

Genetic Algorithms approach for containership fleet management dependent on cargo and their deadlines

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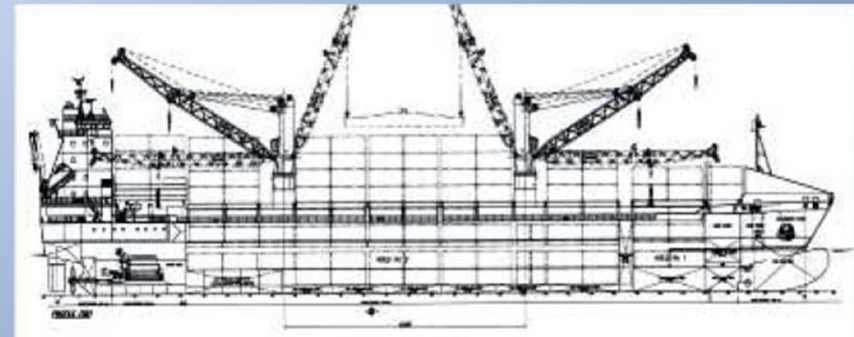


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Short Sea Shipping



Advantages

- Environmentally sound;
- Contribution to road safety;
- Low infrastructures cost (motorway is the sea);
- Reaches “peripheral” regions.

Disadvantages

- Customs bureaucracy;
- Port costs and efficiency;
- Dependency of environmental factors
- Travel duration;
- Inflexibility of routes.

State of the art - optimization



- Container stowage problem (CSP)
 - ✓ Avriel *et al.* – binary programming (1993)
– suspensory heuristics (2000)
 - ✓ Wilson and Roach – two phase method (1999)
– branch and bound application (2000)
 - ✓ Several authors – different implementation methods
- Ship routing and Scheduling
 - ✓ Christiansen and Nygreen (1998)
 - ✓ Agarwal and Ergun (2008)
- Vehicle routing problem with time windows - supply chains
 - ✓ Gendrau *et al.* (2006)
 - ✓ Moura and Oliveira (2009)

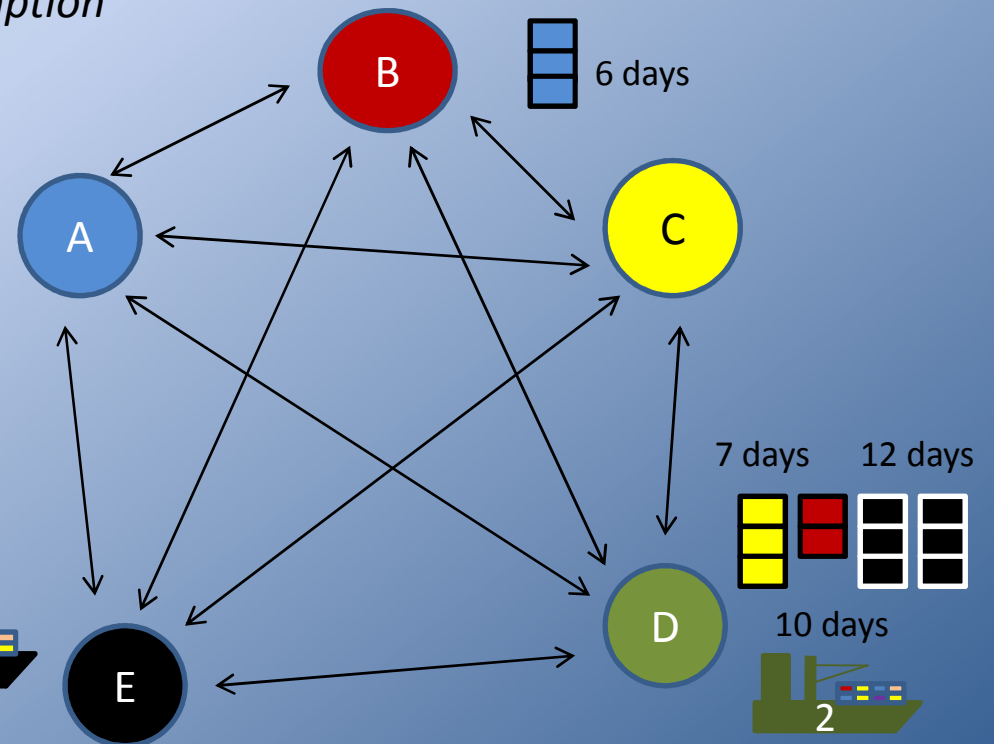
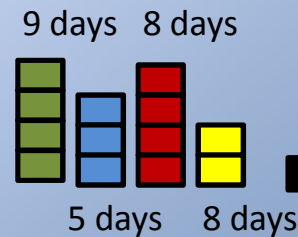
Problem definition (1)



