

A polynomial algorithm for a simple scheduling problem at cross docking terminals

Ruslan Sadykov

► **To cite this version:**

Ruslan Sadykov. A polynomial algorithm for a simple scheduling problem at cross docking terminals.
[Research Report] RR-7054, 2009. inria-00412519v2

HAL Id: inria-00412519

<https://hal.inria.fr/inria-00412519v2>

Submitted on 6 Oct 2009

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

A polynomial algorithm for a simple scheduling problem at cross docking terminals

Ruslan Sadykov*

INRIA Research Report

October 6, 2009

Abstract

At cross docking terminals, products from incoming trucks are sorted according to their destinations and transferred to outgoing trucks using a small temporary storage. Such terminals allow companies to reduce storage and transportation costs in supply chain. This paper focuses on the operational activities at cross docking terminals.

We consider the trucks scheduling problem with the objective to minimise the storage usage during the product transfer. We show that a simplification of this NP-hard problem in which the arrival sequences of incoming and outgoing trucks are fixed is polynomially solvable and propose a dynamic programming algorithm for it.

1 Introduction

Cross docking terminal is a distribution center carrying a considerably reduced amount of stock in contrast to traditional warehouses. Incoming shipments delivered by incoming trucks are unloaded, sorted and loaded onto outgoing trucks waiting at the dock, which forward the shipments to the respective locations within the distribution system. Compared to traditional warehousing, a cost intensive storage and retrieval of goods is eliminated by a synchronization of inbound and outbound flows. An additional advantage of cross docking is efficient usage of truck capacity (i.e. full loads) and the implementation of a good scheduling system [1].

In this paper, we consider a simplified cross docking terminal with one receiving, one shipping door and a storage. The incoming trucks fully loaded by products of different types arrive at the receiving door where the products are unloaded. Each truck leaves the platform when it is fully unloaded. Each unloaded product is processed and transferred to the shipping door if the outgoing truck currently staying there demands the products of this type and does

*INRIA Bordeaux — Sud-Ouest, France, e-mail: Ruslan.Sadykov@inria.fr

not yet have enough of them. Otherwise, the unloaded product is transferred to the temporary storage. Each outgoing truck can leave the door only if it is fully loaded by products moved either directly from incoming trucks or from the storage. The objective is to find the best arrival (or departure) order for the whole set of incoming and outgoing trucks to increase the efficiency of the cross docking terminal.

There are several papers in the literature which dealt with this scheduling problem. Yu and Egbelu [4] developed a model for scheduling incoming and outgoing trucks to minimise the makespan (i.e. the maximum completion time) and proposed several heuristic algorithms for it. A simpler but similar problem was considered by Boysen, Flidner and Scholl [2]. They have proposed some lower bounds and an exact decomposition approach for it. Maknoon, Baptiste and Kone [3] concentrated on the objective function which minimizes the storage cost in a simplified setting in which the sequences of incoming and outgoing trucks are fixed and any outgoing truck demands the products of only one type.

We now define the problem formally. The number of incoming and outgoing trucks are, respectively, n and m . The number of different product types is T . We will denote the i -th incoming truck as \mathbf{I}_i and the o -th outgoing truck as \mathbf{O}_o . There are several types of products. Each incoming truck \mathbf{I}_i supplies a_{it} products of type t . Each outgoing truck \mathbf{O}_o demands b_{ot} products of type t . Let also T_o be the set of product types demanded by \mathbf{O}_o , i.e. $T_o = \{t : b_{ot} > 0\}$. We suppose that each outgoing truck demands products of at most q different types, i.e. $q = \max_{1 \leq o \leq m} |T_o|$. Note that, for each type t , the total number of supplied products of this type should be not less than the total number of demanded products of this type, otherwise at least one outgoing truck would not be able to depart fully loaded.

A product of type t can be moved directly from an incoming truck at the receiving door to an outgoing truck \mathbf{O}_o at the shipping door if the number of products of type t which are already in \mathbf{O}_o is less than b_{ot} . Otherwise the product should be transferred to the temporary storage at a cost c_t . One product of type t occupies a volume d_t in the storage. The total volume of products in the storage should not exceed its total capacity which is D . The problem consists in finding a policy for unloading and loading products such that the total cost is minimized. For convenience, an equivalent objective is used which is the maximization of the total cost of the directly transferred products. This problem is NP-hard in the strong sense even for a very restricted case, as shown in Appendix A.

Economically, the storage cost can be also interpreted as the difference between the time needed to transfer a product directly from an incoming to an outgoing truck and via the intermediate storage. So, our objective function is equivalent to the total processing cost. Although this objective is similar to the makespan objective used in [2] and [4], the problems studied there are quite different. The main difference is that we do not allow any concurrent operations. In contrast to this, for example in [4], one can at the same time discharge a product from an incoming truck to the storage and load a product from the storage onto an outgoing truck. As a result, transferring a product via the storage does

not always increase the objective function.

