## Steel stacking

A problem in inventory management in the steel industry

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Part of the presentation concerns joint work with Rui Rei and Mikio Kubo.

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## Informal problem description

## Context

A steel producer has a warehouse where the final product is stocked

- large steel bars enter the warehouse when production finishes
- bars leave the warehouse on trucks or ships for transporting them to the final customer
- there is a crane in the warehouse, which moves the bars one at a time
- the warehouse has $p$ different places
- each place can be empty, or keep a stack of steel bars


## Informal problem description

## Assumptions

- capacity of the stacks is infinite
- no delays on crane movements
- crane can move only one item at a time
- only the item on the top of the stack can be moved
- item on top of each stack may have to be relocated (reshuffling)


## Objective

- minimize the number of movements made by the crane


## Steel stacking



## Steel stacking

## Data

$$
\begin{cases}p \in \mathbb{N} & \text { number of stacks on the warehouse } \\ n \in \mathbb{N} & \text { number of items } \\ R_{i} \in \mathbb{N}, i=1, \ldots, n & \text { release dates } \\ D_{i} \in \mathbb{N}, i=1, \ldots, n & \text { delivery dates }\end{cases}
$$

## Steel stacking

## Constraints

- crane can move only the item on top of the stack
- release and delivery dates must be satisfied
- the valid movements depend on $R, D$, and on the choices made up to the moment.


## Solution representation

List of movements from a stack (o) to another (d)
$M=\left[\left(o_{1}, d_{1}\right), \ldots,\left(o_{k}, d_{k}\right)\right]$

- $0 \leq o_{i} \leq p$ and $\left.1 \leq d_{i} \leq p+1\right\}$
- stack 0 represents the production facility
- stack $p+1$ represents the customer track/ship.
- we want to minimize the number of movements (size of $M$ )


## Example

| ${ }^{A} 16$ |
| :--- |
| ${ }^{B} 10$ |
| ${ }^{C} 12$ |
| ${ }^{C} 14$ |

## stack 0 stack 1 stack 2 stack 3 (production) <br> (client)

Movements:
[]

## Example - step 1



## Movements:

[]

## Example - step 1



Movements:
[(0, 1)]

## Example - step 2



Movements:
[ $(0,1)$ ]

## Example - step 2



Movements:
$[(0,1),(0,1)]$

## Example - step 3



Movements:
[(0, 1), (0, 1)]

## Example - step 3

| ${ }^{D} 14$ | ${ }^{B} 10$ |  |  |
| :---: | :---: | :---: | :---: |
|  | ${ }^{\text {A } 16}$ | ${ }^{\text {c }} 12$ |  |
| stack 0 (production) | stack 1 | stack 2 | stack 3 <br> (client) |

Movements:
[(0, 1), (0, 1), (0, 2)]

## Example - step 4



Movements:
[(0, 1), (0, 1), (0, 2)]

## Example - step 4



Movements:
$[(0,1),(0,1),(0,2),(0,2)]$

## Example - step 5



Movements:
[(0, 1), (0, 1), (0, 2), (0, 2)]

## Example - step 5



Movements:
$[(0,1),(0,1),(0,2),(0,2),(1,3)]$

## Example - step 6



Movements:
$[(0,1),(0,1),(0,2),(0,2),(1,3)]$

## Example - step 6



Movements:
$[(0,1),(0,1),(0,2),(0,2),(1,3),(2,1)]$

## Example - step 7



## Movements:

$[(0,1),(0,1),(0,2),(0,2),(1,3),(2,1), \rightarrow(2,3),(1,3),(1,3)]$
This information is complemented with release and due dates.

## Solution representation: MIP

If we want to solve the problem with standard optimization tools: MIP formulation:

## Sets

$T \in \mathbb{N}$ - time horizon (the number of periods in the model)
$N \in \mathbb{N}$ - number of items
$W \in \mathbb{N}$ - the number of stacks in the warehouse (warehouse width).
$H \in \mathbb{N}$ - the maximum number of items that can be in a stack at any given instant (warehouse height).
$R \in \mathbb{R}^{N}$ - item release dates ( $R_{i}$ denotes the release date of item i).
$D \in \mathbb{R}^{N}$ - item due dates ( $D_{i}$ denotes the due date of item $i$ ).

## Solution representation: MIP

Problem: number of periods that have to be considered


Worst case: $T=2 N+\sum_{n=1}^{N-1} n$

## Solution representation: MIP

## MIP formulation

## Variables

$x_{i j n t}-1$ if item $n$ is released into position $(i, j)$ at period $t$ $y_{i j k l n t}-1$ if item $n$ is relocated from position $(i, j)$ into $(k, l)$ at period $t$
$z_{i j n t}-1$ if item $n$ is delivered from position $(i, j)$ at period $t$ $a_{n t}-1$ if item $n$ has not entered the warehouse yet at period $t$
$b_{i j n t}-1$ if item $n$ is in row $j$ of stack $i$ at period $t$ $c_{n t}-1$ if item $n$ has already left the warehouse at period $t$

## Solution representation: branch-and-bound

## Branch-and-bound

- When an item is released/relocated:
- Check all stacks where it can be placed
- Create a branch for each of them
- When an item is delivered from top: move without branching.


