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An introduction to the VISIR-2 ship weather routing model

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latest update: 2024-08-05

VISIR-2 resources

open access - open review journal paper

<https://doi.org/10.5194/gmd-17-4355-2024>



Geosci. Model Dev., 17, 4355–4382, 2024
<https://doi.org/10.5194/gmd-17-4355-2024>
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Geoscientific
Model Development
Open Access
EGU

Model description paper



VISIR-2: ship weather routing in Python

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open language - open source model code

<https://zenodo.org/communities/visir2>



April 11, 2024 (v3) Software Open

[VISIR-2 ship weather routing model] source code (Python)

Salinas, Mario Leonardo ; Carelli, Lorenzo ; Mannarini, Gianandrea 

Source code of the VISIR-2 ship weather routing model. The Python-refactored VISIR-2 model considers currents, waves, and wind to optimise routes. The model was validated, and its computational performance is quasi-linear. For a ferry sailing in the Mediterranean Sea, VISIR-2 yields the largest percentage emission savings for upwind navigation...

Part of VISIR-2 ship weather routing model (Python)

Uploaded on April 11, 2024

2 more versions exist for this record



1016 131

Outline

1. *Introduction*
2. *Model workflow*
3. *Case studies*
4. *Discussion*

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Why an open-source ship weather routing model?

- *Reducing costs and emissions from shipping is crucial for both the economy and the environment*
- *weather routing is a an operational tool to achieve this goal*
- *open-source models can achieve superior performance, foster cross-disciplinary developments, provide a public good for science and technology*

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VISIR-2 workflow

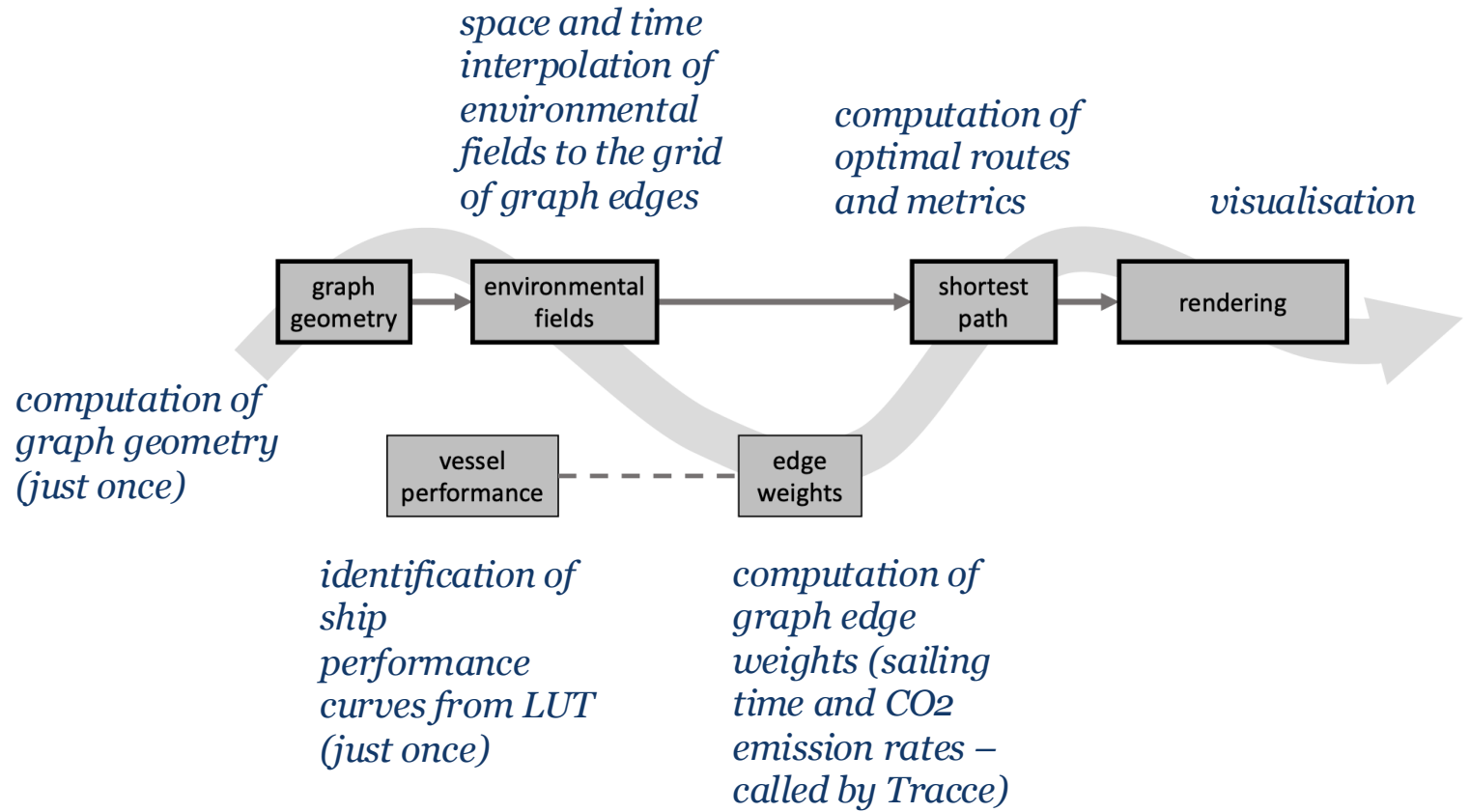
- *Greater modularity with respect to VISIR-1*
- *conda virtual environment ("visir-venv") for portability*

```

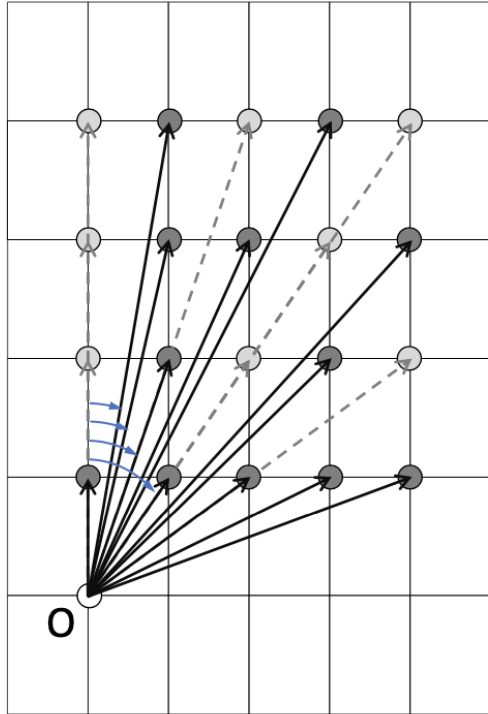
Run: MAIN_Tracce
/Users/gmannarini/opt/anaconda3/envs/visir-venv-gmd2023/bin/python /Users/gmannarini
Choose an option from the list below
or type Q to quit.
=====
n. | Tracce namelist
-----
0. | myTracce.yaml
=====
Type a number in [0, 0], or Q to quit: 0
08:41:06 [INFO] Starting Tracce job: myTracce...
08:41:06 [INFO] -----** Plevel = 0.7 **-----
08:41:06 [INFO] -----* Delay of 0 hours *-----
08:41:06 [INFO] Reading edge weights.
08:41:06 [INFO] Weight file read. Populating nx graph
08:41:06 [INFO] -----Populating networkX graph-----
08:41:06 [INFO] init nx nodes...
08:41:06 [INFO]: 100%|██████████| 2165/2165 [00:01<00:00, 2041.62it/s]
08:41:08 [INFO] assign edge weights to graph...
08:41:08 [INFO]: 100%|██████████| 87346/87346 [00:00<00:00, 262179.08it/s]
08:41:08 [INFO] -----Least distance route-----
08:41:08 [INFO] Least distance metrics:
08:41:08 [INFO]   - navigation distance: 49.015 [nmi]
08:41:08 [INFO]   - navigation duration: 6.003 [hrs]
08:41:09 [INFO] -----Least time route-----
08:41:09 [INFO] Least time metrics:
08:41:09 [INFO]   - navigation distance: 49.989 [nmi] (+2.0%)
08:41:09 [INFO]   - navigation duration: 5.758 [hrs] (-4.1%)
08:41:09 [INFO] -----* Completed run with delay of 0 hours *-----
08:41:09 [INFO] -----** Completed run with Plevel = 0.7 **-----

Process finished with exit code 0

```



Graph structure



Graph stencil for connectivity up to 4th order neighbours ($v=4$)

Graph nodes are joined through edges, up to v hops

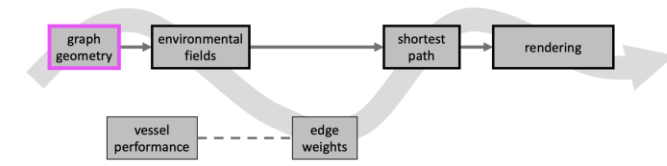
the higher the v , the smaller the minimum resolved angle and the smoother the ship's route

in VISIR-2, multi-hop collinear edges (dashed) are pruned.

