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Monitoring udder health and milk quality using somatic cell counts

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Abstract – In this article the use of somatic cell counts for monitoring udder health and milk quality is discussed. Somatic cell count dynamics at quarter, cow, herd and population level are discussed and illustrated with examples. Quarter and cow somatic cell counts directly represent the inflammatory status of the mammary gland. Herd and population somatic cell count are related to the inflammatory process in individual cows but much more reflect the udder health status of the herd and the quality of the raw milk in the herd and the population. Application of monitoring tools in herd health management are illustrated using a case study. Understanding infection dynamics requires precise longitudinal data. Monitoring tools are required to find the areas of risk in the herd. It is inevitable that more complete udder health programs and monitoring systems are to be developed and implemented. These programs are necessarily dynamic and complex. Implementation of complete udder health programs should be accompanied by research efforts to further fine-tune these complete udder health control and monitoring programs.

somatic cell count / mastitis / milk quality / monitoring / epidemiology

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1. INTRODUCTION

Throughout the world, the importance of udder health programs has increased in the last ten years [16, 34, 39, 40, 45]. There are a number of reasons for this awareness of udder health as a critical production issue on dairy farms. In Europe, the EEC directive 92/46 in April 1992 stated that milk with a somatic cell count (SCC) over 400 000 cells per mL may not be used for fluid milk and starting in 1998 not even for human consumption. In North America limits at 750 000 (USA) and 500 000 cells (Canada) are in place [45]. Another issue is the increased awareness of consumer and dairy organizations with regard to animal welfare issues. Clinical mastitis may be a severe and painful disease that causes distress to the animal. It is therefore important to decrease the clinical incidence of disease. A third, more recent issue are human health concerns regarding milk consumption. This includes antibiotic residues in milk, transfer of antibiotic resistance from animal to human, and transfer of pathogens or products thereof through milk or milk products [30, 39, 40, 60]. Approximately 80% of antibiotic residues in milk can be traced back to mastitis treatments, either during lactation or during the dry period [27, 43]. Hence, monitoring programs will need to address these three components. Their ultimate goal is to aid the producer in herd management and to guarantee the quality of the raw product to the consumer [6].

The objective of this review is to summarize the widely available tools for evaluating udder health and milk quality on dairy farms. First, the inflammatory process in the mammary gland will be observed at quarter, cow, herd and population level (see also [37]). Thereafter monitoring available to the dairy producer in a veterinary herd health program will be summarized. Monitoring the inflammatory process in the mammary gland can be done using several diagnostic tools including but not limited to somatic cell counts, conductivity, California mastitis tests (CMT) [46], N-acetyl-β-D-glucosaminidase (NAGase) and many others [42] for a review of diagnostic tests. In this review somatic cell count will be emphasized because of their use in regulatory systems and wide availability and use throughout the dairy industry in the world.

2. SOMATIC CELL COUNT PATTERNS DURING THE INFLAMMATORY PROCESS

2.1. Somatic cell count patterns at quarter level

Somatic cells are mostly cells of the immune system (80% in uninfected quarters, 99% in mastitic quarters) [54]. These somatic cells are part of the natural defense mechanism and include lymphocytes, macrophages, polymorphonuclear cells and some epithelial cells [38]. Somatic cells are therefore a reflection of the inflammatory response to an intramammary infection or another trigger of the immune system. Somatic cell count, or a parameter derived from this count, is often used to distinguish between infected and uninfected quarters. There is a general agreement between infection status and the inflammatory response to this infection as measured by an increased SCC. As with any diagnostic test, errors will occur when solely depending on a single test. To minimize the amount of error, diagnostic test parameters such as sensitivity and specificity are calculated at various cut-off values in the SCC continuum [47]. Research from North America and Europe has shown that uninfected quarters have a mean SCC of approximately 70 000 cells [10, 11, 27, 47]. There is of course variation around this mean, and it was also shown that the mean SCC of uninfected quarters increases with age, decreasing milk production and days in milk [47]. Hence, to be able to distinguish between infected and uninfected quarters it was repeatedly shown that a cut-off of approximately 200 000 to 250 000 cells was
optimal to reduce diagnostic error [11, 24, 27, 47]. At this cut-off value, diagnostic sensitivity was shown to be approximately 75%, while specificity was approximately 90% [47]. Throughout this paper a cut-off of 200 000 cells/mL will be used. The 200 000 cut-off is not considered a physiological cell concentration in milk distinguishing “healthy” from unhealthy quarters or udders, but that it is an operational threshold of practical value under field conditions (minimizing diagnostic error), not the ultimate goal for udder health and production of the best quality milk. Other thresholds (such as 100 000 or 500 000) are advocated by others. Any threshold of cell counts to indicate intramammary infection will have its advantages and disadvantages, we have selected 200 000 to minimize classification error. A parameter based on somatic cell count that is often used is the Linear Score (LS). The LS is a base 2 logarithmic conversion of SCC [50]. Cow LS is calculated as LS = \log_2 (SCC/100) + 3, where SCC is cells/μL. The conversion of LS to SCC is calculated as SCC = 100 \times 2^{(LS – 3)}.

