

Evaluating Imputation Models for Totals Considering Missing Patterns

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Session E-6: The Missing Data Puzzle: Exploring Imputation Method:
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Outline



INTRODUCTION

NCFO / Imputation Study

DATA

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METHODS

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RESULTS

Individual Models / Combined Models / Overall Best Models

CONCLUSIONS

INTRODUCTION



National Census of Ferry Operators



- The Safe, Accountable, Flexible, Efficient, Transportation Equity Act—A Legacy for Users (SAFETEA-LU) of 2005 (P.L. 114-94) requires BTS to maintain a national ferry database.
- BTS conducts a biennial census of all ferry operators in the U.S. and its territories.

2020 NCFO Imputation Study



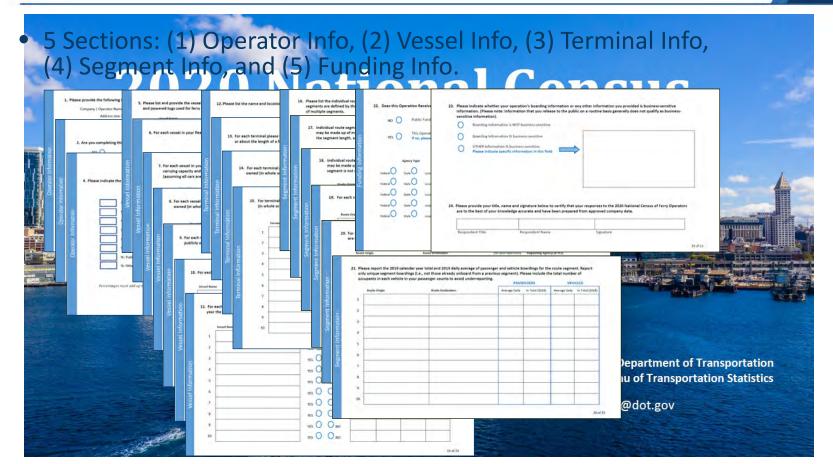
Background:

- ✓ Passenger and vehicle boarding counts are key information.
- √ 34% of ferry segments have missing passenger boarding in 2020 NCFO
- ✓ Missing counts make it difficult to monitor changes in national totals

DATA



2020 NCFO Questionnaire



2020 NCFO Data Release

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548



5 Tables	;	Operator	Operato	or-Segment	Segmer	nt T	erminal	Vesse	el			
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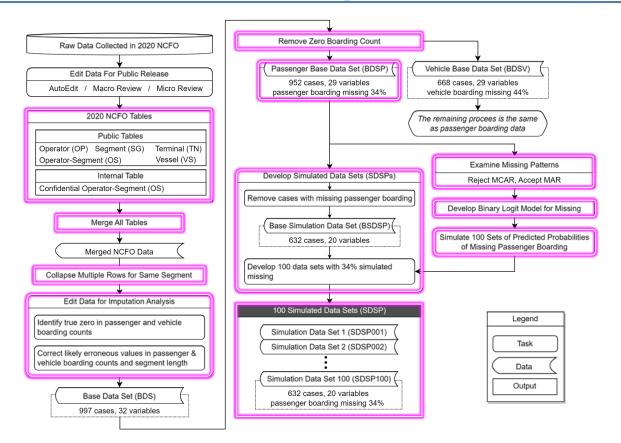
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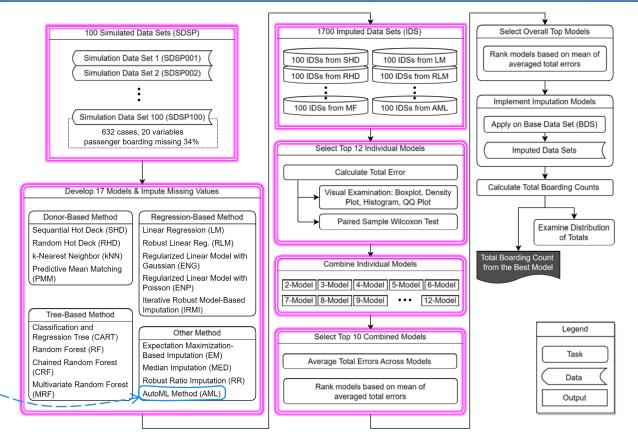
METHODS



