

# WRM: Weather Routing Metaheuristic

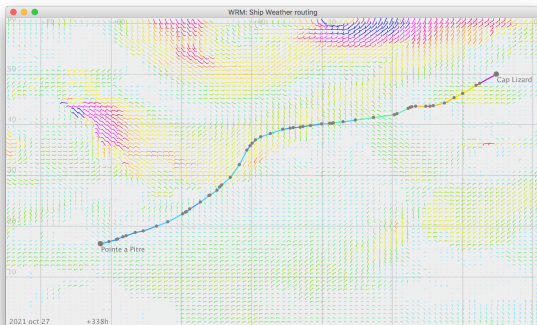
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June, 2022

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# Ship weather routing



Search optimised routes for ships in an environment that changes with the time.

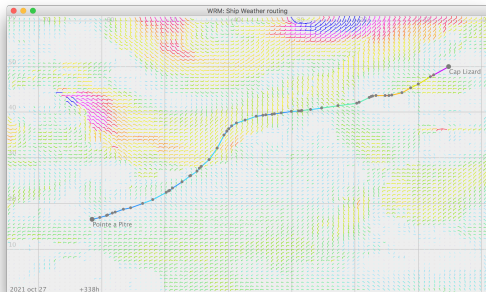
# Plan

Ship weather routing

WRM : a new approach for ship weather routing

WRM running

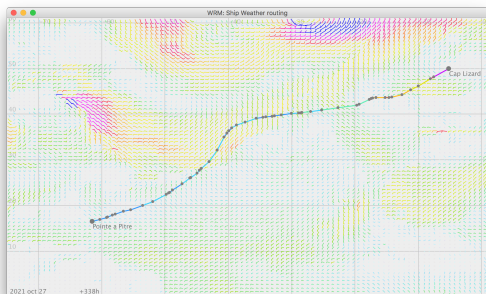
# Ship weather routing



Search optimised routes for ships in an environment that changes with the time.



# Ship weather routing

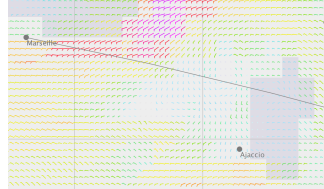


## Plan

- ▶ the problem,
- ▶ WRM approach,
- ▶ experimentations.

Search optimised routes for ships in an environment that changes with the time.

# Ship weather routing

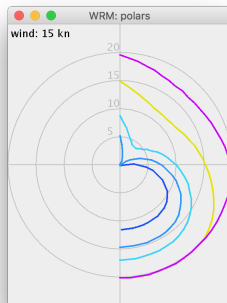


## Environmental factors

- ▶ wind (speed, direction),
- ▶ waves (direction, height, . . . ),

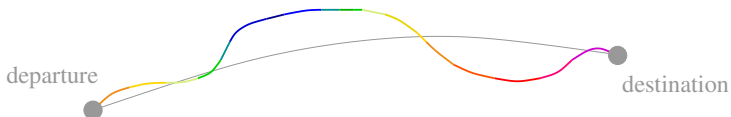
## affect ship performances

- ▶ *consumption*,
- ▶ *progression*,



SOG with 15-knots wind, power levels 5 kW, 10 kW, 15 kW, 25 kW, 50 kW

# Ship weather routing : Route



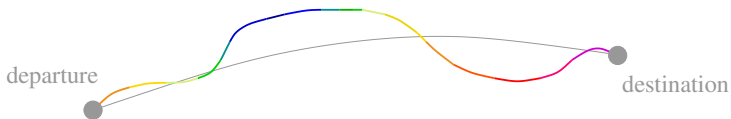
## Route :

- ▶ a trajectory,
- ▶ a value representing the *speed* at each point of the trajectory.

## speed parameter

- ▶ SOG, **engine power level**, shaft rotation speed,...

# Ship weather routing



## Optimize :

- ▶ cost,
- ▶ safety, green gazes emissions, comfort, duration,...