Scenario definition:

- Five ports A , B , C , D, E
- Distances between ports < 1000 nm - *10 possible arcs*
- Fleet of Two containerhips
 - i. dimensions* *ii. deadweight*
 - iii. speed* *iv. fuel consumption*
 - v. stability characteristics*
- Several containers
 - i. origin* *ii. destination*
 - iii. weight* *iv. deadline*
- Costs at each port

- | |
|---------------|
| Harbour fee € |
| Berthing € |
| Facilities € |
| Pilot € |
| Duration € |
| ETC ... |



Problem definition (2)



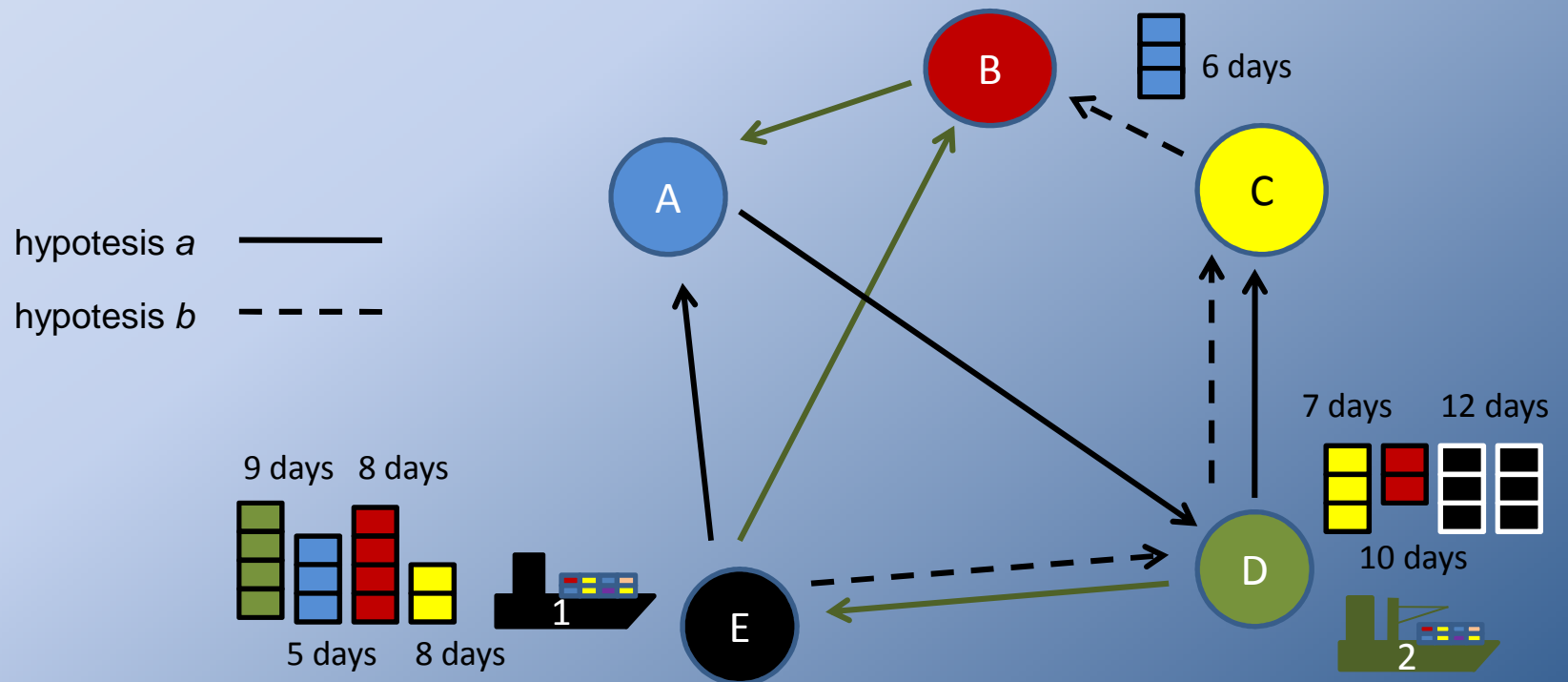
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Stage 1

