

# The Tactical Berth Allocation Problem with QC Assignment and Transshipment Costs

## *Models and Heuristics*

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# Outline

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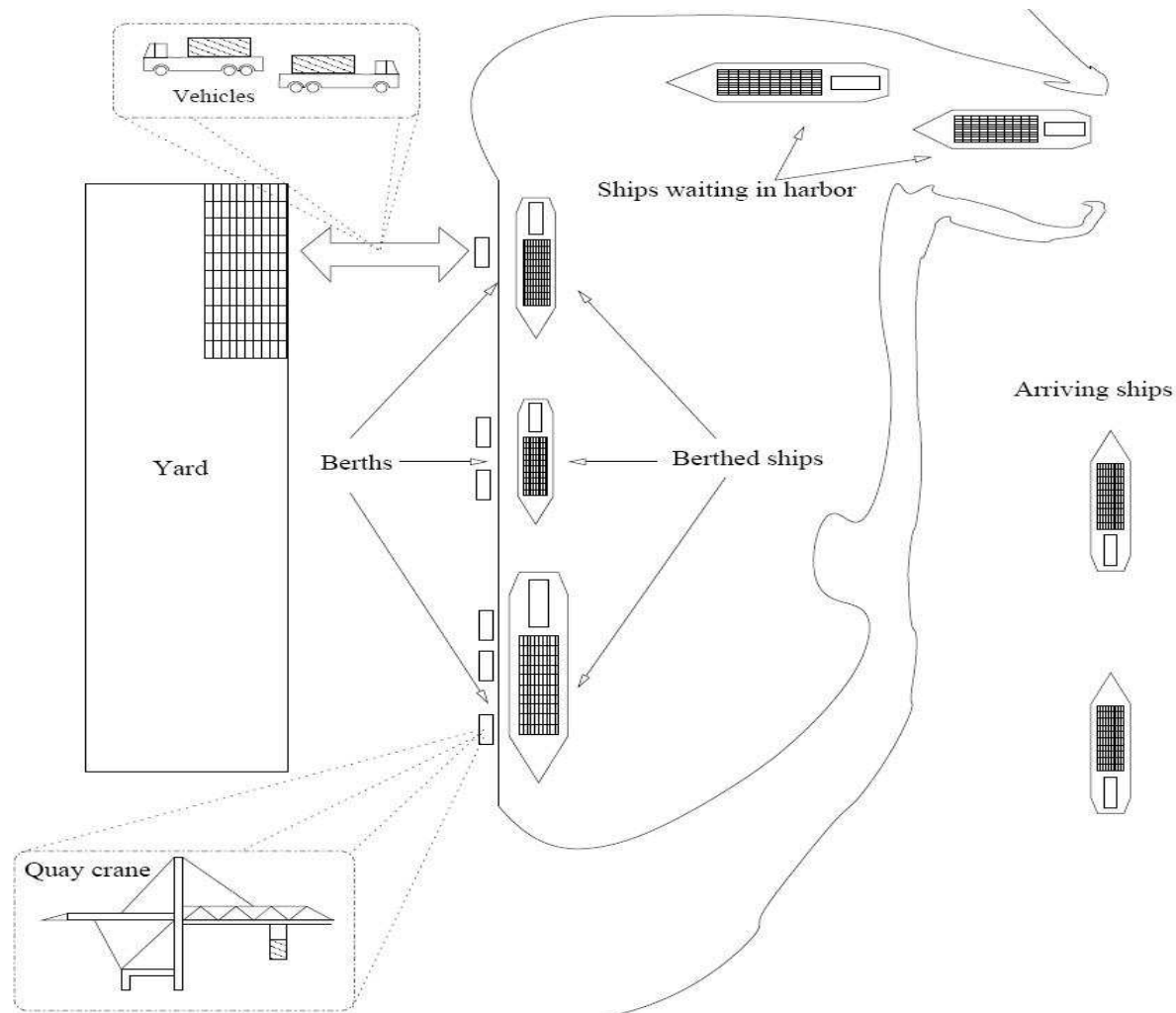
- Container terminals
- Tactical Berth Allocation Problem (TBAP) with Quay Crane Assignment
- MILP and MIQP models
- Heuristics for TBAP: Tabu Search & Math Programming
- Computational results
- Conclusions