**Constraints :** dates, forbidden/penalized regions, bad weather conditions,...

**The search space is continuous in space as in time**

# Ship weather routing

## Publications :

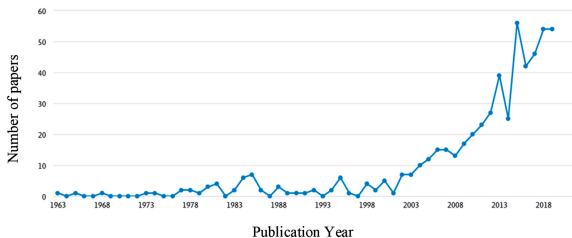


Fig. 2. Publishing trend in the area of ship weather routing and voyage optimization.

Source: Scopus February 2020.

Zis, T.P., Psaraftis, H.N., Ding, L. : Ship weather routing : A taxonomy and survey. Ocean Engineering 213, 107697 (2020).

# Ship weather routing

Existing approaches :

- ▶ **isochrones, isopones** : time/fuel consumption,
- ▶ **dynamic programming** : 2D, 3D,
- ▶ **Dijkstra based algorithms**,
- ▶ **genetic algorithms**,
- ▶ ...

# Ship weather routing

Existing approaches :

- ▶ **isochrones, isopones** : time/fuel consumption,
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- ▶ ...

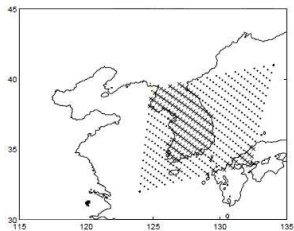
**WRM** : metaheuristic

- ▶ launch many times a simple forward algorithm on *simplified versions of the problem*

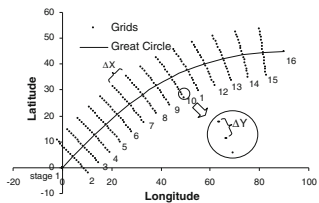
Grandcolas, S., A metaheuristic algorithm for ship weather routing, to appear in *Operations Research Forum*, 2022.

DOI : 10.1007/s43069-022-00140-0

# Existing approaches



"Multi-Objective Optimization of Motor Vessel Route."  
Marie, S., Courteille, E. 2009



"Development of a novel forward dynamic programming method for weather routing"  
Shao, W., Zhou, P., Thong, S. 2011



# Existing approaches

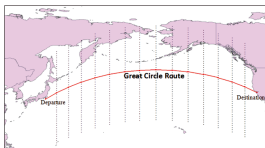
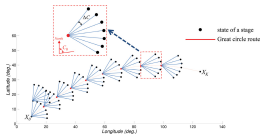
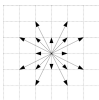
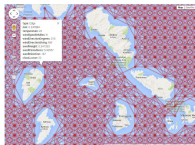


Fig. 1. Grid system and Great Circle Route

"An Ant Colony Algorithm for efficient ship routing."  
Tsou, M.C., Cheng, H.C. 2013



"The optimization of ship weather-routing algorithm based on the composite influence of multi-dynamic elements."  
Fang, M.C., Lin, Y.H. 2015



"Minimizing the fuel consumption and the risk in maritime transportation : A bi-objective weather routing approach"  
Veneti, A., Makrygiorgos, A., Konstantopoulos, C., Pantziou, G., Vetsikas, I. 2017

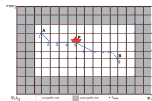


Fig. 2 The grid as navigational chart

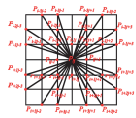
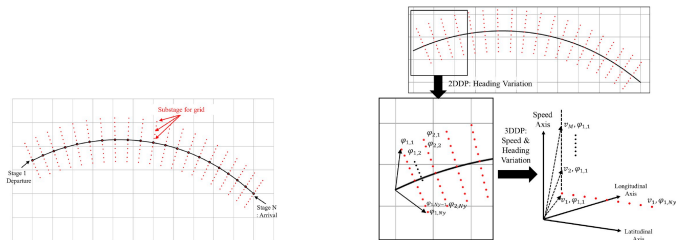


Fig. 3 Illustration of all vessel movement possibilities in a grid

"Multi-Objective Weather Routing of Sailing Vessels."  
Życzkowski, Marcin and Rafał Szlapczyński. 2017

# Existing approaches : 3D dynamic programming



Kim, K.I., Lee, K.M. : Dynamic Programming-Based Vessel Speed Adjustment for Energy Saving and Emission Reduction. *Energies* 11(5) (2018).

