

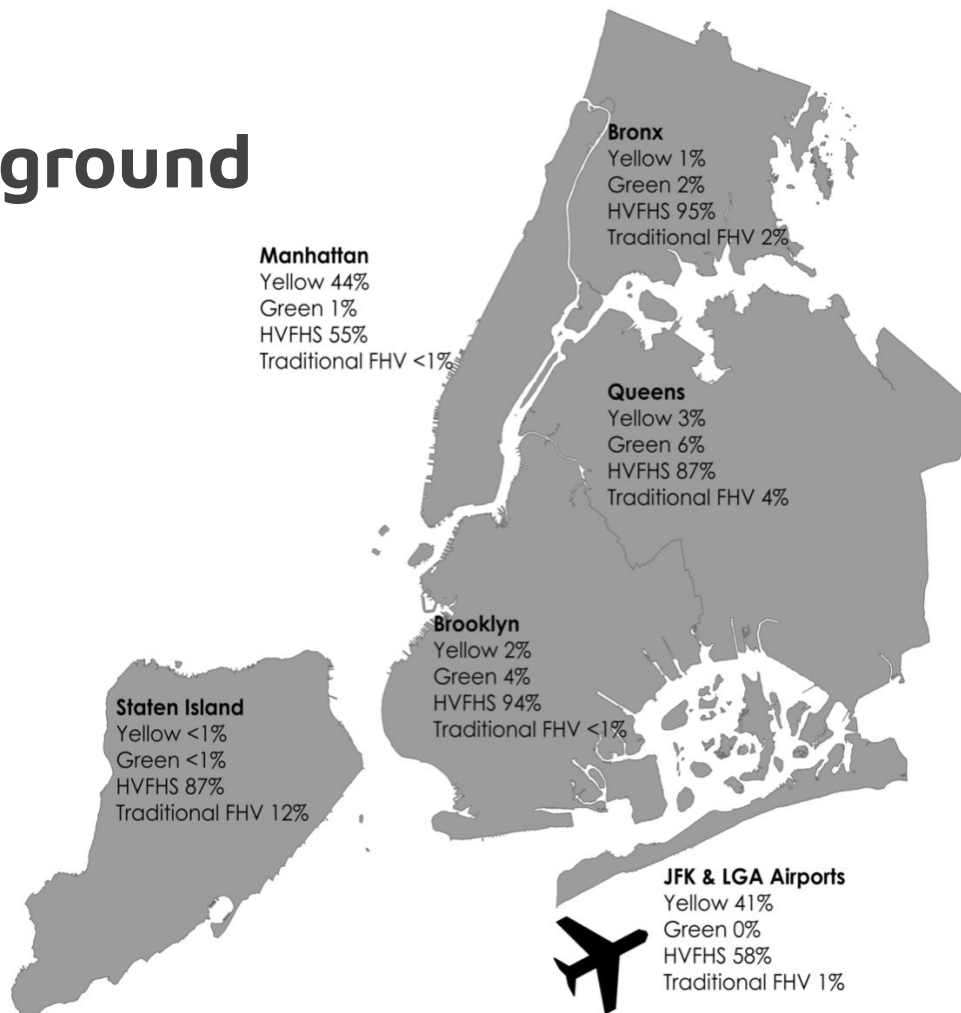
# STAT 453: NYC Taxi Tip Prediction

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# Background



