

Decarbonizing Maritime Operations: A Data-Driven Revolution



NEURAL INFORMATION
PROCESSING SYSTEMS

NeurIPS 2023 Workshop: Tackling Climate Change with Machine Learning

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UQAR



1. Introduction

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The maritime industry serves as a crucial conduit for global trade and commerce.



Urgent need for the maritime sector to reduce its carbon footprint as part of the global decarbonization challenge.



International Maritime Organization regulations have instituted stringent rules aimed at emission reduction.



The advent of Maritime 4.0 marks an era characterized by digitalization, automation, and optimization.



Machine learning and artificial intelligence are pivotal tools that are facilitating the achievement of sustainability goals.



**Maritime CO2 Emissions:
More than 2% of Global Total in 2018**

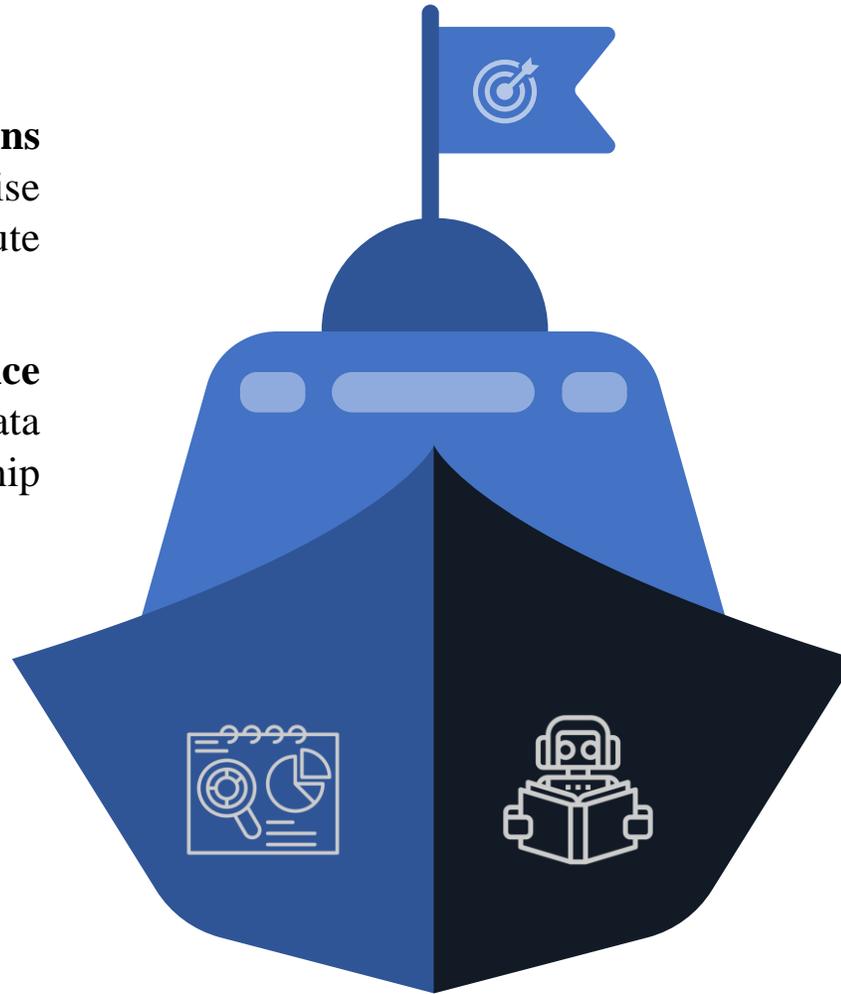
1.

2. Prior studies

3.

Data Analytics Approaches

- **AIS data analysis for emissions estimation:** Utilizing AIS data for precise emissions estimation and route optimization.
- **Big data analytics for ship performance monitoring:** Employing Big Data techniques to monitor and enhance ship performance.



Machine Learning Models

- **ML-based estimators for fuel consumption:** ANNs optimize fuel efficiency in diverse conditions for greener operations.
- **ML for ship performance prediction:** Statistical and ML models enhance ship performance prediction and operational efficiency.
- **Data-driven approaches for sustainable shipping:** ML frameworks analyze AIS/weather data for energy-efficient routing.
- **ML and statistical analysis for fuel efficiency:** ML and statistical analysis jointly optimize routing and fuel efficiency considering environmental data.

1.

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3. Proposal



Ship speed optimization



01

02

03

Data collection

- Historical Voyage Data
- Ship Specifications Data
- Real-time Environmental Data
- Operational Data

Machine Learning

- Regression Models for predicting optimal speed settings
- Reinforcement Learning for real-time speed control adjustments

Output

- Optimal speed and route recommendations



- **Significantly reduces fuel consumption, directly cutting down emissions.**
- **Provides a benchmark for environmental efficiency in maritime operations.**

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3. Proposal



Ships' Arrival Times Estimation



01

Data collection

- AIS Data
- Port Schedules
- Historical Performance Data
- Environmental Data



02

Machine Learning

- Time Series Models for forecasting arrivals
- Classification Models for estimating arrival probabilities



03

Output

- Accurate predictions of ships' Estimated Time of Arrival



- **Reduces unnecessary idling emissions by optimizing port arrival schedules.**
- **Lowers the overall carbon footprint of maritime voyages.**

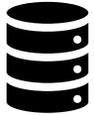
1.

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3. Proposal



Quayside Planning



01

Data collection

- Port Operation Data
- Cargo Data
- Logistical Data



02

Machine Learning

- Predictive Modeling using supervised ML for berth allocation
- Clustering Algorithms for resource allocation patterns



03

Output

- Predictions of berthing times and durations



- **Decreases emissions from ships by reducing idle time at the port.**
- **Enhances air quality around ports by enabling faster turnaround.**

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3. Proposal



Resource Allocation and Scheduling



01

Data collection

- Resource Inventory Data
- Labor Schedules Data



02

Machine Learning

- Deep Neural Networks for complex, multi-resource scheduling
- Reinforcement Learning for dynamic and adaptive scheduling



03

Output

- Schedules and allocation plans for port resources to minimize idle time



- **Diminishes emissions from port machinery through improved scheduling.**
- **Cuts down on indirect emissions by reducing the operational footprint.**

1.

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3. Proposal



Supply Chain Monitoring



01

02

03

- Aids in preventing emissions spikes by smoothing supply chain operations.
- Facilitates lower emissions through improved logistical efficiency.

Data collection

- Tracking Data
- Transportation Schedules Data

Machine Learning

- Graph Neural Networks for optimizing supply chain networks

Output

- Predictive insights on cargo flow and inventory

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Thank You

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