Benefits:

- saving RAM memory
- more faithful representation of the environmental fields

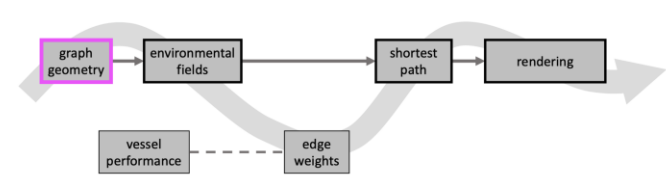
Mercator projection of the graph grid (`pyproj` library) used



v	$v(v+1)$	N_{q1}
1	2	2
2	6	4
3	12	8
4	20	12
5	30	20
6	42	24
7	56	36
8	72	44
9	90	56
10	110	64

- $v(v+1)$: number of edges in the first quadrant
- N_{q1} : number of edges upon pruning of collinear ones
- N_{q1} affects the memory allocation

Graph computation

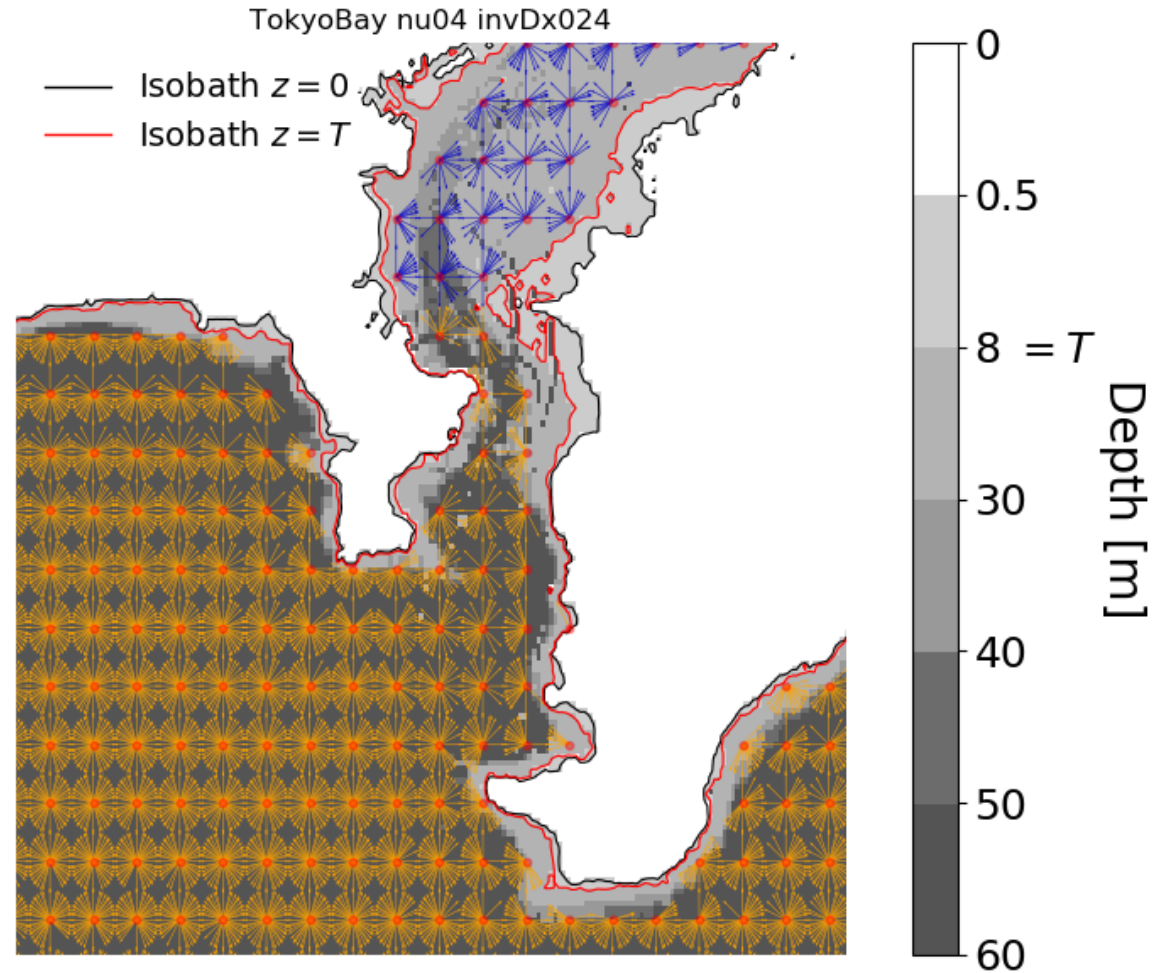


Graph edges indexed via a K -dimensional tree

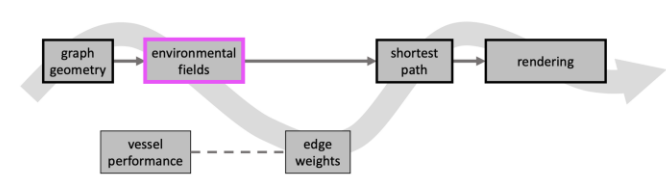
it can effectively be queried for:

- nearest neighbours (coast proximity of nodes)*
- range queries (coast intersection of edges)*

*implementation in Python:
`scipy.spatial.KDTree`*



Input metocean data

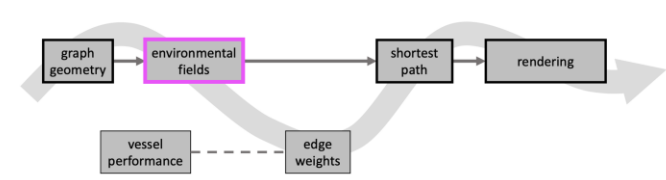


dynamic environmental fields from data-assimilative models:

<i>type</i>	<i>product</i>	<i>Spatial resolution</i>	<i>Time resolution</i>
<i>Waves</i>	<i>MED-SEA_ANALYSISFORECAST_WAV_006_017</i>	$(1/24)^0$ 2.5 miles	<i>1 hour</i>
<i>Currents</i>	<i>MEDSEA_ANALYSISFORECAST_PHY_006_013</i>	$(1/24)^0$ 2.5 miles	<i>1 hour</i>
<i>Wind</i>	<i>Set I - HRES</i>	$(1/10)^0$ 6.0 miles	<i>6 hours</i>



Sea over land

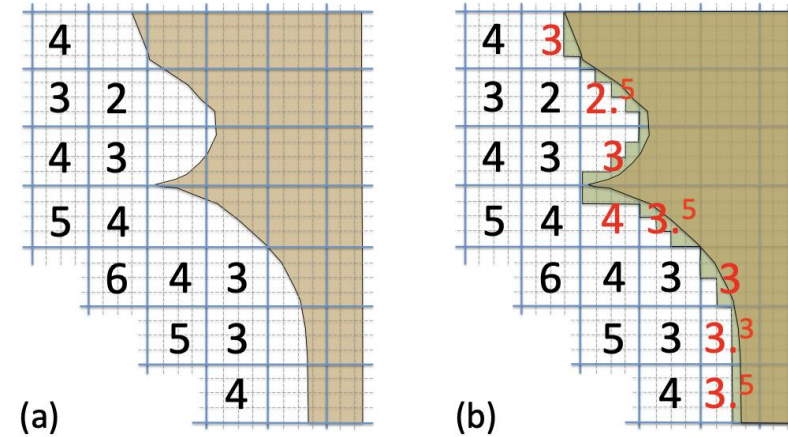


Limited spatial resolution of marine fields (waves, currents) leaves blank gridpoints in the vicinity of the physical shoreline

→ in VISIR-2, marine fields are extrapolated inshore via «sea over land»:

