

Physics-guided ML to build digital twin for wind-assisted propulsion ships

Xiao Lang¹, Torsten Wik², Scott Mackinnon¹, Wengang Mao^{1*}

¹Chalmers University of Technology, Dep. of Mechanics and Maritime Sciences

²Chalmers University of Technology, Dep. of Electrical Engineering

email: xiao.lang@chalmers.se, torsten.wik@chalmers.se, scottm@chalmers.se, wengang.mao@chalmers.se

Problem and Purpose

Wind-assisted propulsion systems (WAPS) (Figure 1) are receiving renewed attention as the shipping industry confronts stricter greenhouse gas (GHG) emission regulations on the pathway toward net-zero emissions by 2050.

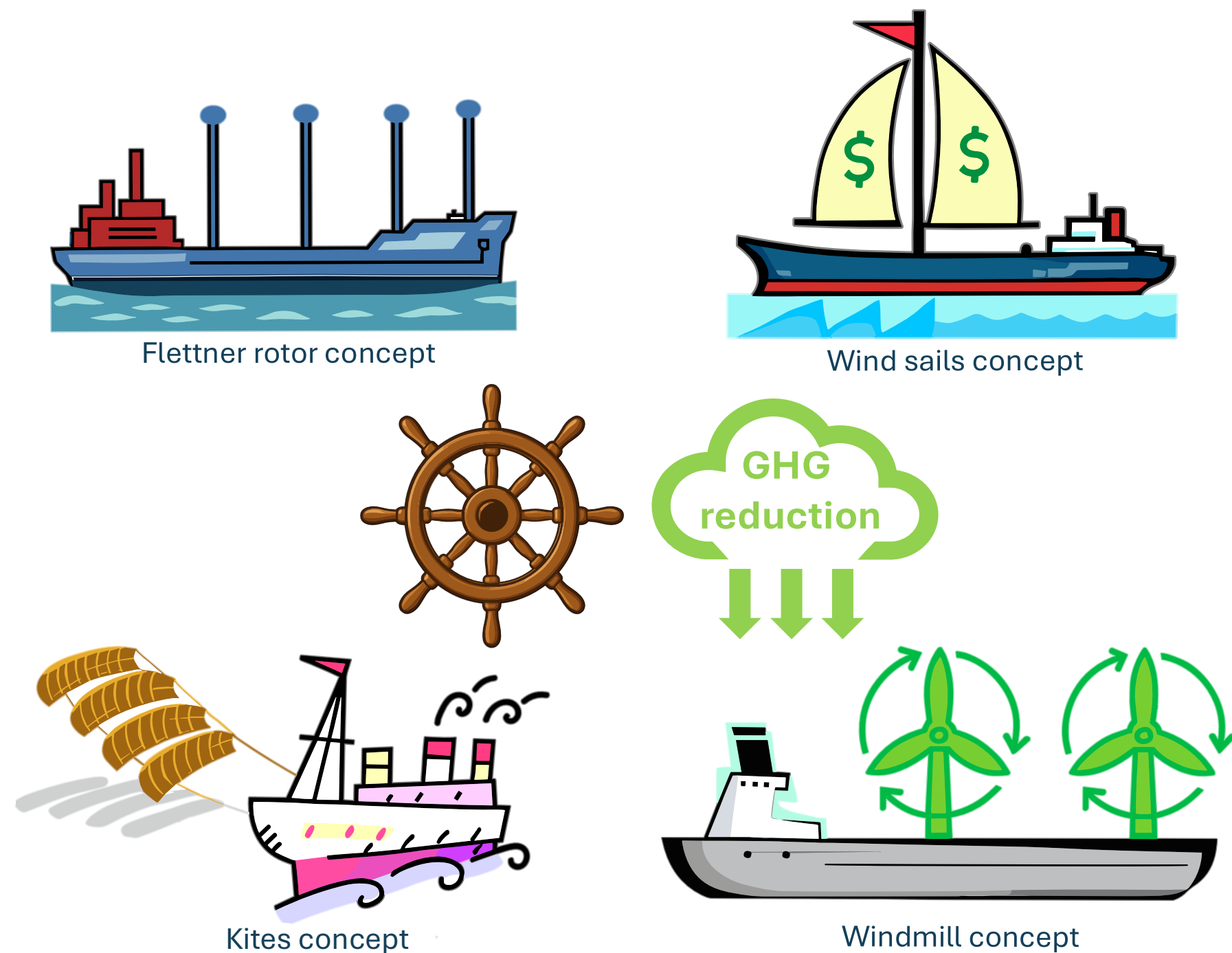


Figure 1: Illustration of key wind-assisted propulsion concepts.