# Context: container terminals





# Container terminal operations



# Tactical Berth Allocation with QCs Assignment

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Giallombardo, Moccia, Salani and Vacca (2009)

## Problem description

- *Tactical Berth Allocation Problem (TBAP)*: assignment and scheduling of ships to berths, according to time windows for both berths and ships; tactical decision level, w.r.t. negotiation between terminal and shipping lines;
- *Quay-Cranes Assignment Problem (QCAP)*: a quay crane (QC) profile (number of cranes per shift, ex. 332) is assigned to each ship;
- *Housekeeping Quadratic Yard Costs*: take into account the exchange of containers between ships, in the context of transshipment container terminals.

# The concept of QC assignment profile

TIME	ws=1	ws=2	ws=3	ws=4	ws=5	ws=6	ws=7	ws=8
berth 1	ship 1				ship 2			
	3	2	2		4	4	5	5
berth 2		ship 3				ship 4		
		4	5			3	3	3
berth 3			ship 5					
			3	3	3	2	2	
QCs	3	6	10	3	7	9	10	8



# Tactical Berth Allocation with QCs Assignment

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## Issues

- the chosen profile determines the ship's handling time and thus impacts on the scheduling;
- feasible profiles can vary in length (number of shifts dedicated to the ship) and in size (number of QCs dedicated to the ship in each active shift).



# Tactical Berth Allocation with QCs Assignment

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## Find

- a berth allocation
- a schedule
- a quay crane assignment

## Given

- time windows on availability of berths
- time windows on arrival of ships
- *handling times dependent on QC profiles*
- values of QC profiles

## Aiming to

- maximize total value of QC assignment
- minimize housekeeping costs of transshipment flows between ships

# TBAP with QCs assignment: the model

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- $N$  = set of **vessels**;
- $M$  = set of **berths**;
- $H$  = set of **time steps** (each time step  $h \in H$  is submultiple of the work shift length);
- $S$  = set of the time step indexes  $\{1, \dots, \bar{s}\}$  relative to a work shift; ( $\bar{s}$  represents the number of time steps in a work shift);
- $H^s$  = subset of  $H$  which contains all the time steps corresponding to the same time step  $s \in S$  within a work shift;
- $P_i^s$  = set of feasible QC assignment profiles for the vessel  $i \in N$  when vessel arrives at a time step with index  $s \in S$  within a work shift;
- $P_i$  = set of **quay crane assignment profiles** for the vessel  $i \in N$ , where  $P_i = \cup_{s \in S} P_i^s$ ;

# TBAP with QCs assignment: the model

- $t_i^p$  = **handling time** of ship  $i \in N$  under the QC profile  $p \in P_i$  expressed as multiple of the time step length;
- $v_i^p$  = the **value** of serving the ship  $i \in N$  by the quay crane profile  $p \in P_i$ ;
- $q_i^{pu}$  = number of **quay cranes** assigned to the vessel  $i \in N$  under the profile  $p \in P_i$  at the time step  $u \in (1, \dots, t_i^p)$ , where  $u = 1$  corresponds to the ship arrival time;
- $Q^h$  = maximum number of quay cranes available at the time step  $h \in H$ ;
- $f_{ij}$  = **flow of containers** exchanged between vessels  $i, j \in N$ ;
- $d_{kw}$  = **unit housekeeping cost** between yard slots corresponding to berths  $k, w \in M$ ;
- $[a_i, b_i]$  = [earliest, latest] feasible arrival time of ship  $i \in N$ ;
- $[a^k, b^k]$  = [start, end] of availability time of berth  $k \in M$ ;
- $[a^h, b^h]$  = [start, end] of the time step  $h \in H$ .