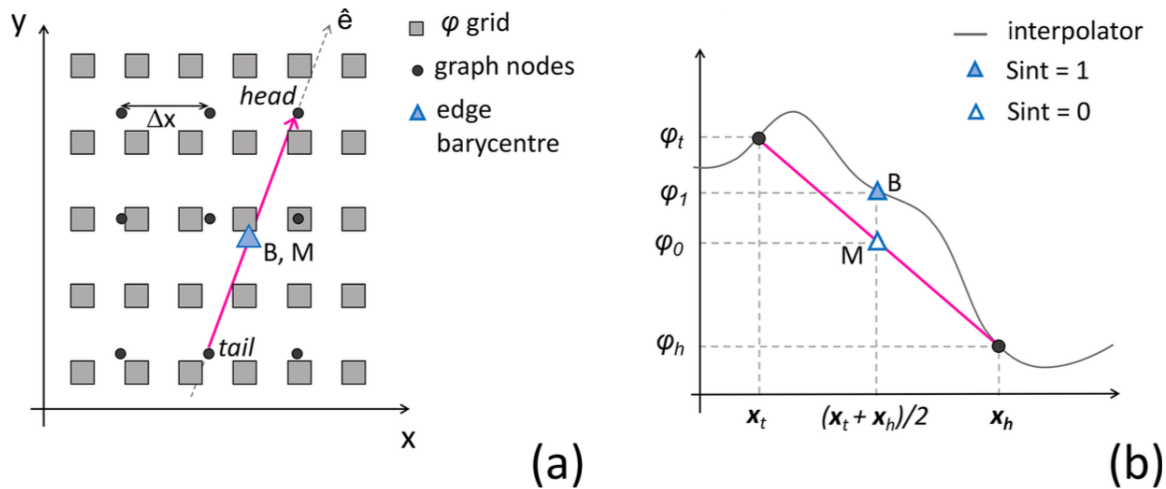
coastal missing values of sea fields are replaced with the average of the first neighbouring grid points

Note: sea over land is applied ahead of field interpolation (s. next slide)



- (a) Numbers represent original field values, with coastline (black line) and landmass (brown area)
- (b) Field values after one sea-over-land iteration (replaced missing values are printed as red numbers). The land-sea mask of the target grid is shown in greenish colour

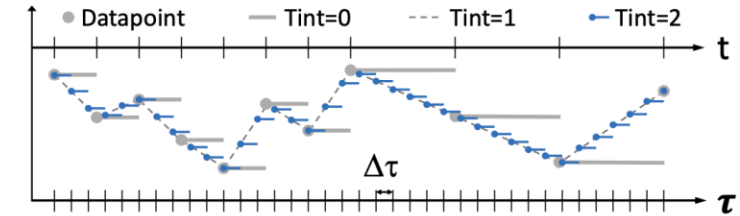
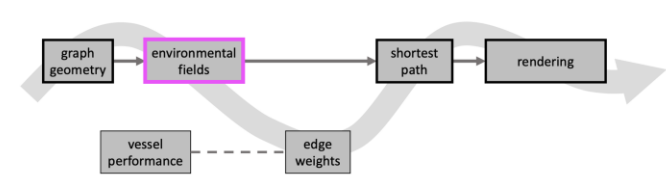
Spatial and temporal interpolation



Remap environmental fields to the graph grid:

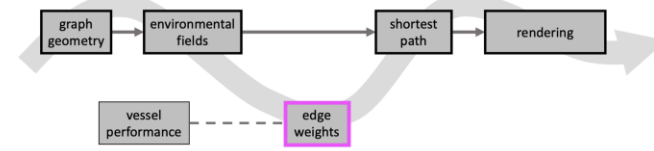
Two options:

- averaging between the edge head and tail's values ($S_{int} = 0$, default)
- interpolating their values to the edge barycentre ($S_{int} = 1$)



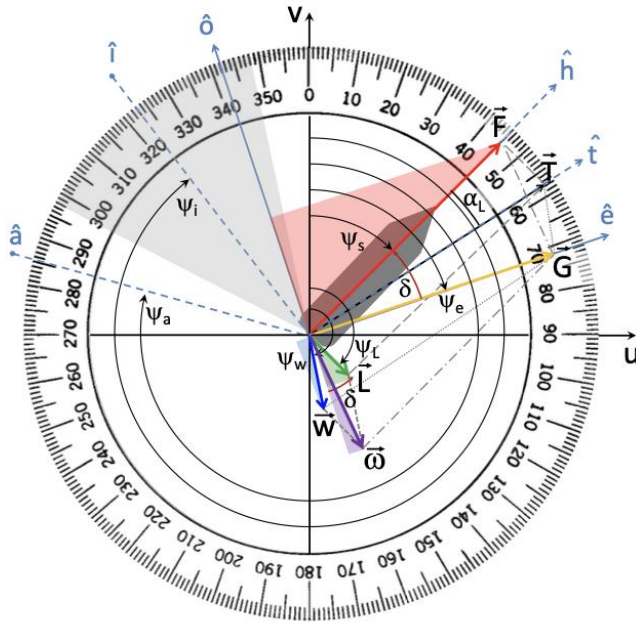
- environmental field values (grey dots) interpolated in time on a finer grid with $\Delta\tau$ spacing ($T_{int} = 2$ or blue dots)
- edge weight at the nearest available timestep (floor function used, blue segments) is selected

Kinematics



Key hypotheses:

- linear superposition of velocities (speed through water, sea currents, leeway)
- ship's motion to occur along a graph edge
- speed through water resulting from sea state only (cf. «vessel performance curve»)



F : forward speed (along vessel's heading)

T : speed through water (differs from F if leeway present)

G : speed over ground (along vessel's course – graph edge)

w : ocean current

L : leeway speed

→ **STW** as a vector sum of **F** and leeway

$$\vec{T} = \vec{F} + \vec{L}$$

→ **SOG** as a vector sum of: **STW** and **current** (no leeway)

or: **F** and **w** (in general)

$$SOG = \omega_{\parallel} + \sqrt{F^2 - \omega_{\perp}^2}$$

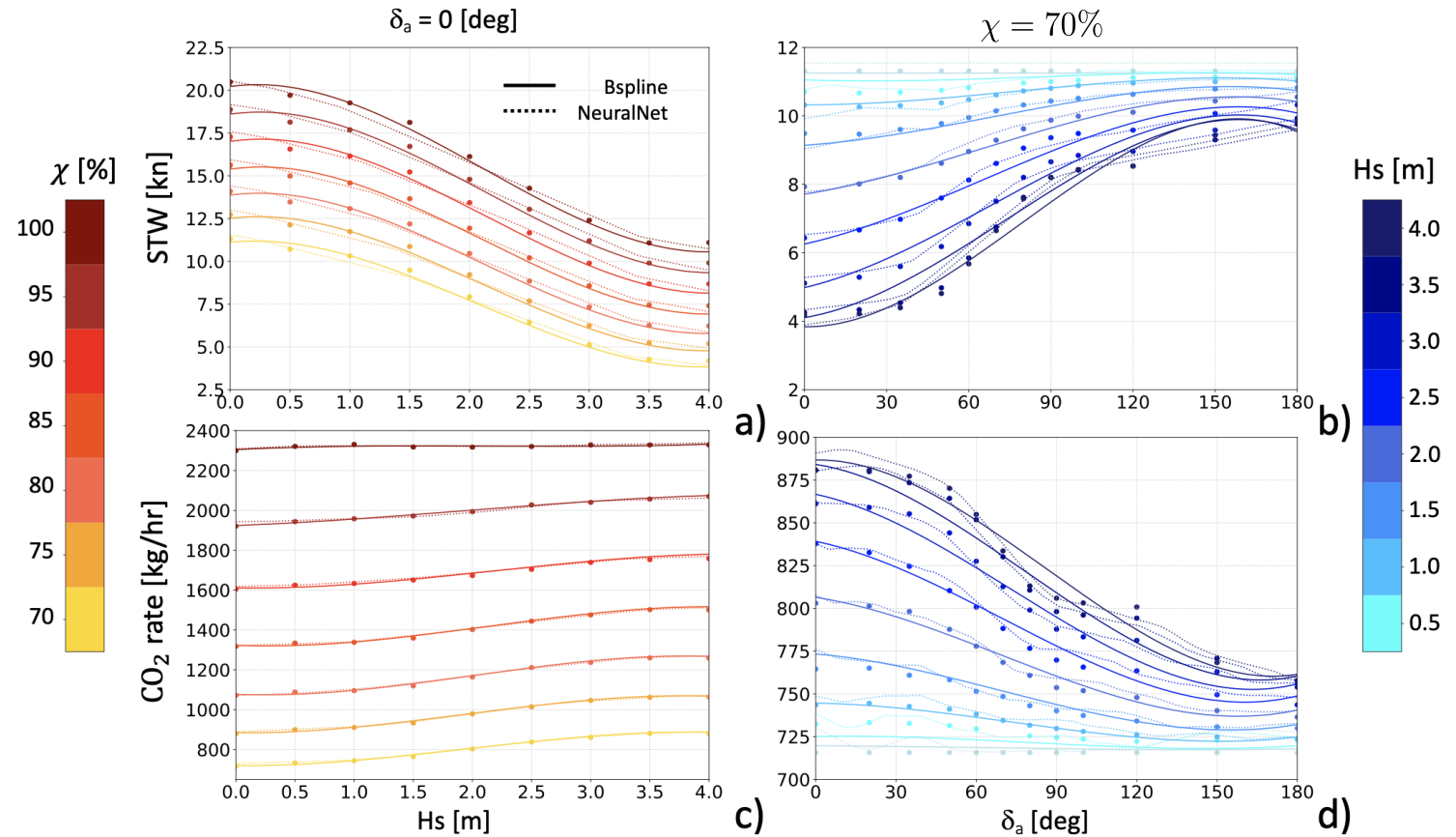
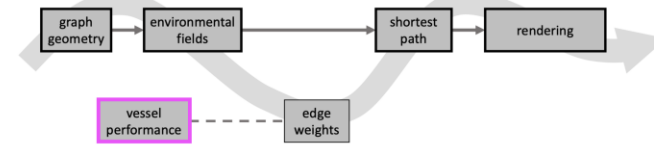
→ angle of attack δ between ship's heading and course for offsetting cross-flow