When a cow gets infected, the resident somatic cells signal to a resting population of white blood cells in the blood stream, and a massive influx of mostly polymorphonuclear cells into the milk takes places [8, 51]. These cells kill bacteria, and when the infection is eliminated then usually within a few weeks cell count of milk returns to normal. An example of such a response is presented in Figure 1A, where data are presented of an experimental E. coli infection. The intramammary infection with E. coli was eliminated in approximately two days [58].

When the immune system is not able to remove the bacteria, a chronic infection within the mammary gland results in a continuous trigger and somatic cell counts are high long term [18]. This process is depicted in Figure 1B, where somatic cell counts in a chronic E. coli infection is shown [12]. Usually, there is fluctuation in cell counts, but cell counts are often above the previously defined cut-off for uninfected quarters. Based on these infection dynamics phenomena, somatic cell counts are particularly useful to follow individual quarters or cows over time. Only quarters and cows with long term high cell counts are indicators of chronic infection in these animals and require further management attention. Short term high counts are not

![Figure 1](image-url)

**Figure 1.** A: Somatic cell count pattern during a successful immune response to an incoming E. coli bacterial infection. A non-infected contra lateral quarter is shown to represent non-infected quarters (data from [58]). B: Somatic cell count pattern of a quarter chronically infected with E. coli (data from [12]).
necessarily a reason for concern since an apparent immediate cure occurred. Clearly, SCC data are used as a proxy for infection. In some situations quarters with low SCC may still harbor infection (see for example [7, 12, 18]). Only with repeated bacteriological culture an accurate diagnosis of cure of IMI can be made.

2.2. Somatic cell count patterns at cow level

Somatic cell counts are usually measured in composite cow milk. It is particularly important to characterize the relationship between presence of an intramammary infection (IMI) with the cell count response of the cow in the composite milk. Knowledge of this relationship will enable inferences about IMI prevalence, from somatic cell count data, which is routinely recorded in 91% of herds that are participating in milk recording in the United States of America [35]. The most accurate relationship between IMI and SCC exists at quarter level. Cow composite samples of SCC and IMI are a composite of four quarters with dependent but separate infection status and inflammatory response. Most dairy producers and dairy veterinarians have only access to cow composite information and therefore the relationship between these two parameters is of great practical importance. There are essentially two approaches to relating cow-SCC and IMI. First, known infected or non-infected cows can be followed, and the mean cell count for these two groups of cows can be calculated, or diagnostic parameters such as sensitivity and specificity can be obtained [24, 28]. Second, cows can be classified based on their cell count level and the probability of IMI can be calculated. An example of this second approach will be presented in some more detail. Here, the AVELS was the Average of Monthly Linear Scores in the current lactation up until the time of bacteriological culture (an average of 150 days in milk). The AVELS was available for 65,229 cows whose composite milk was cultured for bacteriological diagnosis of IMI at 914 herd visits from March, 1992 to April, 2000 in New York State. Sampling methodology and bacteriology were described in detail by [61]. Association between AVELS category (0.1–0.9; 1.0–1.9; 2.0–2.9, etc. to 9.0–9.9) and whether cows within each score had IMI defined by positive milk culture (also a categorical variable) were evaluated using Chi-square. Table I shows the prevalence of IMI within each one-log interval of AVELS for major pathogens (Streptococcus agalactiae, Staphylococcus aureus, environmental streptococci, and Mycoplasma spp.), other major pathogens (Escherichia coli, Klebsiella, Serratia, Arcanobacter pyogenes) and minor mastitis pathogens (Corynebacterium bovis, coagulase negative staphylococci (CNS)). For all types of IMI, cows with AVELS < 3.0 had significantly less IMI than the overall population mean. Those with AVELS > 4.0 and for each higher log category of AVELS up through ≥ 9.0 had higher prevalence of major pathogens (chi-square, P < 0.001, Tab. I). Minor pathogens were significantly less present in AVELS < 2.0 and in AVELS > 6.0, and showed a higher prevalence between AVELS from 2.0 to 5.9.

