Soshianest-ICS Canada – Spring 2025 Lunch & Learn Artificial Intelligence Modelling in the Maritime Space March 6th, 2025

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Institute of Chartered Shipbrokers Canada



## Soshianest Collaborates With

















Soshianest was selected by CIIP for the 2024 PDA delegation to Singapore





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# **Our Mission**

At Soshianest, we help maritime businesses navigate international trade complexities. By leveraging AI and machine learning, we provide accurate forecasts, optimize operations, and enhance market responsiveness, helping clients remain competitive.

Supporting shipping stakeholders in adopting AI for better operations and decision-making with AI Data Driven models.





# Artificial Intelligence Modelling in the Maritime Space

Why AI Deep Learning Outperforms Traditional Models in Freight Rate Prediction

Artificial intelligence and deep learning are transforming maritime freight rate prediction, enabling more accurate forecasting and strategic decision-making

ICS-Canada – Ship Owners and Charterers

Dr. Mehdi Hazrati

Dr. Payman Eslami



# Volatile Market-Freight Rate & Prediction

- Freight rate prediction plays a critical role in the maritime industry, influencing contract negotiations, strategic decision-making, and profitability.
- Maritime freight markets experience high volatility and cyclicality, with complex dependencies on geopolitical, economic, and seasonal factors.
- Understanding market fluctuations is essential for effective risk management and long-term planning.

### BCTI TC 10-TCE 40,000mt, S Korea-Vancouver



# The Strategic Impact of Freight Rate Forecasting on Maritime Operations



Sustainability and Compliance



# Strategic Decision Points Impacted by Freight Rate Prediction







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# Traditional Model, Limitations & Strategic Impact

| Traditional Model  | Strength  | Limitation  |  |  |  |  |  |
|--|---|---|--|--|--|--|--|
| Linear Regression  | Simple & fast   | Can't handle market shifts                                  |  |  |  |  |  |
| ARIMA  | Good for time-series trends                               | Weak for non-linear markets                                 |  |  |  |  |  |
| Bayesian Forecasting   | Probabilistic modeling                                    | Requires extensive data                                     |  |  |  |  |  |
| VAR (Vector Autoregression)  | Captures relationships<br>between multiple variables      | Assumes stationarity,<br>struggles with long-term<br>trends |  |  |  |  |  |
| Exponential Smoothing  | Captures short-term trends Too simple for comple patterns |   |  |  |  |  |  |
| Traditional models struggle with maritime market complexities, leading to inaccurate predictions and |   |   |  |  |  |  |  |

Traditional models struggle with maritime market complexities, leading to inaccurate predictions and higher financial risk, due to Static Assumptions human intervention, leading to bias and incomplete feature selection etc.



# Key Machine Learning Architectures for Maritime Al

Ensemble Learning Methods: Random Forest & Gradient Boosting Improves predictive accuracy by combining multiple decision trees, making them effective for freight rate forecasting. Classification & Pattern Recognition: Support Vector Machines (SVM) Detects patterns and classifies maritime data points, helping with vessel tracking and demand forecasting. **Clustering & Anomaly Detection: K-Nearest Neighbors (KNN)** Used for clustering and anomaly detection in shipping routes and freight rate predictions, leveraging proximitybased learning.



# What is Deep Learning?



# Key AI & Deep Learning Architectures

### LSTM (Long Short-Term Memory Networks)

LSTM networks capture long-term dependencies and temporal patterns, making them ideal for sequential maritime data like historical freight rates.

#### **CNNs (Convolutional Neural Networks)**

CNNs extract spatial-temporal patterns from complex maritime datasets, enabling applications in vessel tracking and maritime maps.

#### Transformers

Transformers improve upon traditional deep learning architectures by allowing parallel processing of data, making them ideal for large-scale maritime forecasting and anomaly detection.



