



# Simulation-assisted machine learning for yard departure prediction: a benchmarking study between European and North American contexts

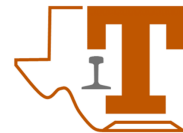
- Niloofar Minbashi
- PhD Candidate
- KTH Royal Institute of Technology



Minbashi, N., Zhao, J., Dick, C. T., Bohlin, M. (2023, April 25-28). Application of simulation-assisted machine learning for yard departure prediction. Presented at the 10th International Conference on Railway Operations Modelling and Analysis (ICROMA), RailBelgrade 2023, Belgrade, Serbia.



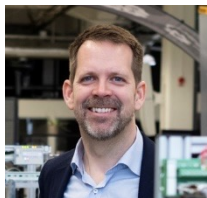
# Agenda



- Motivation
- Overview
- Research Questions
- Method
- Results
- Conclusion



- PhD thesis: **Application of Predictive Analytics for Shunting Yard Delays**
- October 2018 – June 2023
- Licentiate defense November 2020
  
- Supervisors:



Markus Bohlin  
Visiting Professor, KTH

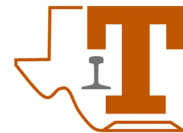


Behzad Kordnejad  
Researcher, KTH



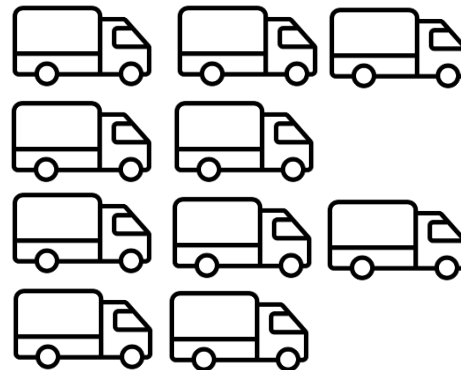
Carl-William Palmqvist  
Researcher, KTH