Results demonstrated that AVELS was a good indicator of IMI prevalence in this dairy cow population. The strongest increases in IMI prevalence occurred with increases in AVELS between 2.0 and 5.9. Within this range, prevalence of all pathogens (major and minor) increased by on average 12% for each one-point increase in AVELS. From the standpoint of improved management, total IMI prevalence would be expected to decrease by 12% for each one-point drop in AVELS within the above range.

Mean AVELS for cows without IMI was just below 3.0, corresponding to a geometric average SCC of 96,000 cells/mL. This is higher than the average SCC reported in another study [24] of cows that were repeatedly and consistently cultured negative for IMI each month during lactation, which was approximately 50,000 cells/mL. For
the sum of all the difference in prevalence between the lowest AVELS category (0.1 to 0.9) and the second lowest category (1.0 to 1.9) was small. However, for cows with AVELS between 2.0 and 2.9, prevalence for all pathogens was 34%, substantially above the prevalence of 20% for AVELS between 1.0 and 1.9. Most of the IMI at these low AVELS ranges was due to CNS and \textit{C. bovis}; their prevalence decreased steadily among cows with AVELS > 4.0. This agrees with previous reports that CNS and \textit{C. bovis} have a relative small effect on increasing SCC \cite{10, 22, 25, 31, 61}. These facts suggest that a goal for AVELS should be around 2.0 (50,000 cells/mL) to attain the lowest practical level of IMI. Little benefit would be gained by striving for AVELS below 2.0. More than 20% of the cows in this study were below AVELS 2.0. Striving for cell counts below this level has been reported to lead an increased risk of clinical mastitis \cite{3, 17, 55}, although this is still somewhat controversial \cite{44}.

### 2.3. Somatic cell count patterns at herd level

Monitoring somatic cell counts at herd level requires longitudinal data over time. Given the variability in inflammatory responses between all cows that make up a herd. An average bulk milk SCC of 50,000 cells/mL is therefore not a realistic goal \cite{23, 33}. Prevalence of infection increases with mean bulk milk SCC but this is not tight relationship due to the lognormal nature of cow SCC. At herd level it is especially important to follow trends over time, and interfere when the cell counts appear to increase above a given threshold. In Figure 2, cell counts of a herd are shown during approximately three years. Mean bulk milk SCC in this herd is approximately 200,000 cells/mL, with a standard deviation of approximately 35,000 cells/mL. Using process control algorithms, 95% confidence limits around a given mean can be calculated \cite{41}. Observations outside these intervals would represent true deviations of