In the rest of the paper we concentrate on the problem in which the sequences of incoming and outgoing trucks are fixed. Formally this means that, for each $i \in \{1, \dots, n-1\}$, the truck \mathbf{I}_i departs before the arrival of \mathbf{I}_{i+1} , and for each $o \in \{1, \dots, m-1\}$, the truck \mathbf{O}_o departs before the arrival of \mathbf{O}_{o+1} .

Note that fixing sequences of incoming and outgoing trucks in problems considered in [2] and [4] make them trivial, and thus polynomially solvable. However, determining the computational complexity of our problem with fixed sequences is not trivial at all. In [3], Maknoon et al. considered this problem with some restrictions including $q = 1$, but left the complexity question open.

In this paper, we answer this question by presenting a polynomial dynamic programming algorithm for the problem considered.

2 Preliminary observations

In the problem, we need to find an optimal policy which is characterized, among others, by the departure order of trucks. Note that the departure order is fixed within the set of incoming or outgoing trucks but not fixed within the set of all trucks. A departure order of all trucks can be represented by a path in the graph depicted in Figure 1. In this graph, each node (i, o) represents the situation in which the trucks \mathbf{I}_i and \mathbf{O}_o are at the doors.

The path shown on this picture corresponds to the policy which is characterized by the following departure order of trucks:

$$(\mathbf{I}_1, \mathbf{I}_2, \mathbf{I}_3, \mathbf{O}_1, \mathbf{O}_2, \mathbf{I}_4, \mathbf{I}_5, \mathbf{O}_3, \mathbf{O}_4, \mathbf{I}_6, \dots, \mathbf{I}_{n-1}, \mathbf{O}_6, \dots, \mathbf{O}_{m-1}, \mathbf{I}_n, \mathbf{O}_m).$$

Some paths in the departure order graph are not feasible due to the following storage constraints. First, for each t , the number of products of type t in the storage should always be non-negative (all demanded products should be first supplied). Second, the overall volume of products in the storage can never exceed its capacity. We say a node is *infeasible* if it does not belong to any feasible path. Formally, a node (i, o) is infeasible if and only if

$$\exists t : \sum_{k=1}^i a_{kt} < \sum_{p=1}^{o-1} b_{pt} \quad \text{or} \quad \sum_{t=1}^T d_t \cdot \left(\max \left\{ 0, \sum_{k=1}^{i-1} a_{kt} - \sum_{p=1}^o b_{pt} \right\} \right) > D. \quad (1)$$

From (1) it follows that

1. if two nodes (i', o) and (i'', o) such that $i' < i''$ are feasible then all nodes (i, o) such that $i' \leq i \leq i''$ are feasible;
2. if two nodes (i, o') and (i, o'') such that $o' < o''$ are feasible then all nodes (i, o) such that $o' \leq o \leq o''$ are feasible.

A possible situation of infeasible nodes (which are in black) is given in Figure 1. For a fixed o^* , we denote by $fi(o^*)$ and $li(o^*)$ the smallest and the largest

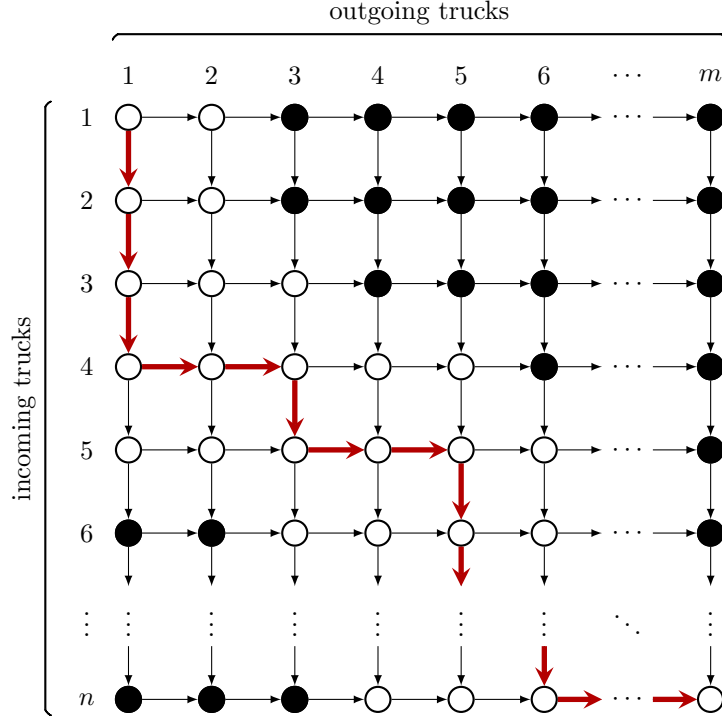


Figure 1: The departure order graph and an example of the path

values of i such that node (i, o^*) is feasible. In the same manner, for a fixed i^* , we denote by $fo(i^*)$ and $lo(i^*)$ the smallest and the largest values of o such that node (i^*, o) is feasible.

