

MIT Center for Transportation & Logistics



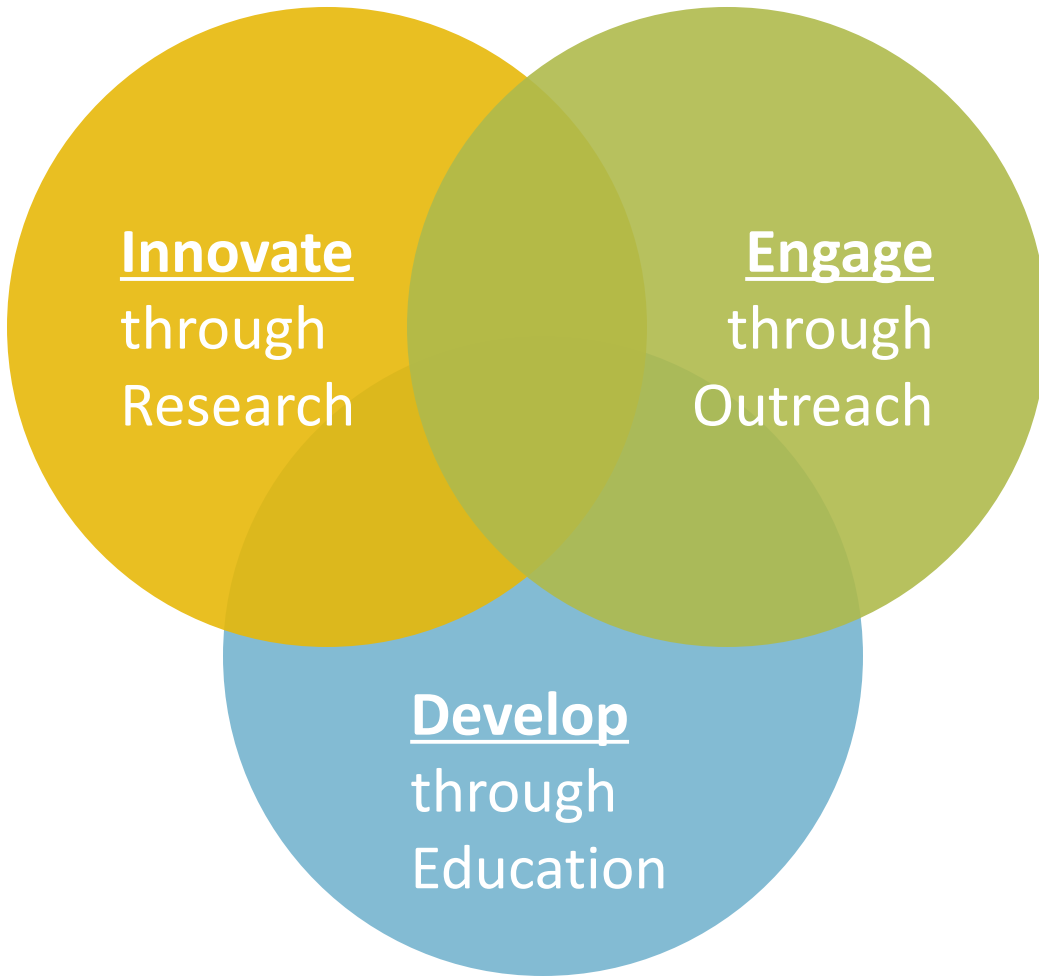
MIT Center for
Transportation & Logistics



**Deep Knowledge
for SCM Lab**

What is MIT CTL?

“Create supply chain innovation and drive it into practice.”



- Roundtables
- Capstone projects
- Industry-driven workshops & symposia

- Supply Chain Management masters program
- Graduate Certificate in supply chain management
- MicroMasters Program in supply chain management
- Customized courses
- Executive Education

- 15 Research Labs
- 6 Research Centers worldwide
- 20+ Active research projects
- 60+ Faculty & researchers across MIT

