presented as the following:

- The output layer
 - Each neuron corresponds to one plastic sample. With 8 samples we have 8 neurons
 - No optimisation was required here
- The input layer was optimised with 13 neurons
 - Each neurons is associated to a spectral line intensity
 - Optimisation was done manually through cross-validation
- The hidden layer was optimised with 12 neurons
 - Optimisation was done through cross-validation where the minimal MSE value provided us with the most efficient number of neurons
- Back-propagation limit value was fixed at $MSE = 0.007$
 - Optimisation was done through cross-validation where the lowest MSE value avoided under and over-training
- The threshold value $T_{value} = 0.45$
 - Optimisation was done through cross-validation where the lowest MSE validation value avoided under-learning and over-learning.

After these optimisations, we expect the ANN to provide efficient results. All the parameters have been defined and optimised. Our next aim in this chapter is to provide high quality single spectrum results before the last phase, which consists in a testing phase with 10 random chosen PVC samples that the ANN procedure has never treated before.

4. VALIDATION PHASE AND DISCUSSIONS

4.1 SINGLE SPECTRUM IDENTIFICATION

Now that all the different parameters have been optimised and established, we can continue with the identification of the polymer samples. The different spectra are analysed one by one by the ANN. In the learning phase, the teacher provides to the ANN the learning spectral data along with the information of their corresponding sample. In the validation phase, the ANN is fed with statistical data that it has never treated before with the aim of identifying the corresponding sample without any help from the teacher. The ANN does not know the sample providing the validating data. But if the learning has been properly done, the ANN should be able to identify the sample from the spectrum. Even though the learning and validating data are both statistical data taken under the same conditions, and there are a few differences between them.

A statistical series means that all the spectra were recorded with the same experimental parameters one after the other. The spectra are not perfectly identical due to the experimental fluctuation that can influence the spectral results, such as laser energy variation, local inhomogeneity of the sample, and fluctuations related to the slight changes in the position of the sample with respect to the focus of the laser beam. If the measurements are spaced out by a few days, the environmental factors such as the humidity and the cleanness of the air, can also influence the spectra.

The aim is to prepare the ANN to be reactive to other fluctuations that can come from the exterior environment, for example, the human errors when placing a new sample, along with the usual experimental fluctuations. The results of the validation phase are presented in Figure 23. It is normal to expect such good results because the spectra, even though independently taken, are recorded from the same samples. ANN is known to be error-tolerant and we could have identified a threshold value that would have been ideal for each polymer for even better results. But since a global value was needed, the optimisation is not perfect for all samples.

This is the case especially for POM and PMMA. Their threshold value chosen was below their limit, which resulted in wrongly identified spectra. With this value, correct identification ranged from 94% to 100% for all samples except for POM with 81% (Figure 23). 100% correct

identifications were found for PP, PC and PVC. Only POM and PMMA were wrongly identified at 17% and 4%, while PE, PTFE and PA 6,6 were only unidentified at 6%, 1% and 1% respectively.