$$\delta = \psi_h - \psi_e$$

Vessel performance curve: ferry

use of University of Zadar ship command-bridge/ engine-room coupled simulator:

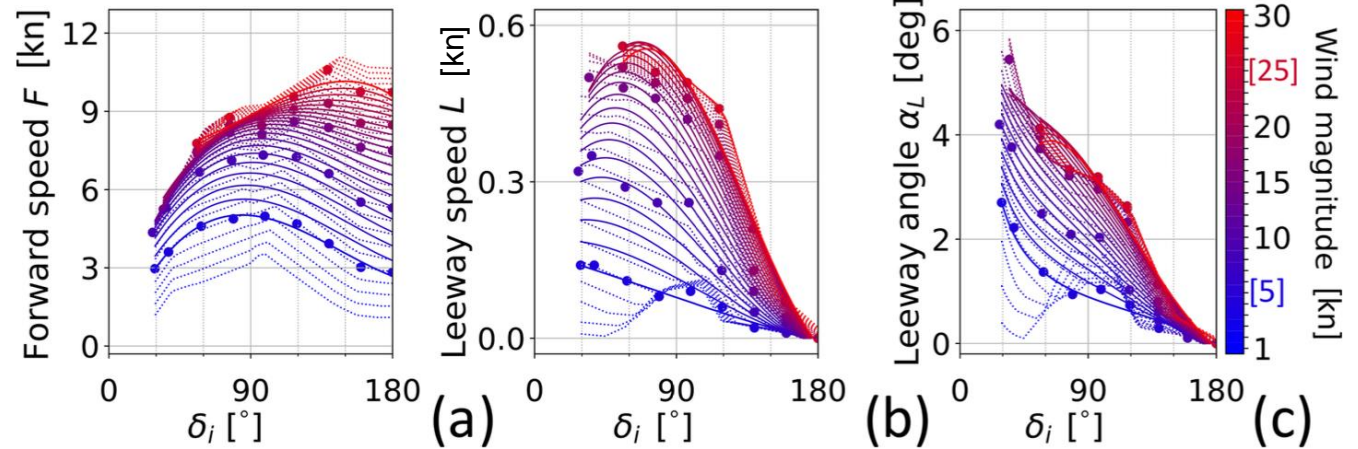
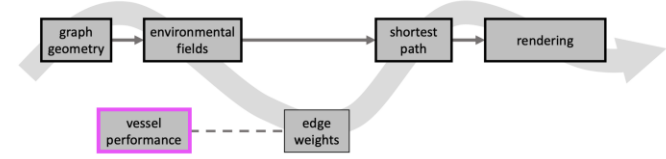
- wind waves
- no leeway
- explored dependence of **STW** on
 - engine load (χ)
 - significant wave height (H_s)
 - relative wave direction (δ_a)
- outcome interpolated via a neural network (multi-layer perceptron)



Vessel performance curve: sailboat

WinDesign velocity prediction program run at University of Genoa:

- considered wave added resistance by means of “Delft method”
- response amplitude operators derived from CFD
- choice of sail:
 - upwind: main sail and jib sail
 - otherwise: main sail and spinnaker



no-go angle varies from 27 to 53° as wind speed increases from 5 to 25 kn

forward speed **F** increases with wind intensity

peak **F** attained for broad reach ($\delta_i \approx 135^\circ$)

Leeway is max for points of sail between no-go zone and beam reach ($\delta_i = 90^\circ$)

As the point of sail from the no-go zone to running conditions, leeway angle reduces from 6 to 0°

Dynamic least-CO₂ algorithm

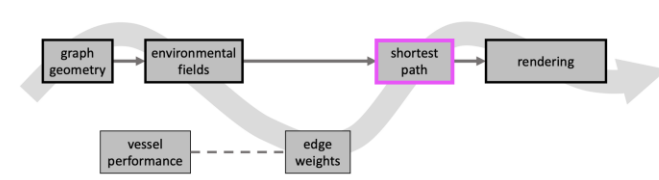
Dijkstra's algorithm generalized for dynamic edge weights

under FIFO hypothesis, same complexity of a static algorithm

built on the `single_source_Dijkstra` function of the `networkX` library

use of data structures (heaps) to achieve ideal performance

key advancement for least-CO₂ paths is accessing edge weight at a specific time step



Algorithm 2 GET_TIME_INDEX.

Require: (*paths*, *d*, *wT*, *Ntau*, *Dtau*), a dictionary of paths, node costs, type of edge weight, maximum number of time steps, and time resolution, respectively

Ensure: *t_idx*, the time step at which the costs *d* are realised along the paths

```
1: if wT = "time" then
2:   t_idx ← min(Ntau, ⌊d/Dtau⌋)
3: else
4:   # compute cTime cumulative time
5:   cTime ← 0
6:   t_idx ← 0
7:   for edge in paths do
8:     # evaluate edge delay at time step t_idx
9:     cTime ← cTime + edge.cost.at_time(t_idx, "time")
10:    t_idx ← min(Ntau, ⌊time/Dtau⌋)
11:   end for
12: end if
```

Algorithm 1 _DIJKSTRA_TDEP.

Require: (*G*, *source*, *target*, *wT*, *Ntau*, *Dtau*), a NetworkX graph, *source* and *target* nodes, type of edge weight, maximum number of time steps, and time resolution, respectively

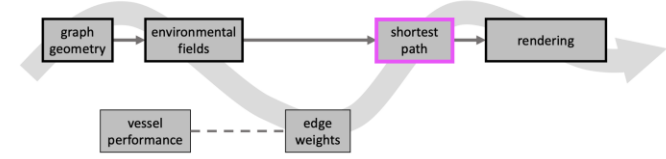
Ensure: (*costs*, *paths*), Two dictionaries keyed by node ID: path costs from the source (e.g. cumulated CO₂) and corresponding optimal paths

```
1: costs ← {}
2: seen ← { source:0 }
3: paths ← { source:[source] }
4: # fringe is a min-priority queue of (cost, node) tuples
5: fringe ← heap()
6: fringe.push(0, source)
7: while fringe ≠ ∅ do
8:   (d, v) ← fringe.pop()
9:   if v ∈ costs then
10:    # Already visited node
11:    skip
12:   end if
13:   costs[v] ← d
14:   if v = target and ∀n ∈ G.neigh(target), n ∈ seen then
15:     exit
16:   end if
17:   t_idx ← get_time_index(paths[v], d, wT, Ntau, Dtau)
18:   # Iterate on v's forward-star
19:   for (u, cost) in G.succ(v) do
20:     # evaluate edge weight of wT type at time step t_idx
21:     c ← cost.at_time(t_idx, wT)
22:     vu_cost ← costs[v] + c
23:     if u ∉ seen or vu_cost < seen[u] then
24:       seen[u] ← vu_cost
25:       fringe.push(vu_cost, u)
26:       paths[u] ← paths[v] + [u]
27:     end if
28:   end for
29: end while
```

Validation

VISIR-2 routes and metrics were compared to:

