

# Predicting WRTA Bus Travel Time

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Final Project – Data  
Science 2010

By: Daniel Kwan



## Methodology Workflow

1. Load and Clean Dataset
2. Select Key Features
3. Convert Time Columns
4. One-Hot Encode Categorical Variables
5. Train-Test Split (80/20)
6. Train Models (Linear, RF, XGBoost)
7. Evaluate Models ( $R^2$ , MSE)
8. Predict Stop-to-Stop Travel Time

Goal:

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To predict actual WRTA bus travel time between stops using historical data found on Kaggle.

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What are the questions?:

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Can Travel time be predicted using route, time of day, speed, and stop data?

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How do different regression models compare in accuracy?

# Dataset Overview

- Rows: 115,147
- Features: 44
- Target Variable: Running Time Act
- Key features:

```
columns_to_keep = [  
    'TIMEPOINT_MILES', 'TRIP_START_TIME', 'ROUTE_NAME', 'DIRECTION_NAME',  
    'SERVICE_PERIOD', 'TIME_SCHEDULED', 'SPEED_SCHEDULED', 'SPEED_ACTUAL', 'FIRST_LAST_STOP',  
    'STOP_ID_1', 'STOP_ID_2', 'RUNNING_TIME_ACT', 'TIME_ACTUAL_DEPART'  
]
```

- 17. 📅 TIME\_PERIOD\_SORT: Numeric code for sorting the time periods.
- 18. 📅 SORT\_ORDER\_1: Sort index for sequence of observations.
- 19. 🚻 SORT\_ORDER\_2: Secondary sort order (usually 0).
- 20. 🚻 TIMEPOINT\_ID\_1: ID for the origin stop of this segment.
- 21. 🚻 TIMEPOINT\_ID\_2: ID for the destination stop of this segment.
- 22. 🚻 TIMEPOINT\_NAME\_1: Text name of the origin stop.
- 23. 🚻 TIMEPOINT\_NAME\_2: Text name of the destination stop.
- 24. 🚻 STOP\_ID\_1: Alternate ID for the origin stop.
- 25. 🚻 STOP\_ID\_2: Alternate ID for the destination stop.
- 26. 🚻 STOP\_KEY\_1: Another stop key for the origin stop.
- 27. 🚻 STOP\_KEY\_2: Another stop key for the destination stop.
- 28. 🕒 TIME\_SCHEDULED: Scheduled time of arrival at the destination stop.
- 29. 🚻 TIMEPOINT\_MILES: Distance in miles between the two stops.
- 30. 🕒 READ\_TIME: Time the data collection system recorded the bus.
- 31. 📅 READ\_DATE: Date the reading was captured.
- 32. 🕒 RUNNING\_TIME\_SCH: Scheduled duration for this segment (in minutes).
- 33. 🕒 RUNNING\_TIME\_ACT: Actual duration the bus took.
- 34. 🕒 RUNNING\_TIME\_DIFF: Difference = Actual - Scheduled (positive = late).
- 35. 🚻 TIMEPOINT\_DWELL\_1: Time bus spent waiting at the first stop.
- 36. 🚻 TIMEPOINT\_DWELL\_2: Time spent at the second stop.
- 37. 🚻 SPEED\_SCHEDULED: Planned average speed between the two stops (mph).
- 38. 🚻 SPEED\_ACTUAL: Actual average speed.
- 39. 🚻 FIRST\_LAST\_STOP: Flag (1 or 2) indicating if it's the start or end of the route.
- 40. 🕒 TIME\_ACTUAL\_ARRIVE: Actual arrival time at the destination.
- 41. 🕒 TIME\_ACTUAL\_DEPART: Actual departure time.
- 42. ✖️ ONTIME\_METHOD\_1: On-time performance category (1 = on time, 2 = late, etc.).
- 43. 📅 ONTIME\_METHOD\_2: Alternative method of determining on-time status.
- 44. 📅 TRIPS\_COUNT: Number of trips aggregated into this row (often 0 = individual trip).