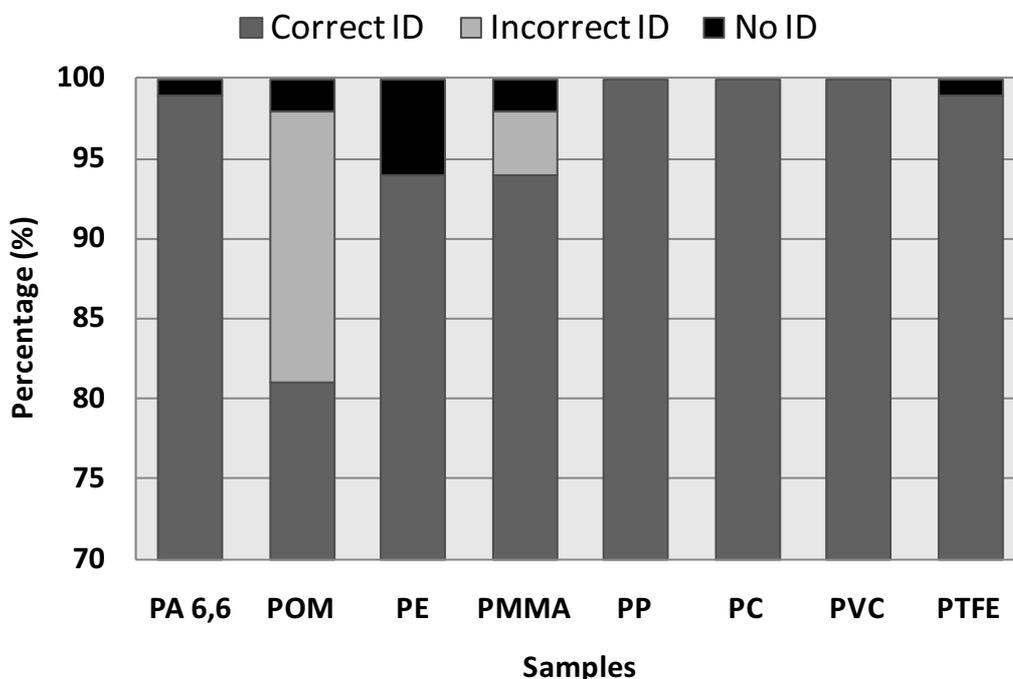


Figure 23. Validation phase results of the optimised ANN. Note that the graph starts at 70% correct for single spectra identification

These results were not real-time results because the algorithm needed to be optimised first but they remain single spectra analysis, which means the spectra were not accumulated or averaged together. The spectra of some of the problematic samples, such as POM, PE and PMMA, were averaged together to improve the results. Figure 24 shows a graph which demonstrates that only an average of 2 PMMA spectra was needed for 100% successful identification and only 5 for POM and PE.

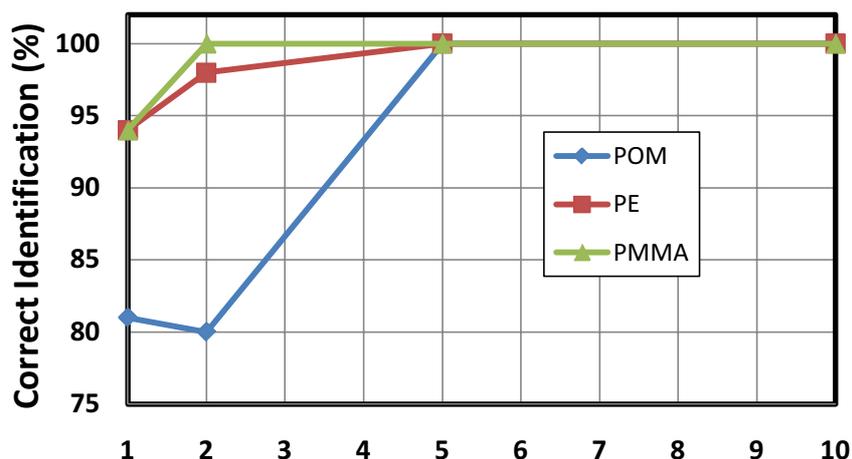


Figure 24. Average number of spectra needed for 100% correct identification. PMMA required 2 while PE and PMMA required 5 spectra.

4.2 THE FINAL PHASE: TESTING

In this phase, the aim is to test the efficiency of our parameter optimisation by feeding the ANN spectra from samples it has never seen before. It already has PVC incorporated in its memory base but the samples we are asking it to identify are completely new to it.

4.2.1 PVC identification

PVC is a dangerous plastic in the sense that its mixture amongst other plastic samples can release toxic gases during the recycling process. It is why we will focus our attention in this section on the identification of PVC samples of different origins. The aim was to test the ANN identification performance for a large number of samples. In order to test the efficiency of our ANN optimisation, a selection of PVC samples were collected and fed to it. The ANN has the memory of what a PVC sample should look like but it has never seen the following samples which may contain different elemental traces. New batches of statistical data sets were taken on randomly collected PVC samples, which varied in colour and uses. 10 different samples were collected from diverse areas such as plumbing, mechanical equipment, flooring and bathroom accessories. The collected samples are presented in Figure 25.



Figure 25. PVC testing samples.

Statistical series were recorded in the same way as before. Since we only need to test the ANN, we do not need the use of independent data sets but only a series of 60 spectra for each sample. 60 spectra covered a large area of the surface of a sample to dampen out any local inhomogeneous concentrations. These data sets were new to the ANN and were not used during its training phase. To be more robust, the ANN must prove that it is able to recognise other PVC samples that were not part of training and that it has never treated before. The identification results along with the different PVC properties are presented in Table 7. As can be seen, the ANN correctly identified all the samples at a rate of 100% except for PVC flooring sample # 4. Further examinations concluded that this PVC material was on the bottom of the PVC flooring which contained a mixture of different polymers. The results show again the capacity of the ANN to recognise the texted materials.