**Can we predict how well  
people tip for yellow taxi  
trips in Manhattan in  
June of 2021?**



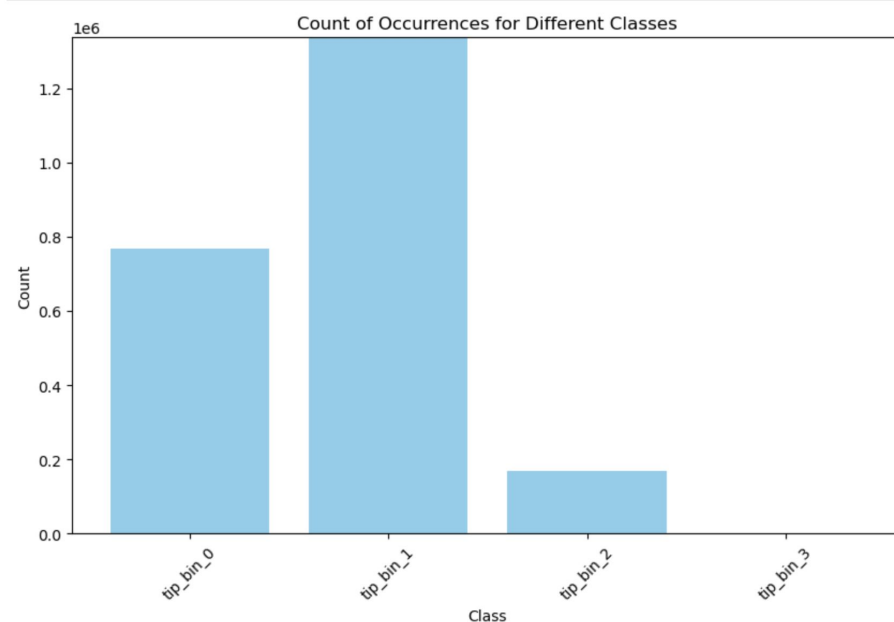
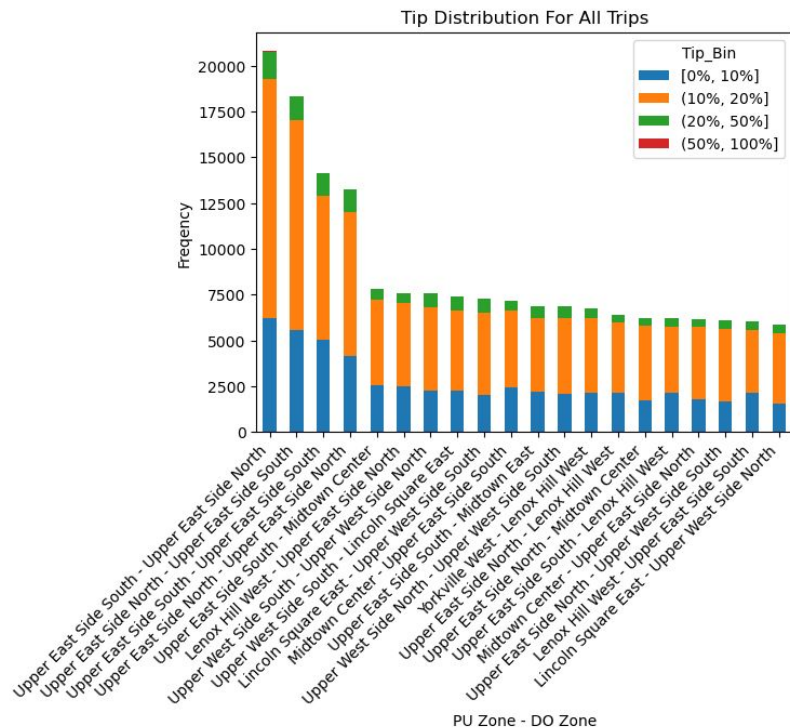