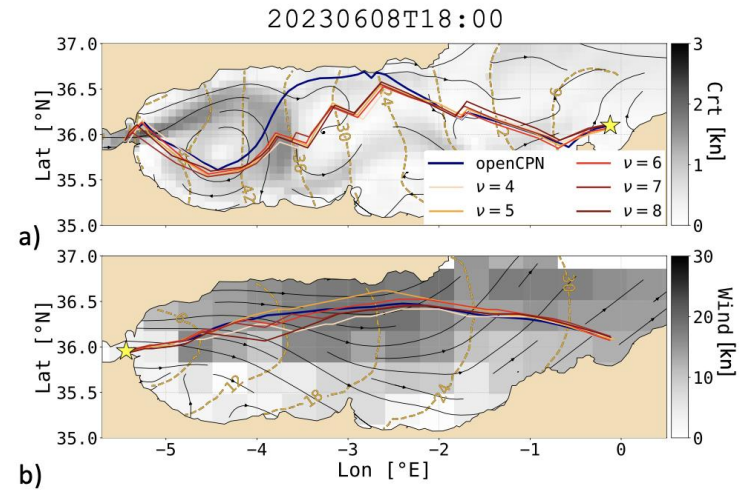
- semi-analytical results (cycloid, Techy)
- MIT model based on partial differential equations (LSE, *)
- openCPN (dynamic programming)



Oracle	ν	$1/\Delta x$ 1/°	$\Delta \tau$ min	L_0 nmi	T_0 h	$T^{(e)}$ T_0	T^* T_0	dT^* %
Cycloid	2	60	5				1.738	0.691
	5	60	5	56.38	2.672	1.726	1.726	0.012
	10	50	5				1.732	0.342
Techy	5	25	5				1.057	0.076
	5	100	5	140.11	6.640	1.056	1.046	-0.956
	10	50	5				1.050	-0.599

1 nmi = 1852 m.

Model	ν	$1/\Delta x$ 1/°	$\Delta \tau$ min	T^* h	dT^*_{120} %	dT^*_{240} %
VISIR-1.b	6	129	-	13.73	-1.58	-0.43
VISIR-2	6	129	30	13.62	-2.36	-1.23
VISIR-1.b	3	134	-	13.79	-1.12	0.03
VISIR-2	3	134	30	13.71	-1.73	-0.59
VISIR-1.b	2	142	-	13.90	-0.36	0.79
VISIR-2	2	142	30	13.85	-0.74	0.42
LSE	-	120	-	13.95	-	-
	-	240	-	13.79	-	-

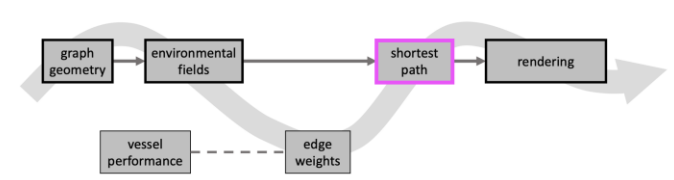


Version	ν	$1/\Delta x$ [1/°]	Wind				Current + wind			
			T^* [h]	dT^* [%]	T^* [h]	dT^* [%]	T^* [h]	dT^* [%]	T^* [h]	dT^* [%]
VISIR-2	4	12	51.9	3.4	34.4	-0.4	54.0	-2.6	32.2	0.0
	5	15	52.0	3.5	34.5	-0.2	53.9	-2.7	31.7	-1.5
	6	18	51.2	1.9	33.6	-2.9	53.4	-3.7	30.9	-3.9
	7	21	50.7	1.0	32.8	-5.0	52.8	-4.8	30.9	-4.1
	8	23	51.0	1.6	32.8	-5.0	53.1	-4.2	30.8	-4.5
OpenCPN			50.2		34.6		55.4		32.2	

*) Mannarini et al 2019, doi.org/10.1109/TITS.2019.2935614



Numerical performance: optimal paths



Three variants of the algorithm:

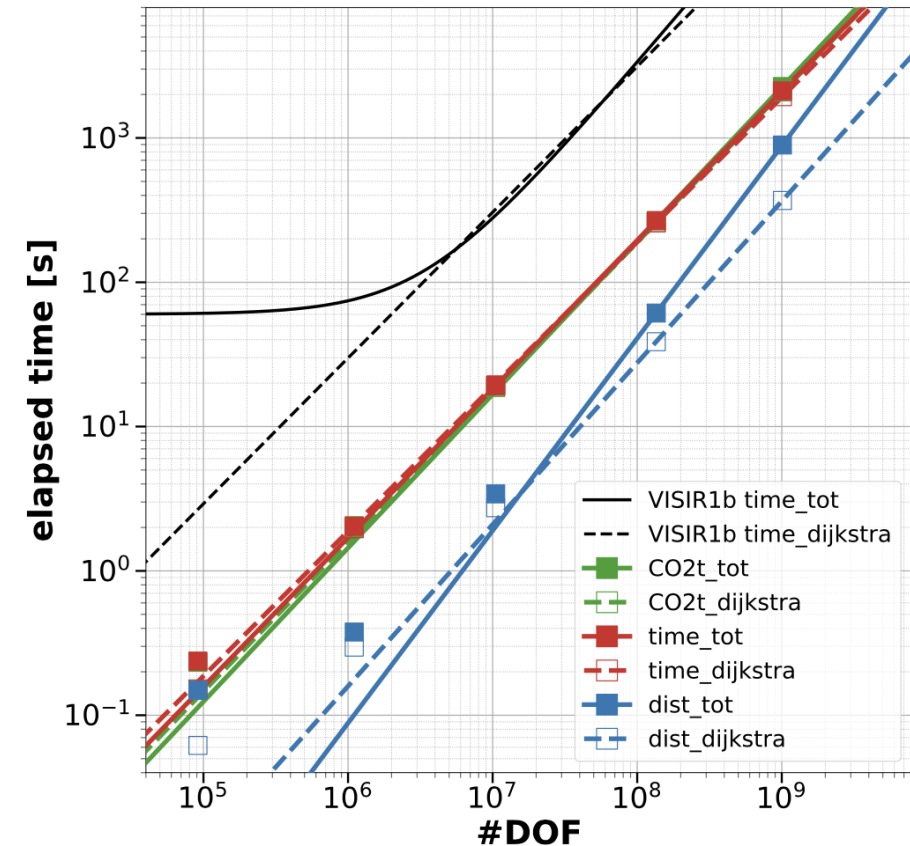
- *least-distance*
- *least-time*
- *least-CO₂*

Assessment for:

- *"Dijkstra": optimal sequence of graph nodes*
- *"total": "Dijkstra" + marine and vessel dynamical information along the path*

Outcome:

- *linearity in the number of DOF*
- *10x faster than VISIR-1*
- *least-distance routine still to be improved*
- *RAM: 420B per DOF (5x more than VISIR-1, to be improved e.g via single precision)*



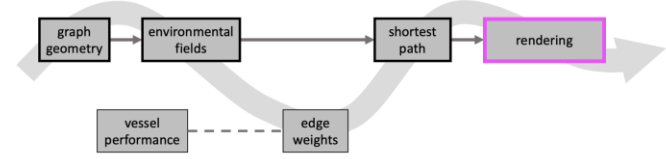
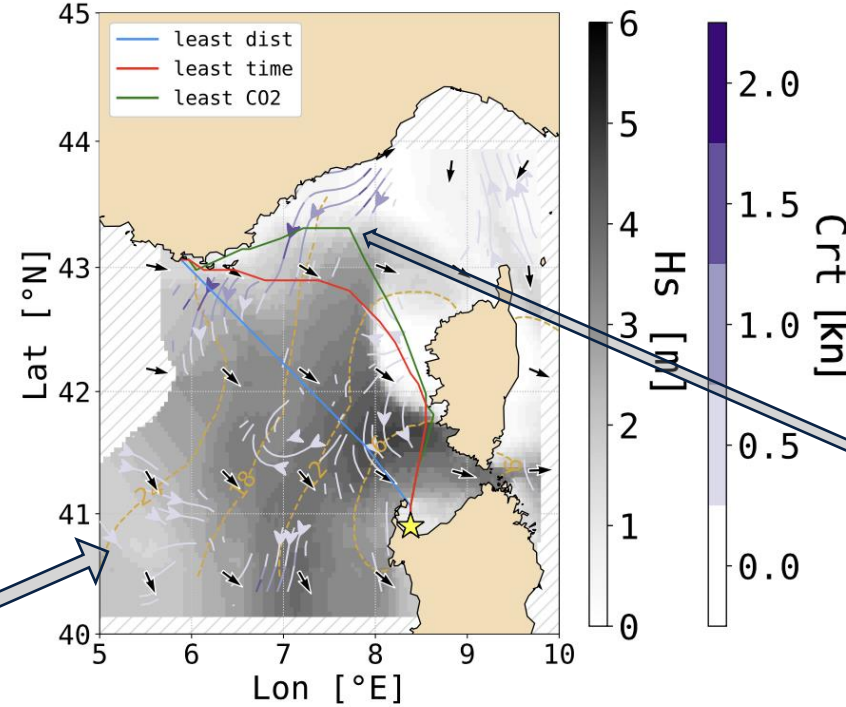
Visualisation

dynamic environmental fields rendered via

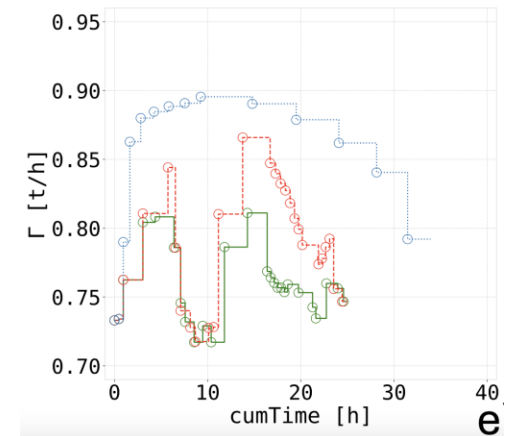
- concentric shells originating at the departure location
- shape of shells defined by isochrones

saving of 1 dimension (can be used for departure date or engine load)