<i>N°</i>	<i>Type</i>	<i>Colour</i>	<i>Correct</i>	<i>Misclassified</i>	<i>Unclassified</i>
PVC 1	Mechanical purposes	Grey ++	100	0	0
PVC 2	Mechanical purposes	Grey +++	100	0	0
PVC 3	Mechanical purposes	Grey ++++	100	0	0
PVC 4	PVC flooring	Colourful	0	51.7	48.3
PVC 5	Window pane	White	100	0	0
PVC 6	Bathroom accessory	Blue	100	0	0
PVC 7	Mechanical purposes	White	100	0	0
PVC 8	Mechanical purpose	Grey +++	100	0	0
PVC 9	Water pipe	Grey +	100	0	0
PVC 10	Water pipe	Grey ++	100	0	0

Table 7. PVC real-time single spectra test analysis. Light +, medium ++, dark +++, very dark ++++.

The PVC spectra also gave forth different substances that are not part of the basic molecules. According to the utility of the materials, PVC manufacturers add what are known as fillers or additives [36]. For example, titanium oxide (TiO_2) has whitening properties but also increases the opacity and improves the durability of the material. In addition, it absorbs UV light which avoids colour change, a particularly useful attribute for PVC when used outdoors (windowpanes, pipes etc). Without this property, they would become all yellow with time. Calcium and magnesium detected in the spectra can originate from their carbonate form (CaCO_3 , MgCO_3 respectively), which are the most common known fillers. These additives and fillers can drastically change in concentration from one sample to another added according to the future uses of the plastics in question.

Fillers are very important in the plastic industry for they reduce the production costs. Calcium carbonate (CaCO_3) is the most commonly used filler. CaCO_3 is the most important of all the fillers. It also has whitening properties and is cheaper to use than TiO_2 . It has other properties but mainly it is used to increase the stiffness of a material and to avoid shrinkage and distortion. Quantitative analysis of these elements would allow recycling industries not only to sort according to polymer type but also according to their charges (additives + fillers) to improve the quality of recycled end products. The PVC samples collected come from different applications. Figure 26 shows a wide variety of ‘concentrations’ of titanium (from TiO_2), calcium arising from its carbonate form CaCO_3

Qualitative PVC charge detection

Qualitative charge detection was performed on the different PVC samples, where spectra were normalised with respect to oxygen to reduce matrix effects. The results are shown in Figure 26. Despite the normalisation, the variation on the normalised emission intensity from chlorine shows the influence of the matrix effect. The concentration of chlorine has normally to be constant with respect to that of oxygen for different PVC materials. For the analytical purpose, focus was made on metallic elements such as calcium or titanium. Our results show that the samples can be separated into 3 groups with high, medium and low calcium concentration. As previously mentioned, calcium carbonate is often added to PVC materials. Most of the PVC materials with medium calcium concentration belong to the group of materials for ordinary mechanical use (PVC 1, PVC 2, PVC 3, and PVC 8). PVC 7 along with

the entire ready-to-use PVC types (PVC 5, PVC 6, PVC 9, and PVC 10) are materials with high calcium content. These materials need to be resistant to changes external constrains such as weather, temperature and pressure and need to be reinforced. PVC flooring showed to have both very low calcium and titanium concentrations.

The PVC sample is supposed to contain the same matrix for all and therefore the chlorine content must be the same. This not being the case, we can assume that the matrix effects have an effect on the spectra line intensities. PVC 4 was not identified as PVC however it shows normal chlorine content compared to the others. The calcium and titanium content can be used to produce different categories according to their concentrations.