The next fact is very important and serves as the base for the algorithm.

Observation 1 *There exists an optimal policy in which, each time trucks \mathbf{I}_k and \mathbf{O}_j are at the doors, for each t , \mathbf{I}_k transfers directly to \mathbf{O}_j as many products of type t as possible, i.e. the minimum between the number of products of type t still available in \mathbf{I}_k and the number of products of type t which are still demanded by \mathbf{O}_j .*

It is easy to see that this observation is correct. Suppose, in an optimal policy, \mathbf{I}_k does not transfer to \mathbf{O}_j z products of type t it can and “saves” them for consequent outgoing truck(s). Then \mathbf{O}_j is obliged to take these z products from the storage. In the modified policy, these z products are transferred directly from \mathbf{I}_k to \mathbf{O}_j , and products, transferred directly from \mathbf{I}_k to consequent outgoing truck(s) in the original policy, are taken from the storage. The cost of the modified policy which comply with the observation do not increase.

We will call a policy which complies with Observation 1 *direct first*. Each path in the departure order graph corresponds to exactly one direct first policy.

3 The algorithm

We now present a dynamic programming algorithm for the problem which finds an optimal direct first policy.

Each state in our dynamic programming algorithm does not correspond to a node in the departure order graph but to an edge, which a path can follow just after “turning”. To clarify the presentation, we first present in Figure 2 the underlying directed graph of the dynamic programming algorithm. For each pair of incoming truck \mathbf{I}_i and outgoing truck \mathbf{O}_o , there is the “square node” (i, o) and the “circle node” (i, o) . Each node corresponds to a set of states of the algorithm, each edge defines all possible moves between a state in the origin of this edge and a state in the destination of this edge. The paths shown in Figures 1 and 2 correspond to the same policy. Again, the infeasible nodes are shown in black.

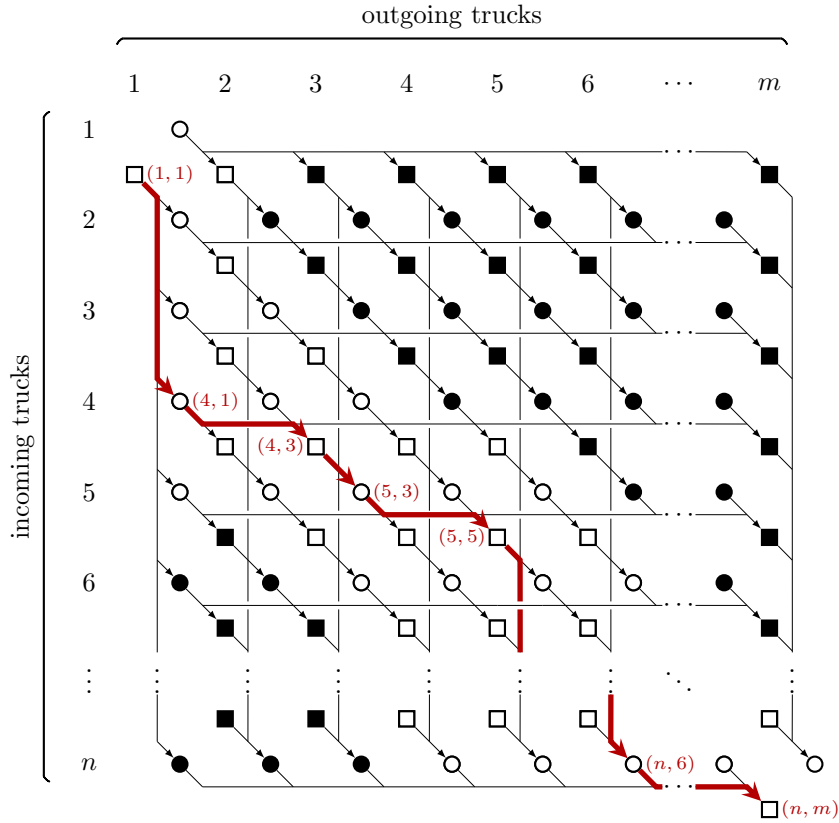


Figure 2: The underlying directed graph for the dynamic programming algorithm

Each “square node” (i, o) in the graph in Figure 2 represents the situation

in which truck \mathbf{I}_i has departed, truck \mathbf{I}_{i+1} is going to arrive, truck \mathbf{O}_o is at the shipping door, and \mathbf{O}_o had arrived after the departure of the truck \mathbf{I}_{i-1} . This means that \mathbf{O}_o could get products directly only from \mathbf{I}_i .