# Data Cleaning

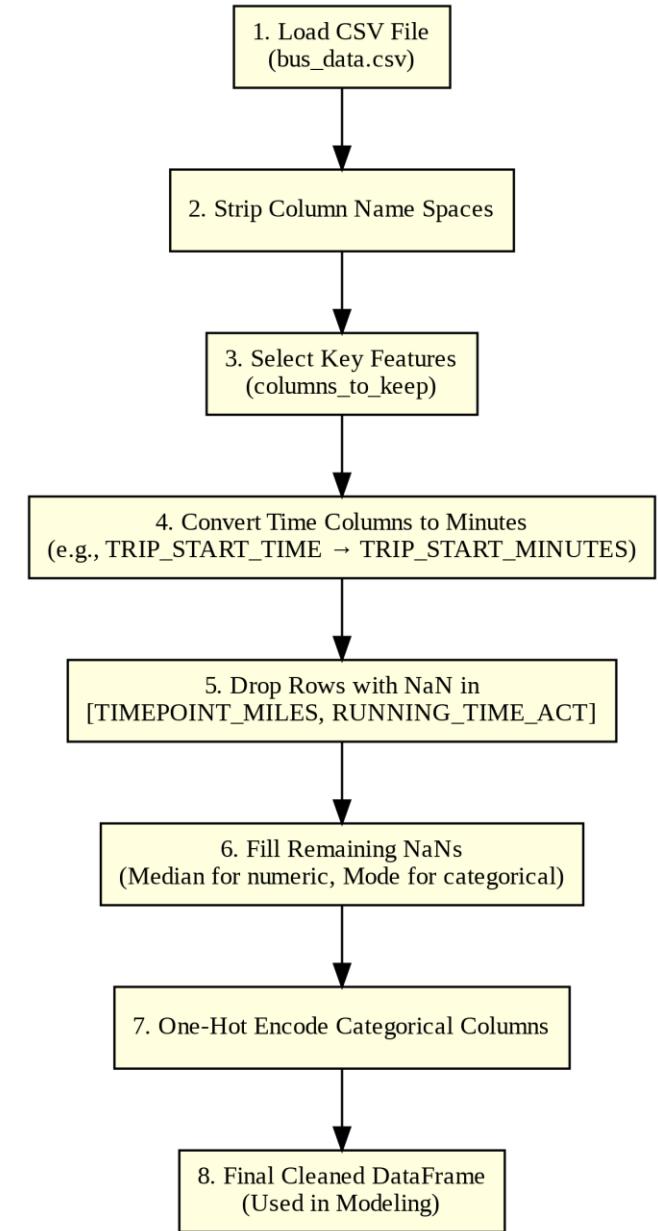
Dropped unnecessary columns,  
retained 13 key features.

Converted time fields to minutes  
after midnight.

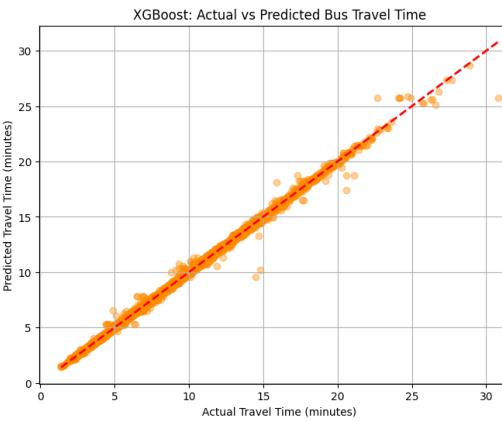
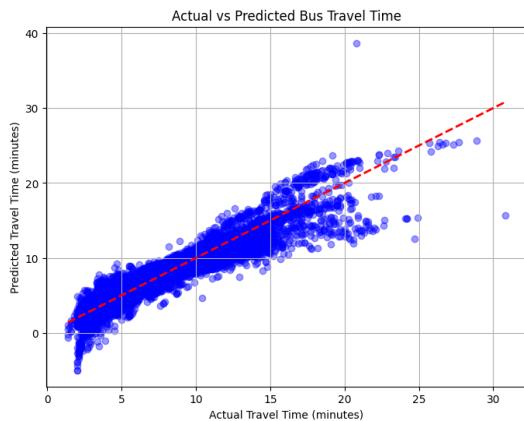
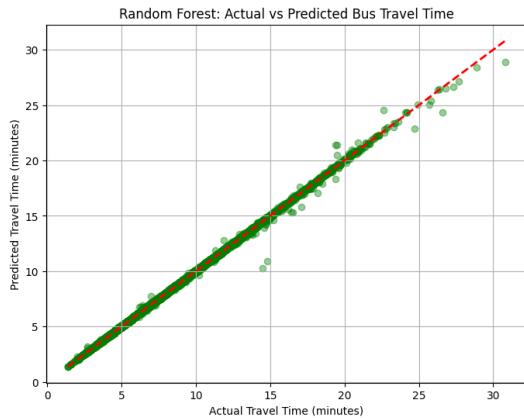
One-hot encoded categorical  
columns (ROUTE\_NAME,  
DIRECTION\_NAME,  
SERVICE\_PERIOD)

Removed rows with NaNs in key  
fields, filled others with  
median/mode.