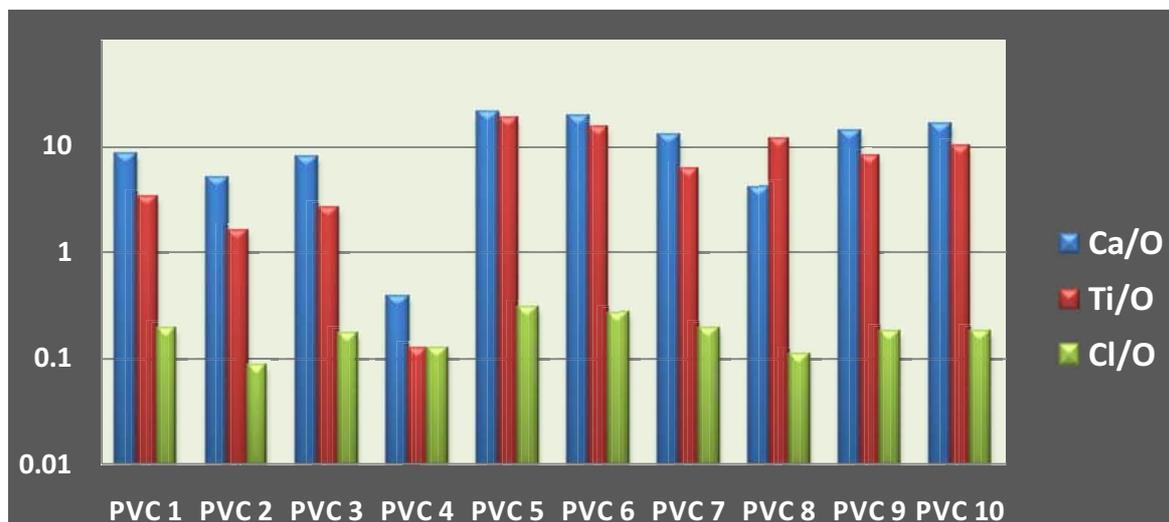


Figure 26. PVC metallic charge content per sample.

5. CONCLUSION

In this chapter, eight different polymers were used to effectively train and carefully optimise the different parameters within the 3-layer artificial neural network (ANN). Further successful identification tests were performed on 10 unknown PVC materials. Semi-quantitative measurements regarding the charges present in these samples, allowed the classification of sub-categories according to the concentrations of the charges. However, the matrix effect does not permit precise quantitative determination of these charges for more acute sorting.

With our results, we have shown that broad spectral range LIBS combined to ANN processing can meet reliable and high speed identification and classification required for the industry of plastic waste recycling. With the proper supervised training and optimisation of its parameters, the ANN processing has demonstrated a good performance when used to identify different types of polymer samples. A particular attention was paid to PVC samples due to their potential hazardous impact during recycling. In our study, the test was conducted using series of 100 spectra in order to optimise and evaluate the performance of the method. However, this processing is easily adaptable to real time and online analysis. Once the network is trained and validated, identification answer can be provided in only few milliseconds with a common computer and with no need for human supervision. However before its final validation for real industrial sorting application, it is necessary to run the ANN for a much larger number of materials collected in industrial environments as well in daily life. Taking into account of the influence of the quantity of additives on the identification performance of a same type of polymeric materials represents still a challenge.

Perspectives

Meanwhile, we think from these results that LIBS combined to ANN could be used to complement the NIR spectrometry. Today the NIR is largely used in the industry of plastic waste sorting but is known to be not suited for sample with contaminated surface or dark-coloured polymers [14]. Besides, LIBS could also provide additional information such as the concentrations of the additives. The combination of the both techniques will then provide the industry with more accurate and precise information allowing them to indentify dangerous polymers while providing a more detailed means of sorting with regards to grades and additives. In order to retrieve the quantitative composition of the additives, two methods can be used. First, once the sample has been identified, and so the class of polymer known,

calibration curves can be plotted with the help of the reference samples. The approach of calibration free LIBS (CF-LIBS) [16] can also be implemented for the deduction of the quantitative analytical data concerning the additives independently to the analysed plastics matrix. The efforts should be provided for fast time response of the algorithm and the analytical performance of the procedure in order to meet the requirements of real time and online sorting.

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GENERAL CONCLUSION AND PERSPECTIVES

The main aim of this thesis was to demonstrate and evaluate the performance of the LIBS technique for the identification and classification of polymeric materials. The potential applications will be the automated sorting of polymer wastes. The basis of laser ablation and plasma formation were recalled in the first chapter along with the key elements for plasma diagnostics (determinations of the electronic density and temperature). The criterion for the local thermal equilibrium (LTE) and an introduction to the procedure of the calibration free LIBS (CF-LIBS) for quantitative analysis without the use of external standards was also discussed.