MIT Global SCALE Network

10 + EDUCATIONAL PROGRAMS

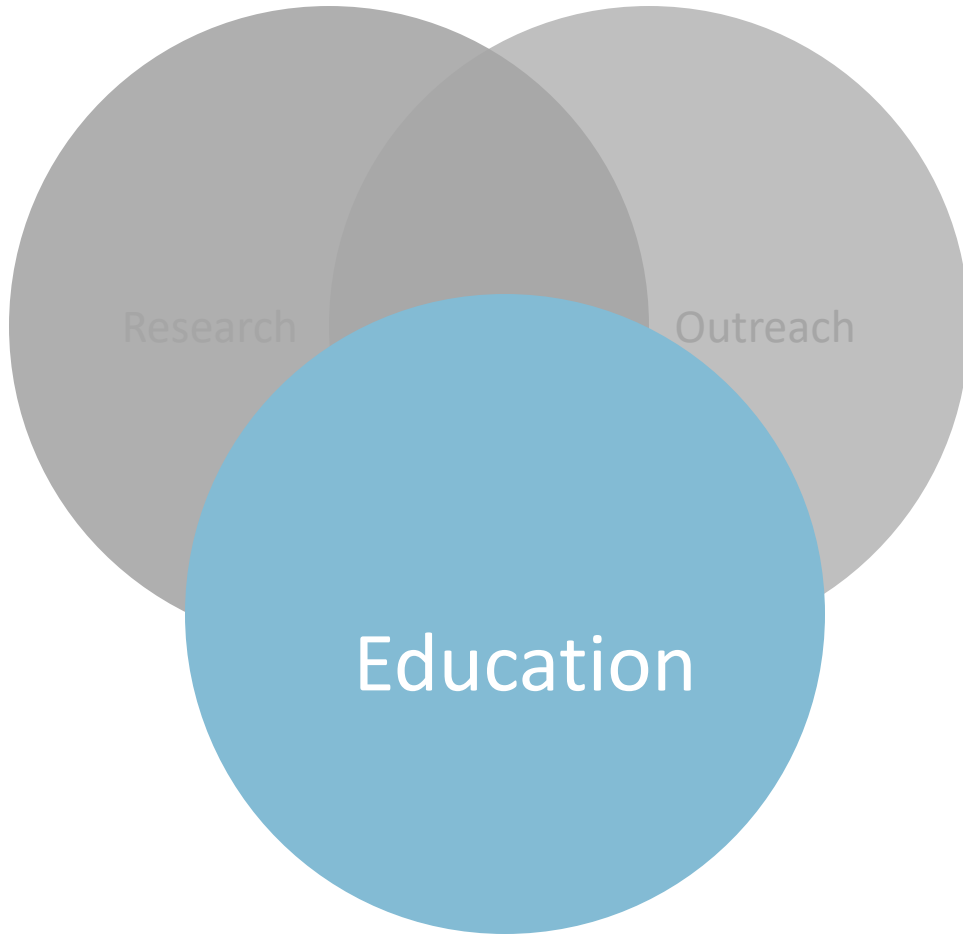
80 + RESEARCHERS & FACULTY

150 + CORPORATE PARTNERSHIPS

200 STUDENTS ANNUALLY

2000 + ALUMNI GLOBALLY

Education



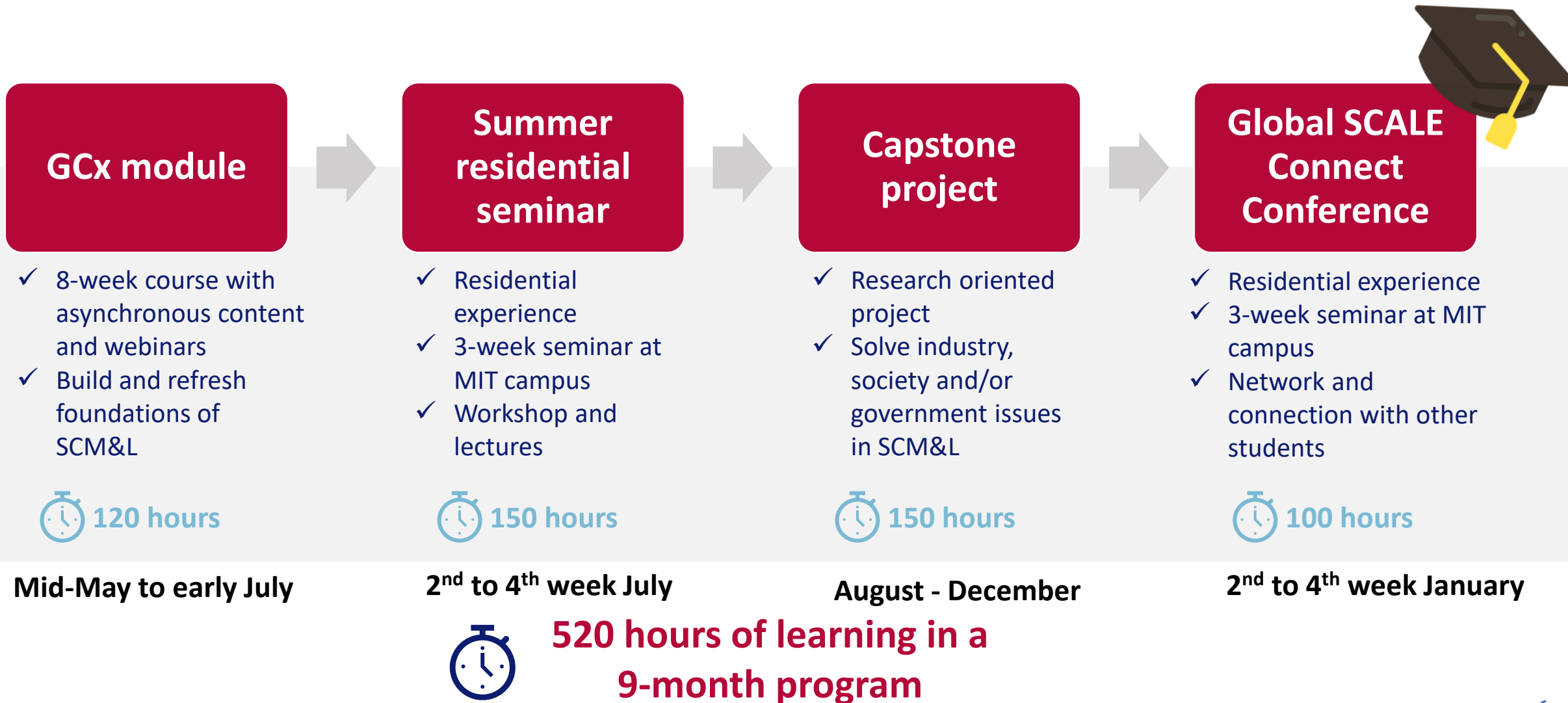
GCLOG - Mission and eligibility criteria

“...A life
changing
experience...”