This project aims to **pioneer physics-informed machine learning (PIML)-based WAPS ship modeling**^{1,2} by using key technological enablers, such as physics-informed neural networks (PINNs), and by addressing the following research questions (Figure 2):

- 1) How can a **PIML framework** be developed to combine physical principles with real-world sailing data for WAPS ships?
- 2) How can **WAPS devices be controlled** to optimize WAPS ship operation?
- 3) How can **onboard crews be trained** to operate the new WAPS ships efficiently and safely?

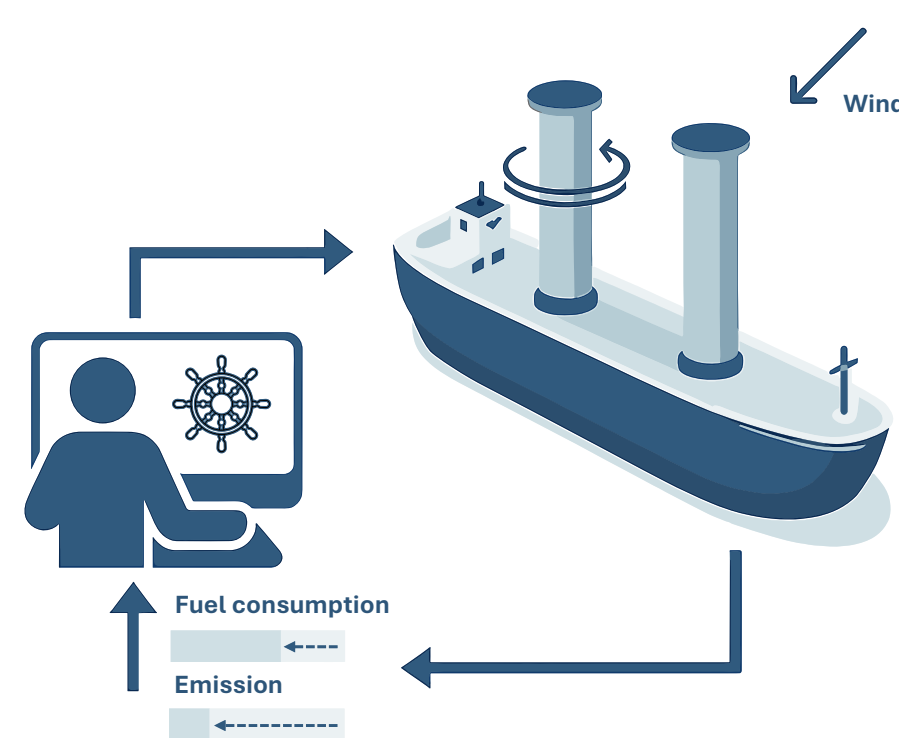
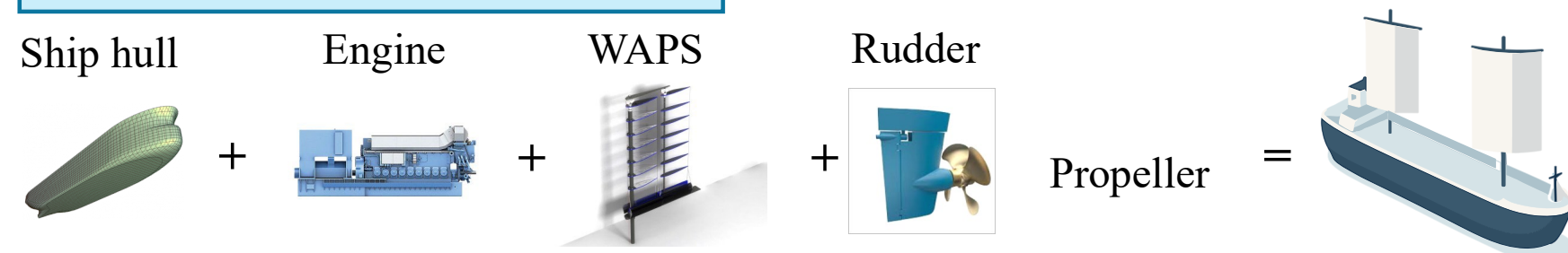


Figure 2: Illustration of how onboard crews control the WAPS device to maximize wind-assisted thrust, leading to reduced fuel consumption and emissions.

Realization

This project establishes four research steps with cross-disciplinary methods (Figure 3). **PIML approaches** will develop the ship performance models. The models will be embedded into the **weather routing**³ system, combined with **control algorithms**⁴ to enable optimal control. The **human factor** of the WAPS ship operation will also be investigated.

