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VISIR-2 ship weather routing model: an introduction

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Outline

Introduction Numerical features Vessel performance curves Computational performance Validation Case studies Operational service Discussion



Motivation

New (IMO MEPC-80) decarbonisation strategy

- 2030: uptake of low-carbon fuels (5-10%) ٠
- 2050: a zero-carbon shipping

EU-ETS for shipping

- all calls at EU ports included
- starts 2024, progressive application ٠
- CO2, CH4, N20 •

MEPC 80

role of weather routing

- saving money ٠
- saving emissions
- hardly quantified so far, open models needed ٠









[1] https://www.imo.org/en/MediaCentre/PressBriefings/pages/Revised-GHG-reduction-strategy-for-global-shipping-adopted-.aspx [2] https://climate.ec.europa.eu/eu-action/transport/reducing-emissions-shippingsector_en#:~:text=Inclusion%20of%20maritime%20emissions%20in,of%20the%20flag%20they%20fly.



VISIR-2 resources

open access – open review manuscript

https://egusphere.copernicus.org/preprin ts/2023/egusphere-2023-2060/ https://doi.org/10.5194/egusphere-2023-2060 Preprint. Discussion started: 16 November 2023 © Author(s) 2023. CC BY 4.0 License.



VISIR-2: ship weather routing in Python

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open language - open source model code

https://zenodo.org/records/8305527





Kinematics: geometry



F: forward speed (along vessel's heading)
STW: speed through water (differs from F if leeway present)
SOG: speed over ground (along vessel's course – graph edge)
C: ocean current

L: leeway velocity

Key hypotheses:

- linear superposition of velocities (STW, sea currents, leeway)
- *ship's motion to occur along a graph edge*

 \rightarrow angle of attack δ between ship's heading and course

$$\delta = \psi_s - \psi_e$$

- → SOG as a vector sum of:
 STW and C (no leeway) or:
 F and ω (general)
- $S_g = F\cos(\delta) + \omega_{\parallel}$ $0 = -F\sin(\delta) + \omega_{\perp}$



Kinematics: angle of attack

as F depends on environmental field's angle,

need to solve a transcendental equation for angle of attack δ :

$$\sin \delta = \frac{\omega_{\perp}(\delta, \delta_i(\delta))}{F(|\delta_i(\delta))|, |\delta_a(\delta)|)} \quad \Leftrightarrow \quad F \neq 0$$

solve numerically (scipy.optimize.root, not vectorizable) or via iteration (vectorizable):



$$egin{array}{rcl} \delta^{(0)} &=& 0 \ \delta^{(k)} &=& h(\delta^{(k-1)}) & \mbox{for} & k=1,2,\ldots \end{array}$$

$$h(x) = \arcsin\left(rac{\omega_{\perp}(\delta = x, \delta_i = x - \gamma)}{F(|x - \gamma|)}
ight)$$



Graph structure

Graph stencil for connectivity

up to 4th order neiabhours (v=4)



benefits of pruning of collinear edges (dashed) :

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

connectivity and number N_{q_1} of edges in the first quadrant

u	$\Delta heta$ [°]	N_{q1}	$\nu(\nu+1)$
1	45.0	2	2
2	26.6	4	6
3	18.4	8	12
4	14.0	12	20
5	11.3	20	30
6	9.5	24	42
7	8.1	36	56
8	7.1	44	72
9	6.3	56	90
10	5.7	64	110



Graph computation

indexing via a K-dimensional Tree

a spatial data structure which can effectively be queried for:

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

implementation in Python: scipy.spatial.KDTree





Pseudo-shoreline

pseudo-shoreline: avoiding too shallow water for a given vessel

high resolution bathymetry dataset → compute as a zero contour line of under-keel clearance:

UKC = z - T

retain in the graph just edges with UKC > 0



- GSHHG "high" res shoreline: 200m
- *GEBCO_2022* bathy : 463 m
- EMODnet bathy: 116 m



Space and Time interpolation





Remap environmental fields to the graph grid: Two options:

- averaging between the edge head and tail's values ("Sint = 0")
- interpolating their values to the 315 edge barycentre ("Sint = 1", default)

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



VISIR-2 suite modules

Greater modularity with respect to VISIR-1

facilitating both R&D and operational applications

conda virtual environment ("visirvenv") for portability

Run	4	MAIN_Tracce ×
	\uparrow	/Users/gmannarini/opt/anaconda3/envs/visir-venv-gmd2023/bin/python /Users/gmannar
r	\downarrow	Choose an option from the list below
	=	or type Q to quit.
	-1	
==		n. Tracce namelist
	÷	
~	Î	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] X 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



https://zenodo.org/records/8305527



Vessel performance curves: ferry

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

- wind waves
- no leeway
- explored dependence of STW on
 - engine load
 - significant wave height
 - relative wave direction
- outcome interpolated through a neural network (multi-layer perceptron via the scikit-learn package)