- “Amplifying graduate SC education for **emerging markets**”
- An **elite academic program** geared towards **outstanding professionals**
 - **At least 3 years of working experience in SCM**
 - **Preferably, admitted to, enrolled in or recently graduated from a graduate program**
- Taught by **senior industry experts, MIT and SCALE-recognized researchers**

GCLOG structure



GCLOG - Key takeaways

Hybrid, high-quality,
and flexible program

**#1 graduate program in
SCM worldwide
(EdUniversal, 2025)**

*Industry-centered and
international capstone
projects in EME*



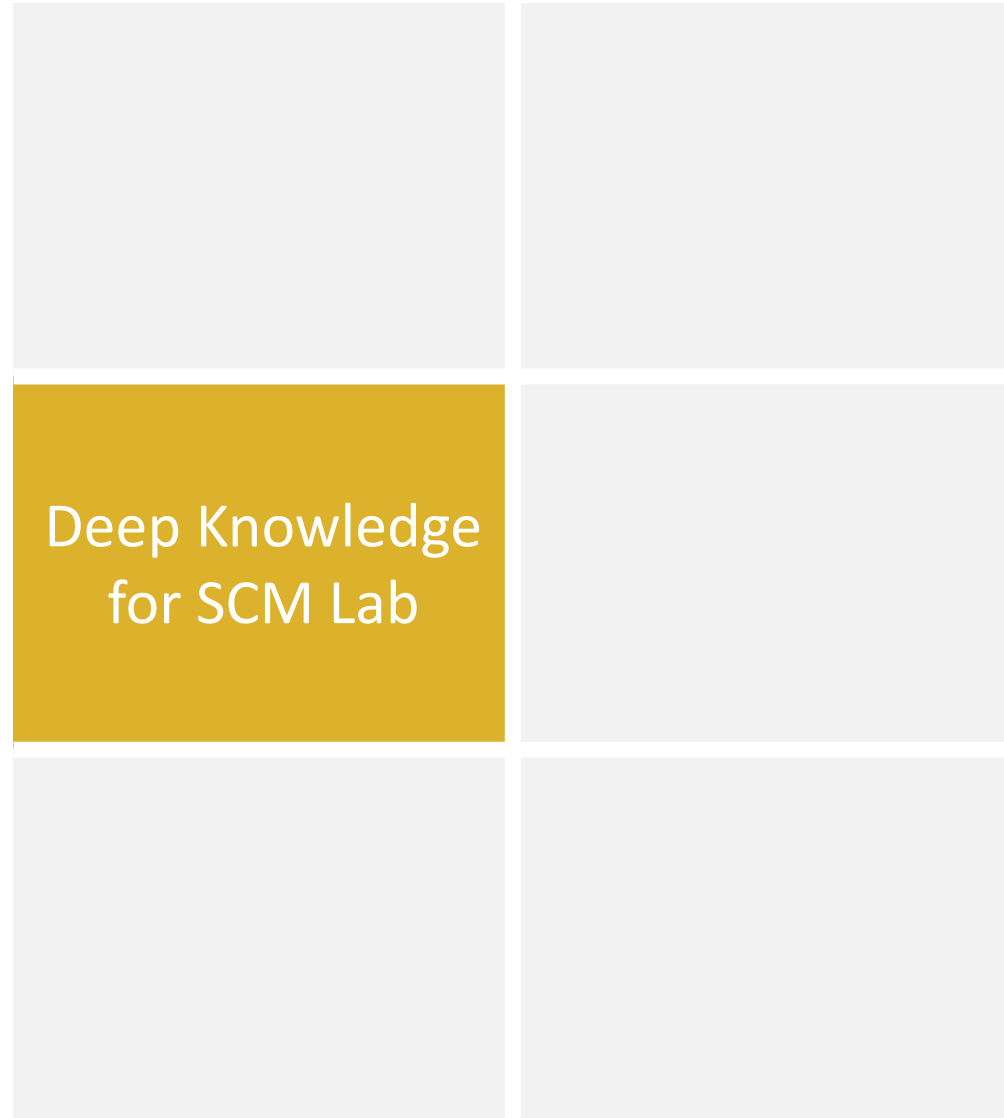
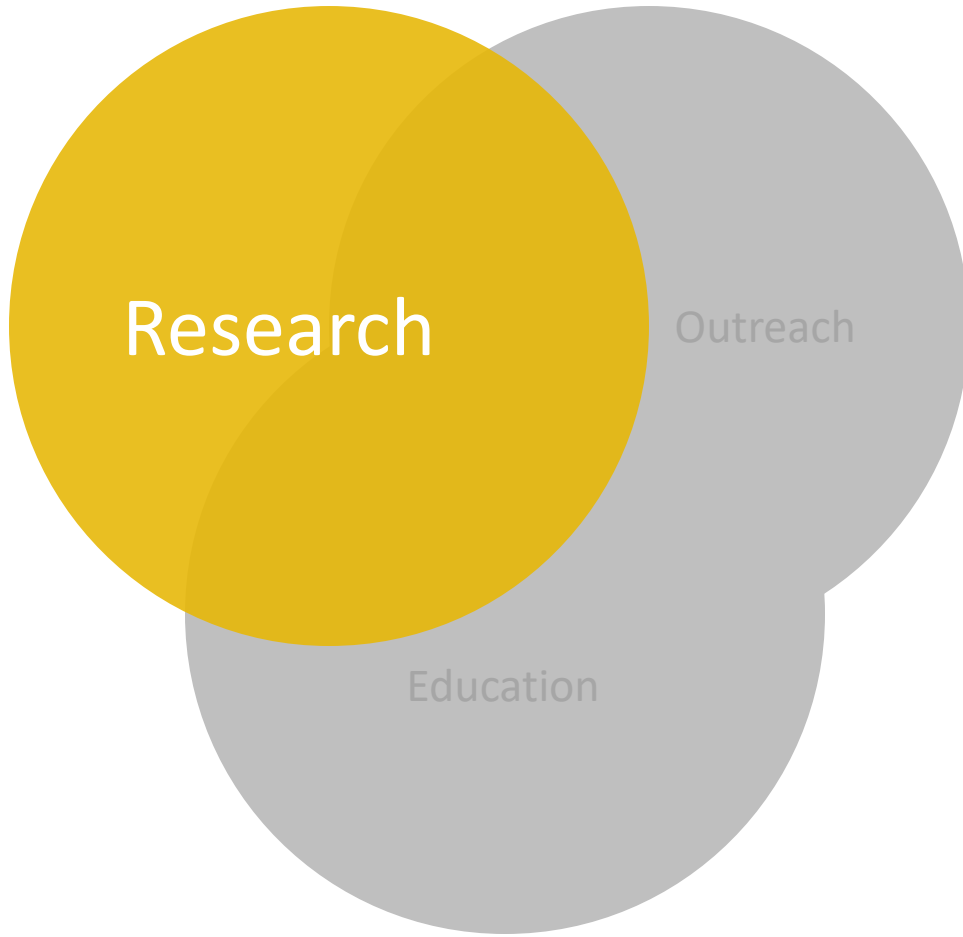
Applied knowledge and
multicultural networking