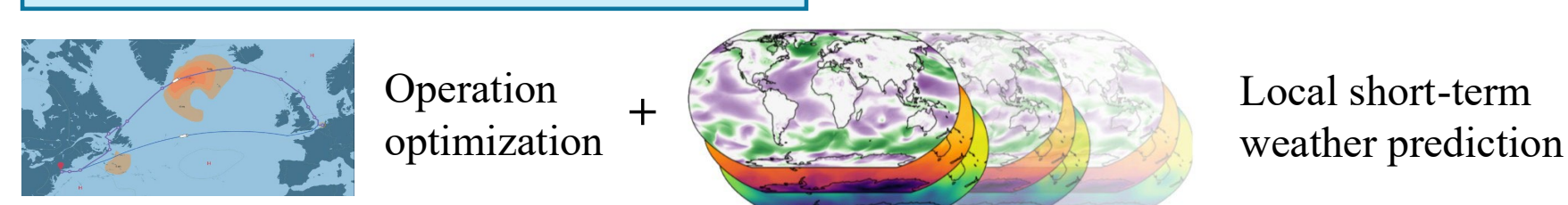
Step 1: Low-fidelity physical models



Step 2: High-fidelity physics-informed machine learning models



Step 3: Optimal operation for WAPS ship



Step 4: Simulator-based training

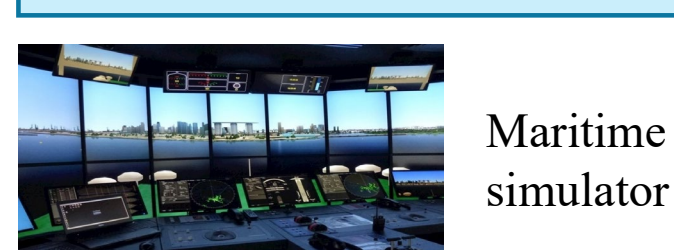


Figure 3: Schematic of the four research steps framework for PIML-based WAPS ship modeling and optimal operation.

Physics-guided ML to build digital twin for wind-assisted propulsion ships is funded by Transport Area of Advance



Multiple industry partners cooperatively support this project. The reference ship and full-scale measurements are provided by **Terntank** and **Odfjell**. **Berg Propulsion** is responsible for the engine and propeller modeling, while **DNV** offers guidance on industry standards and ensures a transparent evaluation. **Chalmers' maritime simulators** and **Wärtsilä** will support the virtual training.

Results

At present, the physical model has been established and embedded into a decision support system to optimize the control of the WAPS and the controllable pitch propeller (CPP) settings, thereby minimizing the power demand from the main engine. Figure 4 presents a 10-day case study voyage for the reference ship.

