# Yard traffic and congestion in container terminals 

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## Outline

- Introduction and motivation
- Modeling
- Congestion measures
- Optimization
- Computational results
- Future work



## Container Terminals (CT)

- Zone in a port to import/export/transship containers
- Different areas in a terminal: berths, yard, gates
- Different types of vehicles to travel between the yard and the berth
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## Motivation

- Along the quay, containers are loaded/unloaded onto/from several boats
- Containers' transfer lead to a high traffic in the yard zone
- The berth\&yard allocation plan assigns ships to berths and containers to yard blocks
- Terminal planners usually minimize the total distance travelled by the carriers, disregarding:
- Congestion issues (operations slowdowns because of bottlenecks)
- Alternative solutions (symmetries)


## Aim of this study:

$\checkmark$ Model the terminal and develop measures of congestion
$\checkmark$ Evaluate the impact of the optimization of such measures on the terminal

## Assumptions

- We take into account flows of containers from the quayside to the yard
- Given a berth\&yard allocation plan, we define a path as an OD pair:
- origin (berth)
- destination (block)
- number of containers
- We consider flows of containers over a working shift
- Decisions could be taken on:
- the berth allocation plan (berths and ships)
- the yard allocation plan (destination blocks)
- demand splitting over blocks
$\rightarrow$ In this study: given a set of $p$ paths, determine the destination blocks fedirall di laushnne


## Literature

- Layout:
- Kim et al. An optimal layout of container yards, OR Spectrum, 2007.
- Congestion:
- Lee et al. An optimization model for storage yard management in transshipment hubs, OR Spectrum, 2006.
- Beamon. System reliability and congestion in a material handling system, Computers Industrial Engineering, 1999.


## Modeling the terminal

Basic element




## Modeling the terminal

- ( $m \times n$ ) basic elements of 2 blocks each compose the yard
- coordinates system for OD pairs $\left(x_{o}, y_{o}\right)-\left(x_{d}, y_{d}\right)$
- only berth-to-yard and yard-toberth paths are considered

yard
berth


## Routing rules

- Horizontal lanes are one way
- Vertical lanes are two way
- Toward the block, closest left vertical lane, turn right.
- Toward the quay, turn right at the first vertical lane.
- Back to origin berth position.
- Distance travelled, closed formula (Manhattan)

yard


## Symmetries

## Minimize distance:

in a $2 \times 2$ yard with 2 paths, no capacity on blocks


## Congestion measures

- Aim of the study:
- estimate the state/congestion of a yard when implementing a plan
- provide simple closed formulas, to be used as secondary objectives
- Factors taken into account:
- interference between blocks sharing the same lane
- lane congestion
- interference between paths ECOLE POLYTRCHNIQU


## 1. Block congestion

- congestion among blocks sharing the same lane
- "area": blocks with the same entrance node
- \# of areas: $s=2 n+n(m-1)$
- $c_{j}$ : \# of containers on path $j=1$...p
- $\boldsymbol{N}_{i}$ : \# of containers allocated to area $i$
- $\boldsymbol{N}^{*}$ : \# of containers in each area in the optimal solution (even distribution among areas)

$$
C_{b}=\frac{D}{D_{\max }}=\frac{\sum_{i=1}^{s}\left|N_{i}-N^{*}\right|}{\frac{2(s-1)}{s} \sum_{j=1}^{p} c_{j}}
$$

- 1-norm and 2-norm w.r.t. the best over the worst case



## 1. Block congestion

- 3 paths in a $2 \times 3$ yard ( 12 blocks) $\rightarrow$ possible solutions : $12^{3}=1728$
- number of solutions with same block congestion (distribution of 2-norm $\mathrm{C}_{\mathrm{b}}$ ) :
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## 2. Edge congestion

- this indicator simply measures the average traffic over an edge

$$
\begin{array}{lll}
\theta=\max _{k} f_{k} \\
\mu=\min _{k} f_{k} & \theta_{\max }=\sum_{j=1}^{p} c_{j} & \mu_{\min }=0
\end{array}
$$

$$
C_{e}=\frac{\theta-\mu}{\theta_{\max }-\mu_{\min }}=\frac{\theta-\mu}{\sum_{j=1}^{p} c_{j}}
$$

- the best traffic situation is when flows are spread over the network: $\mu^{*}=\frac{\sum_{j=1}^{p} c_{j}}{n}$

$$
C_{e}=\frac{\theta-\mu^{*}}{\theta_{\max }-\mu^{*}}=\frac{\theta-\mu^{*}}{\sum_{j=1}^{p} c_{j}-\frac{\sum_{j=1}^{p} c_{j}}{n}}=\frac{(n) \theta-\sum_{j=1}^{p} c_{j}}{(n-1) \sum_{j=1}^{p} c_{j}}
$$