# LSTM and CNN Data Processing Pipelines



## 2010s-2020s AI & Deep Learning Evolution in Maritime Trade







# AI & Deep Learning vs Machine Learning – Key Differences

| Aspect                   | Machine Learning (ML) in Maritime   | AI & Deep Learning (DL) in Maritime  |  |  |
|--------------------------|---|--|--|--|
| Definition               | Uses statistical models to learn from historical shipping & trade data    | Uses deep neural networks to analyze <b>vast,</b><br><b>complex maritime datasets</b>            |  |  |
| Feature Engineering      | Requires manual selection of features (e.g., port congestion, fuel costs) | Automatically extracts features from raw data (satellite images, weather reports, market trends) |  |  |
| Data Processing          | Works well with structured tabular data (shipping logs, trade records)    | Handles <b>unstructured &amp; multimodal data</b> (text reports, images, IoT sensor data)        |  |  |
| Complexity Handling      | Struggles with dynamic & non-linear market behaviors                      | Excels at <b>recognizing nonlinear dependencies</b> (global supply chain disruptions)            |  |  |
| Adaptability             | Learns from past data but requires frequent retraining                    | <b>Continuously adapts to real-time maritime data</b> , optimizing forecasts dynamically         |  |  |
| Use Cases in<br>Maritime | Freight rate estimation, route optimization based on historical trends    | Al-driven vessel scheduling, anomaly detection, multimodal trade forecasting                     |  |  |
|                          |   | Coshia   |  |  |





## **Future Evolution of AI in Industry & Maritime**



Economic Impact: McKinsey & Company estimates that AI could contribute an additional \$13 trillion to the global economy by 2030, Productivity Growth: Generative AI, a subset of AI technologies, has the potential to increase U.S. labor productivity by 0.5 to 0.9 percentage points annually through 2030, depending on the rate of technology adoption and effective integration into business processes.



# Why we need to act: what is Projection (2025-2030+)?

- Maritime Industry Adopts New Technology More Slowly Than Other Sectors
- Al in finance, e-commerce, and manufacturing moves quickly.
- But maritime & logistics have longer adoption cycles due to:
- Infrastructure limitations (ships, ports need modernization).
- Regulatory requirements (IMO rules, emissions compliance).
- Risk aversion (shipping companies prioritize reliability over fast AI adoption).



# **Challenges in Al-Driven Maritime Forecasting**

## Explainable AI for Decision Transparency

Computational Complexity & Data Needs

## **Risk of Overfitting**

Ensuring AI predictions are interpretable and trusted by industry stakeholders. Deep learning requires high processing power and large datasets for accurate predictions.

Balancing model complexity to avoid overfitting on historical maritime data.

Data Quality and Availability: Reliable data sources are critical for accurate maritime freight forecasting.



# Ensuring AI Reliability in Maritime









# Why Ethical Al Matters

AI plays a growing role in maritime decision-making.

Ethical AI ensures trust, fairness, and compliance.



# **Overall Industry Challenges**

## Port Congestion & Supply Chain Disruptions

Global supply chains face bottlenecks due to port congestion, extreme weather, and labor strikes, causing inefficiencies and delays

## Market Volatility & Uncertainty

Fluctuating freight rates, economic downturns, and geopolitical risks make it difficult for stakeholders to predict costs and revenues..

## Regulatory & Sustainability Compliance

Shipping companies must meet IMO regulations, reduce emissions, and adapt to changing environmental policies, requiring significant investment.

## Operational Inefficiencies & Cost Pressures

Rising fuel prices, inefficient fleet deployment and unexpected maintenance costs put financial strain on maritime businesses.

## Technological Adaptation & Digital Transformation

Many maritime companies struggle to implement AI and data-driven solutions due to legacy systems and lack of technical expertise.



## AI in Maritime Operations-Tackling Industry Challenges

# Al in Sustainability & Compliance

Increasing regulatory pressure to reduce carbon emissions.

Al models optimize fuel consumption and reduce  $CO_2$  emissions.

Machine learning simulates and optimizes energyefficient routes.

Al assists in monitoring IMO compliance and alternative fuel adoption.

## AI in Vessel Performance Monitoring Maintenance

Al-powered optimization enhances fuel efficiency and maintenance scheduling.

IoT sensors collect real-time data on engine health, fuel efficiency, and hull conditions

Al-driven predictive maintenance detects early signs of equipment failure, reducing vessel downtime and saving maintenance costs."