- Containers characteristics and deadlines
- Containership characteristics
- Containership operation cost

What route for each ship?
What containers to carry?

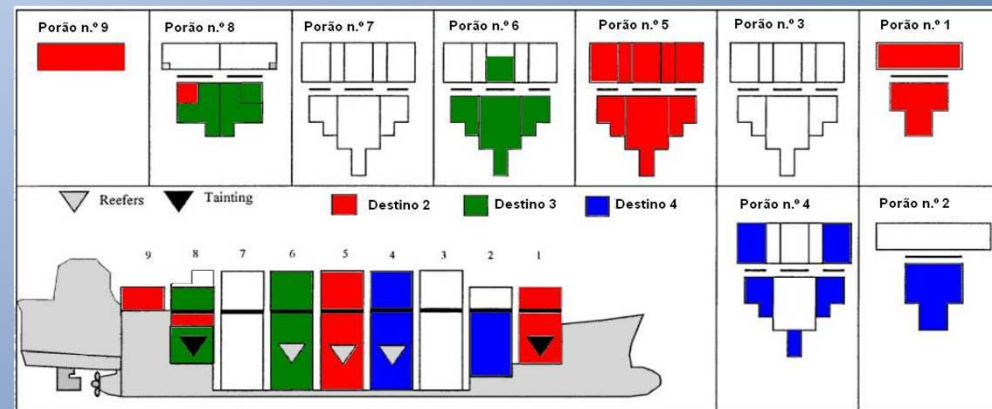


Problem definition (3)

Stage 2

- Containers characteristics and deadlines
- Containership characteristics
- Containership operation cost
- Containerships routes
- Containers carried in each journey