**Change mindset and
provide practical tools
to solve issues**

*Be part of the MIT alumni
community*



Research



Maritime Logistics



MIT Center for
Transportation & Logistics



1. Published Research
2. Current Research
3. Future Research

1. Published Research

Sponsor: DOT Freight Logistics Optimization Works (FLOW)

Data Sources: AIS Data, Import Genius

1

DBSCAN: Ship congestion points and infrastructure

ANT 2024

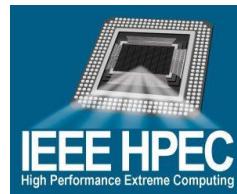


2

Predicting port congestion for East Coast Ports



GRAPH EXPLOITATION SYMPOSIUM
MAY 20–21, 2025



3

Ship sequencing in ports and prediction of stay



TRANSPORTATION
RESEARCH
JOURNAL



*Under
Submission

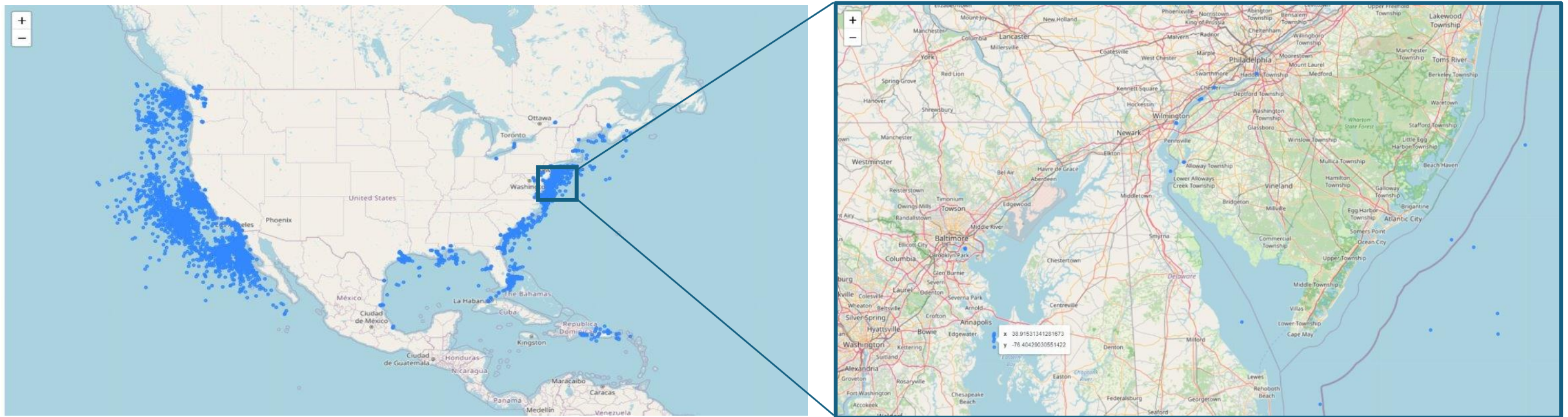
1. Published Research

1

2

3

DBSCAN: Ship congestion points and infrastructure from AIS data



1. Published Research

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2

3

DBSCAN: Ship congestion points and infrastructure from AIS data

- Congestion points automatic classification
- Automatic ship to terminal assignment
(New York/New Jersey, Los Angeles/Long Beach)