Shipowners face high maintenance costs and unexpected breakdowns.

## Al in Port & Terminal Congestion Prediction

Port congestion leads to delays, inefficiencies, and increased costs.

Al analyzes real-time AIS data, satellite imagery, and historical congestion trends.

Machine learning predicts vessel arrival times and berth availability.

Al-powered re-routing recommendations help reduce waiting times.



# Most Commonly Foundational Concepts in Modern Al-Text Processing



Transformers

Leverage self-attention (understand, interpret, and generate human language) to capture global dependencies in NLP (Natural Language Processing) and other fields, making them ideal for maritime forecasting. Examples: BERT (NLP), GPT-4 (NLP), Vision Transformer (VIT - Computer Vision), Time Series Transformer (TST - Forecasting).



LLMs (Large Language Models) Analyze unstructured data like news and social media, using NLP and sentiment analysis to improve predictions. Examples: ChatGPT, LLaMA 2, Claude (Anthropic), Mistral AI.



Generative AI (GenAI) Generates scenario-based forecasts, offering best, worst, and average-case freight rate predictions with explainability. Examples: DALL·E (Image GenAI), Gemini (Multimodal GenAI), OpenAI Codex (Code GenAI), MidJourney (Image GenAI).



## Automatic Feature Engineering & Extraction and Adaptive Learning Loop



## Core Advantage

Al-driven models automatically learn hidden patterns without manual

intervention, eliminating human bias in feature selection.

## **Market Adaptability**

Deep learning adapts dynamically to market disruptions, improving predictive accuracy even in volatile maritime conditions.

## Strategic Impact

By continuously refining insights, AI enables strategic agility and provides a competitive edge in maritime decision-making.

# Modeling Process



# Factors Influencing The Market

## **Model Inputs**

| Vessel<br>Supply                 |           | Orderbook                | Deliveries         | Demolitions      | Scrap Value        | Secondhand<br>Price      | Newbuild<br>Price/Capacity         |
|----------------------------------|-----------|--------------------------|--------------------|------------------|--------------------|--------------------------|------------------------------------|
| Global<br>Economic<br>Conditions | <b>\$</b> | GDP                      | Interest Rate      | Exchange<br>Rate | Oil Price          | Industrial<br>Production | Commodity Price                    |
| Seaborne<br>Trade                | 000       | Seaborne<br>Trade Volume | Import &<br>Export | Bunker Price     | Port<br>Congestion | Regulations              | Deployment &<br>Port Call Activity |





# Performance Comparison: Deep Learning vs. Traditional Models

#### **Comparison Metrics**

Deep learning models outperform traditional models across key metrics such as accuracy, Mean Absolute Percentage Error (MAPE), and Root Mean Square Error (RMSE).

#### **Accuracy & Error Reduction**

Al-driven models significantly reduce forecasting errors, making them more reliable for freight rate predictions.

#### **Strategic Benefit**

Higher accuracy leads to better decision-making in maritime freight forecasting, ensuring improved market competitiveness.



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# Future Outlook of AI in Maritime



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Multi-Modal Learning: Integrating diverse data sources for improved predictions.

**Generative AI:** Simulating future

market scenarios for proactive

planning.

 Explainable AI: Enhancing model
03 transparency and decision trustworthiness.

AI-Driven Automation: Reducing
04 manual efforts and improving operational efficiency.

Maritime AI evolution is shaping smarter and more adaptive forecasting solutions.





# Al in Maritime: Future, Opportunities & Call to Action

## Strategic Advantage

## Emerging Trends

## **Call to Action**

Al models provide a competitive edge, improving risk management and decisionmaking. Future advancements include Generative AI, Multi-Modal Learning, and Explainable AI for enhanced transparency and adaptability. Maritime stakeholders should embrace Al-driven models to optimize operations, reduce risks, and stay ahead in dynamic markets.



## **Engagement Opportunities**

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#### Engage

Engage with Soshianest : Collaborate with us to harness AI technologies for transformative improvements in maritime logistics operations.

#### **Tailored Solutions**

Address your unique challenges with data-driven insights.

#### **Pilot Program**

Work with us for customized AI solutions for your organization.

# Thank You

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