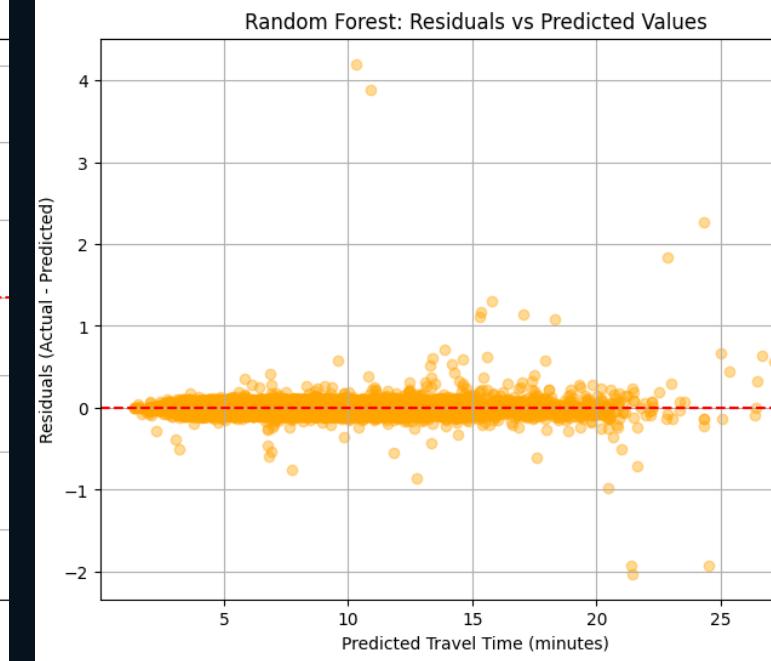
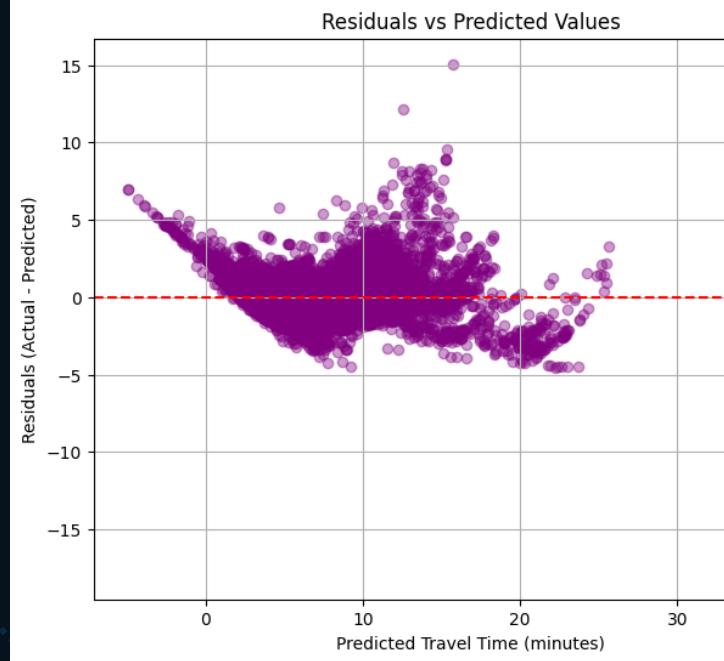
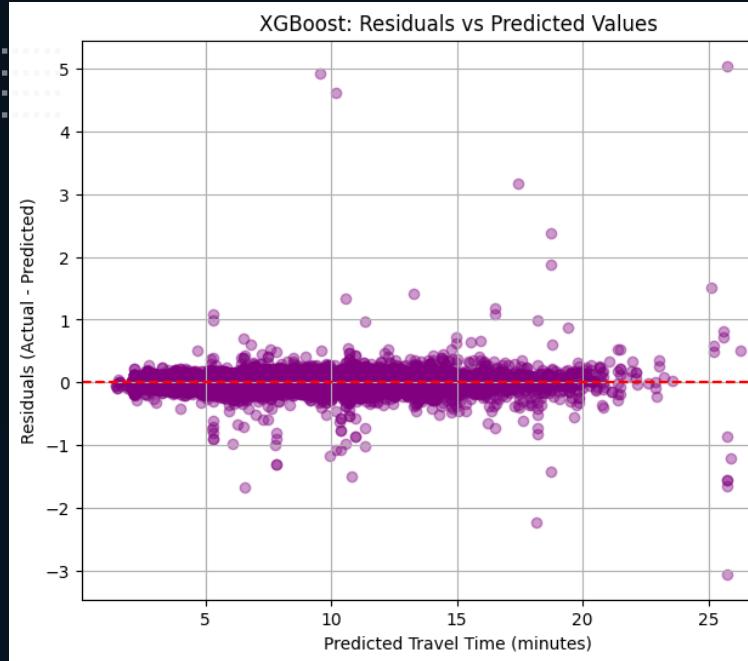
Verified all data was numeric and  
ready for modeling.