1. Published Research

1

2

3

Predicting port congestion for East Coast Ports:



LSTM
(Long Short
Term
Memory)

XGBoost
(Gradient
Boosted
Trees)

Liquid
S4
(Structural
State-Space
Model)

ST-GNN
(Spatio-Temporal Graph
Neural Network)

01/02/
2015

08/31/
2023

09/30/
2023

Training

Testing

1. Published Research

1

2

3

Terminals at Port of New York / New Jersey

New York / New Jersey Port Terminals



- Primary features:
 - 31 Terminals
 - 22 Operators
 - 7 Ship Size Groups
 - 7-day History

Other features:

- Time
- Node-specific features

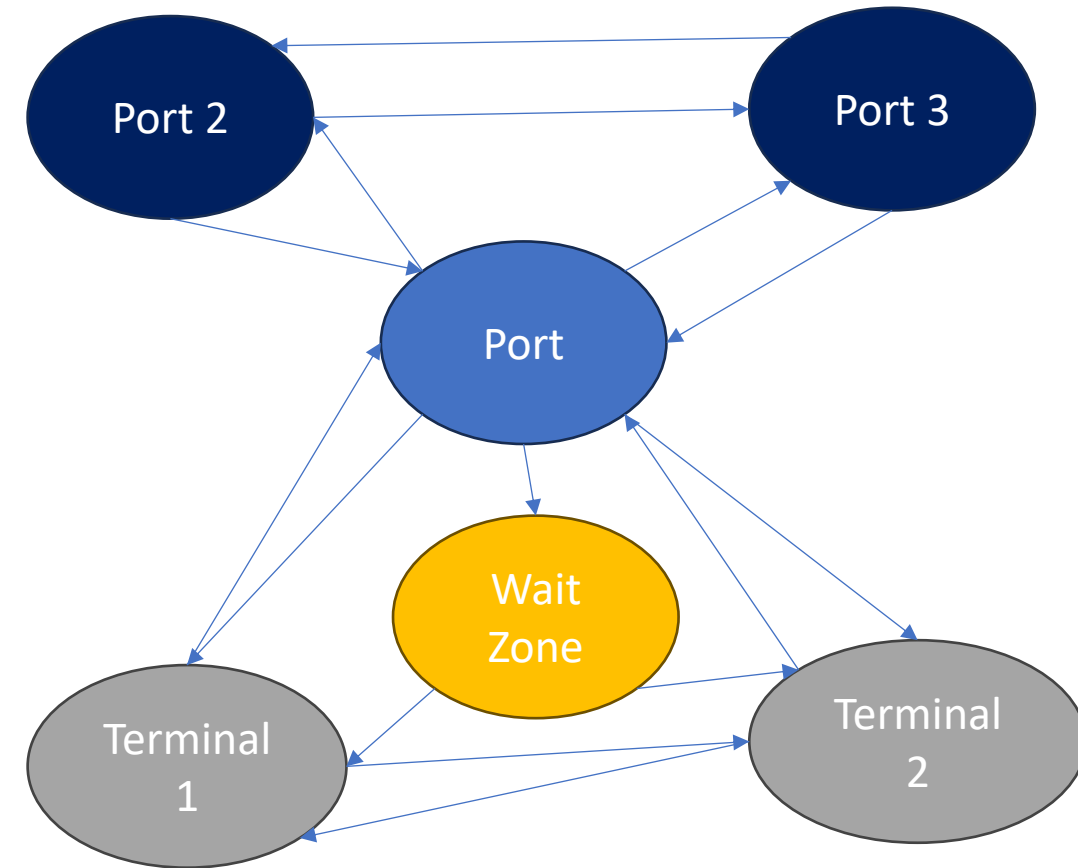
1. Published Research

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- Used 5 ports, 5 wait zones and available terminals at each of the 5 ports
 - Ports connected to each other
 - Ports connected to internal nodes
 - Wait Zone
 - Terminals
 - Wait Zones connected to Terminals at port
 - Terminals connected to other Terminals
- Adjacency matrix used the distance between ports, and fixed distance inside ports, normalized by standard deviation



1. Published Research

1

2

3

Predicting port congestion for East Coast Ports

Port Terminal	Model	MAE	MAPE
APM Terminals (high throughput)	LSTM	1.57	37.64
	STGNN	1.52	55.16
	XGBoost	0.52	12.28
	Liquid	0.93	29.26
Port Liberty Bayonne Terminal (medium throughput)	LSTM	0.70	50.68
	STGNN	0.51	37.75
	XGBoost	0.67	58.92
	Liquid	0.63	32.36
The Red Hook Container Terminal (low throughput)	LSTM	0.36	57.94
	STGNN	1.39	79.62
	XGBoost	0.94	72.32
	Liquid	0.5	59.58