# TBAP with QCs assignment: the model

Consider a graph  $G^k = (V^k, A^k) \forall k \in M$ , where  $V^k = N \cup \{o(k), d(k)\}$ , with  $o(k)$  and  $d(k)$  additional vertices representing berth  $k$ , and  $A^k \subseteq V^k \times V^k$ .

## Decision variables:

- $x_{ij}^k \in \{0, 1\} \forall k \in M, \forall (i, j) \in A^k$ , 1 if ship  $j$  is scheduled after ship  $i$  at berth  $k$ ;
- $y_i^k \in \{0, 1\} \forall k \in M, \forall i \in N$ , set to 1 if ship  $i$  is assigned to berth  $k$ ;
- $\lambda_i^p \in \{0, 1\} \forall p \in P_i, \forall i \in N$ , set to 1 if ship  $i$  is served by the profile  $p$ ;
- $T_i^k \geq 0 \forall k \in M, \forall i \in N$ , representing the berthing time of ship  $i$  at the berth  $k$  i.e. the time when the ship moors.

## Linking variables:

- $\gamma_i^h \in \{0, 1\} \forall h \in H, \forall i \in N$ , set to 1 if ship  $i$  arrives at time step  $h$ ;
- $\rho_i^{ph} \in \{0, 1\} \forall p \in P_i, \forall h \in H, \forall i \in N$ , set to 1 if ship  $i$  is served by profile  $p$  and arrives at time step  $h$ ;
- $T_{o(k)}^k, T_{d(k)}^k \geq 0 \forall k \in M$ , representing the starting and ending operation time of berth  $k$  respectively.

# TBAP with QCs assignment: the MIQP model

## Objective function

Maximize total value of QC profile assignments + Minimize the (quadratic) housekeeping yard cost of transshipment flows between ships:

$$\max \sum_{i \in N} \sum_{p \in P_i} v_i^p \lambda_i^p - \frac{1}{2} \sum_{i \in N} \sum_{k \in M} y_i^k \sum_{j \in N} \sum_{w \in M} f_{ij} d_{kw} y_j^w \quad (1)$$

# TBAP with QCs assignment: the MIQP model

## Berth covering constraints

$$\sum_{k \in M} y_i^k = 1 \quad \forall i \in N, \quad (2)$$

## Flow and linking constraints

$$\sum_{j \in NU\{d(k)\}} x_{o(k),j}^k = 1 \quad \forall k \in M, \quad (3)$$

$$\sum_{i \in NU\{o(k)\}} x_{i,d(k)}^k = 1 \quad \forall k \in M, \quad (4)$$

$$\sum_{j \in NU\{d(k)\}} x_{ij}^k - \sum_{j \in NU\{o(k)\}} x_{ji}^k = 0 \quad \forall k \in M, \forall i \in N, \quad (5)$$

$$\sum_{j \in NU\{d(k)\}} x_{ij}^k = y_i^k \quad \forall k \in M, \forall i \in N, \quad (6)$$



# TBAP with QCs assignment: the MIQP model

## Precedence constraints

$$T_i^k + \sum_{p \in P_i} t_i^p \lambda_i^p - T_j^k \leq (1 - x_{ij}^k)M \quad \forall k \in M, \forall i \in N, \forall j \in N \cup d(k) \quad (7)$$

$$T_{o(k)}^k - T_j^k \leq (1 - x_{o(k),j}^k)M \quad \forall k \in M, \forall j \in N, \quad (8)$$