### Table I. Prevalence of mastitis pathogens among cows in each Average lactation Linear Score (AVELS) category.

<table>
<thead>
<tr>
<th>AVELS</th>
<th>Total</th>
<th>Negative</th>
<th>Major pathogens$^a$</th>
<th>% Other major pathogens$^a$</th>
<th>% Minor pathogens$^a$</th>
<th>% Other</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.1–0.9</td>
<td>2643</td>
<td>2237</td>
<td>85$^+$</td>
<td>110</td>
<td>4$^+$</td>
<td>12</td>
<td>0$^+$</td>
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<tr>
<td>1.0–1.9</td>
<td>11245</td>
<td>9029</td>
<td>80$^+$</td>
<td>484</td>
<td>4$^+$</td>
<td>66</td>
<td>1$^+$</td>
</tr>
<tr>
<td>2.0–2.9</td>
<td>14727</td>
<td>9736</td>
<td>66$^+$</td>
<td>1156</td>
<td>8$^+$</td>
<td>93</td>
<td>1$^+$</td>
</tr>
<tr>
<td>3.0–3.9</td>
<td>12609</td>
<td>6465</td>
<td>51$^+$</td>
<td>1998</td>
<td>16$^+$</td>
<td>122</td>
<td>1$^+$</td>
</tr>
<tr>
<td>4.0–4.9</td>
<td>9661</td>
<td>3939</td>
<td>41$^+$</td>
<td>2772</td>
<td>29$^+$</td>
<td>162</td>
<td>2$^+$</td>
</tr>
<tr>
<td>5.0–5.9</td>
<td>7226</td>
<td>2307</td>
<td>32$^+$</td>
<td>2850</td>
<td>39$^+$</td>
<td>136</td>
<td>2$^+$</td>
</tr>
<tr>
<td>6.0–6.9</td>
<td>4431</td>
<td>1186</td>
<td>27$^+$</td>
<td>2042</td>
<td>46$^+$</td>
<td>134</td>
<td>3$^+$</td>
</tr>
<tr>
<td>7.0–7.9</td>
<td>1946</td>
<td>456</td>
<td>23$^+$</td>
<td>989</td>
<td>51$^+$</td>
<td>70</td>
<td>4$^+$</td>
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<tr>
<td>8.0–8.9</td>
<td>592</td>
<td>146</td>
<td>25$^+$</td>
<td>294</td>
<td>50$^+$</td>
<td>34</td>
<td>6$^+$</td>
</tr>
<tr>
<td>9.0–9.9</td>
<td>148</td>
<td>39</td>
<td>26$^+$</td>
<td>64</td>
<td>43$^+$</td>
<td>13</td>
<td>9$^+$</td>
</tr>
<tr>
<td>Total</td>
<td>65228</td>
<td>35540</td>
<td>54$^+$</td>
<td>12759</td>
<td>20</td>
<td>842</td>
<td>1</td>
</tr>
</tbody>
</table>

$^+$ Significantly more than expected, \( P < 0.05 \); $^-$ Significantly less than expected, \( P < 0.05 \).

$^a$ Major pathogens are \textit{S. agalactiae}, \textit{S. aureus}, environmental streptococci, and \textit{Mycoplasma}; other major pathogens are \textit{E. coli}, \textit{Klebsiella}, \textit{Serratia}, \textit{A. pyogenes}; minor mastitis pathogens are \textit{C. bovis}, and CNS.
the udder health situation. In this example (Fig. 2) such deviations from acceptable are observed in May–June 2001, and in August–September 2002. During these periods, a further investigation into the reasons for this increase in bulk milk SCC is warranted (see further).

However, udder health and milk quality is defined by more parameters then prevalence of subclinical mastitis or SCC. Clinical mastitis is in herds an equally important issue. Recently, Elbers and co-workers reported an increase in herd level incidence of clinical mastitis with a decreasing bulk milk SCC [13, 29]. Barkema [1, 2] also reported a slight but non-significant increase in clinical mastitis in herds with a low bulk milk SCC (Fig. 3), but they observed a greater proportion of clinical cases with systemic signs of illness in herds with a low bulk milk SCC compared to herds with a higher bulk milk SCC [2]. Average rate of clinical cases in the latter (Dutch) study was 26 cases per 100 cow-year at risk. Clinical mastitis incidence rates have increased in Denmark in the last four years (from 29% to 39%), decreased in Finland (from 32% to 23%) and approximately level in Norway (45%) and Sweden (21%) [34]. Clearly, a decrease in prevalence of subclinical mastitis has not resulted in an associated decrease in clinical mastitis [14, 20].