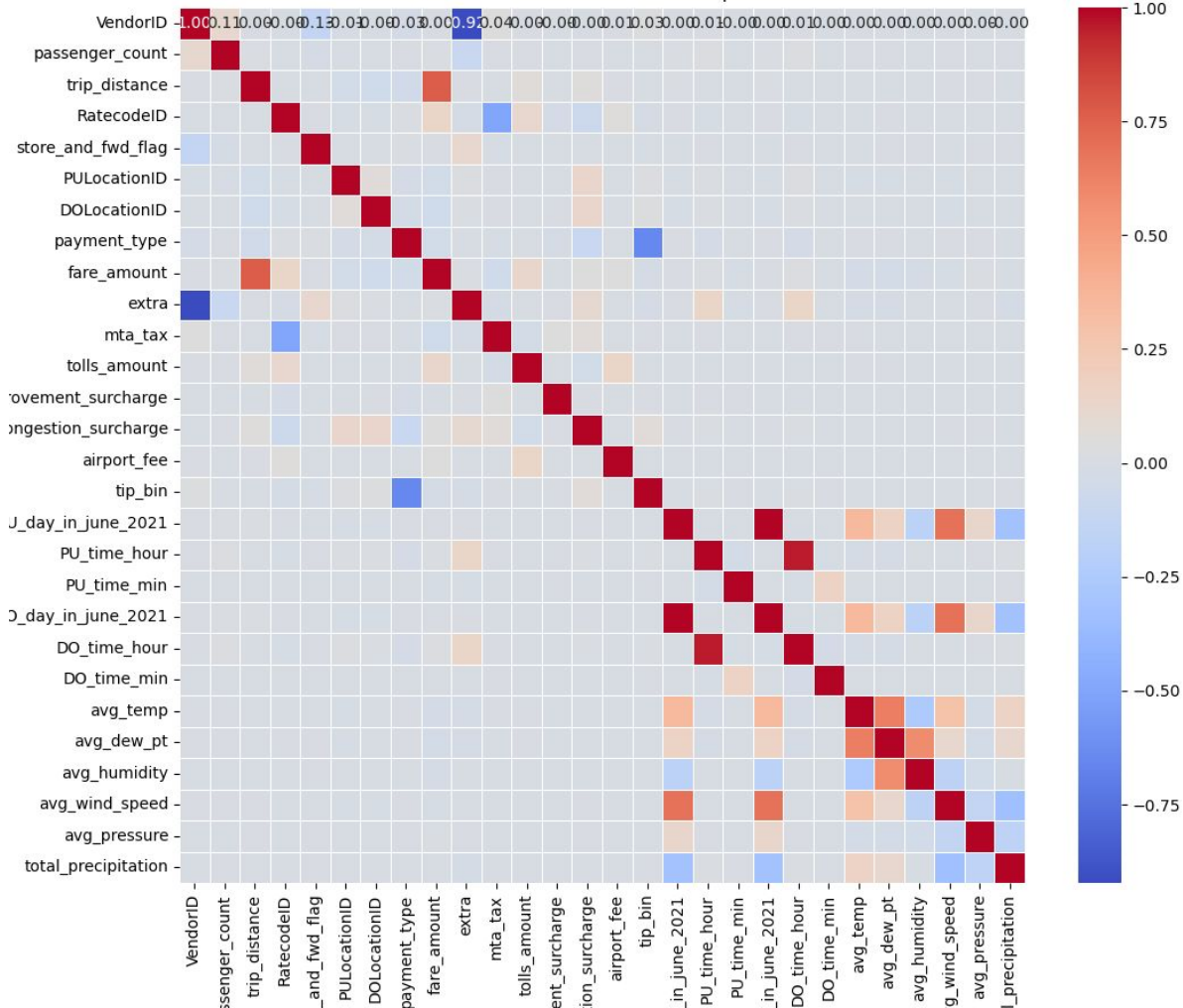
## What does the data look like?

- **Taxi Info:** Vendor, Rate Code ID, Store And Forward Trip
- **Passenger Trip Info:** Pickup Date/Time, Pickup Location, Drop-off Date/Time, Drop-off Location, Trip Distance, Passenger Count
- **Payment Info:** Payment Type, Fare Amount, Extra Cost, MTA Tax, Tolls Amount, Improvement Surcharge, Congestion Surcharge, Airport Fee, Tip Range
- **Daily Average Weather:** Temperature, Dew Point, Humidity, Wind Speed, Pressure, and Total Precipitation