# Existing approaches : Dijkstra's algorithm

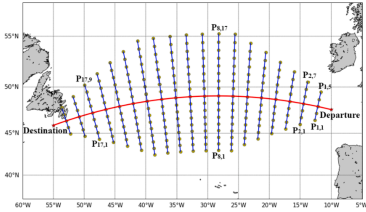


Fig. 2. Waypoints between departure and destination.

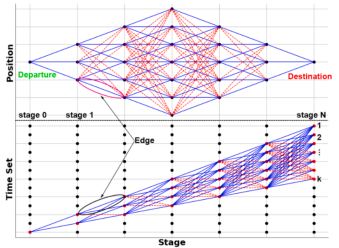


Fig. 3. An illustration of the 3D graph (a grid of waypoints) system for voyage optimization.

Wang, H., Mao, W., Eriksson, L. : A Three-Dimensional Dijkstra's algorithm for multi- objective ship voyage optimization. Ocean Engineering 186, 106131 (2019).

# WRM : Weather Routing Metaheuristic

Ship weather routing

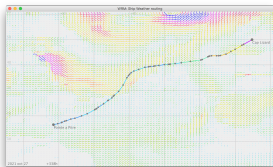
WRM : a new approach for ship weather routing

WRM running

# WRM

## Problem :

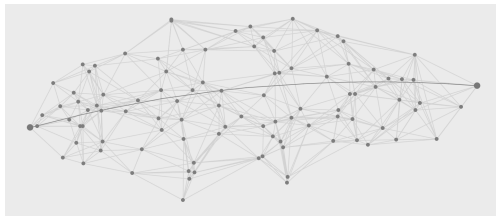
- ▶ *departure, destination,*
- ▶ a time window,
- ▶ a weather forecast,
- ▶ *speed parameter* : engine power level



**Goal** : find a route whose total cost is minimal

**The idea** : solve a series of **simplified versions of the problem** gradually focusing on the most promising area.

# WRM : simplified version of the problem



$$P = \{p_0, p_1, \dots, p_n\}$$

*points*

$$L = \{(p_i, p_j), p_i, p_j \in P\}$$

*legs*

$$W = \{l_1, l_2, \dots, l_m\}$$

*power level values*

$$[t_{min}, t_{max}]$$

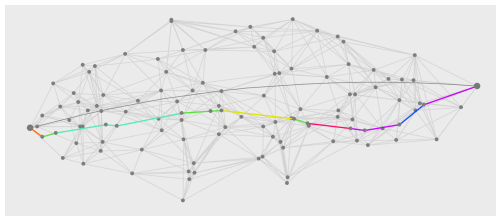
*time window*

$$\mathcal{F}$$

*weather forecast*

$(P, L)$  : directed, acyclic.

# WRM : simplified version of the problem



$P = \{p_0, p_1, \dots, p_n\}$	<i>points</i>
$L = \{(p_i, p_j), p_i, p_j \in P\}$	<i>legs</i>
$W = \{l_1, l_2, \dots, l_m\}$	<i>power levels</i>
$[t_{min}, t_{max}]$	<i>time window</i>
$\mathcal{F}$	<i>weather forecast</i>

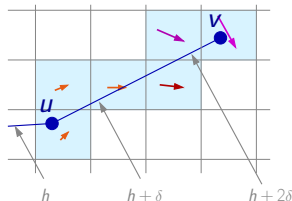
**Route :**  $\langle t_{dep}, (u_0, u_1, \dots, u_k), (w_0, w_1, \dots, w_{k-1}) \rangle$

for each  $i$ ,  $(u_i, u_{i+1}) \in L$  and  $w_i \in W$ , constant power level on each leg.

# WRM : simplified version of the problem

## Evaluation of crossings :

- ▶  $duration(u, v, w, t, \mathcal{F})$
- ▶  $cost(u, v, w, t, \mathcal{F})$



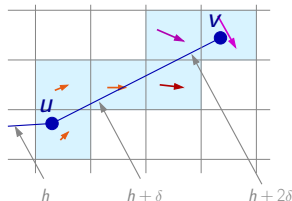
duration and cost to cross from  $u$  to  $v$  at the date  $t$  at engine power level  $w$ , given the forecast  $\mathcal{F}$ .