2.4. Somatic cell count patterns at population level

Since the introduction of a standard mastitis prevention program by Neave [32], an enormous progress in decreasing the prevalence of infection, and in decreasing average bulk milk somatic cell count (BMSCC) in national milk production has been achieved. As an illustrative example Honkanen-Buzalsky and Myllys [22] and Myllys et al. [31] reported the prevalence of intra mammary infections in Finnish cattle in both 1988 and 1995. An important decrease in prevalence was seen with S. agalactiae (from 0.78% to 0.12%), S. dysgalactiae (0.79 to 0.08), S. uberis
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(1.18 to 0.72), and *S. aureus* (5.12 to 3.50). A slight increase was observed in coliforms (0.26 to 0.29). Similarly, minor pathogens such as CNS (6.57 to 11.24) and coryneforms (0.82 to 3.49) have also increased in prevalence. During these years an associated decrease in bulk milk SCC was also observed: from 330 000 in 1988 to 170 000 cells/mL in 1995. In other countries in the same period a decrease in bulk milk SCC was also observed [35, 39, 40, 46]. Apparently, the standard mastitis prevention program has indeed been successful to reduce the prevalence of infections.

Monitoring milk quality over time provides an opportunity to evaluate progress, study relationships between milk quality parameters and estimate the efficacy of control programs [5, 45, 49].

The monitoring data that is summarized here was obtained from five of the largest milk plants operating in New York State (Agri-Mark, Allied Federated Cooperatives Inc., Dairylea Cooperatives Inc., Dairy Farmers of America, Upstate Farms Cooperatives) [57]. Data included monthly milk loads (in kg) and test results for SCC, bacteria count (reported as Plate Loop Count, PLC), antibiotic residue violations, freezing point, butter fat, protein and lactose. In this study, only milk loads, SCC, PLC and antibiotic residue violations were used. To analyze consequences related to farm size (milk load), farms were classified into milk load categories month by month, according to quartiles of total observations. Q1 included farms with ≤ 23 000 kg in a specific month, Q2 had farms with 23 000 < kg ≤ 34 000, Q3 had farms with 34 000 < kg ≤ 68 000, and Q4 had farms with milk loads > 68 000 kg in a specific month. To evaluate the contribution of an individual farm to overall SCC in the milk pool, a new parameters, termed SCC contribution was calculated for each farm in each month SCC contribution was determined as the excess SCC over the previously defined cut-off level of 200 000. SCC contribution was calculated for a particular farm in a particular month weighted by the amount of milk the farm produced that month out of the total amount of milk in the milk pool that month [49].

The average weighted SCC (weights according to amount of milk sold) was $308 \times 10^3$ cells/mL, the average PLC amounted $24.4 \times 10^3$ bacteria/mL, and the average number of antibiotic residue violations in the pool of milk was 3.9 per 1000 producers. Each month between 72%
and 88% of the milk pool had SCC levels in compliance with the EU requirements (SCC < 400 × 10^3 cells/mL). Larger farms had lower SCC and PLC but more antibiotic violations. However, the larger farms contribute most to the SCC of the total pool of milk. Farms with high SCC also had higher PLC and more antibiotic violations but contributed as a group relatively few somatic cells to the milk pool (Fig. 4). As shown in Figure 4, the group of farms with moderately elevated SCC (200–400 × 10^3) or more elevated but still legal (in USA) cell counts (500–750 × 10^3) contribute most to the overall SCC in the pool of milk in New York State. SCC contribution reflects both SCC levels and the size of the farms (kg of milk sold per month). Total SCC contribution per category reflects, in addition, the number of farms in that category. Hence the highest SCC contributions come from farms that are not necessarily in the highest SCC level category. Farms in the 200 < SCC < 400 and 500 < SCC < 750 categories contribute most to the SCC level in the milk of NYS. Figure 5 shows that farms with SCC levels below 200 × 10^3 cells/mL had low PLC levels (equal or less than 25 × 10^3 bacteria/mL). Furthermore, farms with high SCC levels more often had high PLC levels. This finding may imply that subclinical mastitis cases cause an increase in bulk milk bacteria count [19]. However, herds with subclinical mastitis may also have problems in the area of general hygiene and milking equipment cleaning and disinfection [1, 2].