# Motivation



$\approx 10 \times 10^9$  tonne – kilometers

$79 \times 10^6$  tonnes of  $CO_2$



$\approx 20 \times 10^9$  tonne – kilometers

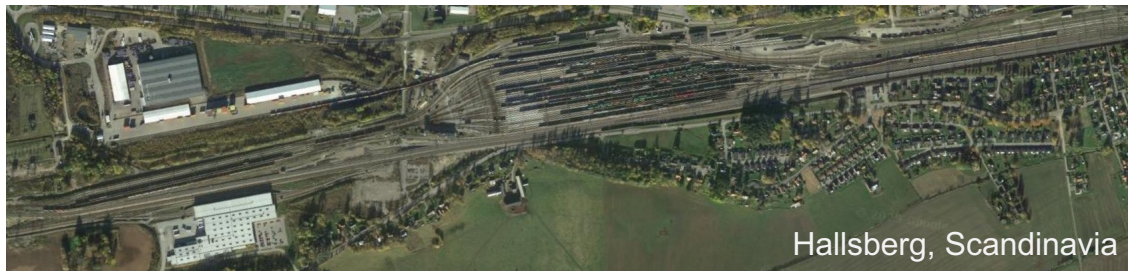
$1800 \times 10^6$  tonnes of  $CO_2$



# 50 by 2050



# Yards



Hallsberg, Scandinavia

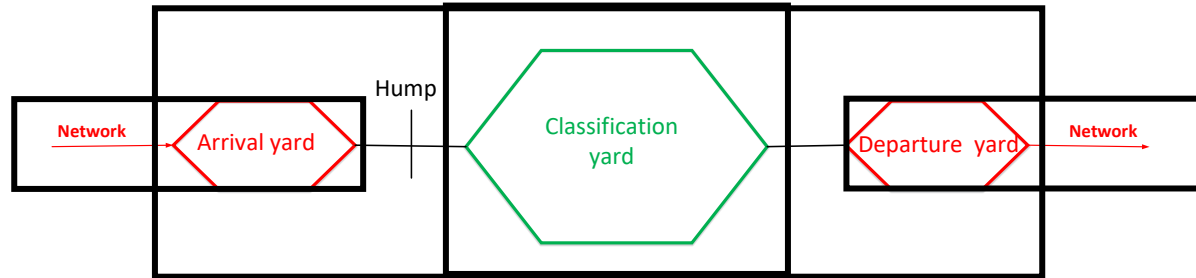


Maschen, Europe



Bailey, North America

# Research Problem





# Research Gaps

## 1. Methods: Data-driven

1. Descriptive Analytics
2. Predictive Analytics
3. Hybrid Modeling

## 2. Yard performance:

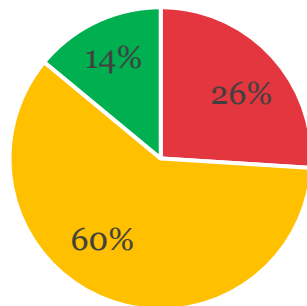
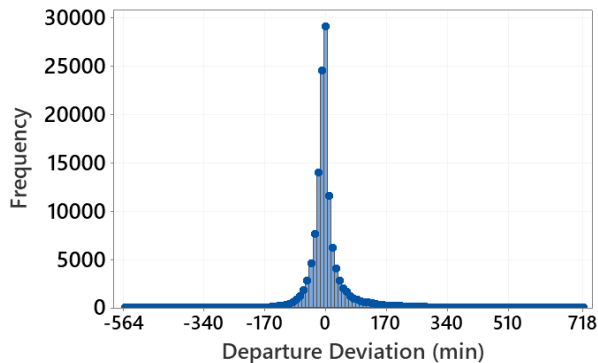
1. Analyzing yard departure deviations
2. Analyzing the network impact on departure delays

## 3. Yard models:

1. Yard departure prediction models
2. Hybrid yard and network models
3. Hybrid yard models: event-driven and data-driven models



# Research Problem (Swedish Context): Yard Departure Prediction



■ Delayed ■ Early ■ On Time

Infrastructure Manager  
Yard Operator



# Descriptive Analytics: Probability Distribution



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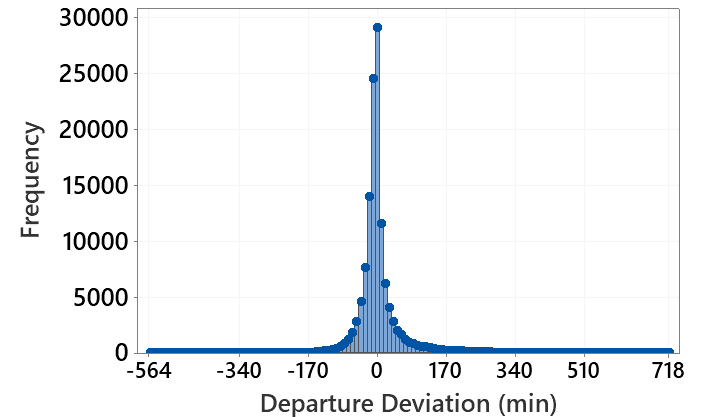
Statistical analysis of departure deviations from shunting yards:  
Case study from Swedish railways

Niloofer Minabshi<sup>a,\*</sup>, Carl-William Palmqvist<sup>b,c</sup>, Markus Bohlin<sup>a</sup>,  
Behzad Kordnejad<sup>a</sup>

<sup>a</sup> Division of Transport Planning, School of Architecture and Built Environment, KTH, SE-100 44, Stockholm, Sweden

<sup>b</sup> Division of Transport and Roads, Department of Technology and Society, Lund University, P.O. Box 118, 221 00, Lund, Sweden

<sup>c</sup> K2 Swedish Knowledge Centre for Public Transport, Bruksgatan 8, 222 36, Lund, Sweden