In our dynamic programming algorithm, for each “square node” (i, o) , we define the states $\mathbf{S}^{out}(i, o, \{f_t\}_{t \in T_o})$, where f_t is the number of products of type t transferred directly by truck \mathbf{I}_i to \mathbf{O}_o .

Each “circle node” (i, o) represents the situation in which truck \mathbf{O}_o has departed, truck \mathbf{O}_{o+1} is going to arrive, truck \mathbf{I}_i is at the shipping door, and \mathbf{I}_i had arrived after the departure of the truck \mathbf{O}_{o-1} . This means that \mathbf{I}_i could ship products directly only to \mathbf{O}_o .

In our dynamic programming algorithm, for each “circle node”, we define the states $\mathbf{S}^{inc}(i, o, \{f_t\}_{t \in T_o})$, where f_t is the number of products of type t transferred directly by truck \mathbf{I}_i to \mathbf{O}_o .

Now, every policy corresponds to exactly one sequence of states or a states path. Therefore, in the following, we will use the same notation for a states path and the corresponding policy. A path which corresponds to a direct first policy, we will call a direct first path. For each state $\mathbf{S}^{out}(i, o, \{f_t\}_{t \in T_o})$ and $\mathbf{S}^{inc}(i, o, \{f_t\}_{t \in T_o})$, we keep and update the objective function value of the best path to this state. Let $V^{out}(i, o, \{f_t\}_{t \in T_o})$ and $V^{inc}(i, o, \{f_t\}_{t \in T_o})$ be these values. To solve the problem, we need to find the best path terminating at a state $\mathbf{S}^{out}(n, m, \{f_t\}_{t \in T_o})$ or $\mathbf{S}^{inc}(n, m, \{f_t\}_{t \in T_o})$.

To simplify the presentation, when there is no ambiguity, we will use the shortened notations $\mathbf{S}^{inc}(i, o, f)$, $\mathbf{S}^{out}(i, o, f)$, $V^{inc}(i, o, f)$, and $V^{out}(i, o, f)$, where f is a vector.

Note that, in the algorithm, we handle only the states belonging to at least one direct first path. The total number of such states can be different even for instances of the same size.

Suppose we are in a state $\mathbf{S}^{inc}(i, o, f)$. From this state it is only possible to move to a state $\mathbf{S}^{out}(i, o', f')$, where $\max\{o+1, fo(i+1)\} \leq o' \leq lo(i)$ and $fo(n+1) = m$. When we make such a move, truck \mathbf{I}_i transfers directly to every truck \mathbf{O}_j , $o < j \leq o'$, as much products as possible.

The formal procedure for making moves from a state $\mathbf{S}^{inc}(i, o, f)$ is presented in Algorithm 1. The complexity of this procedure is $O(m(q+\rho))$, where ρ is the number of operations needed to check if a state has been already encountered earlier in the algorithm. We will compute ρ later in the paper.

Suppose now we are in a state $\mathbf{S}^{out}(i, o, f)$. From this state it is only possible to move to a state $\mathbf{S}^{inc}(i', o, f')$, where $\max\{i+1, fi(o+1)\} \leq i' \leq li(o)$, and $fi(m+1) = n$. When we make such a move, truck \mathbf{O}_o receive directly from every truck \mathbf{I}_k , $i < k \leq i'$, as much products as possible.

The formal procedure for making moves from a state $\mathbf{S}^{out}(i, o, f)$ is presented in Algorithm 2. The complexity of this procedure is $O(n(q+\rho))$.

In the full algorithm, presented as Algorithm 3, we look through all the created states and make all possible moves from them as described above. To obtain an optimal policy, it suffices to store, for each state, along the value V , the path which gives it. At the end of the algorithm, the best path stored for one of the states $\mathbf{S}^{out}(n, m, f)$ and $\mathbf{S}^{inc}(n, m, f)$ corresponds to an optimal policy.

```

1 for t = 1 to T do r[t] ← ait;
2 for t ∈ To do r[t] ← ait - ft;
3 v ← Vinc(i, o, f);
4 for j ← o + 1 to lo(i) do
5   for t ∈ Tj do
6     dt[t] ← min{r[t], bjt};
7     v ← v + ct · dt[t];
8   if state Sout(i, j, {dt[t]}t∈Tj) does not exist then
9     create it: Vout(i, j, {dt[t]}t∈Tj) ← -∞;
10  if j ≥ fo(i + 1) and v > Vout(i, j, {dt[t]}t∈Tj) then
11    Vout(i, j, {dt[t]}t∈Tj) ← v;
12  for t ∈ Tj do r[t] ← r[t] - dt[t];