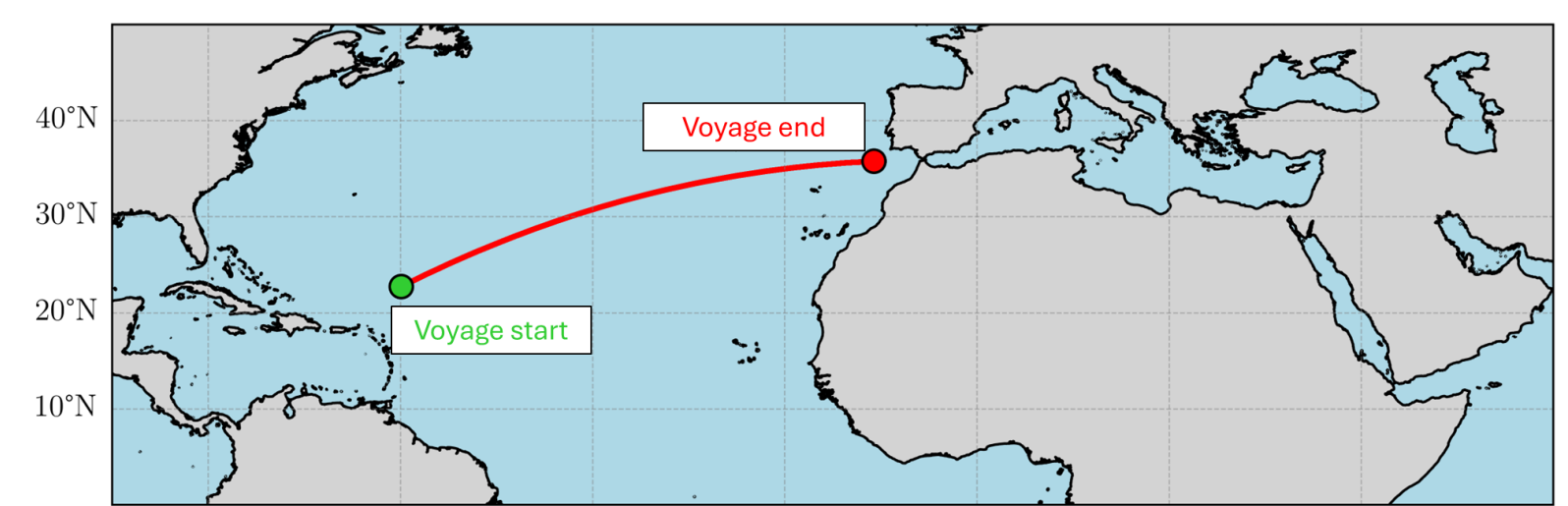


Figure 4: A 10-day case study voyage showing the trajectory of the reference ship from voyage start to voyage end.

For the case study voyages, if two suction sails are installed, **without active control** of the WAPS and the CPP settings, the thrust contributed by the WAPS can reach up to **20%** of the total required thrust when the wind direction is favorable (see Figure 5, upper). When the two installed suction sails are **actively controlled** to rotate for maximum thrust, and combined with CPP pitch control, the thrust contribution from the WAPS can increase to as much as **50%** of the total required thrust (see Figure 5, lower).

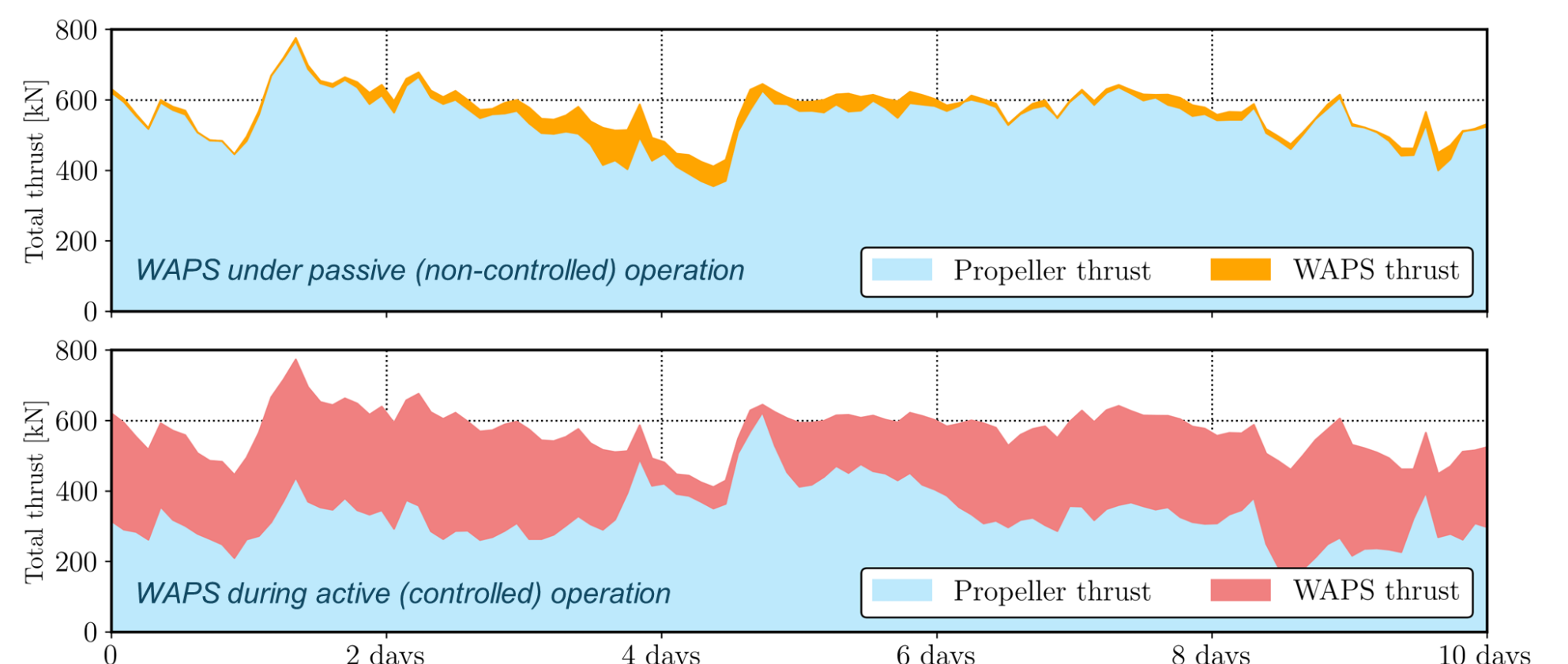


Figure 5: Thrust distribution of WAPS under passive (non-controlled) operation with two suction sails installed (upper); thrust distribution of WAPS under active (controlled) operation with two suction sails (lower).