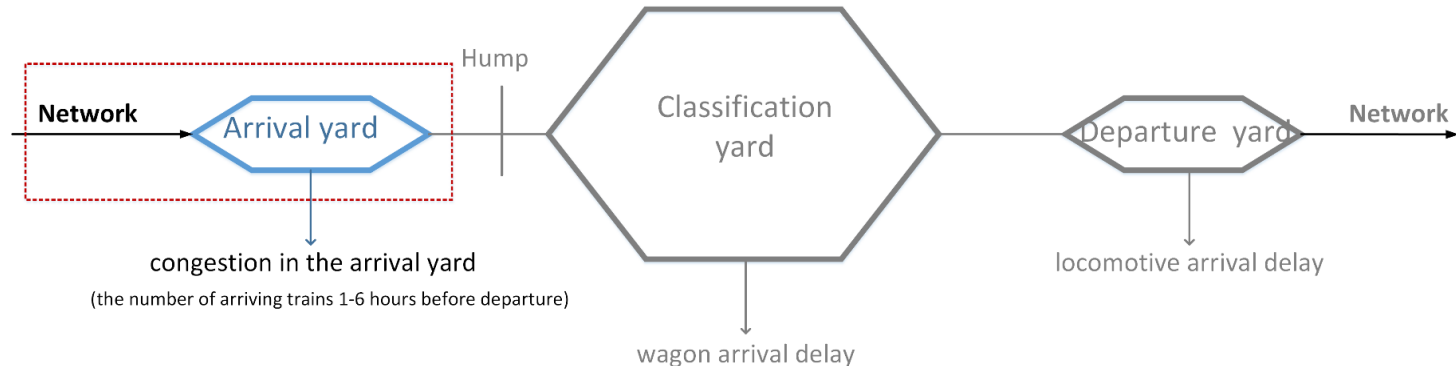
# Descriptive Analytics: Network Impact



Minbashi, N., Bohlin, M., Kordnejad, B. (2020). A departure delay estimation model for freight trains. TRA 2020, the 8th Transport Research Arena: Rethinking transport – towards clean and inclusive mobility, Helsinki, Finland.



Minbashi, N., Bohlin, M., Kordnejad, B. (2021, February 3-4). Analysis of railyard congestion and departure delays relationship: a case study from Swedish railways. hEART 2020: 9th Symposium of the European Association for Research in Transportation, Lyon, France.





# Predictive Analytics: Yard Departure Prediction Model

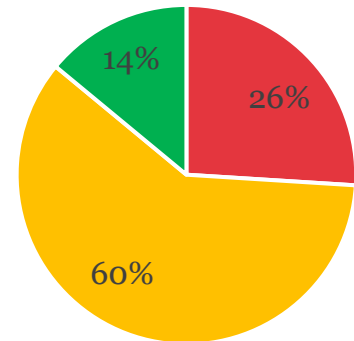
*Research Article*

## The Application of Tree-Based Algorithms on Classifying Shunting Yard Departure Status

**Niloofer Minbashi** <sup>1</sup>, **Markus Bohlin** <sup>1</sup>, **Carl-William Palmqvist** <sup>2</sup>,  
**and Behzad Kordnejad** <sup>1</sup>

<sup>1</sup>Division of Transport Planning, KTH Royal Institute of Technology, 100 44 Stockholm, Sweden

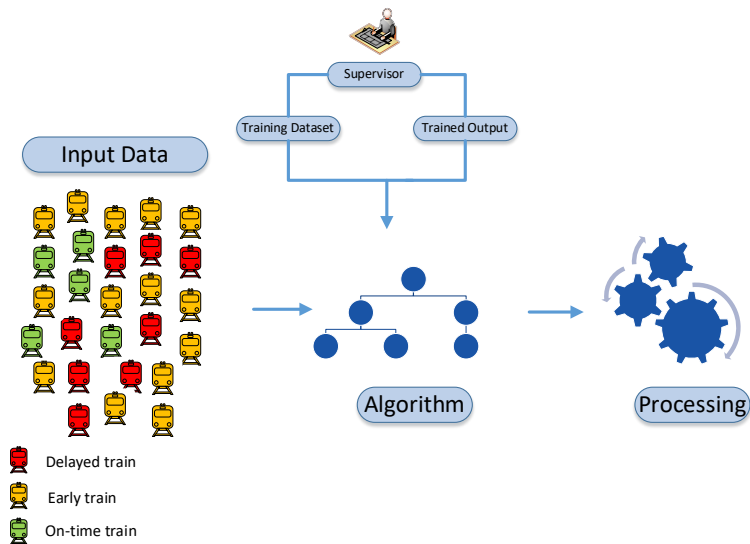
<sup>2</sup>Division of Transport and Roads, Lund University, P.O. Box 118, 221 00 Lund, Sweden