20220109T03:00 -dCO2 = 35.1 [%]

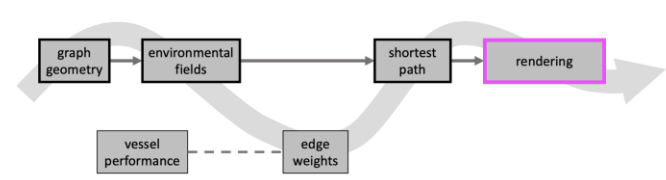


least-CO2 route takes a larger diversion than least-time to sail where emission rate is lower:

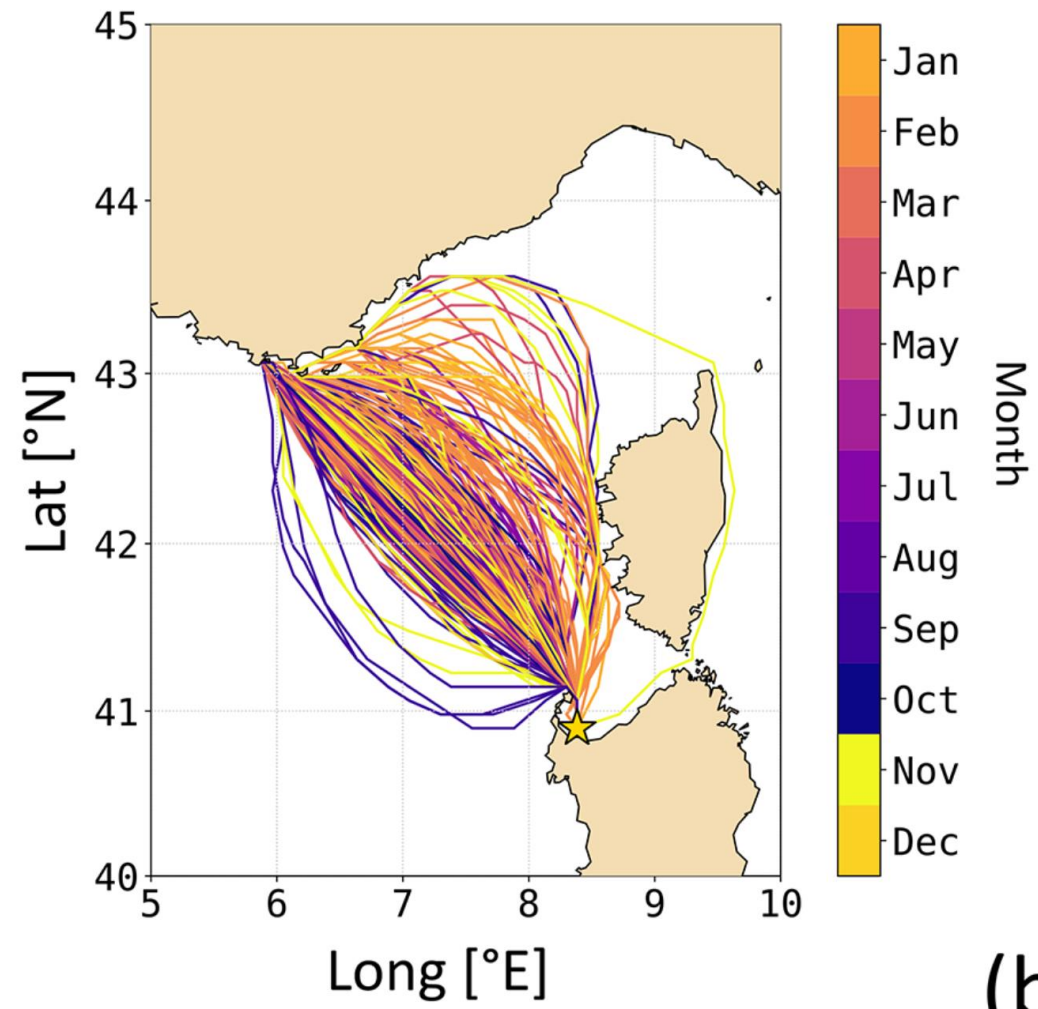


<i>type</i>	<i>meaning</i>	<i>bulging</i>
<i>isometres</i>	<i>equal distance</i>	<i>at obstructions (shoals, islands, landmass in general)</i>
<i>isochrones</i>	<i>equal duration</i>	<i>against gradients of 1/STW</i>
<i>isopones</i>	<i>equal emissions</i>	<i>against gradients of emissions</i>

Visualisation



#routes = 1460



(b)

bundles of least-CO₂ routes:

- *route colour by departure month*
- *seasonal dependence*
- *topological change is possible*

Outline

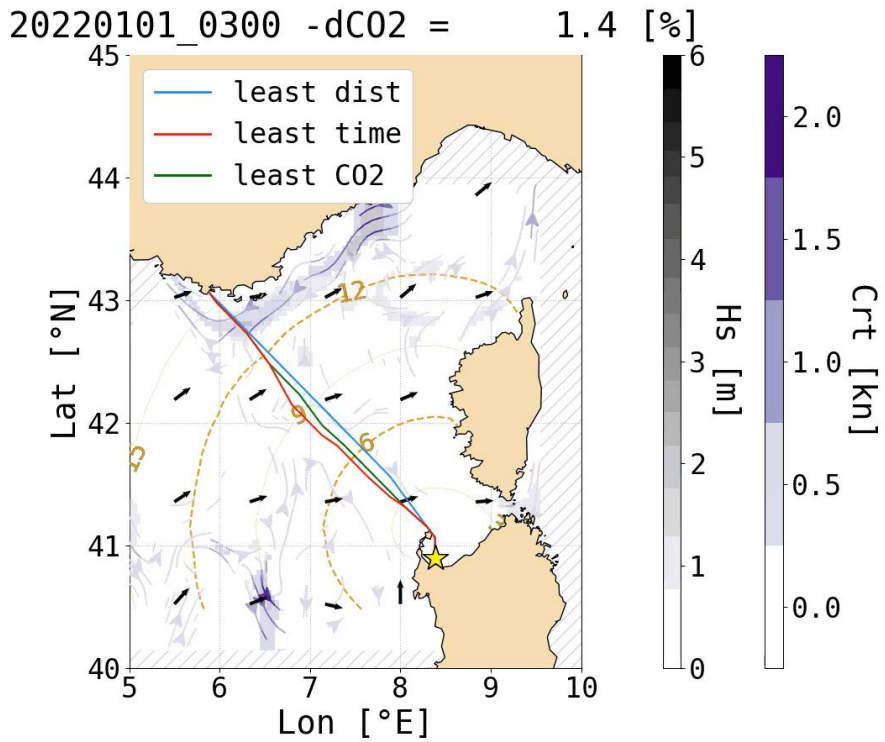
1. *Introduction*
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Case study: ferry

dynamic environmental fields rendered via

- concentric shells originating at the departure location
- shape of shells defined by isochrones

saving of 1 dimension (can be used for departure date or engine load)