# Exploratory Data Analysis







# List of Models

Multilayer Perceptron Neural Network

Decision Tree Classifier/ Random Forest

Softmax

Multinomial Logistic Regression

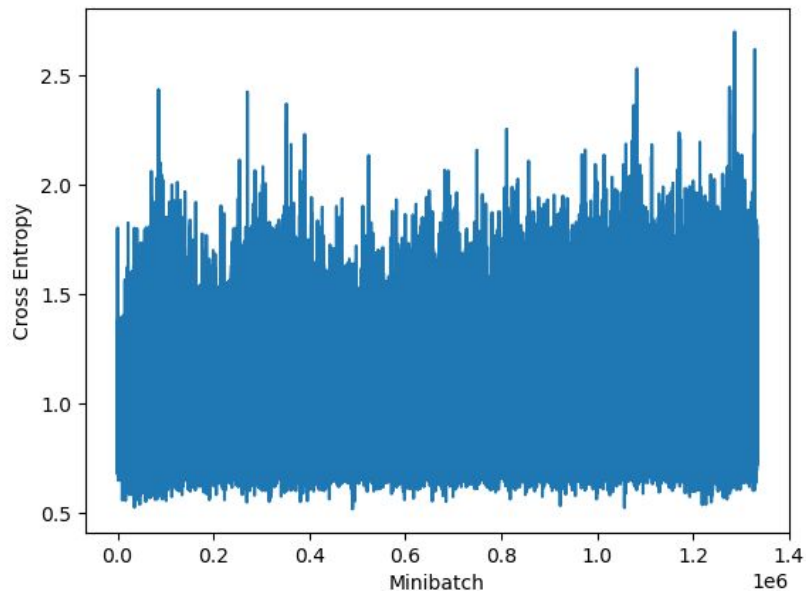


# Neural Network

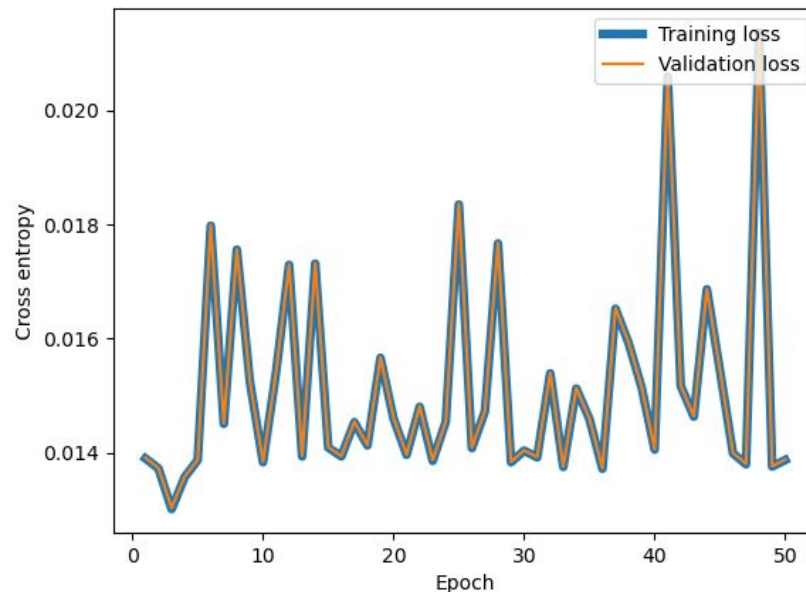
Training Accuracy	58.72
Validation Accuracy	58.69
Testing Accuracy	58.77

Table 1: Optimizer: SGD

Minibatch Cost (SGD optimizer)



Training vs Validation Loss; SGD Optimizer





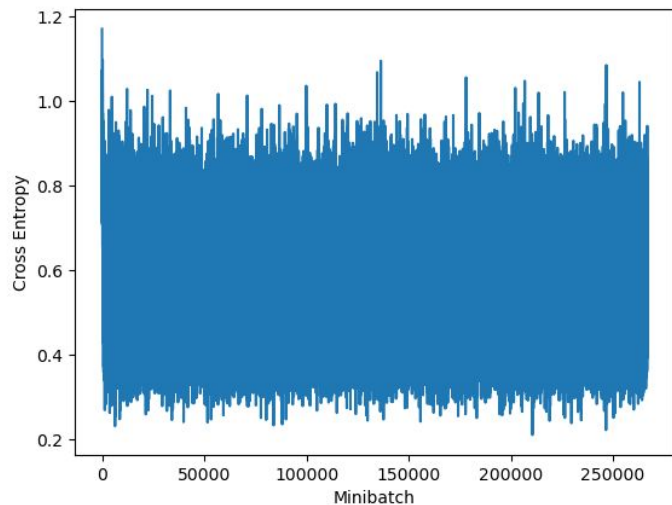


# Neural Network

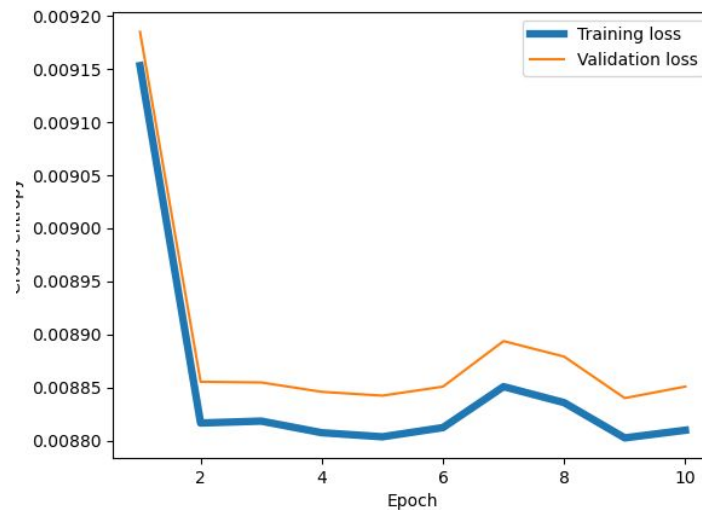
Training Accuracy	80.67
Validation Accuracy	80.53
Testing Accuracy	80.49