■ Delayed ■ Early ■ On Time



# Model Framework



## Application of Simulation-assisted Machine Learning for Yard Departure Prediction

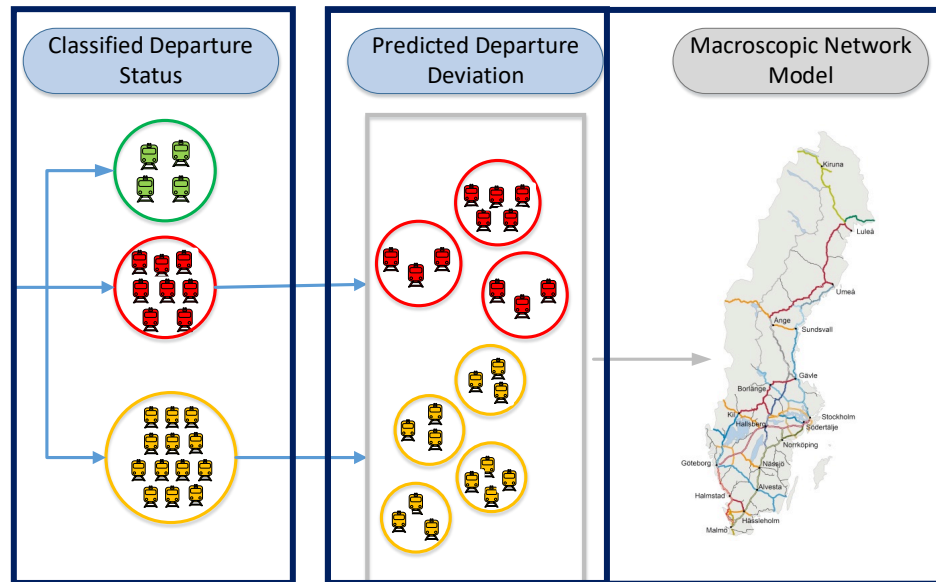
Niloofer Minbashi <sup>a,1</sup>, Jiayi Zhao <sup>b</sup>, C. Tyler Dick <sup>b</sup>, Markus Bohlin <sup>a,c</sup>

<sup>a</sup> Division of Transport Planning, KTH Royal Institute of Technology, SE-100 44, Stockholm, Sweden

<sup>b</sup> Department of Civil, Architectural and Environmental Engineering, University of Texas at Austin, Austin, TX, USA

<sup>c</sup> School of Innovation, Design and Technology, Mälardalen University, Eskilstuna, Sweden

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Research Article

### The Application of Tree-Based Algorithms on Classifying Shunting Yard Departure Status

Niloofer Minbashi <sup>1</sup>, Markus Bohlin <sup>1</sup>, Carl-William Palmqvist <sup>2</sup>, and Behzad Kordnejad <sup>1</sup>

<sup>1</sup>Division of Transport Planning, KTH Royal Institute of Technology, 100 44 Stockholm, Sweden

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journal homepage: [www.elsevier.com/locate/jrtpm](http://www.elsevier.com/locate/jrtpm)



Machine learning-assisted macro simulation for yard arrival prediction

Niloofer Minbashi <sup>1</sup>, Hans Sipilä, Carl-William Palmqvist, Markus Bohlin, Behzad Kordnejad

Division of Transport Planning, KTH Royal Institute of Technology, Brinellvägen 23, SE-100 44 Stockholm, Sweden

