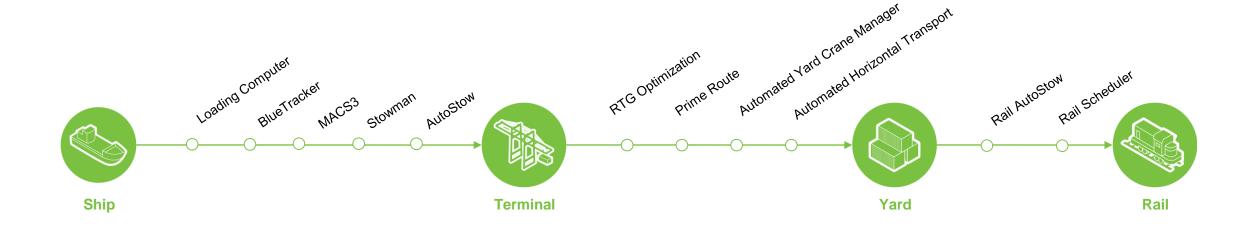


Our Experience in Automation



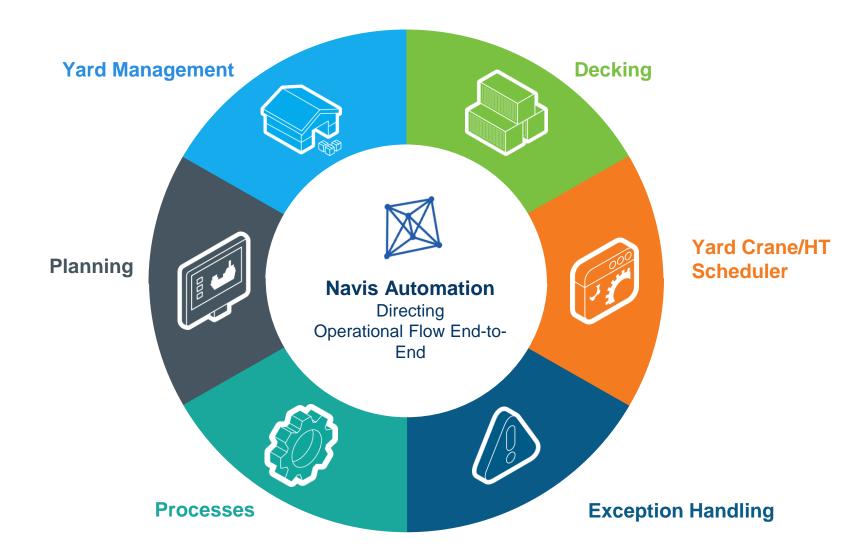
Artificial Intelligence

At every point of a container's journey, we apply Artificial Intelligence to optimize operations and automate decision making