The shadowgraph images presented in Chapter 2 allowed us to observe the interactions involved in the different laser ablation regimes for a polymer sample. The supersonic expansion of the plasma was observed and partial laser shielding was encountered in the nanosecond ablation regime. These measurements allowed us to confirm the choice of nanosecond ultraviolet ablation and its advantages for polymer analysis using LIBS. Despite the higher ablation efficiency in the femtosecond regime, it remains for now, too complicated to put in place in the industry due to its operational complexity. Even though this regime has its advantages, there were not enough to eliminate the nanosecond ultraviolet regime.

In the third Chapter, we focused on the importance of finding the appropriate detection window with the results of the time-resolved plasma emission spectroscopy. The plasma does not display the same emission signal at its birth as it does at its dissipation. Therefore, we performed time resolved measurements to study the evolution of the electronic density and the temperature at various time delays on a polypropylene sample that contained a high amount of metallic traces. The metallic traces present allowed us to calculate the temperature of the plasma by different methods (ionic, atomic as well as molecular due to the organic nature of our sample). Stark broadening of the H_α and Saha-Boltzmann plots provided the most accurate results to calculate, respectively, the electronic density and the temperature of the plasma induced on a polypropylene sample. Due to a satisfactory signal-to-noise ratio the best gate range was $t_d = 600\text{ns}$ with a gate width $t_b = 400\text{ ns}$ was selected. For these parameters, we also took into account the local thermodynamic criteria. This aspect is particularly important if we want to implement calibration free methods e.g. CF-LIBS to provide more precise quantitative

analysis in our future experiments. However, for the time being, it was highly satisfactory for our qualitative analysis required for our industrial application: complementary technique for automated plastic sorting.

With this, we were able to move ahead to evaluate the potential associated with the LIBS technique in the identification and classification of plastic materials. The chemometric data treatment techniques such as principal component analysis (PCA) and artificial neural networks (ANN) were used to extract analytical results from the LIBS spectra. Among these data treatment methods, to answer industrial requirements, we consider the ANN to be more suited for providing real-time single spectra analysis due to the non-linear nature of the calibration curves for organic samples. For this, we studied 8 polymers samples (PVC, PP, POM, PTFE, PE, PA 6,6, PMMA, PC) and then focused on PVC by studying and extra 10 samples collected randomly from various applications. The study resulted in successful predictions ranging from 94 – 100% success for most samples (the exception being POM at 81% successful identification).

LIBS has the potential to be a powerful and important tool but we need to find a way to get it out of the research laboratories and convince the industry of its utility. The applications are extremely vast ranging from homeland security to planetary exploration and historical monument analysis. However, further developments of the technique remain necessary for such applications becoming routine. It is important to note here that the results in this thesis remain preliminary (Chapter 4). In the future, a much wider range of samples (collected from real industrial or domestic wastes) still have to be tested using this procedure to really put forth its capabilities of real-time identification and classification of polymers. The influence of the environment of the data collection needs also to be evaluated, which include the laser energy fluctuations, the sample matrix changes, and the quite random sample surface positioning with respect to the incoming laser beam. The effort has also to be devoted to reduce the time of spectrum acquisition by using spectrometers with more efficient transmission and detectors with higher sensitivity, such as photomultiplier tubes (PMT) to meet the requirements of online detections. The combination with the infrared spectroscopy is certainly pertinent to provide diagnostics based on elemental analysis. LIBS provides complementary elementary information with that of the molecular information from NIR spectroscopy. In addition, when

it comes to specific cases of dark samples or materials with important surface contaminations, the LIBS technique presents to be more efficient.

Another problem lies with the additives in plastics materials that still remain challenging to deal with. The quantitative determination of the concentration of added elements will certainly be required. The CF-LIBS procedure can help by providing an estimation of the orders of magnitude of the concentrations. The improvement of this procedure will be done by taking into account the transient nature of the expanding plasma. It should allow more precise results with regards to the concentrations of the additive elements without the use of external standards. The classical approach with calibration curves will also be more efficient once matrix-matched standard samples are available. Finally, before going to real industrial applications, engineering developments have to be conducted by rendering the LIBS detection system reliable by making them more compact and more robust. With these improvements, LIBS will be better suited for the industrial environment where non-specialists will be operating the system. Therefore, to conclude, once all these studies are conducted and if the results remain successful, we can provide the necessary strong arguments to implement the LIBS technique in the actual plastic waste sorting to prove its capacity in rendering it much more efficient.