## Application of Simulation-assisted Machine Learning for Yard Departure Prediction

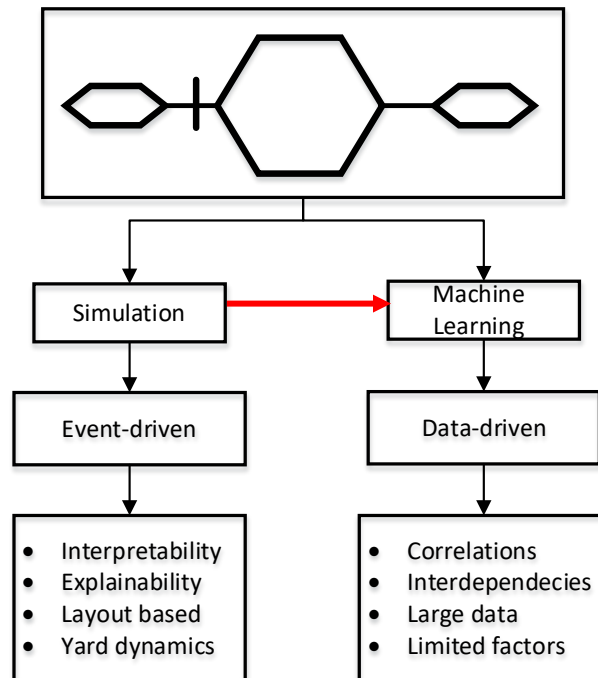
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<sup>a</sup> Division of Transport Planning, KTH Royal Institute of Technology, SE-100 44, Stockholm, Sweden

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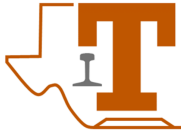
<sup>c</sup> School of Innovation, Design and Technology, Mälardalen University, Eskilstuna, Sweden

<sup>1</sup> E-mail: minbashi@kth.se, Phone: +46 (8) 790 70 54





# Research Questions

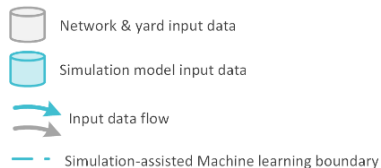
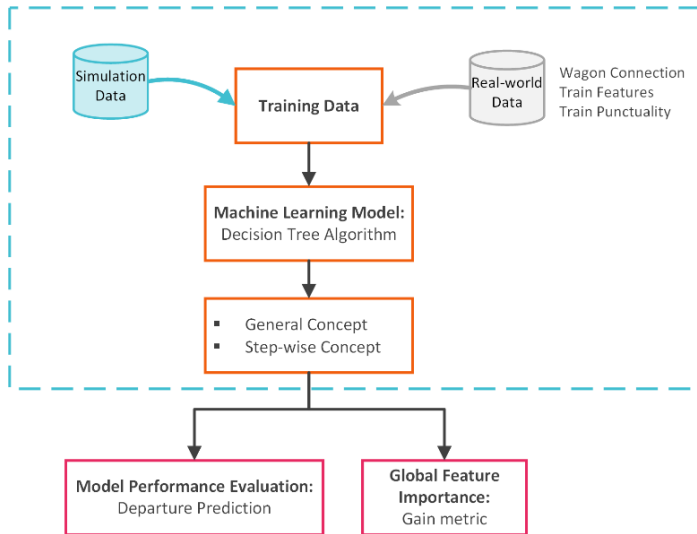


- 1) Can synthetic data from simulation in a North American context validate a machine learning model developed on real-world yard data from the European context?
- 2) Can simulation data improve the machine learning by providing access to more operational factors?
- 3) Can a model developed from simulation data inform the collection of new performance metrics to improve a model developed from the real-world data?



