

Dynamical Handling of Straddle Carriers Activities on a Container Terminal in an Uncertain Environment - Swarm Intelligence Approach -

G. Lesauvage



*Unité de Formation et de Recherche des
Sciences et Techniques*



*Laboratoire d'Informatique et du Traitement de
l'Information et des Systèmes*

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Outline

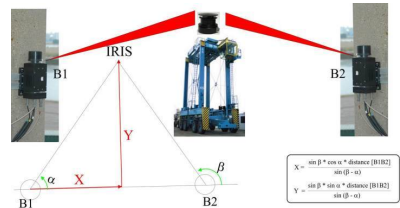
- 1 System description
- 2 Vehicle Routing Problem : state of the art
- 3 Ant Colony and Straddle Carrier Handling
- 4 Simulator
- 5 Preliminary Results
- 6 Conclusion

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The CALAS project

- Laser measure system and software
- 2 main companies :
 - Laser Data Technology Terminal
 - *Terminaux de Normandie*
- 2 main laboratories :
 - LMAH
 - LITIS



Objective of the CALAS project :

To know the state of the terminal, in real time, for both containers and trucks location.

Terminal description

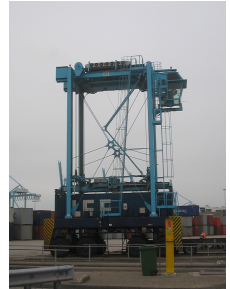
- Container terminal
- 3 main areas :
 - Ship handling
 - Stock area
 - Truck/Train handling



- Stock area contains many long rows of stacked containers

Handling trucks

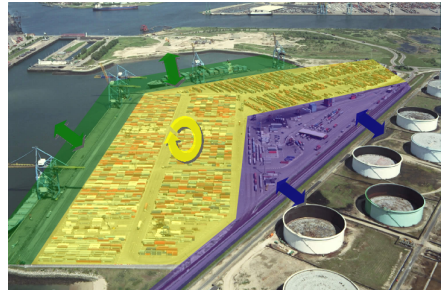
- Straddle carriers are handling trucks which can carry on one container at a time
- They can treadle stacks of containers and take/drop a container from/on the peak of a stack
- One mission consist in moving a container from a place inside the terminal to another one



Container Terminal Activities

3 kinds of missions :

- Preparing a **ship** (un)loading
- Preparing a **truck** (un)loading
- Optimizing **stock** area

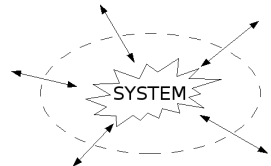


System dynamic

Open system means uncertain environment.

Incoming and outgoing flows :

- do not only depend on the system
- affect the system and lead it into a new state



Uncertain Environnement

There are several unpredictable events such as :

- Incoming missions
- Trucks and ships arriving time
- Road/rail network disconnections
- Straddle carriers failures
- Human behavior

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VRPTW[1]

- **V**ehicle **R**outing **P**roblem
- **T**ime **W**indows

Goal

Optimizing the delivery routes of each truck

example of VRPTW : the toys factory

- The factory produces toys and its vehicles deliver a set of stores
- Stores are spread all over the country and goods are carried by trucks
- Every truck has a restricted capacity and starts from the factory depot
- Deliveries must occur during a time interval and if a truck comes too early, it will have to wait

DVRPTW[6]

- Dynamic

Goal

Optimizing the new routes of each truck without recomputing from scratch

Dynamic toys factory

- Toy factory problem
- While a schedule is running, stores are still allowed to ask for a delivery

(D)PDP

- DVRP where the goods have to be picked-up before being delivered

Goal

Optimizing both pickup and delivery routes

Pickup and Delivery Problem example : mail delivery problem

- A mail company employs a set of postmen
- They have to pickup mails from the mail boxes of the company
- Then, they must deliver them to their recipients as soon as possible

DSCPDPTW

- Pickup and Delivery Problem class
- Unit capacity
- Time windows : depends on both straddle carriers and boats/trucks/trains
- Dynamic :
 - A plan can change at anytime : vehicles may start from elsewhere than the depot
 - Straddle Carriers failures : the number of ressources can change
 - Trucks/trains/boats arriving time : we cannot be sure they will respect their time window
 - ...

A computed solution can be aborted at anytime !

DSCPPDTW (2)

- 2 problems :
 - Minimize straddle carriers moves : shortest path problem
 - Minimize customers delays : scheduling problem

Problem dependencies

appropriate scheduling	⇒	shortest path concept
scheduling shortest paths	⇒	reducing straddle carriers moves

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Ant Colony Optimization[3]

- ACO is a meta-heuristic
- ACO makes a solution appear thanks to the run of artificial ants into the solution space
- ACO is adapted to the dynamic nature of this problem :
 - Positive feedback : ants spread pheromone according to solution quality
 - Negative feedback : pheromone track evaporates progressively

Scheduling with Ant Colony

Ant Colony with **one colony** provides a sorted list of missions to accomplish.

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Colored ants[2] :

- every straddle carrier represents a colony with its own color
- ants are attracted by pheromones of their own colony
- ants are repulsed by pheromones of foreign colonies

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Colored ants[2] :

- every straddle carrier represents a colony with its own color
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Ant Colony with **many colonies** provides a sorted list of missions per straddle carrier.