# WRM : simplified version of the problem

## Evaluation of crossings :

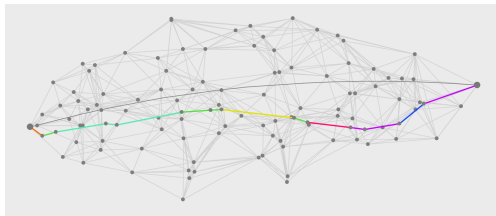
- ▶  $duration(u, v, w, t, \mathcal{F})$
- ▶  $cost(u, v, w, t, \mathcal{F})$



$$P = P_{idle} + (P_{calm} + P_{wind}) \times C_{waves}$$



## WRM : simplified version of the problem



**Valid route :**  $\langle t_{dep}, (u_0, u_1, \dots, u_k), (w_0, w_1, \dots, w_{k-1}) \rangle$

satisfies

$$t_{dep} \geq t_{min} \text{ and } t_{arr} \leq t_{max}$$

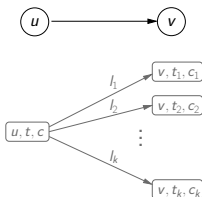
where  $t_{arr}$  is the arrival date

# WRM : searching a good route

**Dated costs :**  $(u, t, c)$

there is a route that reaches  $u$  at date  $t$  whose cost is  $c$

**Propagation :**



**Combinatorial explosion :** discard the dated costs that seem the least promising.

**Best route :** *Arrival* best dated cost.

# WRM : simplified version of the problem

## Improvements :

- ▶ minimal arrival dates (forward exploration),
- ▶ maximal arrival dates (backward exploration),
- ▶ **minimal bounds** of the costs,

## Number of legs evaluations :

$$\mathcal{O}(|L| \times |W| \times K)$$

$K$  : given limit of the number of dated costs for each vertex

# WRM : Metaheuristic algorithm

**algorithm** WRM ( $dep, dest, t_{min}, t_{max}, P_{min}, P_{max}, \mathcal{F}, nb_{runs}$ )

$\pi$  := initial parameters,

**repeat**  $nb_{runs}$  **times**

$\pi$ -generate a simplified problem  $\mathcal{P}$ ,

    solve  $\mathcal{P}$  propagating dated costs,

    update  $bestRoute$ ,

    update  $\pi$ ,

**end loop**

**return**  $bestRoute$ ,

- ▶ *first runs* : unconstrained generation,
- ▶ *following runs* : strengthen  $\pi$  parameters step by step, so as to converge towards the most promising areas

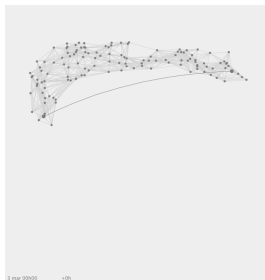
# Metaheuristic : graphs generation

Generate problem



# Metaheuristic : graphs generation

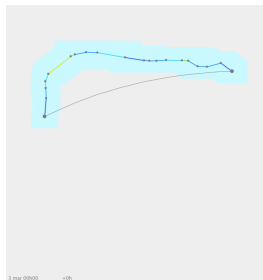
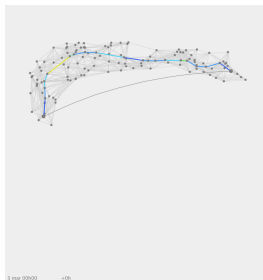
Search a good route





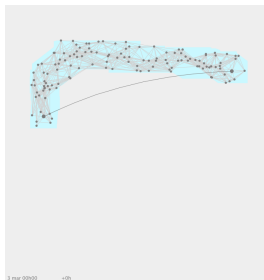
# Metaheuristic : graphs generation

Mark surrounding area



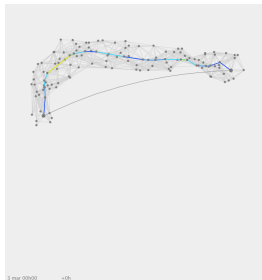
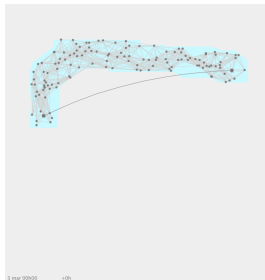
# Metaheuristic : graphs generation