## Ship and Berth time windows

$$a_i y_i^k \leq T_i^k \quad \forall k \in M, \forall i \in N, \quad (9)$$

$$T_i^k \leq b_i y_i^k \quad \forall k \in M, \forall i \in N, \quad (10)$$

$$a^k \leq T_{o(k)}^k \quad \forall k \in M, \quad (11)$$

$$T_{d(k)}^k \leq b^k \quad \forall k \in M, \quad (12)$$

# TBAP with QCs assignment: the MIQP model

## Profile covering & linking constraints

$$\sum_{p \in P_i} \lambda_i^p = 1 \quad \forall i \in N, \quad (13)$$

$$\sum_{h \in H^s} \gamma_i^h = \sum_{p \in P_i^s} \lambda_i^p \quad \forall i \in N, \forall s \in S, \quad (14)$$

$$\sum_{k \in M} T_i^k - b^h \leq (1 - \gamma_i^h)M \quad \forall h \in H, \forall i \in N, \quad (15)$$

$$a^h - \sum_{k \in M} T_i^k \leq (1 - \gamma_i^h)M \quad \forall h \in H, \forall i \in N, \quad (16)$$

$$\rho_i^{ph} \geq \lambda_i^p + \gamma_i^h - 1 \quad \forall h \in H, \forall i \in N, \forall p \in P_i, \quad (17)$$

## Quay crane and profile feasibility

$$\sum_{i \in N} \sum_{p \in P_i} \sum_{u = \max\{h - t_i^p + 1; 1\}}^h \rho_i^{pu} q_i^{p(h-u+1)} \leq Q^h \quad \forall h \in H^{\bar{s}} \quad (18)$$

# TBAP with QCs assignment: the MILP model

## Additional decision variable

$z_{ij}^{kw} \in \{0, 1\} \forall i, j \in N, \forall k, w \in M$ , set to 1 if  $y_i^k = y_j^w = 1$  and 0 otherwise.

## Linearized objective function

$$\max \sum_{i \in N} \sum_{p \in P_i} v_i^p \lambda_i^p - \frac{1}{2} \sum_{i \in N} \sum_{j \in N} \sum_{k \in M} \sum_{w \in M} f_{ij} d_{kw} z_{ij}^{kw} \quad (19)$$

## Additional constraints

$$\sum_{k \in K} \sum_{w \in K} z_{ij}^{kw} = g_{ij} \quad \forall i, j \in N, \quad (20)$$

$$z_{ij}^{kw} \leq y_i^k \quad \forall i, j \in N, \forall k, w \in M \quad (21)$$

$$z_{ij}^{kw} \leq y_j^w \quad \forall i, j \in N, \forall k, w \in M \quad (22)$$

# Generation of test instances

- Based on real data provided by MCT, Port of Gioia Tauro, Italy:
  - container flows
  - housekeeping yard costs
  - vessel's arrival times
- Crane productivity of 24 containers per hours
- Set of feasible profiles synthetically generated, according to ranges given by practitioners:

Class	min QC	max QC	min HT	max HT	volume (min,max)
<i>Mother</i>	3	5	3	6	(1296, 4320)
<i>Feeder</i>	1	3	2	4	(288, 1728)

# Generation of test instances

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- 6 classes of instances:
  - 10 ships and 3 berths, 1 week, 8 quay cranes;
  - 20 ships and 5 berths, 1 week, 13 quay cranes;
  - 30 ships and 5 berths, 1 week, 13 quay cranes;
  - 40 ships and 5 berths, 2 weeks, 13 quay cranes;
  - 50 ships and 8 berths, 2 weeks, 13 quay cranes;
  - 60 ships and 13 berths, 2 weeks, 13 quay cranes.
- 12 scenarios for each class, with high (H) and low (L) traffic volumes;
- each scenario is tested with a set of  $\bar{p} = 10, 20, 30$  feasible profiles for each ship;
- CPLEX 10.2 solver for MILP and MIQP formulations.