## 2. Edge congestion

- 3 paths in a $2 \times 3$ yard ( 12 blocks) $\rightarrow$ possible solutions : $12^{3}=1728$
- number of solutions with same edge congestion (distribution of improved $\mathrm{C}_{\mathrm{e}}$ ):



## 3. Path congestion

- interference among "crossing" paths
- proximity matrix $\boldsymbol{P}(2 p \times 2 p)$
- $p$ berth-to-yard $+p$ yard-to-berth paths
- $P$ is symmetric, 0 on the diagonal, 1 if two paths are "neighbours"
- definition of $P$ is influenced by routing rules
- worst case: all 1 matrix (except diagonal)

$$
C_{p}=\frac{p}{N_{\max }}=\frac{1^{T} . P . c}{(2 n-1) \sum_{i=1}^{2 n} c_{i}}
$$



## Example

- 3 paths in a $2 \times 3$ yard
- Distribution of the objective function $z=\lambda_{b} \cdot C_{b}+\lambda_{e} \cdot C_{e}+\lambda_{p} \cdot C_{p}$




## Example

Objective function: $z=\lambda_{b} \cdot C_{b}+\lambda_{e} \cdot C_{e}+\lambda_{p} \cdot C_{p}$

|  | Nb solutions | Nb different values | MIN | Nb MIN | CPU (s) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| (2x2) - 3 paths | 512 | 46 | 0,4764 | 10 | 0,2 |
| (2x2)-4 paths | 4096 | 282 | 0,3473 | 30 | 1,4 |
| (2x2) - 5 paths | 32768 | 1831 | 0,5068 | 21 | 12,23 |
| (2x2) - 6 paths | 262144 | 7354 | 0,461 | 12 | 112,85 |
| (2x3)-3 paths | 1728 | 52 | 0,4764 | 116 | 0,67 |
| (2x3)-4 paths | 20736 | 470 | 0,3473 | 350 | 7,29 |
| (2x3)-5 paths | 248832 | 4271 | 0,13 | 108 | 121,65 |

## Optimization algorithm: GRASP

- GRASP: Greedy Randomized Adaptive Search Procedure
- Objective: assign a destination to each path such that congestion is minimized
- The algorithm builds a solution iteratively:
- at each step, the destination for one specific path is chosen


## Optimization algorithm: GRASP

|  | MIN | CPU (s) <br> (enumeration) | CPU (s) <br> (algorithm) | Nb iteration to <br> reach optimum |
| :--- | :---: | :---: | :---: | :---: |
| (2x2) - 3 paths | 0,4764 | 0,2 | 0,1 | 5 |
| (2x2) - 4 paths | 0,3473 | 1,4 | 0,2 | 10 |
| (2x2) - 5 paths | 0,5068 | 12,23 | 0,5 | 30 |
| (2x2) - 6 paths | 0,461 | 112,85 | 3 | 150 |
|  |  | 0,67 | 0,1 | 5 |
| (2x3) - 3 paths | 0,4764 | 7,29 | 0,1 | 5 |
| (2x3) - 4 paths | 0,3473 | 121,65 | 0,5 | 25 |
| $\mathbf{( 2 \times 3 ) - 5}$ paths | 0,13 | $\mathbf{? ?}$ |  | 15 |
| (2x3) - 6 paths | 0,1953 |  |  | 1000 |

## Computational tests

## More realistic instances

|  | in 0,1s | in 1s | in 5s | in 10s | in 20s | in 60s |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 3 }}$ | 0,4764 | 0,4764 | 0,4764 | 0,4764 | 0,4764 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 4 }}$ | 0,3473 | 0,3473 | 0,3473 | 0,3473 | 0,3473 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 5 }}$ | 0,13 | 0,13 | 0,13 | 0,13 | 0,13 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 6 }}$ | 0,389 | 0,195 | 0,195 | 0,195 | 0,195 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 7 }}$ | 0,343 | 0,267 | 0,267 | 0,267 | 0,267 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 8 }}$ | 0,26 | 0,1692 | 0,1646 | 0,1646 | 0,1646 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 9 }}$ | 0,304 | 0,2763 | 0,2763 | 0,2763 | 0,2763 |  |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 1 5 }}$ | 0,2446 | 0,1931 | 0,1705 | 0,1582 | 0,1817 | 0,1602 |
| $\mathbf{( 3 \times 1 0 ) - \mathbf { 2 0 }}$ | 0,3275 | 0,2276 | 0,1663 | 0,1624 | 0,1609 | 0,1389 |

## Conclusions and Outlook

- simple closed formulas to evaluate congestion in container terminals
- useful to differentiate symmetric solutions with equal distance

Ongoing work:

- validation of our approach via a CT simulator
- multi-objective optimization problem (explore other than weighted sum)
- improve the algorithm: study an exact approach; relax the assumptions, i.e. extend the set of possible decisions (berth allocation, demand splitting)


## Thanks for your attention!