Name	Symbol	Value	Units
Length overall	LOA	125	m
Draft middle	T	5.3	\mathbf{m}
Deadweight	DWT	4,050	\mathbf{t}
Main engine power	P_{main}	4,000	kW
Main engine rated speed	$n_{ m eng}$	750	rpm
Service speed	v_S	19	\mathbf{kn}



Mannarini et al 2021, https://doi.org/10.3390/jmse9020115



Vessel performance curves: sailboat

Symbol Value Units Name Length of hull L_{hull} 10.68 m Draft T2.2 m m^3 Displacement ∇ 5,773 m^2 Rudder wetted surface 1.42 _ m^2 3.31 Keel wetted surface _ m^2 Main sail area 38 m^2 Jib sail area 3.97 Spinnaker area 95 m^2 -



use of WinDesign Velocity Prediction Program:

- both hydrodynamic and aerodynamic effects
- wave added resistance via "Delft method" on DSYHS series
- same wind-wave relationship of the ferry used
- for upwind, main sail and jib assumed; otherwise: main sail and spinnaker

Claughton et al 1999, 2003

Shortest path problem: least-CO₂ algorithm

Dijkstra's algorithm generalized for dynamic edge weights

same complexity of static algorithm under FIFO hypothesis

built on single_source_Dijkstra function of the networkX library

use of data structures (heaps) to achieve ideal performance

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

Algorithm 2 GET_TIME_INDEX

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

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

1: if wT = "time" then

- 2: $t_idx \leftarrow min(Ntau, \lfloor d/Dtau \rfloor)$
- 3: else
- 4: # compute cTime cumulative time
- 5: $cTime \leftarrow 0$
- 6: $t_idx \leftarrow 0$
- 7: for edge in paths do
- 8: # evaluate edge delay at time step t_idx
- 9: $cTime \leftarrow cTime + edge.cost.at_time(t_idx, "time")$
- 10: $t_idx \leftarrow min(Ntau, \lfloor time/Dtau \rfloor)$
- 11: end for
- 12: end if

Algorithm 1 _DIJKSTRA_TDEP

- Input: (G, source, target, wT, Ntau, Dtau), respectively a networkX graph, source and target nodes, type of edge weight, maximum number of timesteps, and time resolution
 Output: (costs, paths), Two dictionaries keyed by node id: path costs from the source (e.g. cumulated CO₂), and corresponding optimal paths
- 1: $costs \leftarrow \{\}$
- 2: $seen \leftarrow \{source: 0\}$
- 3: $paths \leftarrow \{source : [source]\}$
- 4: # fringe is a min-priority queue of (cost, node) tuples
- 5: $fringe \leftarrow heap()$
- 6: fringe.push(0, source)
- 7: while $fringe \neq \emptyset$ do
- 8: $(d,v) \leftarrow fringe.pop()$
- 9: **if** $v \in costs$ **then**
- 10: # Already visited node
- 11: skip
- 12: end if
- 13: $costs[v] \leftarrow d$
- 14: if v = target and $\forall n \in G.neigh(target), n \in seen$ then
- 15: exit
- 16: end if
- $17: \quad t_idx \leftarrow get_time_index(paths[v], d, wT, Ntau, Dtau)$
- 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: $c \leftarrow cost.at_time(t_idx, wT)$
- 22: $vu_cost \leftarrow costs[v] + c$
- 23: if $u \notin seen \text{ or } vu_cost < seen[u]$ then
- 24: $seen[u] \leftarrow vu_cost$
- 25: $fringe.push(vu_cost, u)$
- 26: $paths[u] \leftarrow paths[v] + [u]$
- 27: end if
- 28: end for
- 29: end while

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)*

Validation

VISIR-2 routes and metrics were compared to

• *MIT model based on partial differential* equations (*LSE*, *)

benchmark	ν	$1/(\Delta x)$	$\Delta \tau$	L_0	T_0	ref_time	V2_time	rel_err
	-	$1/^{\circ}$	min	nmi	hr	T_0	T_0	%
LSE	2	94	30	126.5	7.809	1.762	1.773	0.617
LSE	3	134	30	126.5	7.809	1.766	1.753	-0.774
Techy	5	25	5	140.1	6.640	1.056	1.0563	0.028

- *semi-analytical results (cycloid, Techy)*
- openCPN (dynamic programming)

				wind					current	+ wind	
				Westl	oound	Eastb	ound	Westl	oound	Eastb	ound
version	ν	$1/\Delta x$	$\Delta \Theta$	T^*	dT^*	T^*	dT^*	T^*	dT^*	T^*	dT^*
		[1/deg]	[deg]	[hr]	[%]	[hr]	[%]	[hr]	[%]	[hr]	[%]
VISIR-2	4	12	14	34.6	0.2	57.7	4.0	57.7	4.0	32.3	0.2
	5	15	11	34.5	0.0	57.2	3.2	57.2	3.2	31.6	-1.9
	6	18	9	33.4	-3.4	56.4	1.8	56.4	1.8	31.0	-3.7
	7	21	8	32.9	-4.7	55.4	-0.1	55.4	-0.1	30.8	-4.3
	8	23	7	32.9	-4.7	56.2	1.3	56.2	1.3	30.9	-4.0
openCPN				34.6		55.4		55.4		32.2	
	-							\sim			

