# JFK TAXI-OUT Linear Regression

Nov 2019 – Jan 2020

Deirdre Boland 07 Feb 2025

## Objective and dataset

### **Objective**

Create Linear Regression Model of JFK Taxi-Out, using data scraped from an Academic Paper under Review by IEEE transportation covering Nov 2019- Jan 2020 (D Kansal, Kaggle dataset)

#### **Value**

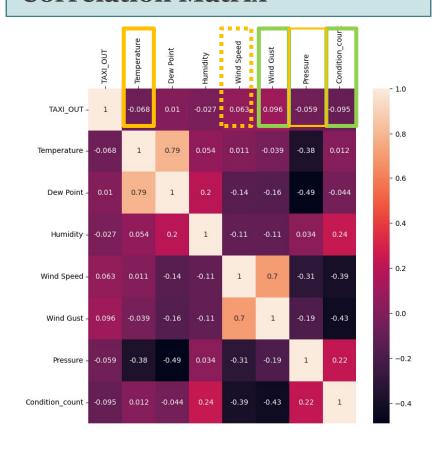
 At JFK airport Taxi-Out prediction is an important concept for calculating runway time and directly impacts the cost of flights.

#### **Dataset**

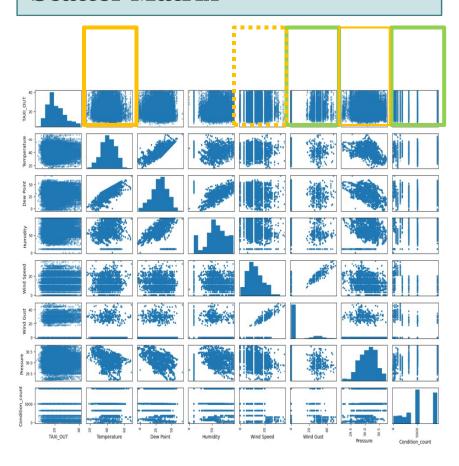
- 5 text based features/variables
  - Airline and flight number indicators (TAIL NUM, OP UNIQUE CARRIER)
  - Destination (DEST) covered by numeric features of distance and scheduled flight time
  - Wind direction (Wind e.g. NW, E etc) have wind speed and gust in numeric
  - Climate/weather (Condition) converted to numeric by using frequency counts (more severe weather less frequent)
- 18 numeric features, including target feature(TAXI\_OUT)
  - Covering weather/conditions at flight time and flight details e.g. Time, duration
  - Scheduled Arrival Time (CRS\_ARR\_M) excluded arrival at another airport, includes time difference offsets. This feature is covered by scheduled duration of flight feature
- Final 18 numeric features (17 original & 1 generated from Condition text feature)
- Reference Kaggle page <u>Flight Take Off Data JFK Airport</u>