```

Algorithm 1: Algorithm to make moves from a state $\mathbf{S}^{inc}(i, o, f)$.

```

1 for t ∈ To do r[t] ← ft;
2 v ← Vout(i, o, f);
3 for k ← i + 1 to li(o) do
4   for t ∈ To do
5     dt[t] ← min{bot - r[t], akt};
6     v ← v + ct · dt[t];
7   if state Sinc(k, o, {dt[t]}t∈To) does not exist then
8     create it: Vinc(k, o, {dt[t]}t∈To) ← -∞;
9   if k ≥ fi(o + 1) and v > Vinc(k, o, {dt[t]}t∈To) then
10    Vinc(k, o, {dt[t]}t∈To) ← v;
11  for t ∈ To do r[t] ← r[t] + dt[t];

```

Algorithm 2: Algorithm to make moves from a state $\mathbf{S}^{out}(i, o, f)$.


```

1 for  $t \in T_1$  do  $dt[t] \leftarrow \min\{a_{1t}, b_{1t}\}$ ;
2  $v \leftarrow \sum_{t \in T_1} c_t \cdot dt[t]$ ;
3  $V^{inc}(1, 1, \{dt[t]\}_{t \in T_1}) \leftarrow v$ ;
4 run Algorithm 1 for the state  $\mathbf{S}^{inc}(1, 1, \{dt[t]\}_{t \in T_1})$ ;
5  $V^{out}(1, 1, \{dt[t]\}_{t \in T_1}) \leftarrow v$ ;
6 run Algorithm 2 for the state  $\mathbf{S}^{out}(1, 1, \{dt[t]\}_{t \in T_1})$ ;
7 for  $i \leftarrow 1$  to  $n$  do
8   if  $i > 1$  then
9     for  $o \leftarrow fo(i)$  to  $\min\{lo(i-1), lo(i)-1\}$  do
10    |   run Algorithm 1 for all created states  $\mathbf{S}^{inc}(i, o, f)$ ;
11   if  $i < n$  then
12     for  $o \leftarrow \max\{fo(i)+1, fo(i+1)\}$  to  $lo(i)$  do
13    |   run Algorithm 2 for all created states  $\mathbf{S}^{inc}(i, o, f)$ ;
14 return  $\max\{V^{out}(n, m, f), V^{inc}(n, m, f)\}$ 

```

Algorithm 3: The full algorithm

We will now determine the complexity of the algorithm. The critical point here is estimating the total number of created states in the algorithm.

To do it, we will need the following lemma. But first we introduce some additional notations. Let $\mathcal{P}^{out}(i, o, f)$ and $\mathcal{P}^{inc}(i, o, f)$ be the sets of paths which terminate at states $\mathbf{S}^{out}(i, o, f)$ and $\mathbf{S}^{inc}(i, o, f)$. Let also $\mathcal{P}^{out}(i, o, f, o', f')$ be the set of paths which terminate at state $\mathbf{S}^{out}(i, o, f)$ and contain state $\mathbf{S}^{inc}(i, o', f')$ such that $o' < o$. Similarly, let $\mathcal{P}^{inc}(i, o, f, i', f')$ be the set of paths with terminate at state $\mathbf{S}^{inc}(i, o, f)$ and contains state $\mathbf{S}^{out}(i', o, f')$ such that $i' < i$. Finally, for a path P , we will denote as P_- its sub-path in which the terminating state of P is excluded.

Lemma 1

1. For any two direct first paths $P' \in \mathcal{P}^{out}(i, o, f')$ and $P'' \in \mathcal{P}^{out}(i, o, f'')$, $f' \neq f''$, if, for some type $t' \in T_o$, $f'_{t'} < f''_{t'}$, then $f'_t \leq f''_t$ for all types $t \in T_o$.
2. Analogously, for any two direct first paths $P' \in \mathcal{P}^{inc}(i, o, f')$ and $P'' \in \mathcal{P}^{inc}(i, o, f'')$, $f' \neq f''$, if, for some type $t' \in T_o$, $f'_{t'} < f''_{t'}$, then $f'_t \leq f''_t$ for all types $t \in T_o$.

Proof. We will prove this lemma by induction.

Suppose that the claim 2 is true for $i \leq i^*$ and $o < o^*$. We will prove the claim 1 for $i = i^*$ and $o = o^*$. Without loss of generality, let $P' \in \mathcal{P}^{out}(i^*, o^*, f', o', \bar{f}')$ and $P'' \in \mathcal{P}^{out}(i^*, o^*, f'', o'', \bar{f}'')$, $f' \neq f''$. As $f'_{t'} < f''_{t'}$, truck \mathbf{O}_{o^*} receives less products of type t' from \mathbf{I}_{i^*} in policy P' than in policy P'' . Therefore, \mathbf{I}_{i^*} transfers to trucks \mathbf{O}_o , $o' \leq o < o^*$ more products of type t' in policy P' than to

all trucks \mathbf{O}_o , $o'' \leq o < o^*$, in policy P'' . As P' and P'' are direct first policies, there are two possible cases:

- $o' = o''$ and $\bar{f}'_{t'} > \bar{f}''_{t'}$. Then, as $P'_- \in \mathcal{P}^{inc}(i^*, o', \bar{f}')$ and $P''_- \in \mathcal{P}^{inc}(i^*, o', \bar{f}'')$, by claim 2, $\bar{f}'_t \geq \bar{f}''_t$ for all types $t \in T_{o'}$.
- $o' < o''$.