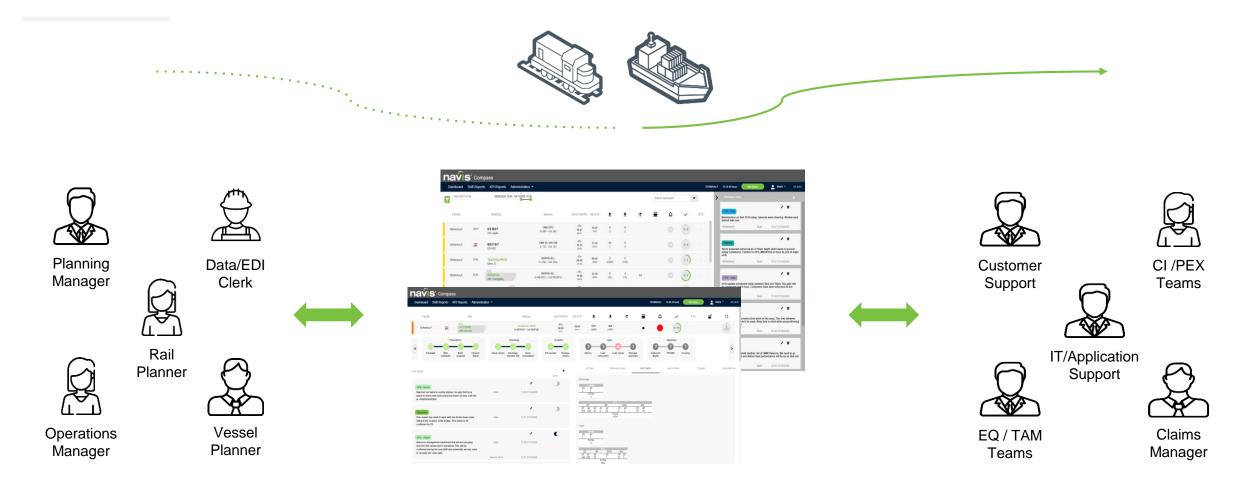


We Build Automation on Optimization





Automating Existing Processes





The Human Factor

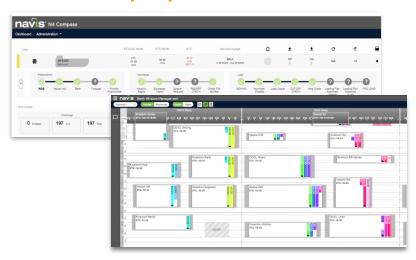
Remove variability and

repetition



Leverage creativity and

people skill set



Focus on a proactive

way of working





New Expertise is Needed



Digitalization & Automation Increases Data Volume











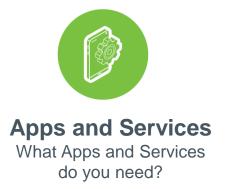


Other Data
Sources:
IoT, SQL, Oracle,
MongoDB,
Firebase, etc.





Navis Smart Data Layer
Data Consolidation, Streaming Service, ETL





AnalyticsHistorical Trends, ML, etc.



OpsViewReal-time Reports



APIsCustom Applications

Actionable Intelligence



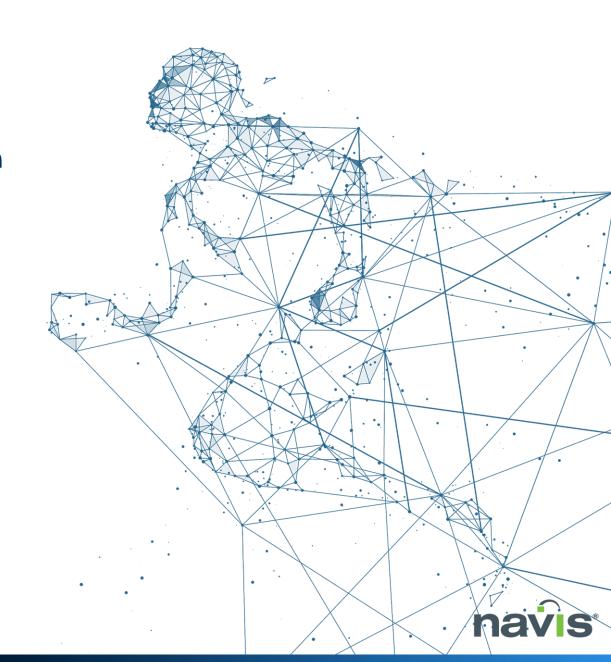
Having the necessary information immediately available...



to deal with the situation at hand quickly and efficiently...



in a 12 to 72 hour period.



We Considered a lot of Different Use Cases

Towards more accurate situational awareness

PREDICT CARGO VOLUME, MODE, TIMING, COST/REVENUE

- Predict equipment needs
- Predict the need to prep for wind/weather
- Predict estimated time of arrival of ships, trains or trucks
- Predict mode of departure
- Predict long term yard utilization
- Suggest appointment quota per hour
- Suggest block space in yard for recurring cargo
- Predict gate visits, pickup time of imports
- Predict estimated time of departure of vessels
- Predict container damage
- Predict weight discrepancies

DYNAMICALLY SET PARAMETERS THAT CONTROL THE SOFTWARE

- Set expert decking parameters
- Update allocations
- Set prime route parameters
- Dynamic drive time calculation
- Improve EDI data accuracy
- Automate gate rule failures (including fail to deck)
- Optimize berth and crane assignment
- Set all manual input needed for automated vessel planning
- Auto create pre-defined work patterns