# CPLEX results

10x3			10x3		
Instance	MILP	MIQP	Instance	MILP	MIQP
H1_10	99.17	98.90	L1_10	97.68	100.00
H1_20	97.91	97.96	L1_20	100.00	99.76
H1_30	97.98	98.76	L1_30	98.64	99.99
H2_10	98.87	99.26	L2_10	98.82	99.63
H2_20	96.97	96.91	L2_20	99.42	99.06
H2_30	96.79	-	L2_30	99.08	100.00

20x5			40x5		
Instance	MILP	MIQP	Instance	MILP	MIQP
H1_10	94.33	-	L1_10	94.92	-
H1_20	93.74	-	L1_20	94.47	-
H2_10	93.52	96.66	L2_20	94.93	-
L2_10	93.87	96.74	L2_30	94.61	-



# CPLEX results

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- Time limits:
  - 1 hour for class 10x3;
  - 2 hours for classes 20x5 and 30x5;
  - 3 hours for classes 40x5, 50x8 and 60x13.
- The objective function value is scaled to 100 with respect to the upper bound:

$$\text{scaled obj} = \frac{\text{obj} * 100}{\text{UB}}$$

A value of 100 means that the solution is certified to be optimal.

- No feasible solution was found for classes 30x5, 50x8 and 60x13;
- However, an upper bound is always provided (although very bad).

# A New Heuristics for TBAP

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- Our heuristic algorithm is organized in two stages:
  1. identify a QC profile assignment for the ships;
  2. solve the resulting berth allocation problem for the given QC assignment.
- The procedure is iterated over several sets of QC profiles;
- New profiles are chosen via reduced costs arguments (MILP formulation).

# A New Heuristics for TBAP

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## Algorithm 1: TBAP Bi-level Heuristics

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**Initialization** : Assign a QC profile to each ship

**repeat**

- 1. solve BAP
- 2. update profiles

**until** *stop criterion* ;

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### TBAP Bi-level Heuristics:

1. BAP solution via Tabu Search
2. Profiles' updating via Math Programming

# 1. Tabu Search for BAP

Adapted from [Cordeau, Laporte, Legato and Moccia \(2005\)](#).

- New objective function: minimization of yard-related transshipment quadratic costs
- New constraints: QCs availability
- Each solution  $s \in S$  is represented by a set of  $m$  berth sequences such that every ship belongs to exactly one sequence.
- Penalized cost function:

$$f(s) = c(s) + \alpha_1 w_1(s) + \alpha_2 w_2(s) + \alpha_3 w_3(s)$$

where  $w_1(s)$  is the total violation of ships' TWs,  $w_2(s)$  is the total violation of berths' TWs and  $w_3(s)$  is the total violation of QCs availability.

- “Move”: ship  $i$  is removed from sequence  $k$  and inserted in sequence  $k' \neq k$ . The new position in  $k'$  is such that  $f(s)$  is minimized.
- Initial solution: randomly built assigning ships to berths and relaxing the QCs availability constraint.

## 2. Profiles' Updating via Math Programming

Basic idea: use information of reduced costs to update the vector of assigned QC profiles in a “smart” way.

- Let  $\bar{X} = [\bar{x}, \bar{y}, \bar{T}]$  be the BAP solution found by the Tabu Search for a given QC profile assignment  $\bar{\lambda}$ .
- We solve the linear relaxation of the TBAP MILP formulation, with the additional constraints:

$$\bar{X} - \epsilon \leq X \leq \bar{X} + \epsilon \quad (23)$$

$$\bar{\lambda} - \epsilon \leq \lambda \leq \bar{\lambda} + \epsilon \quad (24)$$

- As suggested by [Desrosiers and Lübbecke \(2005\)](#), the shadow prices of these constraints are the reduced costs of original variables  $X$  and  $\lambda$ .
- We identify the  $\lambda_i^p$  variable with the maximum reduced cost and we assign this new profile  $p$  to ship  $i$ .
- If all reduced costs are  $\leq 0$ , we stop.