- Different models for terminals
- Improving one terminal makes others worse -> **comparing Liquid to all 3 different winning models results**
- Liquid consistently ranks second across all terminals
 - Further parameter tuning or architectural changes could potentially improve its performance

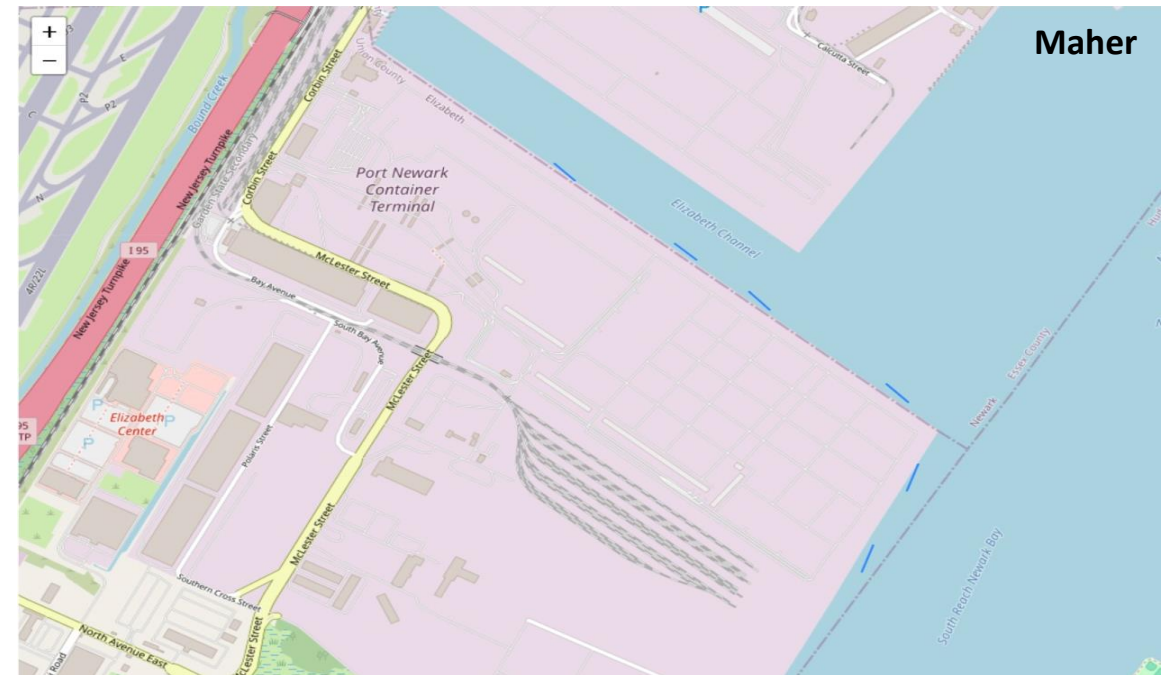
1. Published Research

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3

Ship sequencing and prediction of stay (Inverse Reinforcement Learning)



1. Published Research

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3

Ship sequencing and prediction of stay (Inverse Reinforcement Learning)

Terminal	Window	Model Base	MAE (Mean Absolute Error), number of time windows	Ships, predicted stay difference	
				In 18h %	In 24h %
APM	6h	State-based	1.98	81%	89%
APM	6h	Ship-based	1.94	82%	92%
Maher	6h	State-based	2.73	71%	79%
Maher	6h	Ship-based	3.99	51%	63%
Savannah	6h	State-based	2.82	69%	78%
Savannah	6h	Ship-based	3.32	55%	71%

1. Published Research

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Ship sequencing and prediction of stay (Inverse Reinforcement Learning)

MaHer State-Based model variations

Model variations	MAE (Mean Absolute Error), number of time windows	Ships, predicted stay difference	
		In 18h %	In 24h %
MaHer	2.73	71%	79%
MaHer + 1	2.56	74%	83%
MaHer + Attention	2.61	72%	80%
MaHer + Attention + 1	2.52	75%	83%
MaHer Bayesian small	4.38	51%	58%