For the entire voyage, the accumulated GHG emissions during active (controlled) WAPS operation can be **reduced by up to 50%**, when combined with optimized WAPS and CPP settings, as illustrated in Figure 6.

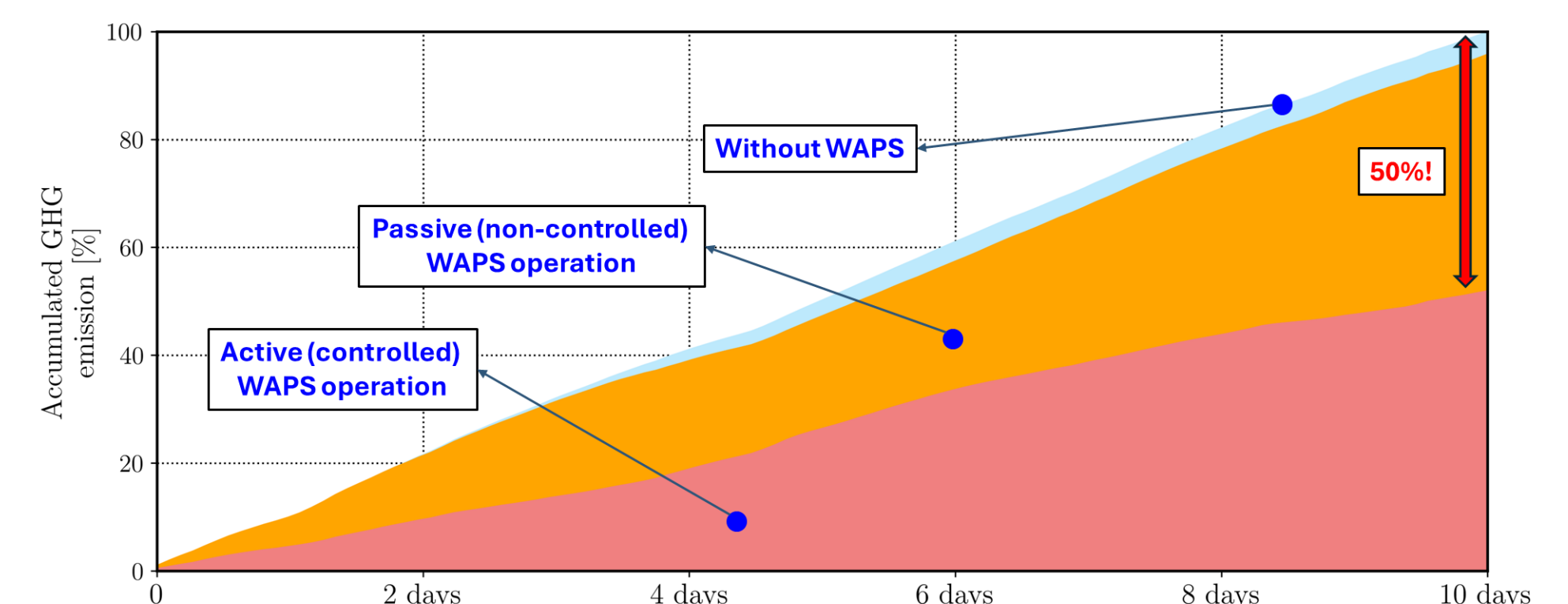


Figure 6: Accumulated GHG emissions for the entire voyage under three scenarios: baseline without WAPS, passive (non-controlled) WAPS operation, and active (controlled) WAPS operation.

References

1. Lang, X., Wu, D. and Mao, W. (2024) Physics-informed machine learning models for ship speed prediction, *Expert Systems with Applications*, 238(A), 121877
2. Lang, X., Zhang, M., Zhang, C., Ringsberg, J.W. and Mao, W. (2024) Physics-guided metamodel for vertical bending-induced fatigue damage monitoring in container vessels, *Ocean Engineering*, 312(Part 3), 119223.
3. Wang, H., Lang, X. and Mao, W. (2021) Voyage optimization combining genetic algorithm and dynamic programming for fuel/emissions reduction, *Transportation Research Part D: Transport and Environment*, 90, 102670.
4. Lang, X., Ge, M. and Mao, W. (2026) A semi-empirical ship performance model for wind-assisted propulsion under dynamic metocean conditions, *OMAE 2026* (accepted).