*) Mannarini et al 2019, <u>doi.org/10.1109/TTTS.2019.293561</u>2

(bug in VISIR-2 manuscript's Tab.6)

Visualization

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)

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

Marine forecast data

dynamic environmental fields from data-assimilative models

type	product	Spatial resolution	Time resolution
Waves	MED- SEA_ANALYSISFORECA ST_WAV_006_017	(1/24)º 2.5 miles	1 hour
Currents	MEDSEA_ANALYSISFOR ECAST_PHY_006_013	(1/24)º 2.5 miles	1 hour
Wind	Set I - HRES	(1/10)º 6.0 miles	3 hours

Case study: ferry

Geography

- Mistral wind
- Liguro-Provençal current •

Numerical experiments

- *graph with* $(v, 1/\Delta x) = (4, 12/^{\circ})$ •
- daily departures, 3 engine loads, two • orientations, with/without currents (5840 runs)
- 4 min/run •

Outcome

- large diversions to avoid upwind sailing • and exploit currents
- *two-digit CO2 savings possible* •
- bundle of optimal solutions shifts N-E in • winter

One year of routes – video:

Case study: ferry

Statistics of **CO2** savings in 2022

	upwind						downwind			
	ITPTO - FRTLN						FRTLN - ITPTO			
	χ [%]					χ [%]				
	70	80	90	100	avg	70	80	90	100	avg
wa	3.1	2.3	1.5	1	2.0	0.9	0.6	0.4	0.3	0.7
wa-cu	3.7	2.8	1.9	1.3	2.5	1.2	0.9	0.6	0.5	0.9

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

- increase in wave height can lead to either substantial or minimal savings
- key is angle of attack of waves
- > 2% for beam or head seas
- >10% once a month, on average

- 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

Geography

- Meltemi wind
- Asia minor current
- archipelagic domain

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

	dou	vnwinc	ł	upwind			
	agains	t curre	ent	with	curren	t	
	GRMC	ON - TRI	MRM	TRMR	M - GR	MON	
	$-dT^*$	$dT^* N_f^{(g)}$		$-dT^*$	$N_f^{(g)}$	$N_f^{(o)}$	
wi	3.0	263	1	3.0	300	1	
wi-le	3.0	274	4	3.1	315	4	
wi-cu	3.2	262	1	3.6	303	2	
wi-cu-le	-le 3.4 273 1		1	3.2	320	6	

- largest savings from currents when along sailing direction
- savings from leeway in downwind routes only thanks avoidance of speed loss along geodetic

- time saving increases with spatial diversion
- max saving for skipping upwind conditions along geodetic route

- currents results in a change in duration (slower/faster) up to about 5%
- *leeway consistently extends the duration of routes (its cross-course component reduces SOG)*

Operational service: GUTTA-VISIR

https://www.gutta-visir.eu

Operational service: GUTTA-VISIR (video tutorial)

https://www.youtube.com/watch?v=-qORsU-Jh_8&t=4s

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
- ✓ variant of the Dijkstra's algorithm developed (minimise not just the CO2 emissions but any figure of merit depending on dynamic edge weights)
- \checkmark quasi-linear computational performance up to 1 billion DOF
- ✓ 10x faster than VISIR-1
- \checkmark Bi-exponential distribution of CO2 savings found for a ferry
- \checkmark sailboat routes: duration savings of about 3% , neglecting leeway would underestimate durations

Possible uses of VISIR-2

□ inter-comparison studies

□ creation of baseline numerical experiments

□ weather routing of vessels with Wind-ASsisted Propulsion (WASP)

□ narrowing the uncertainty about the potential of weather routing for CO2 emission reduction

a exploit generality of its algorithm for minimizing the consumption of costly zero-carbon fuel

□ generate a dataset of optimal routes for the training of AI systems for autonomous vessels

□ educational purposes (ship officials and maritime surveillance authorities, beginner sailors)

Outlook

Computer Science

- computational performance improvements for the least-distance procedure
- reduce the computer's memory allocation
- more use of object-oriented programming principles

Naval architecture

- vessel intact stability
- voluntary speed reduction
- considerations for slamming, green water, lateral acceleration and passenger comfort

Algorithms

- multi-objective optimisation techniques
- consideration of tacking time and motor-assistance for sailboats
- adaptive routing strategies (rerouting)

VISIR-2 resources

open access – open review manuscript

https://egusphere.copernicus.org/preprin ts/2023/egusphere-2023-2060/ https://doi.org/10.5194/egusphere-2023-2060 Preprint. Discussion started: 16 November 2023 © Author(s) 2023. CC BY 4.0 License.

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