Missions graph

The directed graph can be conceptualized as follows :

- Vertices :
 - 1 mission = 1 node
 - 1 straddle carrier = 1 colored node connected to all compatible missions
- Colored Arcs :
 - Compatibility between 2 missions for a straddle carrier

Mission compatibility

We say that mission m_a is **compatible** with mission m_b if the time window of m_a starts before the one of m_b

Example of a mission graph construction (1)

Example

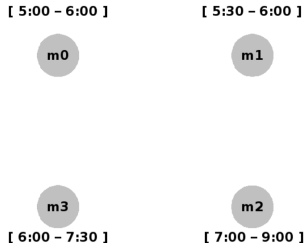
- Missions :

Name	Start	End
m0	5:00	6:00
m1	5:30	6:00
m2	7:00	9:00
m3	6:00	7:30

- Straddle Carriers :

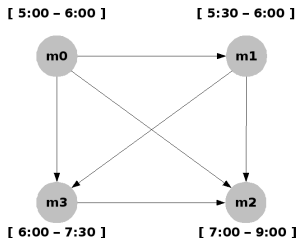
Name	Color	Compatiblility
s0	green	m0, m1, m2, m3
s1	blue	m0,m3

Example of a mission graph construction (2)



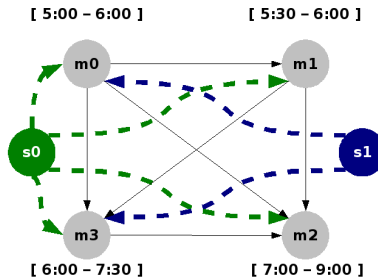
1 mission \iff 1 vertex

Example of a mission graph construction (3)



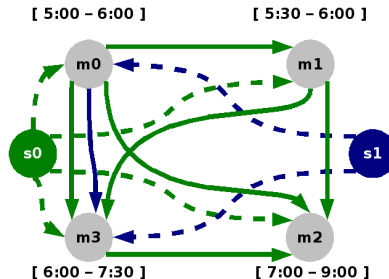
1 arc between two compatible missions

Example of a mission graph construction (4)



Adding nodes modeling straddle carriers and connecting them to every compatible vertices

Example of a mission graph construction (5)



Adding or Coloring edges between nodes according to their connectivity with the vehicles

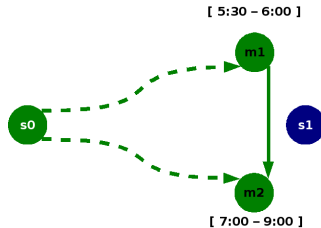
Algorithm description

Main algorithm

```
begin  
| for each colony  $c$  do  
| | for each ant  $a$  of  $c$  do  
| | | choose an unvisited destination according to the pheromone track  
| | | move towards it according to the speed of  $a$   
| | | spread pheromone according to the destination quality  
| | end for  
| end for  
| evaporation  
end
```

Solution

- The solution is the coloring of the nodes.
- When a straddle carrier is free, it chooses the mission of its color which has the highest pheromone rate.
- The chosen missions are removed from the graph and the algorithm continues running.
- The new missions vertices are dynamically added to the graph so they are taken into account by the ant algorithm

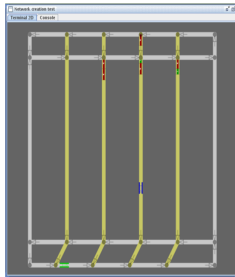


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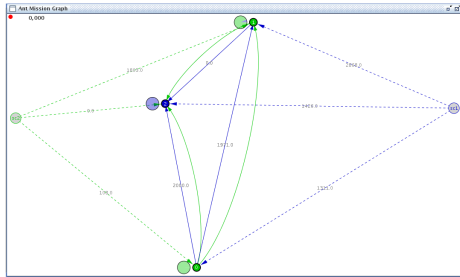
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2 parallel views of the system

- Terminal implementation



- ACO modeling[4]



ACO \iff Terminal

Effects of ACO results must appear on the terminal and the terminal state must affect the ACO setting (mission graph)

Dynamism handling

A scenario file is read all along the execution of the simulation. It contains dynamic events.

Measure of dynamism

According to A. Larsen[5], we can measure how dynamic is a scenario by these two formulas :

- Degree of Dynamism (dod) = $\frac{\eta_d}{\eta_s + \eta_d}$

- Effective Degree of Dynamism (edod) = $\frac{\sum_{i=1}^{\eta_d} t_i}{\eta_s + \eta_d}$

η_s : number of static requests ;

η_d : number of dynamical requests.

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Preliminary results

- Test the relevance of both our modeling and our algorithm on simulated data
- Function of the measures of dynamism

	Static	Half Dynamic	Dynamic
<i>dod</i>	0	0.5	1
<i>edod</i>	0	0.25	1
End time	22693	22276	22693
Number of overrun tw	3	5	7
Overrun time penalty	6467	8477	12485

The exceeded time windows and the time penalties evolve in the same way that *dod* and *edod*.

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Conclusion

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- Dynamic Pickup and Delivery Problem with Time Windows
- Swarm intelligence with :
 - Graph modeling
 - Ant Colony System
 - Colored Ants
- A simulator is being developped and will allow to measure the solution relevance
- Preliminary results confirm that our modelling is able to handle dynamics

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Thank you for your attention



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