How to stow the containers



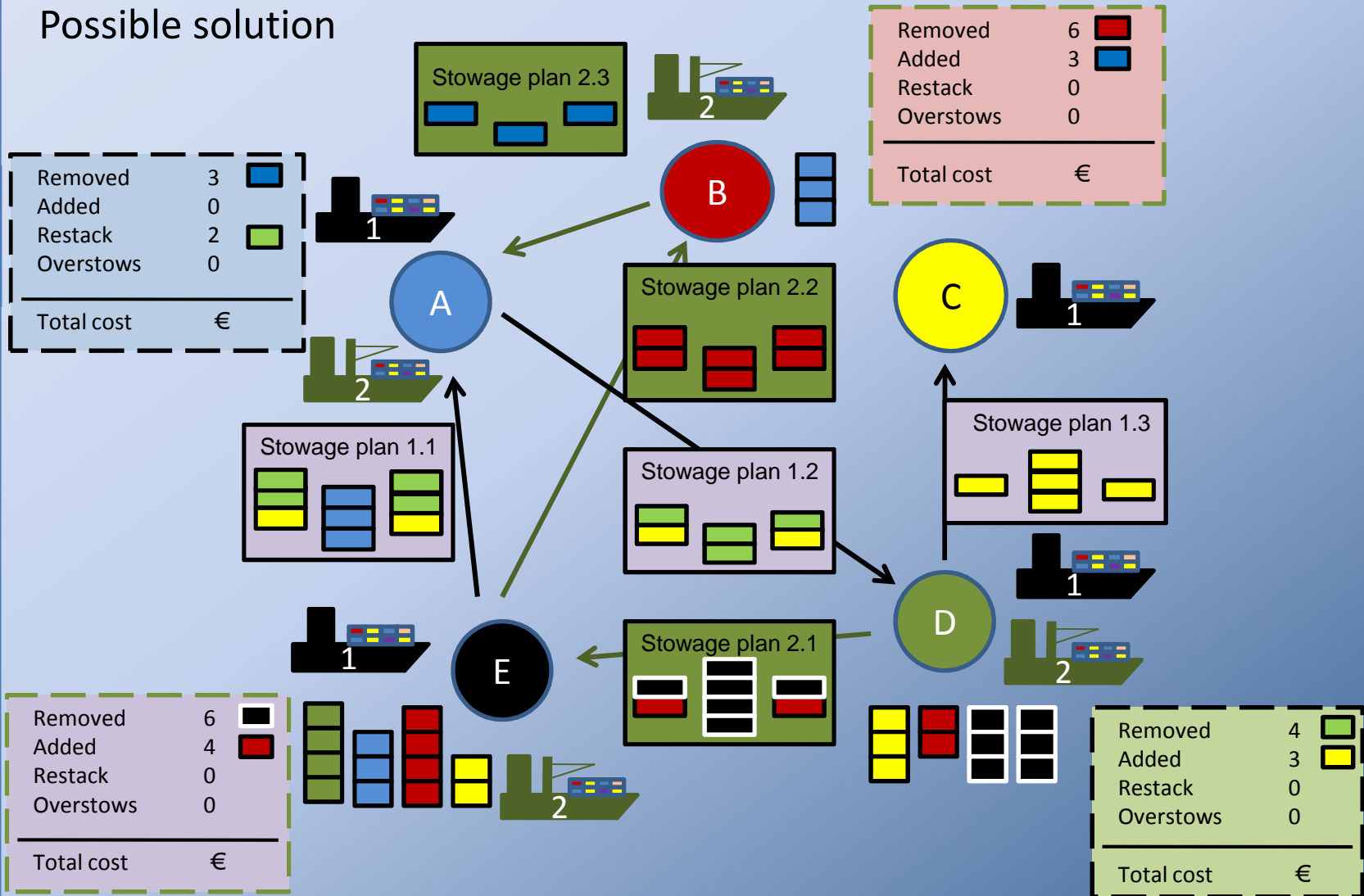
Altered from Wilson and Roach (1999)

Problem definition (4)

Possible solution



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Mathematical formulation (1)

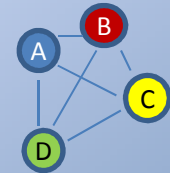


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Input variables

- Ports and Scenario characteristics
- Vessels characteristics
- Containers characteristics
- Associated costs (money and time)



- distances;
- port fees;
- time to move each container



- dimensions
- fuel consumption;
- speed;
- slots arrangement;
- operation costs;

Decision variables

very large binary matrices

- Route selection
 - for each vessel
 - at each journey
 - what's the port of origin
 - what's the port of destination
- Container distribution and stowage
 - for each container
 - what vessel carries it
 - where is it embarked
 - what journeys is it onboard
 - where is it stowed

- weight
- origin;
- destination;
- dimensions;
- delivery deadline;



Mathematical formulation (2)



Objective function :

- minimize cost; AM6
- deliver all cargo within the time limits (accepted loss?) AM5

$$W_1 \sum_{i=1}^p \sum_{j=1 \wedge j \neq i}^p \sum_{k=1}^v \sum_{\phi=1}^{\phi_{\max}} (d_{ij} c_k x_{ijk\phi} + x_{ijk\phi} f_i)$$

Operation costs outside harbour

$$W_2 \sum_{i=1}^p \sum_{k=1}^v \sum_{\phi=1}^{\phi_{\max}} \beta_{ik\phi} \mu_{ik\phi}$$

Containers shift costs

$$W_3 \sum_{i=1}^p \sum_{j=1 \wedge j \neq i}^p \sum_{k=1}^v \sum_{\phi=1}^{\phi_{\max}} \sum_{\alpha=1}^c (C_{orig} r_{jk\alpha\phi} + C_{dest} r_{ik\alpha\phi})$$

Containers load/ unload costs at the ports of origin and destination

Diapositivo 10

AM5

Para retirar

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AM6

- Total cost minimization;

Objective function composed of three componentes

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Mathematical formulation (3)



Constraints:

- route flow conservation - at each journey the vessel's port of origin is the previous port of destination; AM3
- time sequence – the vessel service at each port does not begin before she arrives there; AM2
- containers' deadline;
- vessel's capacity limit;
- containers' exclusivity – only one ship can carry each container at the same time; AM4
- cargo attribution – the vessel that carries a container has to visit its port of destination (simplification);
- slots occupation – if a slot is occupied than the ones bellow are also occupied (cargo hatches not considered).