In both cases, as P' and P'' are direct first policies, for every t , truck \mathbf{I}_{i^*} does not transfer more products of type t to all trucks \mathbf{O}_o , $o'' \leq o < o^*$ in policy P' than to all trucks \mathbf{O}_o , $o'' \leq o < o^*$, in policy P'' , and $f'_t \leq f''_t$ for all types $t \in T_{o^*}$.

Analogously, if we suppose that claim 1 is true for $i < i^*$ and $o \leq o^*$, we can prove the claim 2 for $i = i^*$ and $o = o^*$.

As the base of the reduction for the claim 1, we can take the case $i = 2$. In this case, we can only have $o' < o''$, as, for any fixed o^* , there is only one path which terminates at a state $\mathbf{S}^{inc}(2, o^*, f)$.

Analogously, as the base of the reduction of the claim 2, we can take the case $o = 2$. \square

Proposition 1 *The total number of created states in Algorithm 3 is $O(qnm^2)$.*

Proof. In the following, when a path contains a state $\mathbf{S}^{out}(i, o, f)$ or a state $\mathbf{S}^{inc}(i, o, f)$, we will say that, for each type t ,

- if $t \in T_o$, the path contains triple $(i, o, f)_t^{out}$ or $(i, o, f)_t^{inc}$;
- if $t \notin T_o$, the path contains triple $(i, o, 0)_t^{out}$ or $(i, o, 0)_t^{inc}$.

Let now fix type t^* . We will call triples $(i, o, 0)_{t^*}^{out}$, $(i, o, b_{ot^*})_{t^*}^{out}$, $(i, o, a_{it^*})_{t^*}^{inc}$, and $(i, o, 0)_{t^*}^{inc}$ t^* -canonical. There are $O(nm)$ such triples in total.

Consider a direct first path P' which contains a triple $(i', o', f')_{t^*}^{out}$. In the corresponding policy, truck $\mathbf{O}_{o'}$ receives directly from every truck \mathbf{I}_i , $i' < i \leq i''$, $\nu_{i'o't^*}^{out}(f')$ products of type t^* , where

$$\nu_{i'o't^*}^{out}(f') = \max \left\{ 0, \min \left\{ a_{it^*}, b_{o't^*} - f' - \sum_{k=i'+1}^{i-1} a_{kt^*} \right\} \right\}.$$

Therefore, path P contains a triple $(i'', o', \nu_{i'o't^*}^{out}(f'))_{t^*}^{inc}$, where $i'' > i'$. An important observation is that there is at most one value i'' for which such triple is not t^* -canonical. We can say that any triple $(i', o', f')_{t^*}^{out}$ “generates” at most one triple $(i'', o', f'')_{t^*}^{inc}$ such that $i'' > i'$ which is not t^* -canonical.

Consider now a direct first path P' which contains a triple $(i', o', f')_{t^*}^{inc}$. In the corresponding policy, truck $\mathbf{I}_{i'}$ transfers directly to every truck \mathbf{O}_o , $o' < o \leq o''$, $\nu_{i'o'ot^*}^{inc}(f')$ products of type t^* , where

$$\nu_{i'o'ot^*}^{inc}(f') = \max \left\{ 0, \min \left\{ b_{ot^*}, a_{i't^*} - f' - \sum_{j=o'+1}^{o-1} b_{jt^*} \right\} \right\}.$$

Therefore, path P contains a triple $(i', o'', \nu_{i'o'ot^*}^{inc}(f'))_{t^*}^{out}$, where $o'' > o'$. Again, an important observation is that there is at most one value o'' for which such triple is not t^* -canonical. We can say that any triple $(i', o', f')_{t^*}^{inc}$ “generates” at most one triple $(i', o'', f')_{t^*}^{out}$ such that $o'' > o'$ which is not t^* -canonical.

We now fix a value o^* . From the above analysis, it follows that every t^* -canonical triple $(i, o, f)_{t^*}^{inc}$ or $(i, o, f)_{t^*}^{out}$ such that $o \leq o^*$ “generates” at most one non- t^* -canonical triple $(i', o^*, f')_{t^*}^{inc}$ or $(i', o^*, f')_{t^*}^{out}$ such that $i' > i$. Therefore, the total number of different triples $(i, o^*, f)_{t^*}^{inc}$ or $(i, o^*, f)_{t^*}^{out}$ contained in paths in $\mathcal{P}^{inc}(i, o^*, f)$ and $\mathcal{P}^{out}(i, o^*, f)$ is $O(nm)$.