# Models Tried:



1. Linear Regression
  - Simple, fast, interpretable
  - Struggled with non-linearity
  - $R^2 = 0.89$
2. Random Forest Regressor
  - Handles non-linearity well
  - Best performer ( $R^2 = 0.9995$ )
3. XGBoost Regressor
  - Gradient boosting, slightly below RF
  - $R^2 = 0.9986$



Linear:

# Residual Graphs

Random Forest

XGBoost:

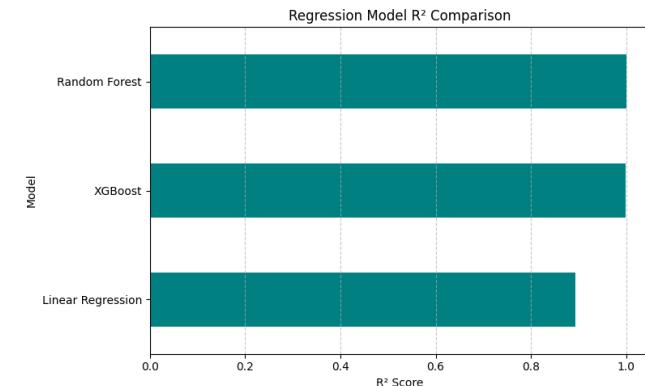
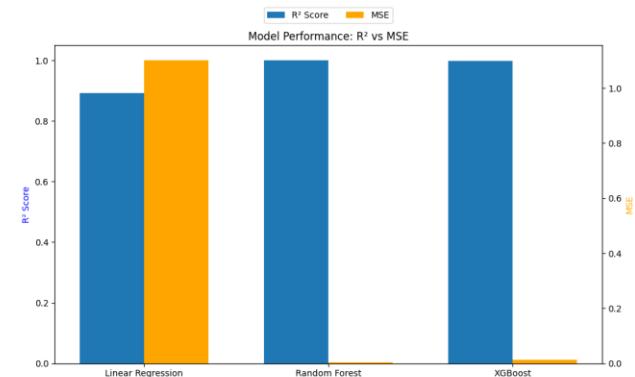
- Linear: High Variance
- Random Forest and XGBoost: Low, mostly centered around 0

# Results Summary

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- Three regression models were evaluated: Linear Regression, Random Forest, and XGBoost.
- Random Forest achieved the best performance with an  $R^2$  score of 0.9995 and an MSE of 0.0048, indicating highly accurate predictions.
- XGBoost also performed exceptionally well ( $R^2 = 0.9986$ , MSE = 0.0137), only slightly less accurate than Random Forest.
- Linear Regression, while still useful, showed significantly lower performance ( $R^2 = 0.8928$ , MSE = 1.10), suggesting that linear assumptions were too simplistic for the data.
- The chart clearly shows that tree-based models significantly outperform linear methods in predicting WRTA bus travel time.