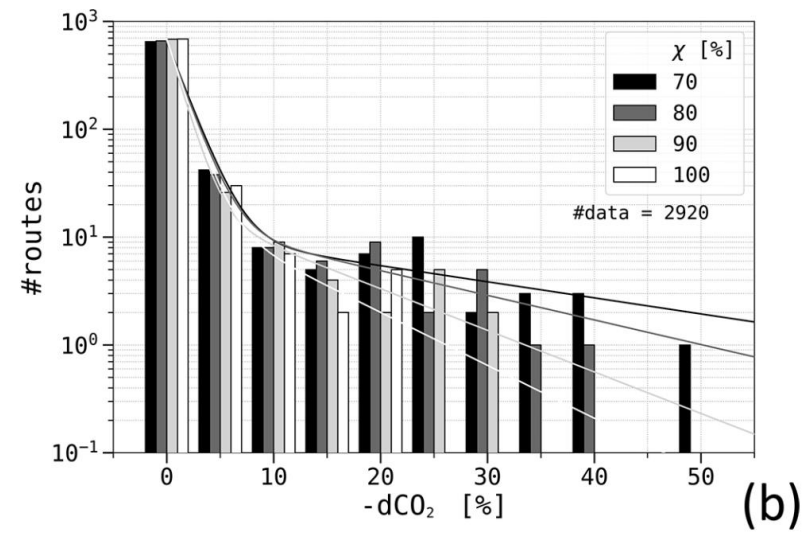
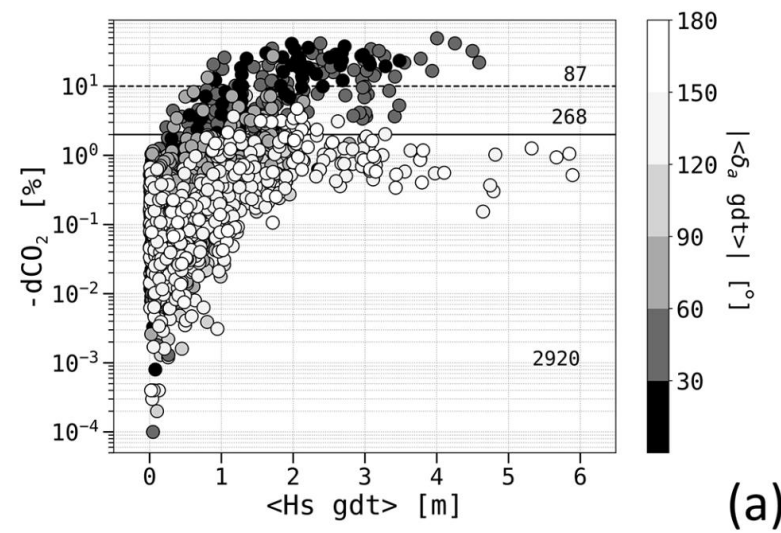


movies with 1-year of daily departures : https://doi.org/10.5446/s_1687

<i>type</i>	<i>meaning</i>	<i>bulging</i>
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<i>isochrones</i>	<i>equal duration</i>	<i>against gradients of 1/STW</i>
<i>isopones</i>	<i>equal emissions</i>	<i>against gradients of emissions</i>

Case study: ferry

Statistics of CO₂ savings in 2022



	upwind					downwind				
	ITPTO-FRTLN χ [%]					FRTLN-ITPTO χ [%]				
	70	80	90	100	Avg	70	80	90	100	Avg
wa	2.9	2.2	1.4	1	1.9	0.9	0.6	0.4	0.3	0.6
wa-cu	3.4	2.5	1.7	1.2	2.2	1.4	1.2	0.7	0.6	1

- largest savings are upwind
- currents increase savings, especially downwind

- wave height alone is not the key to minimal savings
- key is angle of attack of waves
- > 2% for beam or head seas
- >10% for about 10 days in a year

- bi-exponential distribution
- larger decay length inversely proportional to engine load χ
- tail can extend to values ranging between 25 and 50%

Case study: sailboat



movies with 1-year of daily departures :
https://doi.org/10.5446/s_1688

Geography

- *Meltemi wind*
- *Asia minor current*
- *archipelagic domain*

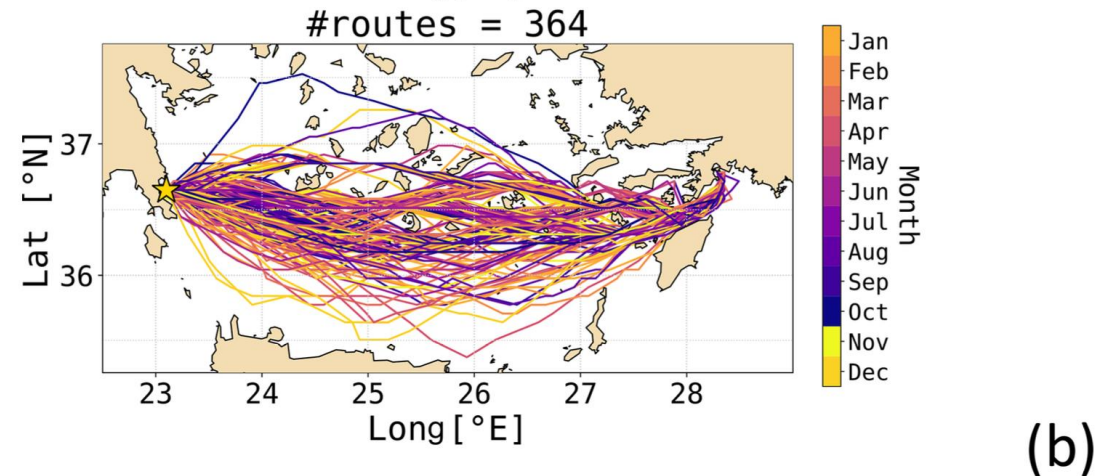
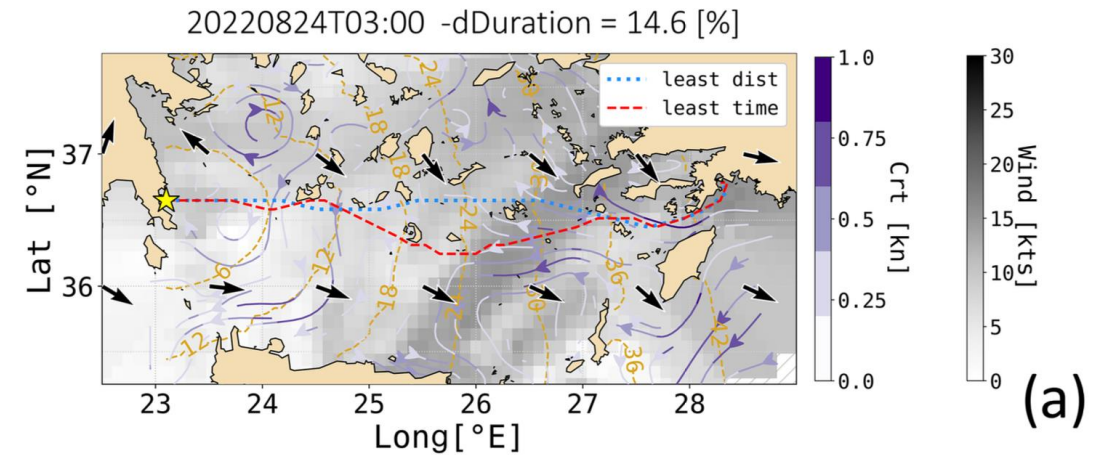
Name	Symbol	Value	Units
Length of hull	L_{hull}	10.7	m
Draft	T	2.2	m
Displacement volume	∇	5.8	m ³
Rudder wetted surface	–	1.4	m ²
Keel wetted surface	–	3.3	m ²
Main sail area	–	38	m ²
Jib sail area	–	4	m ²
Spinnaker area	–	95	m ²

Numerical experiments

- *graph with $(v, 1/\Delta x) = (5, 15/^\circ)$*
- *daily departures, two orientations, with/without currents or leeway (2,920 runs)*
- *7 min/run*

Outcome

- *large diversions to avoid upwind sailing*
- *no clear seasonal trend for diversions*