A straightforward analysis now indicates that the total number of states in the algorithm does not exceed $O((nm^2)^q)$. However, it is possible to get rid of the exponent q by the following reasoning.

We fix values i^* and o^* , and define a complete lexicographic order for states $\mathbf{S}^{out}(i^*, o^*, f)$. A state $\mathbf{S}^{out}(i^*, o^*, f')$ is lexicographically smaller than a state $\mathbf{S}^{out}(i^*, o^*, f'')$, $f' \neq f''$, if and only if $f'_t \leq f''_t$ for all $t \in T_o$. It is always possible to compare in this way two such states, as we cannot have $f'_{t'} < f''_{t'}$ and $f'_{t''} > f''_{t''}$ for any two types $t', t'' \in T_{o^*}$ by Lemma 1. Now, the total number of states $\mathbf{S}^{out}(i^*, o^*, f)$ is equal to the sum, for every type $t \in T_{o^*}$, of different triples $(i^*, o^*, f)_t^{out}$, as, in order to pass from a state $\mathbf{S}^{out}(i^*, o^*, f)$ to the lexicographically next state, at least one of the values f_t , $t \in T_{o^*}$ should be increased. Analogously, the total number of states $\mathbf{S}^{inc}(i^*, o^*, f)$ is equal to the sum, for every type $t \in T_{o^*}$, of different triples $(i^*, o^*, f)_t^{inc}$.

Consequently, for every fixed value o^* , the total number of states $\mathbf{S}^{out}(i, o^*, f)$ and $\mathbf{S}^{inc}(i, o^*, f)$ is equal to $O(qnm)$. From this, we conclude that the total number of states is $O(qnm^2)$. \square

We now return to the question of estimation of ρ , which is the number of operations needed to check whether a state $\mathbf{S}^{out}(i, o, f)$ or $\mathbf{S}^{inc}(i, o, f)$ has been already created. From the proof of Proposition 1, remember that, given fixed values i^* and o^* , the number of different states $\mathbf{S}^{out}(i^*, o^*, f)$ is not more than $O(qnm)$. As these states can be lexicographically ordered, the storage and the search of these states can be done using a binary tree. Therefore, to check whether a state $\mathbf{S}^{out}(i, o, f)$ or, analogously, a state $\mathbf{S}^{inc}(i, o, f)$, has been already created and retrieve the value $V^{out}(i, o, f)$ or $V^{inc}(i, o, f)$ we need $O(\log(qnm))$ operations. This concludes the estimation of the complexity of the algorithm.

Proposition 2 *The complexity of the dynamic programming algorithm is $O(qnm^2(n+m)(q+\log m+\log n))$.* \square

4 Conclusion

In this paper, we presented a polynomial dynamic programming algorithm for a scheduling problem with fixed sequences of incoming and outgoing trucks to

minimise the storage cost in cross docking terminal. This algorithm allows to determine the computational complexity of the problem for the first time.

Although polynomial, the complexity of the algorithm is high. However, a theoretical estimation of the number of created states can be much higher than what we have in reality. Therefore, we hope that the algorithm may be used in practice especially when coupled with some bounding procedure and when q is small. As an indirect evidence of this, we can mention good experimental results reported in [3] for an algorithm proposed for the case $q = 1$ of our problem.

One interesting direction for a future research is to try to find a linear programming formulation for the problem, since it was shown to be polynomially solvable. Such a formulation would help a lot in developing methods for solving more practical generalisation of the problem in which the truck sequences are not fixed.

References

- [1] U. M. Apte and S. Viswanathan. Effective cross docking for improving distribution efficiencies. *International Journal of Logistics Research and Applications*, 3:291–302, 2000.
- [2] N. Boysen, M. Fliender, and A. Scholl. Scheduling inbound and outbound trucks at cross docking terminals. *OR Spectrum*, page to appear, 2008.
- [3] M.Y. Maknoon, P. Baptiste, and O. Kone. Optimal loading and unloading policy in cross docking platform. In *Proceedings of 13th IFAC Symposium on Information Control Problems in Manufacturing*, pages 1263–1268, Moscow, Russia, June 2009.
- [4] W. Yu and P. J. Egbelu. Scheduling of inbound and outbound tracks in cross docking systems with temporary storage. *European Journal of Operations Research*, 184:377–396, 2008.

Appendix A NP-hardness proof

Here we consider the general problem in which the sequences of trucks are not fixed. We show the this problem is NP-hard in the strong sense even for the case in which each incoming truck supplies products of at most two types, each outgoing truck demands products of at most one type, all the storage costs and product volumes are unitary, and the storage capacity is unlimited. We will perform a reduction from the 3-partition problem.