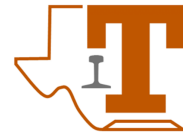
# Method:

## Simulation-assisted Machine Learning





# Method: Machine Learning

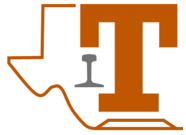


- Original model extended from (Minbashi et al., 2021 & 2023)
- Decision Tree
- 10-fold cross-validation
  
- Evaluation Criteria:
  - R-Squared
  - Mean absolute error
  - Root mean squared error
  - Mean signed difference



# Method:

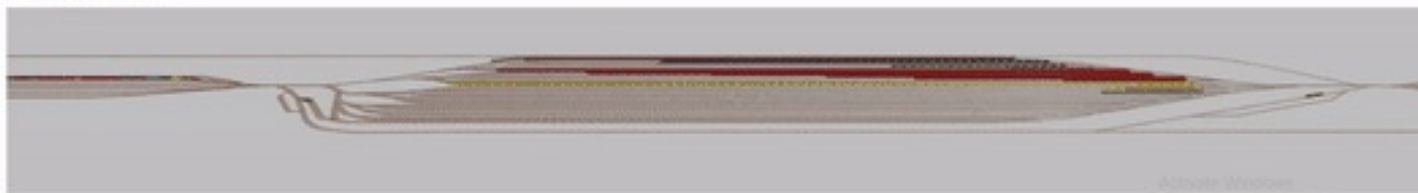
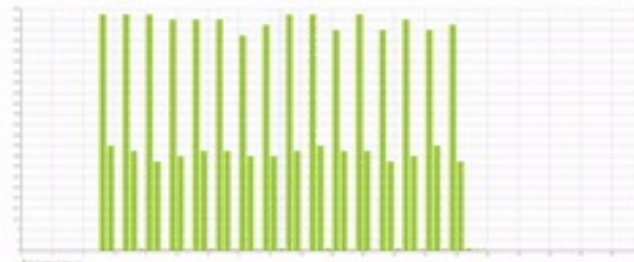
## ML - Predictors



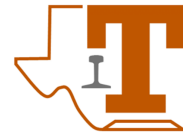
Schedule Data
Departing Train Number
Departure Hour
Departure Week-Day
Departure Month
Min Wagon Dwell Time
Max Wagon Dwell Time
Number of Wagons
Number of Arriving Trains
Maximum Planned Weight
Maximum Planned Length

Operational Data
Maximum Operated Length
Maximum Operated Weight
Min Arriving Deviation
Mean Arriving Deviation

# Method: Discrete Event Simulation



# Method: Simulation - Predictors

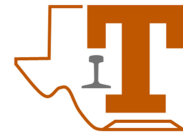


Data Type	Predictors
Schedule	Scheduled departure hour
	Train length
	Number of arriving trains
	Number of wagons
Arrival yard	Minimum hump duration
	Maximum hump duration
	Receiving utilization
	Hump utilization
	Idle time waiting for hump
Classification yard	Minimum bowl idle
	Maximum bowl idle
	Pulldown utilization
Departure yard	Departure utilization
	Departure inspection time
	Idle time waiting for departure



# Results:

## General Concept



<b>Model (General Concept)</b>	<b>R-Squared</b>	<b>Mean Absolute Error (min)</b>	<b>Root Mean Squared Error (min)</b>	<b>Mean Signed Difference</b>
Real-world	0.90	3.4	14.6	-0.004
Baseline simulation	0.87	3.0	6.6	0.002
Ultimate Randomness simulation	0.70	6.6	12.6	0..002