Diapositivo 11

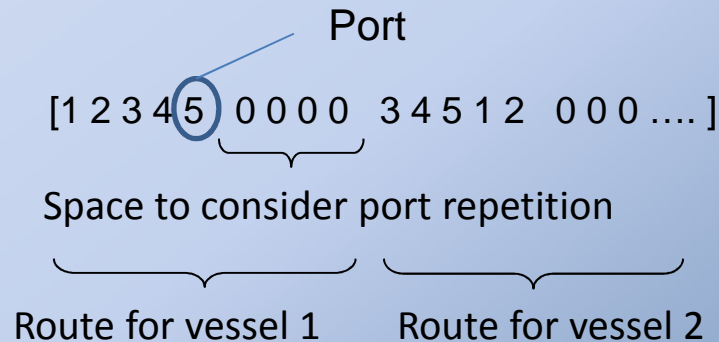
- AM2** at each port, the time service does not begins until the vessel arrives
Ana Moura; 25-06-2010
- AM3** if a vessel arrives at a node then it has to leave from that node to another one.
Ana Moura; 25-06-2010
- AM4** Each container is transported by one and only one vessel between ports.
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Genetic Algorithms Implementation



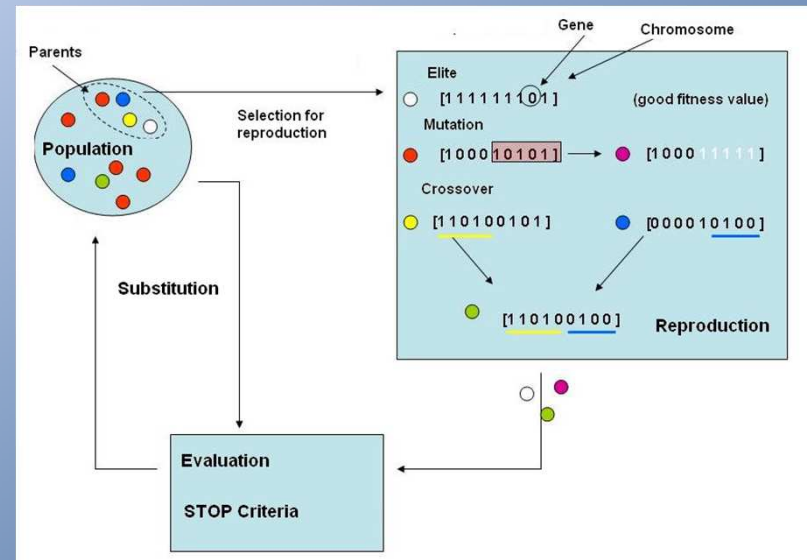
- Formulation of the problem into two different stages:
 - a. Port sequence;
 - b. Containers distribution by vessel;
 - c. Container stowage problem.
- } 1st stage implemented
- } 2nd stage not yet implemented

- First stage chromosome construction:



- Matlab and University of Aveiro Operators used

Biological evolution





Implementation Results

Example – scenario



Scenario definition:

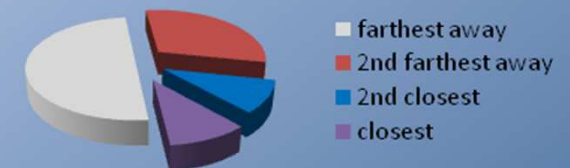
- Five ports A, B, C, D, E
- Fleet of Two containerships
- Several containers
- Operation costs and port fees



Ships	Length (m)	Beam (m)	Draught (m)	Speed (knots)	N.º TEU	Gross tonnage	Fuel (t/h)
AXE	95	15.6	6.15	12.5	348	3814	0.378
HEAVY	132.2	20	7.7	16.5	641	8445	0.945