Farms with higher SCC levels (SCC > 750) showed a much higher rate of antibiotic residue violations. Furthermore, larger farms exhibit a higher rate of antibiotic residue violations (Fig. 6).

3. MILK QUALITY MONITORING WITHIN A HERD HEALTH PROGRAM

The general process of herd health management is depicted in Figure 7. First, the goals of the producer are identified. Based on these goals, the current practices are
evaluated, and the risk profile in the herd is identified. The goals and the current risk profile are used by the veterinarian to identify the most obvious gaps in current management of the farm operation [1, 2, 9, 61]. These gaps are discussed with the producer, and improvements are planned. For each of the planned tasks, the procedures are discussed, and the responsible person is identified. New procedures are implemented, and implementation is documented on the farm. On a regular basis the results are monitored and evaluated, and when things go according to plan, the goals can be further discussed, and continuous improvement is possible. Each of these components will be discussed in somewhat more detail, with a special emphasis on monitoring and evaluation processes.

3.1. Goal setting

This could be a bulk milk somatic cell count below 200,000 with a mastitis incidence of less than 20%. To be able to evaluate true deviations from proposed goals, an estimate of “normal” variability around the goal should also be calculated. Usually, variability in mean SCC increases with the mean SCC, and variability decreases with herd size. Formal quality control procedures can be defined to identify abnormal SCC performance [59]. The goals should be defined in a collaboration between the dairy producer and the herd health veterinarian. Realistic goals are important, when goals are beyond reach, this could actually lead to a loss of motivation for the producer and the employees.

3.2. Risk assessment and planning

When thinking about Risk assessment it is probably essential to keep the biology of udder health dynamics as a central focus. In Figure 8, the dynamics of infection in herds is depicted. The key components in this figure are the rate of new infections,

![Figure 5. Total milk produced cross-classified by PLC category and SCC category.](image)

![Figure 6. Antibiotic residue violations per 1000 producers by SCC category (a) and by milk production quartile (b) (adapted from [57]).](image)
the cure rate, the culling rate, and the entry of infected or non-infected animals into the milking herd [4]. For each of these components of the herd infection dynamics a number of specific risk factors are known [13, 15, 48, 61].

3.3. Monitoring and evaluation

Monitoring and evaluation of the current performance should include analysis of data from the herd, evaluation of farm management and housing through observation, and the observation of clinical data from animals in the herd. For each of the components of udder health dynamics as presented in Figure 8 a number of parameters for monitoring and evaluation are presented.

3.3.1. Entry into the milking herd

Evaluation of culture results from cows purchased into the herd is the preferred option for monitoring the entry of clean animals into the herds. A similar monitoring system should be implemented for incoming heifers. As an alternative (albeit less accurate) the evaluation of cell count data in the first milking after calving can be used to evaluate the udder health status of incoming animals. A goal could be to have less than 10% of incoming heifers with a cell count over 200,000 to 250,000 cells per mL. In the herd depicted in Figure 9, the percentage of infected heifers is much too high (approximately 20%). Monitoring of housing quality and hygiene levels for

<table>
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<th>9.6</th>
<th>9</th>
<th>8.9</th>
<th>8.3</th>
<th>7.7</th>
<th>7.0</th>
<th>6.4</th>
<th>5.8</th>
<th>5.1</th>
<th>4.5</th>
<th>3.9</th>
<th>3.2</th>
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<td>1</td>
<td>1</td>
<td>1</td>
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<td>5</td>
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<td>4</td>
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<td># cows:</td>
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<td>121</td>
<td>66</td>
<td>27</td>
<td>7</td>
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</tbody>
</table>

Figure 9. Linear score at calving for cows in lactation number 1 to 6. The numbers in the graph reflect the number of animals in each lactation at each LS level.
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heifers and dry cows is then an important tool in further analysis of the udder health problems in this herd [36].

3.3.2. New infections

Recording and sampling of clinical mastitis cases is important to obtain data on new clinical infections. Clinical mastitis data can be summarized as the number of cows with at least one clinical case per lactation, or alternatively can be graphed against days in milk at the occurrence of clinical case (Fig. 10).