# Results:

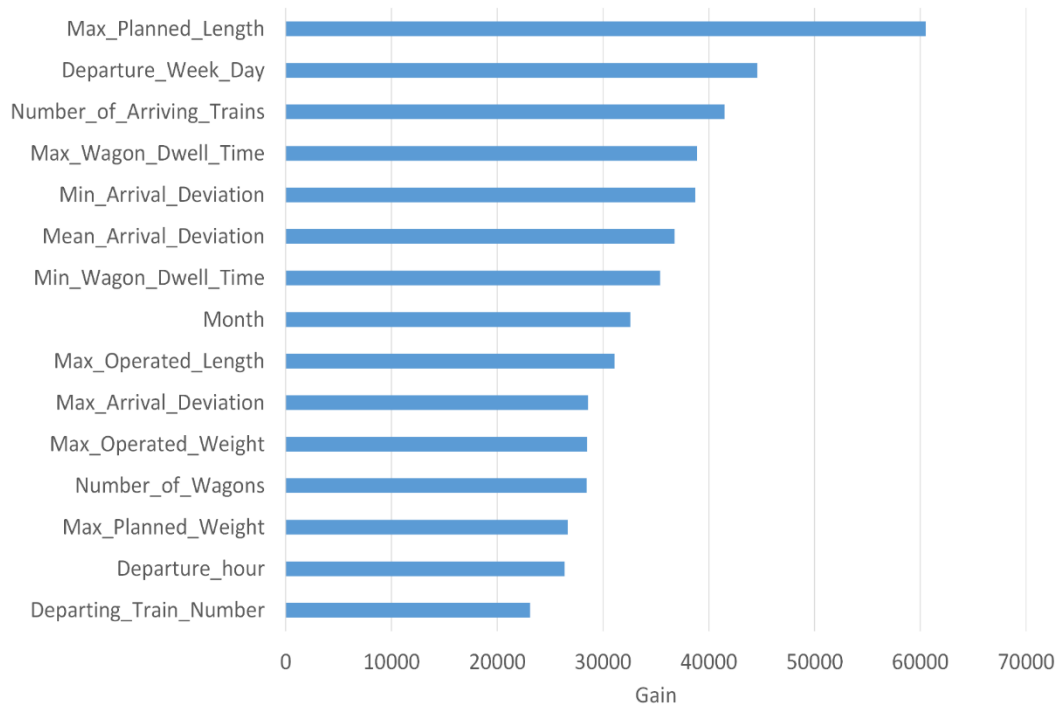
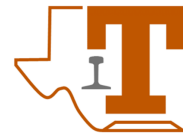
## Step-wise Concept

Model	Step-wise Concept	R <sup>2</sup>	MAE	RMSE	MSD
Real-world	Step 0 Schedule	0.90	3.5	15.18	0.05
	Step 1 Arrival	0.88	3.6	16.4	0.11
	Step 2 Classification	0.90	3.5	15.31	0.06
Baseline simulation	Step 0 Schedule	0.80	4.2	7.80	0.00
	Step 1 Arrival	0.79	4.2	8.40	0.00
	Step 2 Classification	0.79	4.2	8.40	0.002
	Step 3 Departure	0.87	3.0	6.00	0.001
Ultimate Randomness simulation	Step 0 Schedule	0.57	9.0	16.00	0.003
	Step 1 Arrival	0.60	8.0	15.00	0.003
	Step 2 Classification	0.62	9.0	15.00	0.001
	Step 3 Departure	0.70	7.0	13.00	0.003