Generate problem



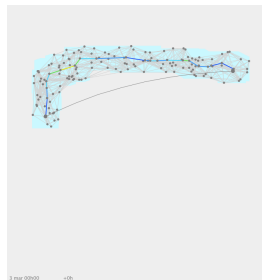
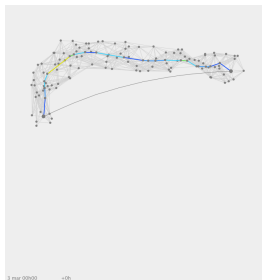
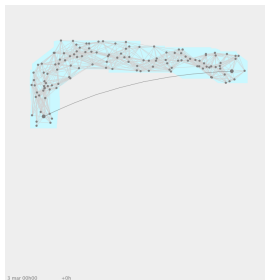
# Metaheuristic : graphs generation

Search a route



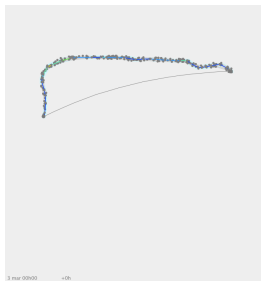
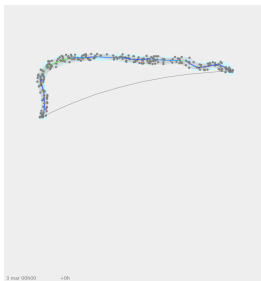
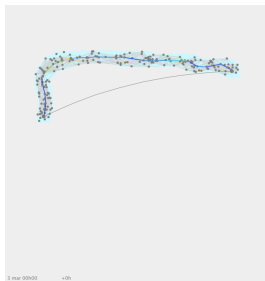
# Metaheuristic : graphs generation

Mark surrounding area



# Metaheuristic : graphs generation

Strengthen generation parameters



- ▶ perimeter,
- ▶ time windows,
- ▶ steps and ranges of the power levels.

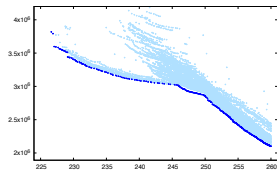
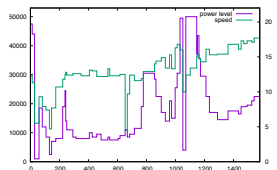
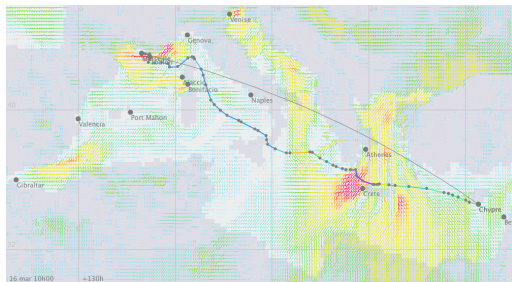
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Ship weather routing

WRM : a new approach for ship weather routing

WRM running

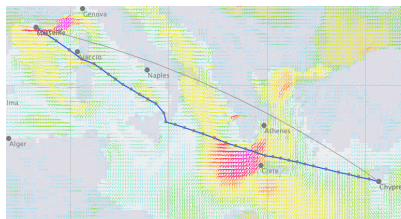
# WRM : Marseille-Chypre [130h, 260h]



**$l=1581.5\text{NM}/1394.1\text{NM}$   $sp=12.2\text{kn}$   $cost=2101399.2$   $130.0\text{h}-260.0\text{h}$  (130.0h) (16 mar 10h00 to 21 mar 19h57) 57 points 141.6s**

Forecast : NOAA (National Oceanic and Atmospheric Administration), GFS atmospheric and waves models, 384h, 3 hours intervals, resolution  $0.25^\circ$ .