Remember that, in the 3-partition problem, we are given an integer B and a set of $3n$ integers r_1, r_2, \dots, r_{3n} such that $\sum_{i=1}^{3n} r_i = Bn$ and $B/4 < r_i < B/2$ for each i . We need to decide whether there exists a partition of the set of indexes $\{1, 2, \dots, 3n\}$ into n sets $\{A_1, A_2, \dots, A_n\}$ such that $\sum_{i \in A_j} r_i = B$, $\forall j = 1, \dots, n$. Note that, if such a partition exists, each subset A_j contains exactly 3 indexes.

Given an instances of the 3-partition problem, we now define the corresponding instance of our cross docking problem. There are $3n$ incoming, $4n$ outgoing trucks ($3n$ of the first type, n of the second type) and two types of products. The supplies and demands are the following:

$$\begin{aligned} a_{i1} &= 1, & i &= 1, \dots, 3n, \\ a_{i2} &= 2n + r_i, & i &= 1, \dots, 3n, \\ b_{i1} &= 1, & i &= 1, \dots, 3n, \\ b_{i2} &= 0, & i &= 1, \dots, 3n, \\ b_{i1} &= 0, & i &= 3n + 1, \dots, 4n, \\ b_{i2} &= 6n + B, & i &= 3n + 1, \dots, 4n. \end{aligned}$$

We claim that there exists a 3-partition if and only if at most n products are transferred via the storage.

Suppose that there exists a 3-partition $\{A_1, A_2, \dots, A_n\}$, where $A_j = \{i_{j1}, i_{j2}, i_{j3}\}$. Then, the trucks are sequenced in n groups. Group j , $1 \leq j \leq n$, includes set A_j of incoming trucks and outgoing trucks \mathbf{O}_{2j-1} , \mathbf{O}_{2j} and \mathbf{O}_{3n+j} . The departure order of group j is the following

$$\mathbf{O}_{2j-1}, \mathbf{I}_{i_{j1}}, \mathbf{I}_{i_{j2}}, \mathbf{O}_{3n+j}, \mathbf{I}_{i_{j3}}, \mathbf{O}_{2j}.$$

As $r_{i_{j1}} + r_{i_{j2}} + r_{i_{j3}} = B$, the incoming trucks transfer $6n + B$ products of type 2 directly to truck \mathbf{O}_{3n+j} , $\mathbf{I}_{i_{j1}}$ transfers 1 product of type 1 directly to \mathbf{O}_{2j-1} , and $\mathbf{I}_{i_{j3}}$ transfers 1 product of type 1 directly to \mathbf{O}_{2j} . Only 1 product of type 1 is transferred to the storage from $\mathbf{I}_{i_{j2}}$. These transfers are depicted in Figure 3. As there are n groups, n products in total are transferred to the storage and then put to the outgoing trucks $\mathbf{O}_{2n+1}, \dots, \mathbf{O}_{3n}$, which are sequenced at the end.

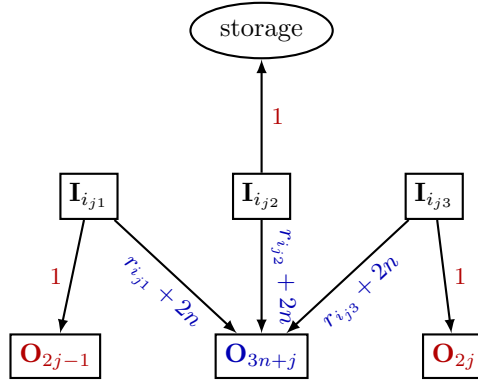


Figure 3: Product transfers within a group j of trucks

Suppose now there is a sequence of trucks such that at most n products are transferred through the storage. Then, every outgoing truck \mathbf{O}_i , $i = 3n + 1, \dots, 4n$ must receive products directly from at least 3 incoming trucks.

Otherwise it would receive from the storage at least $2n$ products of type 2. Therefore, at least 1 product of type 1 goes to the storage while each such truck \mathbf{O}_i is supplied, and the total number of products of type 1 transferred through the storage is at least n , meaning that all products of type 2 should be transferred directly. Then, each incoming truck can transfer directly products of type 2 to exactly one outgoing truck. Otherwise, between two outgoing trucks of type 2, only one outgoing truck can receive a product of type 1 directly, and the total number of products of type 1 transferred through the storage would exceed n . We conclude that there should exist a partition of incoming trucks into triples $\{A_1, A_2, \dots, A_n\}$ such that $\sum_{i \in A_j} r_i = B$. Otherwise there would exist a triple A_j such that $r_{i_{j1}} + r_{i_{j2}} + r_{i_{j3}} < B$, and the outgoing truck which is supplied by the incoming trucks in A_j would need to take at least one product of type 2 from the storage.