Analysis Flow: Data Prep



Analysis Flow: Modeling



Individual Models



Donor-Based Method
Sequantial Hot Deck (SHD)
Random Hot Deck (RHD)
k-Nearest Neighbor (kNN)
Predictive Mean Matching
(PMM)

Tree-Based Method

Classification and Regression Tree (CART) Random Forest (RF) Chained Random Forest (CRF) Multivariate Random Forest (MRF) Linear Regression (LM)
Robust Linear Reg. (RLM)
Regularized Linear Model with
Gaussian (ENG)
Regularized Linear Model with
Poisson (ENP)
Iterative Robust Model-Based
Imputation (IRMI)

Regression-Based Method

Other Method Expectation Maximization-

Based Imputation (EM)
Median Imputation (MED)
Robust Ratio Imputation (RR)
AutoML Method (AML)

Combined Models

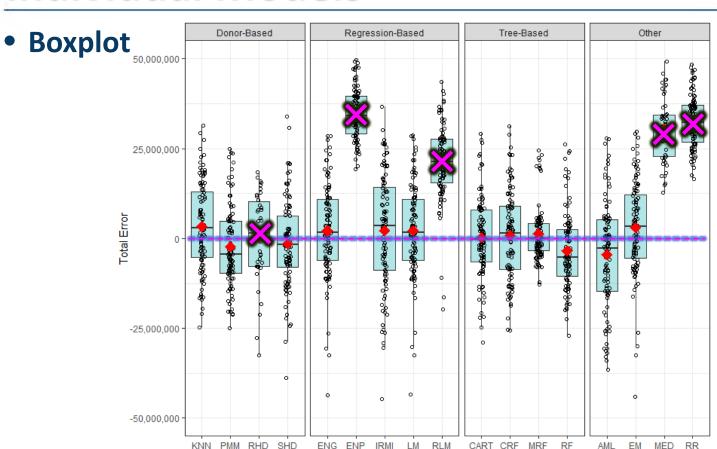


RESULTS



Individual Models





Model

Top 12 Individual Models

1 CART

2 CRF

3 MRF

4 SHD

5 ENG

6 LM

O LIVI

7 IRMI

8 EM

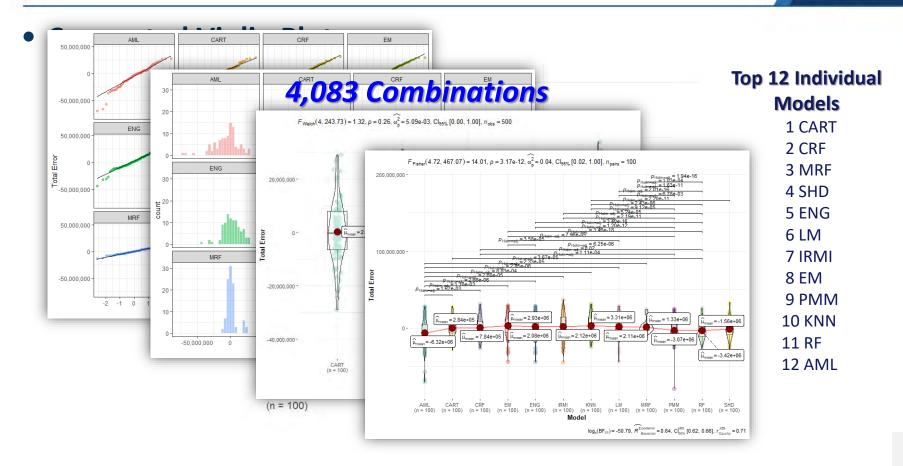
9 PMM

10 KNN

11 RF

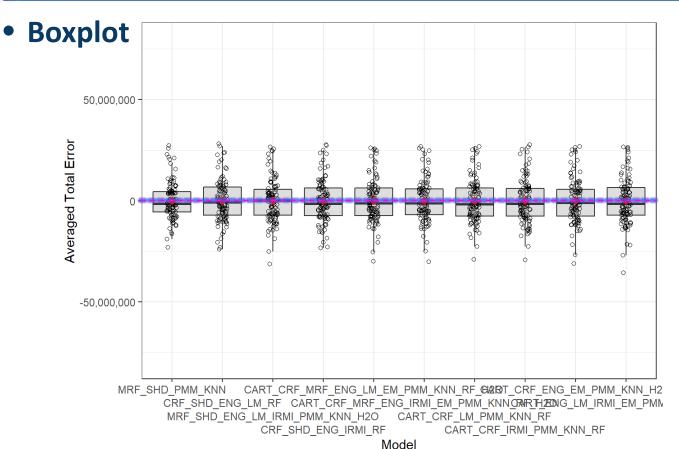