Case study: sailboat

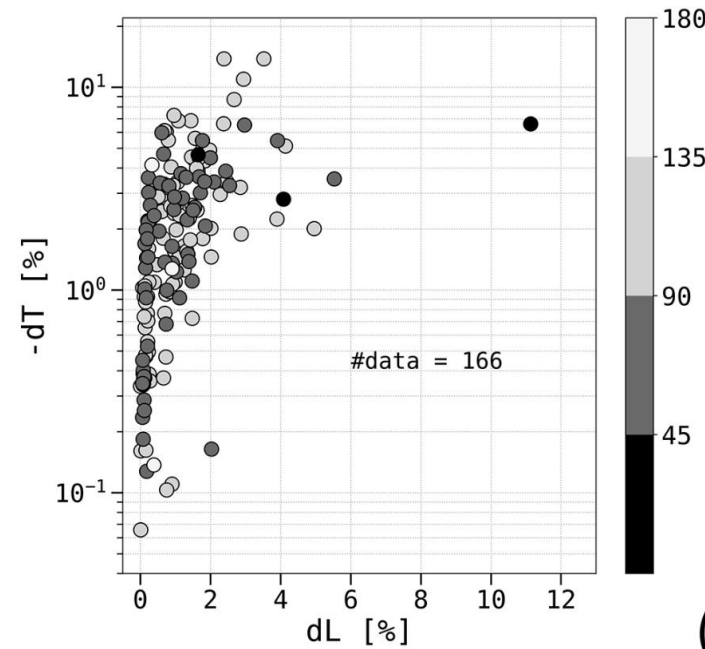
Statistics of time savings in 2022

downwind
against current

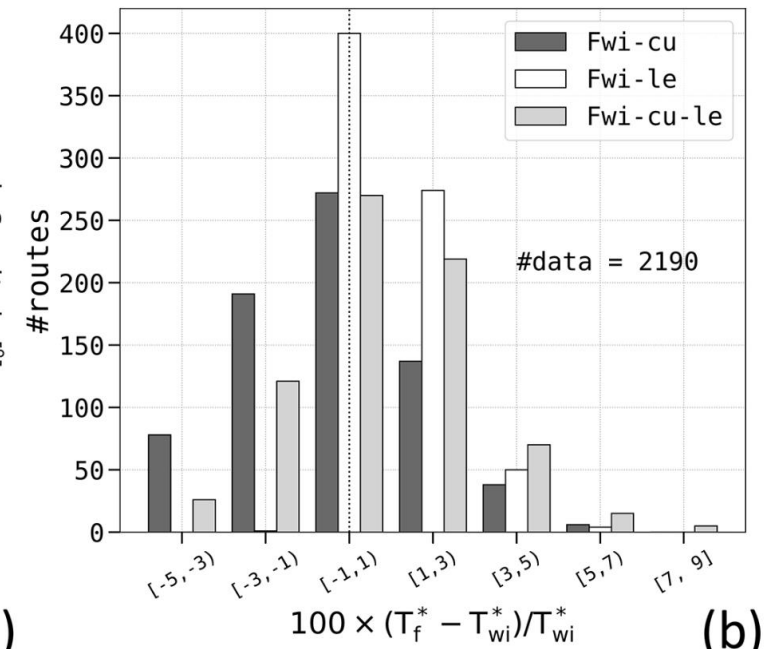
upwind
with current

	GRMON-TRMRM			TRMRM-GRMON		
	$-dT^*$	$N_f^{(g)}$	$N_f^{(o)}$	$-dT^*$	$N_f^{(g)}$	$N_f^{(o)}$
wi	2.5	256	1	2.3	308	1
wi-le	2.3	275	1	2.4	328	3
wi-cu	3.1	254	1	3.1	320	1
wi-cu-le	3.2	267	0	3.5	326	6

- consideration of currents increases duration percentage savings
- leeway results in more failed routes (due to vessel manoeuvrability)



(a)



(b)

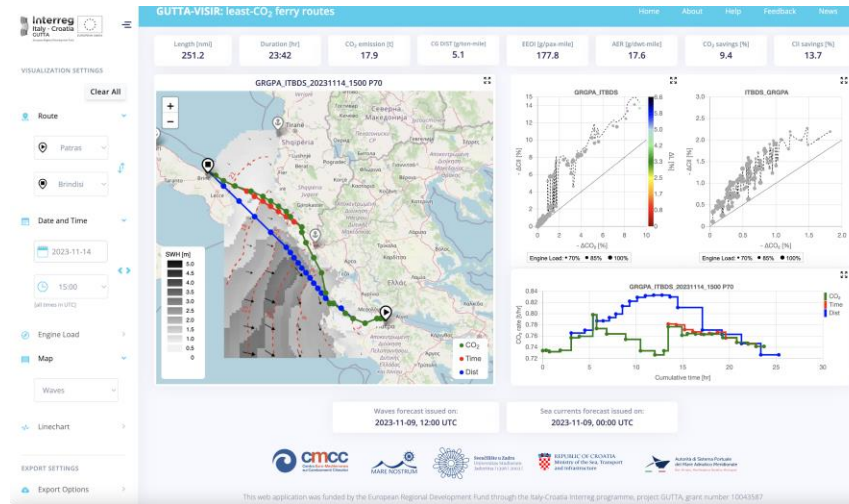
- time saving increases with spatial diversion
- max diversions when skipping upwind along geodetic route

- currents results in a change in duration (up to about 5%, skewed to reduction)
- leeway: consistently extends the duration (leeway cross-course component reduces SOG)

Operational services based on VISIR-2

GUTTA-VISIR

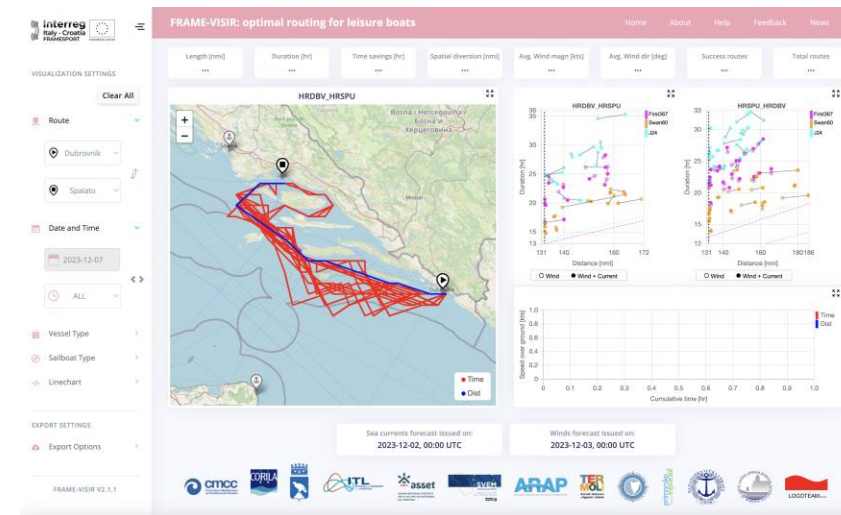
<https://www.gutta-visir.eu>



- *ferry routes*
- *CO2 and CII savings*
- *cross-border routes*
- *7,020 routes/day*
- *operational since Oct.2021*

FRAME-VISIR

<https://www.frame-visir.eu>



- *sailboat and flybridge routes*
- *time and CO2 savings*
- *cross-border and cabotage routes*
- *5,280 routes/day*
- *operational since May 2023*

Outline

1. *Introduction*
2. *Model workflow*
3. *Case studies*
4. *Discussion*

Results



- ✓ *VISIR-2: a modular, validated, documented, and portable model for ship weather routing*
- ✓ *for vessels with an angle-dependent performance curve, an improved level of accuracy in the velocity composition with sea currents*
- ✓ *a variant of the Dijkstra's algorithm developed (minimising not just the CO₂ emissions but any figure of merit depending on dynamic edge weights)*
- ✓ *quasi-linear computational performance up to 1 billion DOF*
- ✓ *10x faster than VISIR-1*
- ✓ *Bi-exponential distribution of CO₂ savings found for a ferry, with savings at times exceeding 10%*
- ✓ *sailboat routes: duration savings of about 3% , neglecting leeway wrongly underestimates route durations*

Novelty



- ❖ *wave, winds, and currents addressed through an unified scheme*
- ❖ *role of currents assessed*
- ❖ *versatile least-CO₂ algorithm (can become a least-X algorithm)*
- ❖ *shape of statistical distribution of CO₂ savings assessed*
- ❖ *consideration of both currents and leeway in the optimization of sailboat routes*
- ❖ *visualisation making use of isochrone-bounded sectors*

VISIR-2 resources

open access - open review journal paper

<https://doi.org/10.5194/gmd-17-4355-2024>



Geosci. Model Dev., 17, 4355–4382, 2024
<https://doi.org/10.5194/gmd-17-4355-2024>
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Geoscientific
Model Development
Open Access
EGU

Model description paper



VISIR-2: ship weather routing in Python

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open language - open source model code

<https://zenodo.org/communities/visir2>



April 11, 2024 (v3) Software Open

[VISIR-2 ship weather routing model] source code (Python)

Salinas, Mario Leonardo ; Carelli, Lorenzo ; Mannarini, Gianandrea 

Source code of the VISIR-2 ship weather routing model. The Python-refactored VISIR-2 model considers currents, waves, and wind to optimise routes. The model was validated, and its computational performance is quasi-linear. For a ferry sailing in the Mediterranean Sea, VISIR-2 yields the largest percentage emission savings for upwind navigation...

Part of VISIR-2 ship weather routing model (Python)

Uploaded on April 11, 2024

2 more versions exist for this record



1016 131



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