MIT CTL Research Project 2025



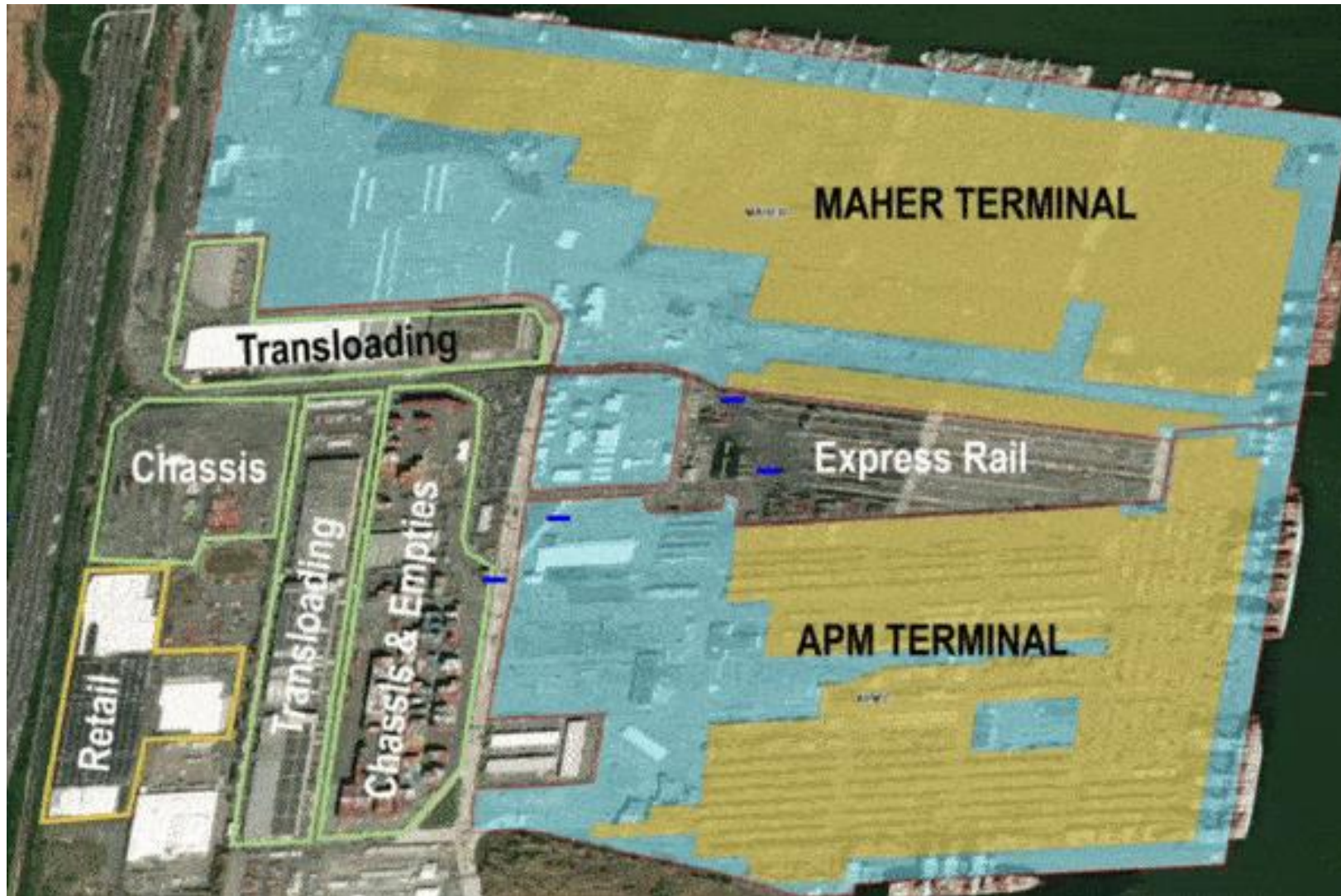
2. Current Research



Simulation at APMT Port Elizabeth with Port Authority NY/NJ

- Developed using open-source software in Python.
- Provides a flexible and scalable framework for future research.
- Discussions with APMT and PANYNJ stakeholders and observations at the terminal have identified hinterland locations suitable for transitioning some truck volume to short-haul rail.
- Using the simulation model, we will test various scenarios of this transition and assess the impact on terminal resources.