For subclinical infections, repeated bacteriological culture of all quarters of all cows in the herd provides excellent data for research in infection dynamics monitoring [4, 26] but is clearly not practical for commercial herds. Repeated cell counting of animals is a relative cheap and more practical method to obtain information on new infections. The proportion of low SCC animals that has a SCC over 250,000 in the next measurement is a key proxy monitoring tool for subclinical infections [47]. This proportion should generally be less than 10%. In Figure 11, the linear score in the last test day (LS) is graphed against the linear score in the previous test day (PLS). A linear score of 4.5 is equal to approximately 250,000 cells. The left-hand upper quadrant shows the cows with a new infection. Cows with new infections, where high SCC remain in subsequent samples should be sampled for evaluation of bacteriological status (now in the upper right-hand quarter of Fig. 11). Observation of milking procedures, scoring of teat-end quality, scoring of cow and udder hygiene, scoring of cubicle hygiene, evaluation of separation of chronically infected animals and evaluation of the cow environment are key risk factors that affect the new infection rate.

In Figure 12, the new infections are plotted against days in milk. In this herd there is a high rate of new infections in the first 100 days of lactation. This was also observed for the clinical mastitis cases (Fig. 10). This points towards management procedures in the dry period, transition period or with the early fresh cows that lead to high infection risks. These risk factors should evaluated in much more detail when data such as shown in this herd example are observed.

3.3.3. Cure

Monitoring of success of treated animals can be done using culturing of treated animals, or by following cell count patterns of animals that have been treated. An example of this is monitoring of dry cow
treatment by graphing of cell counts at dry-off versus cell counts at calving. Usually at least 80% of cows with high cell count at dry-off should be low at calving. A similar analysis of cell count data can be used to evaluate treatments during lactation. Cows with high cell count at previous test day that were treated in lactation should have at least a 50% chance of having a low cell count at current test day. Monitoring of susceptibility patterns of bacteria isolated from clinical and subclinical infections is important to observe whether current treatment protocols are adequate. Monitoring

Figure 11. Graph of previous test day linear score (PLS) versus last test day linear score (LS).

Figure 12. Occurrence of new infections (defined as a cow with a previous LS < 4.5 and next LS > 4.5) graphed against days in milk (DIM) at occurrence of new infection.
the treatment protocols, the proper use of medications, the quality of storage and maintenance of drug cabinets and the prevention of treatment residues should be evaluated on a regular basis (i.e. annually).

In Table II, the recent cases of clinical mastitis are tabulated. Information in this table includes cow number (CowID), current days in milk (DIM), lactation number (Lact), date of calving (FreshDat), date of mastitis (MastDat), treatment remark (Remark), the previous and last test dates (PrevTdat, Tdat), previous and last linear score (PLS, LS) and the number of mastitis cases in this lactation (Nmast). Using the data in Table II, an impression of clinical cure rate per treatment protocol can be obtained. Using linear score after the last clinical case again as a proxy for cure, cow 1678 would be considered a non-cure whereas cow 2957 would be considered a cure.

Another important component here is the evaluation of non-cured or even non-treated animals. These are the long term infected animals, in Figure 11 these are the animals in the upper right hand corner. A number of these animals may be eligible for treatment, especially valuable young animals with a relative short duration of infection should be considered [52, 53].

### 3.3.4. Culling

Monitoring of culling can be done by evaluating the culling data, evaluating the reasons for culling, evaluation the mean

<table>
<thead>
<tr>
<th>CowID</th>
<th>DIM</th>
<th>Lact</th>
<th>FreshDat</th>
<th>MastDat</th>
<th>Remark</th>
<th>PrevTdat</th>
<th>Tdat</th>
<th>PLS</th>
<th>LS</th>
<th>Nmast</th>
</tr>
</thead>
<tbody>
<tr>
<td>1678</td>
<td>235</td>
<td>2</td>
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<td>12/13/00</td>
<td>RFPIRSUE</td>
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<td>2/12/01</td>
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<td>7.2</td>
<td>2</td>
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<td>1</td>
</tr>
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<td>251</td>
<td>5</td>
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<td>4.1</td>
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<td>12/18/00</td>
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<td>2/12/01</td>
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<td>0.9</td>
<td>2</td>
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<td>12/19/00</td>
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<td>2/12/01</td>
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<td>12/26/00</td>
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<td>1/11/01</td>
<td>2/12/01</td>
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<td>2.0</td>
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</tbody>
</table>