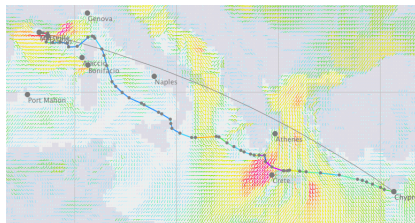
# WRM : Marseille-Chypre [130h, 260h], shortest route



## Direct route (simulation) :

- ▶ 1451.2 nautic miles,
- ▶ 128.5 hours,
- ▶ cost 4315241 (constant power 31.5kw)

cost at idle : 677237

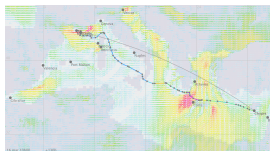
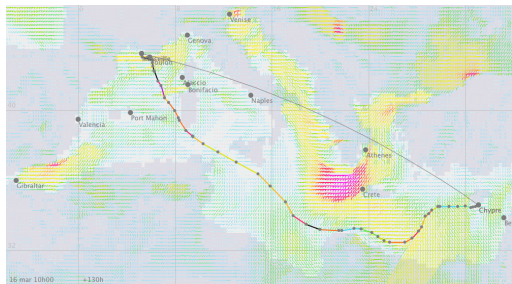


## Route returned by WRM :

- ▶ 1581.5 nautic miles,
- ▶ 130.0 hours (avg speed 12.2 knots),
- ▶ cost 2101399



# WRM : Marseille-Chypre penalize bad weather conditions



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wind speed	$\geq 22$ knots
wind heading	$\leq 120^\circ$

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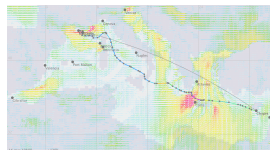
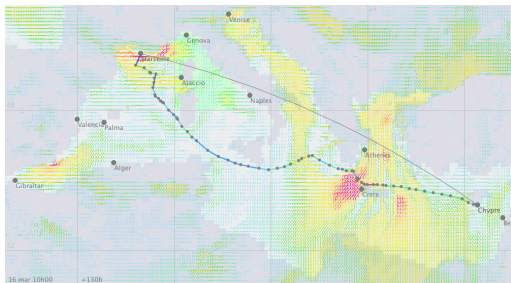
waves height	$\geq 1.5$ m
waves heading	$\geq 10^\circ$

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$l=1703.9\text{NM}/1394.1\text{NM}$   $sp=16.3\text{kn}$   $cost=3704002.4$   $155.5\text{h}-260.0\text{h}$  (104.5h) (17 mar 11h27 to 21 mar 19h59) 50 points 309.8s

$l=1581.5\text{NM}/1394.1\text{NM}$   $sp=12.2\text{kn}$   $cost=2101399.2$   $130.0\text{h}-260.0\text{h}$  (130.0h) (16 mar 10h00 to 21 mar 19h57) 57 points 141.6s

# WRM : Marseille-Chypre ECA [130h, 260h]

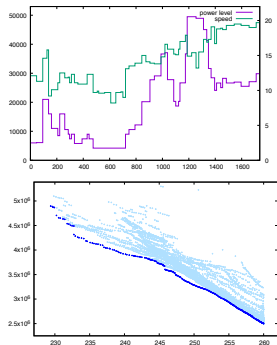
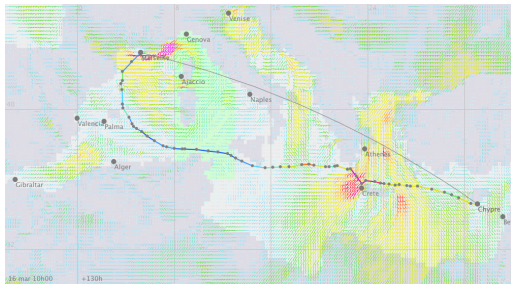


ECA (Emission Control Areas) : cost increase 20%

$l=1631.4\text{NM}/1394.1\text{NM}$   $sp=12.8\text{kn}$   $cost=2191371.0$  132.5h-260.0h (127.5h) (16 mar 12h27 to 21 mar 19h59) 57 points 103.2s

$l=1581.5\text{NM}/1394.1\text{NM}$   $sp=12.2\text{kn}$   $cost=2101399.2$  130.0h-260.0h (130.0h) (16 mar 10h00 to 21 mar 19h57) 57 points 141.6s

# WRM : Marseille-Chypre ECA [130h, 260h]



$l=1737.9\text{NM}/1394.1\text{NM}$   $sp=13.4\text{kn}$   $cost=2501611.2$  130.0h-260.0h (130.0h) (16 mar 10h00 to 21 mar 20h01) 59 points 54.4s

Merci !