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## MIP

- formulate the problem, create model
- read an instance
- send it to a solver


## Branch-and-bound method

(1) create empty queue $Q$
(2) push root node into $Q$
(3) while $Q$ is not empty
(4) $\quad \mu \leftarrow$ pop a node from $Q$
(5) $\quad i \leftarrow$ item to be placed next on node $\mu$
(6) foreach stack $s$ where $i$ can be placed
(7) $\quad \mu^{\prime} \leftarrow$ a copy of node $\mu$
(8) $\quad$ place item $i$ in stack $s$ on node $\mu^{\prime}$
(9)
(10)
execute deliveries from top of stacks in $\mu^{\prime}$
if $\mu^{\prime}$ is a leaf node
check if $\mu^{\prime}$ contains a better solution
push node $\mu^{\prime}$ into $Q$

## Discrete event simulation

- type of simulation used on systems where the state variations are discrete
- computationally "inexpensive"


## In our case:

- each simulation run involves some randomness: stack for each item is selected randomly from list of candidates
- different runs lead to different solutions
- at the end, choose the best run (a large number of runs may be required for obtaining good results)
- several stacking strategies (heuristics) can be used and tested


## Simulation strategy:

Heuristics for choosing a stack for an item are applied
(1) on arrival of an item
(2) when moving an item from the top, to reach another below it (reshuffling).

## Processing order

- delivery has precedence over stacking released items
- "simultaneous" releases are processed by inverse delivery date
- "simultaneous" deliveries are processed from top to bottom of the stack


## "Simultaneous" deliveries:



No specified order on simultaneous deliveries:
(1) $\{C, D\}: s_{2} \rightarrow s_{1}$
(2) $E: s_{2} \rightarrow$ client
(3) $\{D, C, A\}: s_{1} \rightarrow s_{2}$
(9) $B: s_{1} \rightarrow$ client
$\hookrightarrow 7$ movements

Deliveries processed by item depth:
(1) $A: s_{1} \rightarrow s_{2}$
(2) $B: s_{1} \rightarrow$ client
(3) $\{A, C, D\}: s_{2} \rightarrow s_{1}$
(4) $E: s_{2} \rightarrow$ client
$\hookrightarrow 6$ movements

## Positioning heuristics

## When placing an item

- each position (i.e., each stack) is assigned a value, according to some rules
- construct a candidate list of positions with best classification (depends on the heuristics used)
- from this list, randomly choose one for placing the item


## Heuristics: Optimize flexibility

- we define flexibility of a position as the maximum number of items with different delivery dates that can be stacked, without causing an inversion
- candidate stacks for an item are those which:
- maximize flexibility and cause no inversion
- minimize flexibility, if an inversion is unavoidable



## Simulation-based optimization

- For this stacking heuristics:
- do $N$ independent simulation runs
- choose the run that lead to less crane movements
- For each of the $N$ runs:
- for each item, choose a stack according to the selected heuristics
- continue processing item releases and deliveries, until having all items delivered
- return the number of movements required, and the corresponding movement list


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## Problem variants

There are many variants of the problem with practical interest:
(1) stacks are replaced by more general structures, such as double-ended queues; this allows modeling e.g. cases where items are placed in a horizontal configuration, with access from two ends;
(2) crane movements are limited, and thus the number of stacks that can be reached from a given position is limited;
(3) placement in certain stacks limits crane movements, i.e., the graph of connections is dynamically changed when items are placed;
(9) crane movements induce significant delays, making the assumption of immediate delivery unviable;
(3) other objectives.

## More general structures

## E.g., double ended queues



## Limited movements



## Dynamics



Problem description

## Dynamics: allowed movements change 1



Problem description

## Dynamics: allowed movements change 2



Problem description

## Dynamics: allowed movements change 3



Problem description

## Dynamics: times change



## Other objectives

Another interesting variant:
(1) minimize time of service of track/ship;
(2) items are rearranged before track of ship arrive;
(3) aim: having the minimum number of movements for loading the track/ship;
(4) in some variants: more than one item may be moved a the same time.

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## Conclusions

- Problem tackled: difficult (decision in one step may have consequences much later).
- For tackling it:
- MIP model;
- Specialized branch-and-bound;
- Simulation-based optimization.
- Problem has many interesting variants.
- Dynamics: configuration may change upon decisions taken.
- Modeling issues: language for formalizing problem description.
- New solution approaches: finite state automata.