- CowID: cow number, DIM: current days in milk, Lact: lactation number, FreshDat: date of calving, MastDat: date of mastitis, Remark: treatment remark, PrevTdat, Tdat: previous and last test dates, PLS, LS: previous and last linear score, Nmast: number of mastitis cases in this lactation.
SCC and the number of clinical events of culled animals are useful tools to monitor the udder health impact on culling. Probably a more important part of monitoring is the evaluation of cows that have not been culled and should have been placed on the culling list. A list of cows in the herd sorted by the number of mastitis cases and also by the length of high SCC in the current and previous lactation is a tool for analyzing the presence of animals that should have been considered for culling. An example of such a list is shown in Table III. Information in this table includes cow number (CowID), lactation number (Lact), current days in milk (DIM), economic value of this cow in US $ compared to a replacement heifer (CwVal), relative production level compared to the average cow in the herd (Relv), linear score after calving (LS1), the last four linear scores (PLS4, PLS3, PLS, LS), average linear score up until the last test date (AVLS) and the number of mastitis cases in this lactation (Nmast).

### Table III. List of cows with a high linear score, multiple cases of clinical mastitis, and sorted by their economic value.

<table>
<thead>
<tr>
<th>CowID</th>
<th>Lact</th>
<th>DIM</th>
<th>CwVal</th>
<th>Relv</th>
<th>LS1</th>
<th>PLS4</th>
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<td>4.7</td>
<td>3</td>
</tr>
</tbody>
</table>

*a* CowID: cow number, Lact: lactation number, DIM: current days in milk, CwVal: economic value of this cow in US $ compared to a replacement heifer, Relv: relative production level compared to the average cow in the herd, LS1: linear score after calving, PLS4, PLS3, PLS, LS: last four linear scores, AVLS: average linear score up until the last test date, Nmast: number of mastitis cases in this lactation.
4. CONCLUSION

Somatic cell counts are a valuable component of monitoring programs. Somatic cell count measure the inflammatory response to an IMI. We have shown that with some care SCC can be used as a proxy for measuring IMI and milk quality at cow, herd and population level. In this review we have argued that udder health is more than low somatic cell counts. It also includes low incidence of clinical mastitis, minimize potential hazards for human health such as prevention of residues in milk, potential transfer of antibiotic resistance to human pathogens, and transfer of pathogens through dairy products. Addressing consumer demands with regard to product safety, transmission of infectious diseases, welfare and eco-system health [56] becomes a constraint on the dairy production systems. It is therefore inevitable that more complete udder health programs and monitoring systems have been developed and implemented.

Udder health monitoring is an essential component of preventive veterinary medicine. Preventive programs that address all important udder health components are necessarily complex and dynamic. These programs will need to include components of the standard mastitis prevention plan (milking technique and milking machine performance, post-milking teat disinfection, culling policy for chronically infected animals, antibiotic treatment at dry-off and clinical events) [32], but additionally address hygiene, nutrition, housing and cow comfort, air and water quality, antibiotic use, health monitoring, breeding policy, and cow characteristics such as immunologic competence, cow conformation (teat and udder) and milk production level [21, 41, 48, 61]. In each of these categories, a number of critical issues should be defined, and included into a comprehensive control scheme. Recently, a number of control programs aimed at all udder health issues have been designed [9, 16]. The efficacy of such programs to optimize all udder health issues has not been shown yet. It may be expected that compliance to actively participate will not be high unless monetary incentives and/or production restrictions are a component of such udder health programs [46]. These programs are dynamic and complex. Implementation of complete udder health programs should be accompanied by research efforts to further fine-tune these complete udder health control and monitoring programs.

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REFERENCES