# Results:

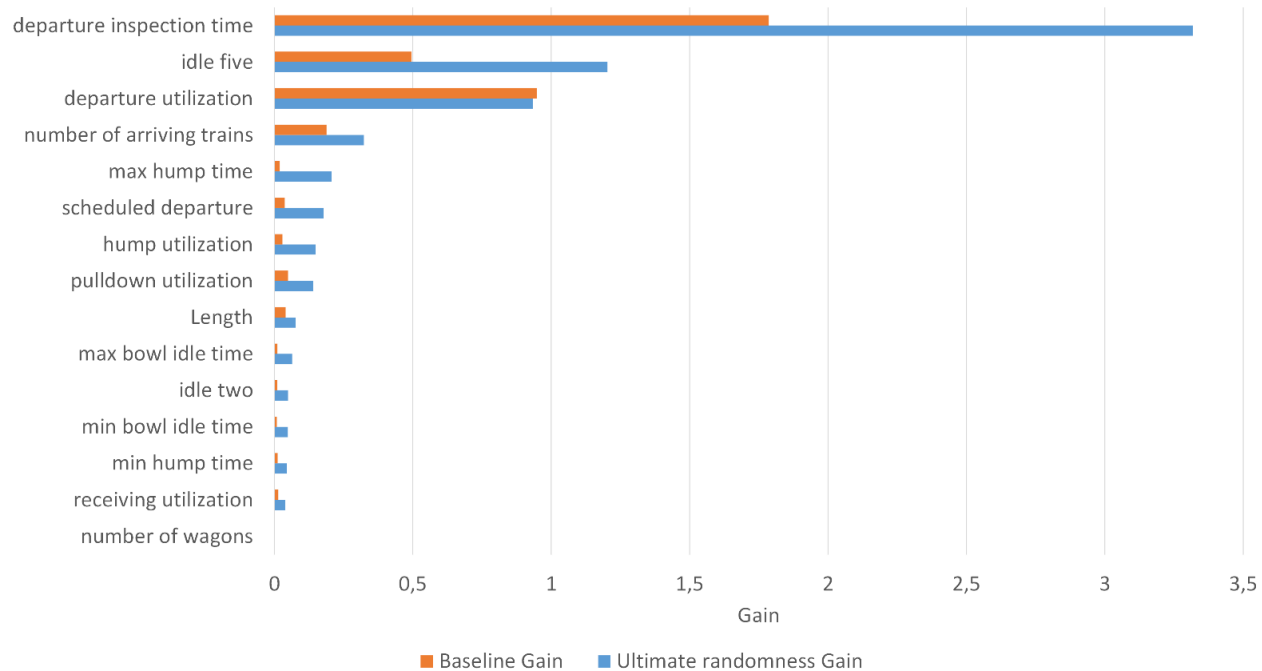
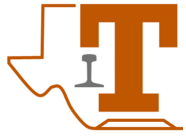
## Global Feature Importance





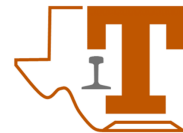
# Results:

## Global Feature Importance





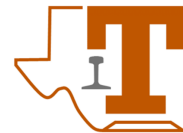
# Conclusion



- Machine learning performs well for both contexts.
- Simulation data can help in validating model performance.
- Simulation data can be beneficial where data is sensitive.
- Schedule predictors have enough information for accurate departure predictions.
- Departure yard predictors add higher information for yard departure prediction.
- Adding randomness to the simulation should be explored in explainability terms.



# Acknowledgement





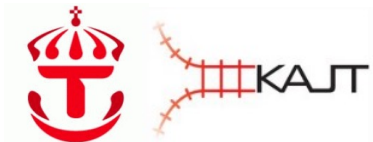
# Questions?

**Niloofer Minbashi**

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[jiayi.zr@utexas.edu](mailto:jiayi.zr@utexas.edu)



Stockholm, 2022-06-13





# List of papers:

- I. Minbashi, N., Palmqvist, C. W., Bohlin, M., Kordnejad, B. (2021). Statistical analysis of departure deviations from shunting yards: case study from Swedish railways. *Journal of Rail Transport Planning & Management*, 18, 100248.
- II. Minbashi, N., Bohlin, M., Kordnejad, B. (2020). A departure delay estimation model for freight trains. *Conference Proceedings of 8th Transport Research Arena TRA*, Helsinki, Finland.
- III. Minbashi, N., Bohlin, M., Palmqvist, C. W., Kordnejad, B. (2021). The application of tree-based algorithms on classifying shunting yard departure status. *Journal of Advanced Transportation*, 2021, 1-10.
- IV. Minbashi, N., Sipilä, H., Palmqvist, C. W., Bohlin, M., Kordnejad, B. (2023). Machine learning-assisted macro simulation for yard arrival prediction. *Journal of Rail Transport Planning & Management*, 25, 100368.
- V. Minbashi, N., Zhao, J., Dick, C. T., Bohlin, M. (2023). Application of Simulation-assisted machine learning for yard departure prediction. Accepted at the 10th International Conference on Railway Operations Modelling and Analysis (ICROMA). RailBelgrade 2023.
- VI. Minbashi, N., Bohlin, M., Kordnejad, B. (2021, February 3-4). Analysis of railyard congestion and departure delays relationship: a case study from Swedish railways. *hEART 2020: 9th Symposium of the European Association for Research in Transportation*, Lyon, France.