# Computational results

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- The heuristic has been implemented in C++ using GLPK 4.31.
- Stopping criteria:
  - $n \times \bar{p}$  iterations;
  - time limit of 1 hour for classes 10x3, 20x5 and 30x5;
  - time limit of 2 hours for classes 40x5, 50x8 and 60x13.
- Results are compared to the best solution found by CPLEX for either the MILP or MIQP formulation.



# Computational results

10x3				20x5			
Instance	CPLEX	HEUR	Time (sec)	Instance	CPLEX	HEUR	Time (sec)
H1_10	99.17	98.52	7	H1_10	-	97.26	81
H1_20	97.96	98.36	15	H1_20	94.33	97.19	172
H1_30	98.76	98.33	27	H1_30	93.74	97.37	259
H2_10	99.26	98.92	7	H2_10	-	97.27	82
H2_20	96.97	98.48	16	H2_20	96.66	97.38	173
H2_30	96.79	98.17	28	H2_30	-	97.26	274
L1_10	100.00	99.12	6	L1_10	-	97.30	74
L1_20	100.00	99.01	15	L1_20	-	97.25	158
L1_30	99.99	98.29	26	L1_30	-	97.06	254
L2_10	99.63	98.92	6	L2_10	-	97.55	80
L2_20	99.42	98.68	15	L2_20	96.74	97.39	170
L2_30	100.00	98.22	27	L2_30	-	97.25	295

# Computational results

30x5				40x5			
Instance	CPLEX	HEUR	Time (sec)	Instance	CPLEX	HEUR	Time (sec)
H1_10	-	95.67	340	H1_10	-	97.38	1104
H1_20	-	95.31	677	H1_20	-	97.38	2234
H1_30	-	95.54	1009	H1_30	-	97.25	3387
H2_10	-	95.88	316	H2_10	-	97.40	1095
H2_20	-	95.81	684	H2_20	-	97.33	2198
H2_30	-	95.30	969	H2_30	-	97.27	3296
L1_10	-	96.55	324	L1_10	94.92	97.41	1421
L1_20	-	96.43	652	L1_20	94.47	97.14	2996
L1_30	-	96.18	966	L1_30	-	96.20	4862
L2_10	-	95.68	308	L2_10	-	97.41	1382
L2_20	-	95.12	614	L2_20	94.93	97.34	3144
L2_30	-	-	920	L2_30	94.61	96.60	4352

# Computational results

50x8				60x13			
Instance	CPLEX	HEUR	Time (sec)	Instance	CPLEX	HEUR	Time (sec)
H1_10	-	96.52	3291	H1_10	-	95.40	6332
H1_20	-	96.37	6020	H1_20	-	95.07	10809
H1_30	-	96.21	9432	H1_30	-	94.76	10807
H2_10	-	96.03	3066	H2_10	-	95.54	6397
H2_20	-	95.64	6180	H2_20	-	94.11	10803
H2_30	-	95.16	9501	H2_30	-	-	10806
L1_10	-	95.97	2752	L1_10	-	95.67	5807
L1_20	-	96.04	6467	L1_20	-	95.40	10803
L1_30	-	95.80	9119	L1_30	-	94.45	10806
L2_10	-	96.18	3157	L2_10	-	95.63	5986
L2_20	-	95.96	5857	L2_20	-	95.64	10809
L2_30	-	96.27	8783	L2_30	-	95.34	10804

# Conclusions

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- The heuristics is able to find feasible solutions in 70 out of 72 instances, whereas CPLEX succeeds at that only on 20 instances, the smaller ones.
- Our algorithm is up to 2 order of magnitude faster than CPLEX, especially on small instances.
- The heuristics performs very well also on the instances of bigger size, where CPLEX generally fails.
- Next step: improve upper bounds using decomposition techniques.

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Thanks for your attention!

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Paper reference:

Giallombardo, G., Moccia, L., Salani, M., and Vacca, I. (2009).  
**Modeling and Solving the Tactical Berth Allocation Problem,**  
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