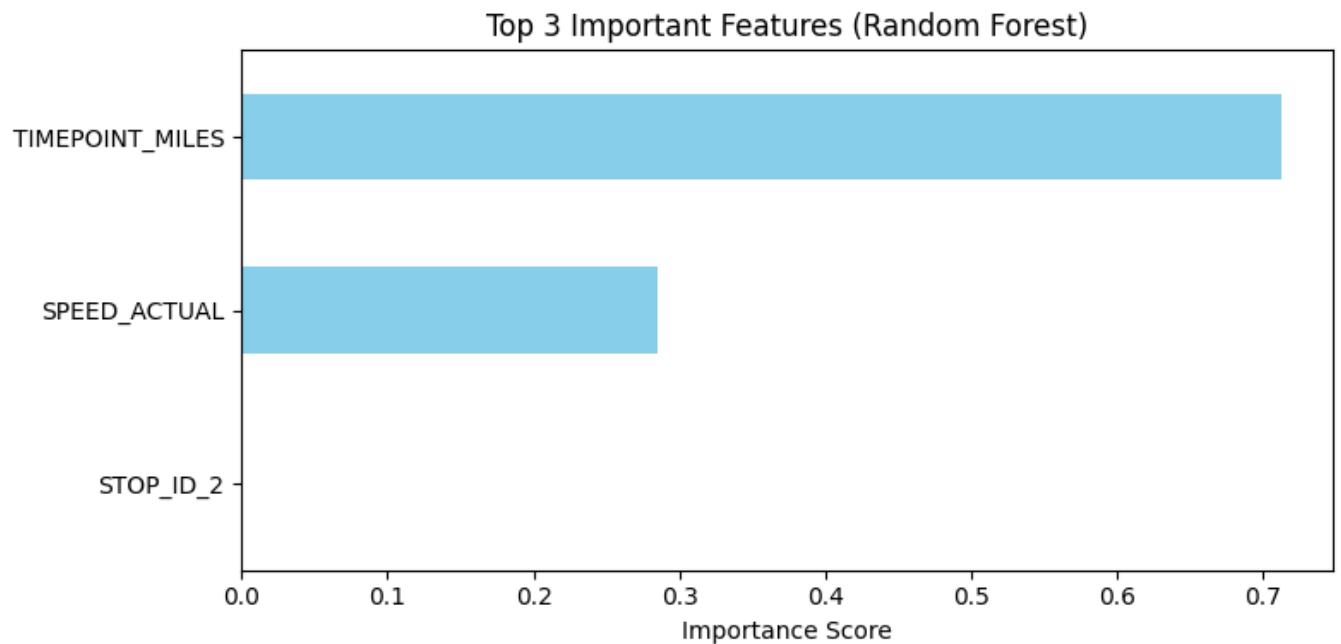
Model	$R^2$	MSE
Linear Regression	0.8928	1.10
Random Forest	0.9995	0.0048
XGBoost	0.9986	0.0137





# Feature Importance

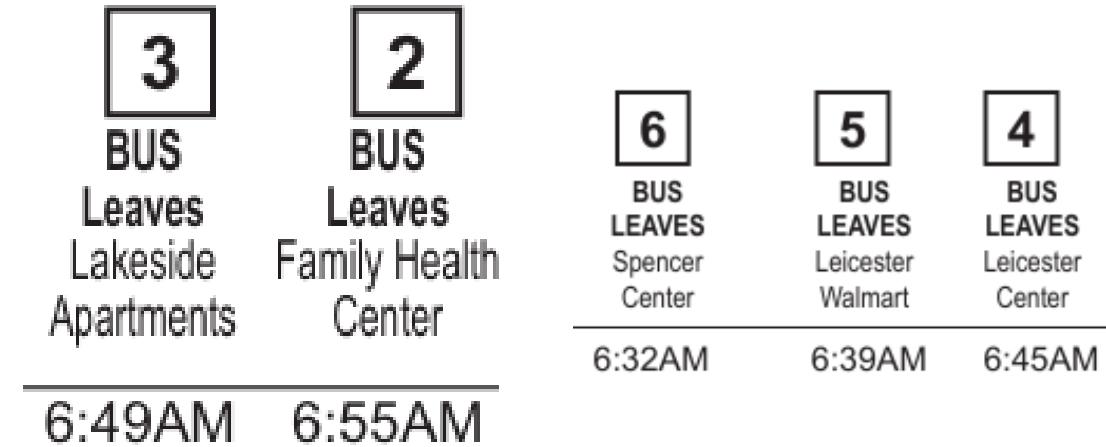
- Most important features:
- TIMEPOINT\_MILES
- SPEED\_ACTUAL
- STOP\_ID\_2



# Prediction Examples

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- Comparison of predicted travel times between key WRTA bus stop pairs using three regression models. Random Forest provides the closest estimates to real-world delays so that is the focus.
- Important observation: the predicted bus time is between two stops is greater than actual scheduled



Stop Pair	Linear Regression	Random Forest	XGBoost
Spencer Center → Leicester Walmart	10.56 min	10.61 min	10.03 min
Leicester Walmart → Leicester Ctr	7.47 min	7.31 min	7.27 min
Lakeside Apartments → Family Health Center	7.58 min	7.40 min	7.41 min

# Conclusions

Buses are consistently late by 1-4 minutes.

Random Forest is the best model.

Scheduled times underestimate actual delays.

Model Could help WRTA improve scheduling



**THANK YOU!**