Port tariffs (ship related costs)	Containers related costs	Voyage costs
<ul style="list-style-type: none"> pilot charges; berthing; space; tugs and towing; 	<ul style="list-style-type: none"> port tariff p/ TEU; container embarkation; container disembarkation; container shift; 	<ul style="list-style-type: none"> fuel; crew related;

Container's Port of destination



Container's deadlines



Container's weight



Example – Problem 2



MatLab Implementation:

- Each square stands for 100 containers
- No shift nor stowage constraints considered
- *Heavy* departs from C and *Axe* departs from A
- After 1 circular route – no solution was found
- After 9 ports following a circular route – 100 containers not delivered
- After 9 ports following a GA route – 315.5h, all delivered but 68 not in time



Containers distribution: A – 400 B – 200 C – 400
 D – 200 E – 200

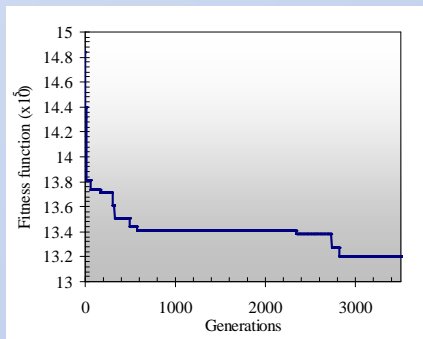
	Ship	Route	Travelled miles	Time interval	Deadlines surpassed	Total cost
Circular routes	Axe	A – B – C – D – E	Not every container was embarked			
	Heavy	C – D – E – A – B				
	Axe	A – B – C – D – E – A – B – C – D	100 containers not delivered			
	Heavy	C – D – E – A – B – C – D – E – A				
GA	Axe	A – B – D – E – B – A – B – D – E	3134	315.5 h	68	634960
	Heavy	C – B – A – B – E – B – D – C	2653	236 h		

Example – Problem 3



University of Aveiro software:

- No shift, deadlines, nor stowage constraints considered
- *Heavy* departs from C and *Axe* departs from A;
- Following a circular route it was found that 15 ports had to be visited by each ship
- Two solutions better were found, with cost reductions of 7% and 14%



Container distribution:	from/to	A	B	C	D	E
	A	0	100	500	300	100
	B	250	0	150	50	50
	C	500	100	0	100	500
	D	250	50	50	0	150
	E	250	50	150	50	0

	Ship	Route	Travelled miles	Time interval	Deadlines surpassed	Total cost
Pre-defined	Axe	A - B - C - D - E - A - B - C - D - E - A - B - C - D - E	6536	637.3 h	-	566039.1
	Heavy	C - D - E - A - B - C - D - E - A - B - C - D - E - A	6844	620.3 h	-	958090.1
	Axe	A - B - E - D - A - B - E - C - A - D	5437	561.7 h	-	551051.4
	Heavy	C - D - B - A - E - D - C - A - C	4367	527.6 h	-	871738.4
GA	Axe	A - C - A - B - E - D - C - B	3831	435.2 h	-	481087.7
	Heavy	C - A - B - C - D - C - E - B - A	3657	451.8 h	-	833674.2

Conclusions



- Conceptual and mathematical models have been built for the containership fleet management considering cargo deadlines, which included:
 - ✓ Route selection;
 - ✓ Cargo distribution per ship;
 - ✓ Container stowage problem solution.
- A Genetic algorithm implementation for route selection and cargo distribution has been done;
- Using a simple scenario of 2 containerships and 5 ports with distances between them smaller than 1000 nm, it was seen that:
 - ✓ There are routes better than circular pattern ones;
 - ✓ It is possible to manage a fleet considering the cargo deadlines and reducing time duration when necessary;
 - ✓ It is possible to turn the maritime transport more flexible with cost reductions.

Future development



- Mathematical model improvement
- Implementation of the CSP in the GA model
- Analysis of different methods efficiency to solve the model other than GA
- Analysis of real scenarios with real data (collaboration required)
- Analysis of the impact of vessel characteristics in the results
- Use the model to define owner requirements for new ship constructions based on known scenarios



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