## Correlation Matrix - Weather/Climate



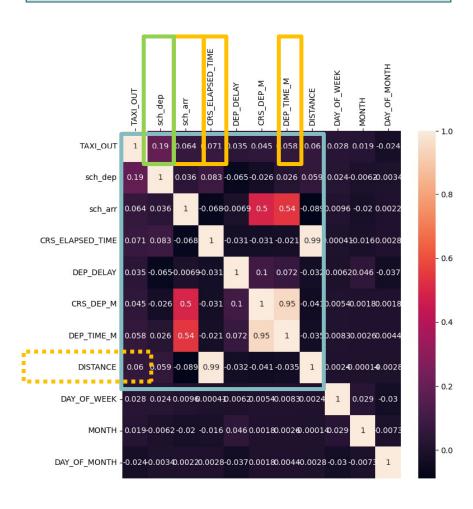


## **Scatter Matrix**



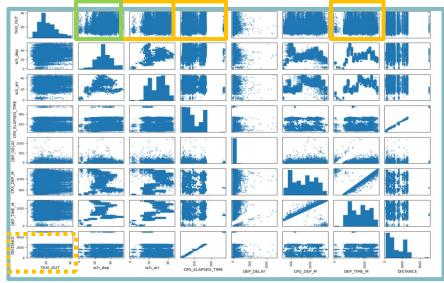
## Correlation Matrix - Flight Details

## **Correlation Matrix**



## **Scatter Matrix**

- Number of flights scheduled for arrival./departure (sch\_arr, sch\_dep)
- Departure delay of the flight (DEP\_DELAY) is calculation of \*Actual Departure Time (DEP\_TIME\_M) subtract Scheduled Departure Time (CRS\_DEP\_M)
  - \* Gate checkout of the flight not the take off time

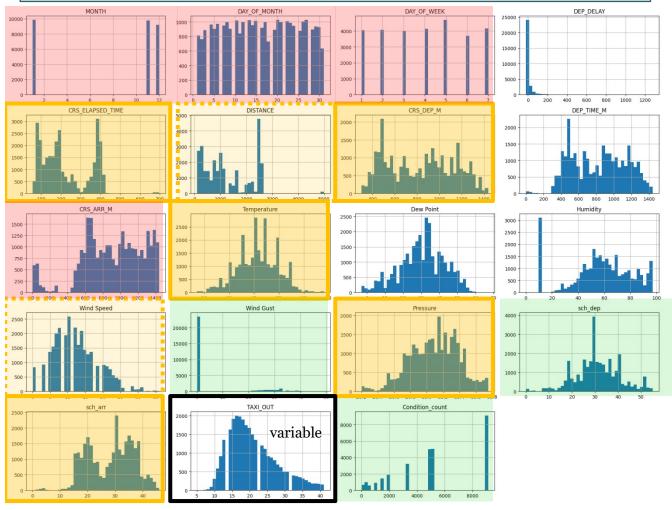


Scheduled journey time of the flight (CRS\_ELAPSED\_TIME)

## Linear regression Model

### **Feature Histograms**

(x- axis, horiz = units bin 40, y- axis, vert = frequency)



- All models scaled using standard scaler. 80% train, 20% test
- R<sup>2</sup> 0-1, higher is better fit

#### 3 features (highest corrl, green)

- Multi Linear Regression (MLR) or Ordinary Least Squares (OLS) – r<sup>2</sup> 0.059
- Lasso r<sup>2</sup> 0.057
- Ridge r<sup>2</sup> 0.059

### 8 features (>0.5 corrl, ambr nondash & green)

- MLR/OLS r<sup>2</sup> 0.077
- Lasso r<sup>2</sup> 0.074
- Ridge r<sup>2</sup> 0.077

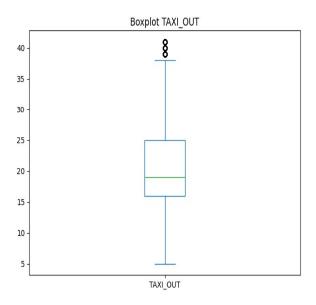
### 14 features (all except red)

- $MLR/OLS r^2 0.083$
- Lasso r<sup>2</sup> 0.080
- Ridge r<sup>2</sup> 0.083

### Alpha tuning (set at 0.1 above)

- Regularisation penalty in ridge and lasso modelling.
- Increasing value on Lasso made model dramatically worse but minimal impact Ridge
- Expected as increasing alpha will get rid of features on Lasso

## Conclusion and next steps



- Linear regression model not viable to model JFK Taxiout with this data set
  - low correlation of features leading to underfitting (erroneous outcomes on new data)
- Limited dataset Nov 2019 Jan 2020, includes holidays and wintry conditions
- Majority Taxi-out time under 25 mins with a min of 5 and max of 40mins
  - Need more information on Taxi-out metric and calculation
    - What is target? Is consistent 5-15 min realistic?
    - What other data not collected, might be impacting? E.g. impact of different terminals (5) and runways (4), staffing levels