AUGMENT EXISTING DECISION-MAKING SOFTWARE WITH INSIGHTS GENERATED THROUGH DATA

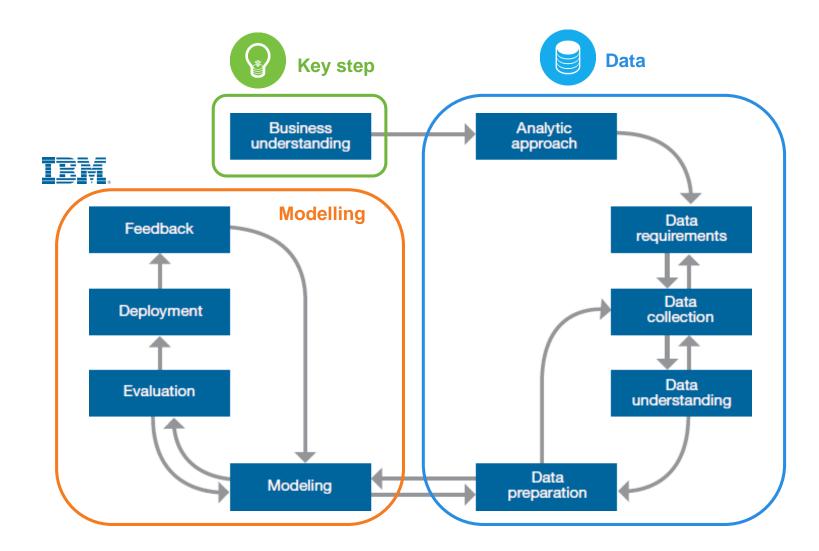
- Optimize AGV/ASC scheduling
- Choose the right AutoStow strategy

RECOMMEND ACTIONS FOR THE CONTROL ROOM

- Predict congestions
- Manage pushrate/MaxPM
- Update CHE ranges
- Hazardous planning
- Suggest time of replanning or work queue balancing without replanning
- Understand cause and duration of crane delays and/or AGV delays
- Recommend multi-lift opportunities
- Learn how user handles exceptions and do it automatically
- Activate optimum automatic work queue activation
- Recommend times for lower energy consumption
- Suggest housekeeping/work for idle equipment

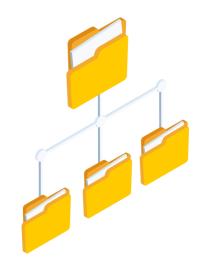


Methodology for Data Science



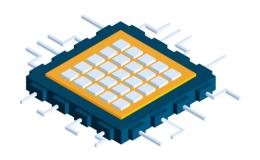


Built Two Models



Extreme Gradient Boosting Model:

- Mean Absolute Error: 5.7 Moves Per Hour
- Accuracy score: 92.4%



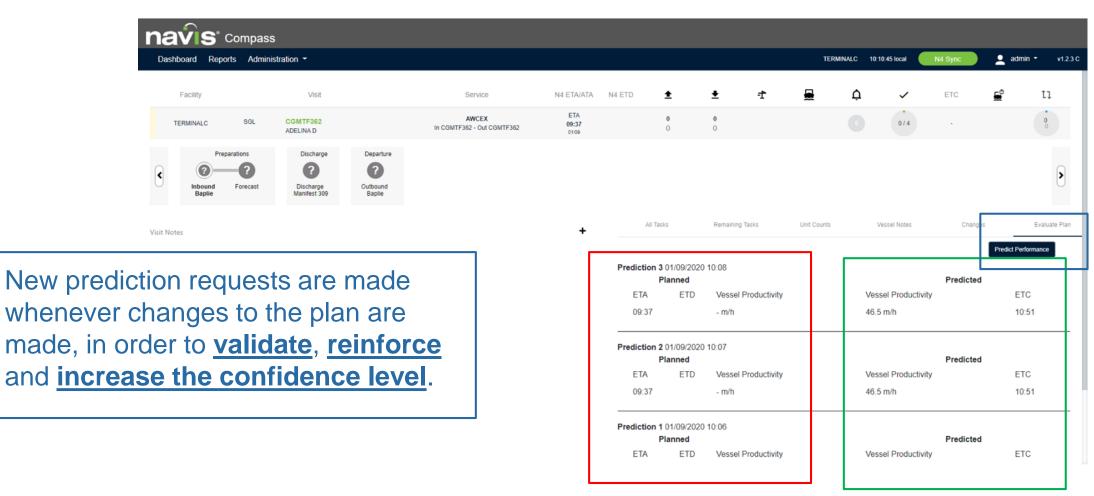
Deep Neural Network (Deep Learning) Model:

- Mean Absolute Error: 1.5 Moves Per Hour
- Accuracy score: 98.4%



Leveraging Data Across Our Platform

Integrating Predictions



Prepare for the Port of the Future



Equipment is the Key Driver

Workers link individual process steps and direct yard operations



Management by Process

- Business processes mapped to TOS
- Decisions made by control tower operators



Automated Equipment

- Scheduling & Optimization Algorithms
- Management by Exception



Al Driven Ecosystem

- Automated Control Rooms with Social Cobots
- Optimizing flow of cargo between Carriers, port complex, and larger logistics supply chain



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THANK YOU!



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