2. Current Research – Example Simulation

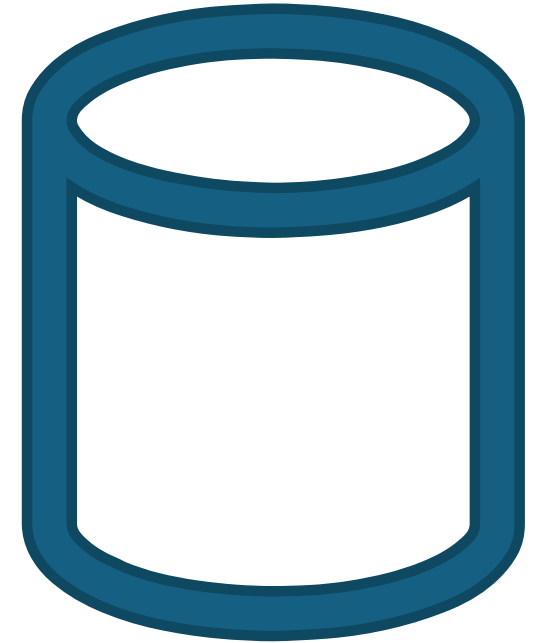


2. Current Research – Data Requirements

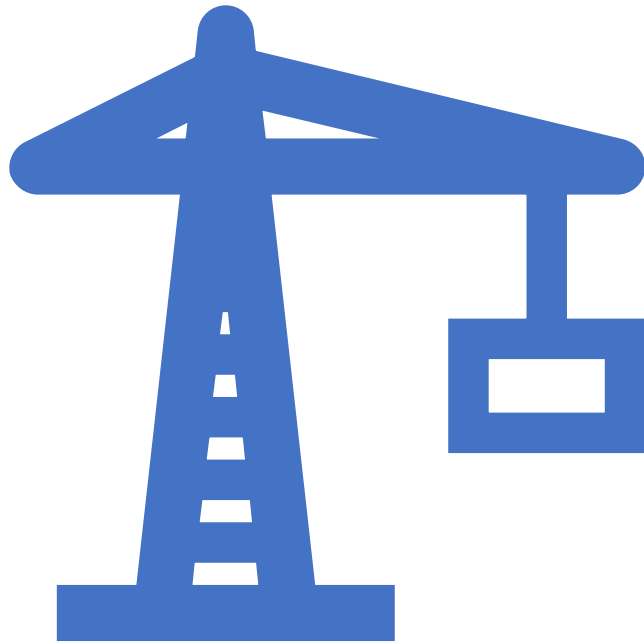
- **Ocean Operations:** Extensive work using AIS data to model vessel arrivals.
- **Terminal Operations:** ImportGenius data (CBP shipping manifest data), filtered to generate container-level insights for the Port of NY/NJ.
- **Combined Approach:** By integrating AIS and ImportGenius, we can develop a comprehensive model of historical import arrivals using publicly available data
- **Taking it a Step Further:** Using the container numbers obtain from ImportGenius analysis, we can automate web-tracking tools online offered by carriers and terminal operators, to download container movements.

This gives us a method to estimate container throughput of individual terminals using only publicly available data.

We gain insights into the volume of containers arriving at the terminal, where those containers are headed, the time each spent in the port, and how the container left the port.



2. Current Research – Analysis



Impact of Operational Changes on Port Efficiency

- **Objective:** Evaluate how changes in resources and container volumes impact port efficiency and dwell times.
- **Key Variables:**
 - **Resources:** Crane availability, yard use, truck/rail throughput.
 - **Container Volumes:** High-peak vs. low-volume scenarios.
- **Outcomes:**
 - Measure yard utilization, throughput and container dwell times.
 - Identify bottlenecks caused by resource or volume constraints.
- **Use Cases:**
 - Support decision-making for resource allocation and infrastructure investments.
 - Optimize operations during peak periods or constrained scenarios.

2. Current Research – Supporting Analysis

Commodity Classification using Natural Language Processing



A semi-fine-tuned Bidirectional Encoder Representations from Transformers (BERT) model is applied for this task



Additional fine-tuning on a labelled dataset of ImportGenius records has shown promising results for commodity classification



Could result in the addition of features to the simulation based on commodity.

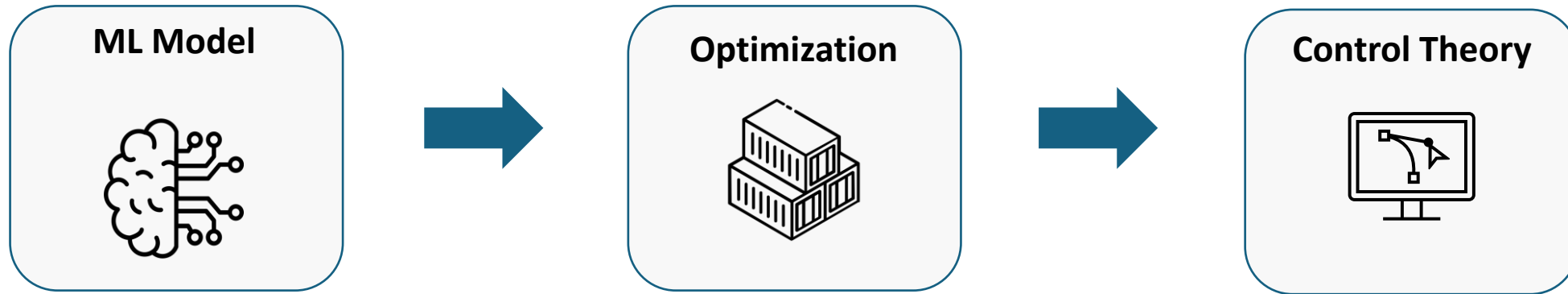
3. Future Research – Container Tracking



1. Smart Container Tracking and Monitoring – Edge AI
2. Supply Chain and Container Flow Prediction – Edge & Cloud AI
3. Optimizing Container Placement and Routing with Reinforcement Learning – Cloud AI

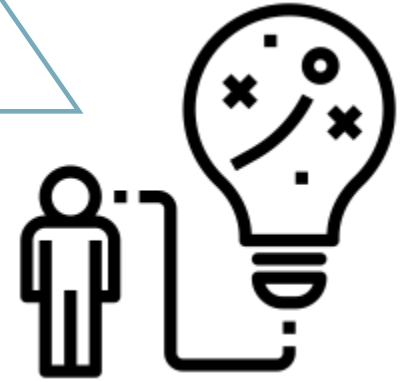
**** Potential Sponsored Research Project***

3. Future Research – Yard Optimization



1. Cost Estimation Model – Machine Learning
2. Optimal Cost – Optimization Model
3. Transition from Current State to Optimal State – Control Theory

**** Potential Sponsored Research Project***



1. **Capstone Projects** with the GCLOG or SCM programs
2. **Professional Development** in the GCLOG or SCM programs
3. **Executive Education** for Middle Eastern professionals
4. **Sponsored Research** with MIT CTL on Maritime Logistics