12 AML

Combined Models



Combined Models 2



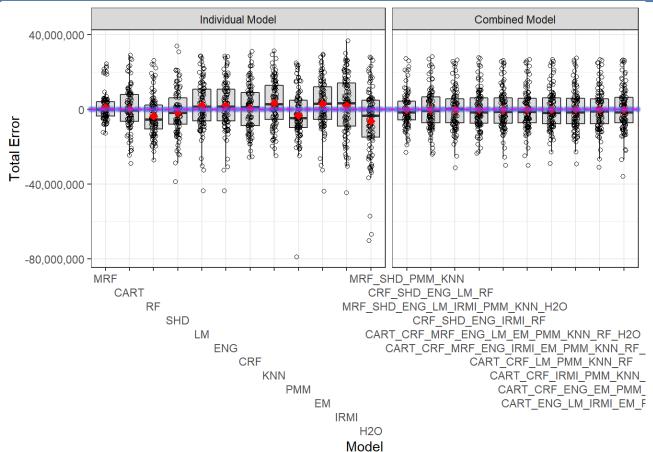


Top 10 Models

- 1 MRF_SHD_PMM_KNN
- 2 CART_CRF_IRMI_PMM_KNN_RF
- 3 CART_CRF_LM_PMM_KNN_RF
- 4 CRF_SHD_ENG_IRMI_RF
- 5 CART_CRF_ENG_EM_PMM_KNN_H20
- 6 CART_CRF_MRF_ENG_LM_EM_PMM_KNN_RF_H20
- 7 CRF_SHD_ENG_LM_RF
- 8 CART_CRF_MRF_ENG_IRMI_EM_PMM_KNN_RF_H20
- 9 MRF_SHD_ENG_LM_IRMI_PMM_KNN_H2O
- 10 CART_ENG_LM_IRMI_EM_PMM_KNN_RF_H20







CONCLUSIONS

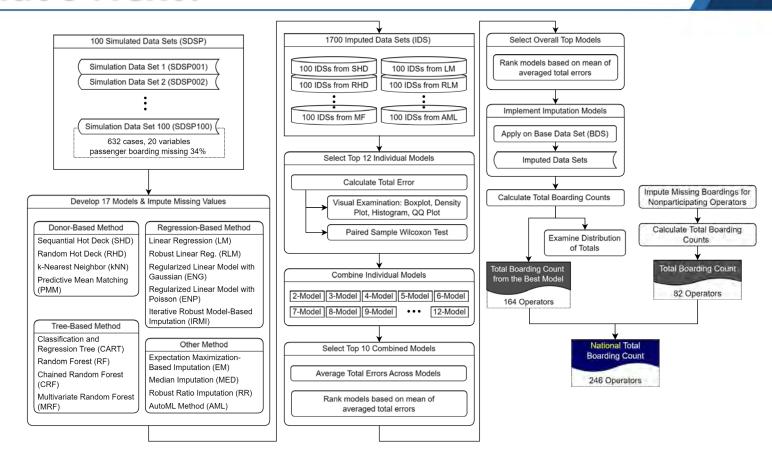


Conclusions



- Imputation for estimating **totals**Limited to this study
- Some models show very poor imputation performance: ENP, RLM, RR.
 - Attributable to the evaluation metric
- AutoML does not live up to expectation
- Averaging results from individual models stabilizes the distribution of total errors. But, majority of variation stays
- Thus, the best model cannot be determined statistically

What's Next?



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Questions?