Table 1: Optimizer: ADAM

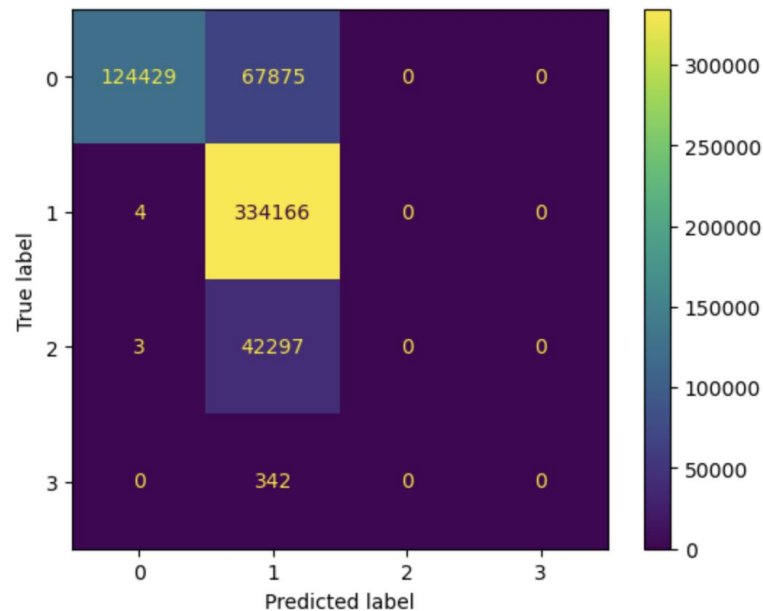
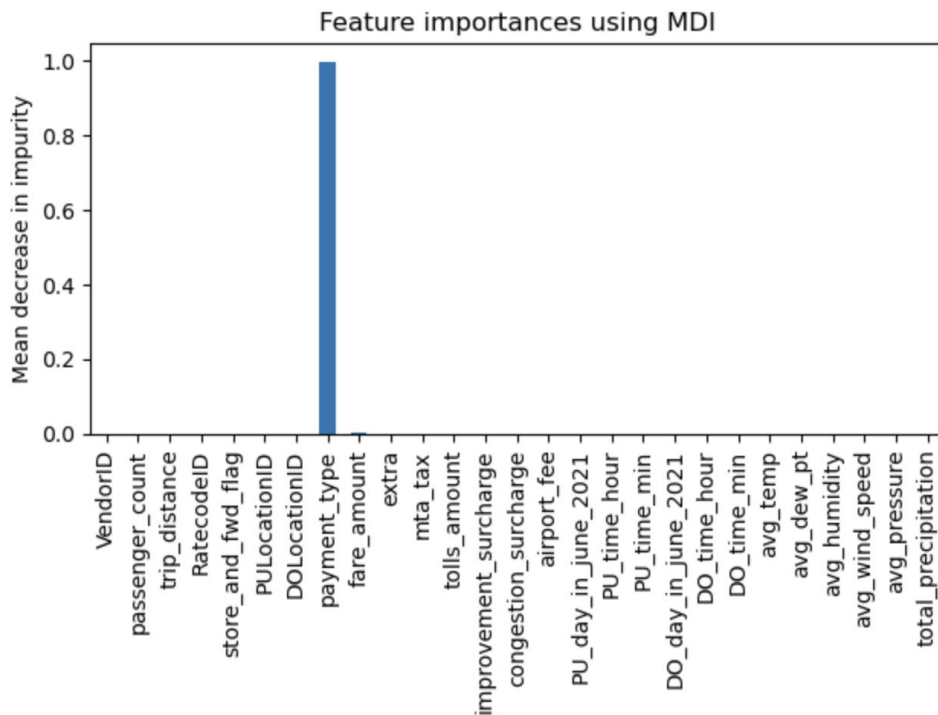
Minibatch Cost (ADAM optimizer)



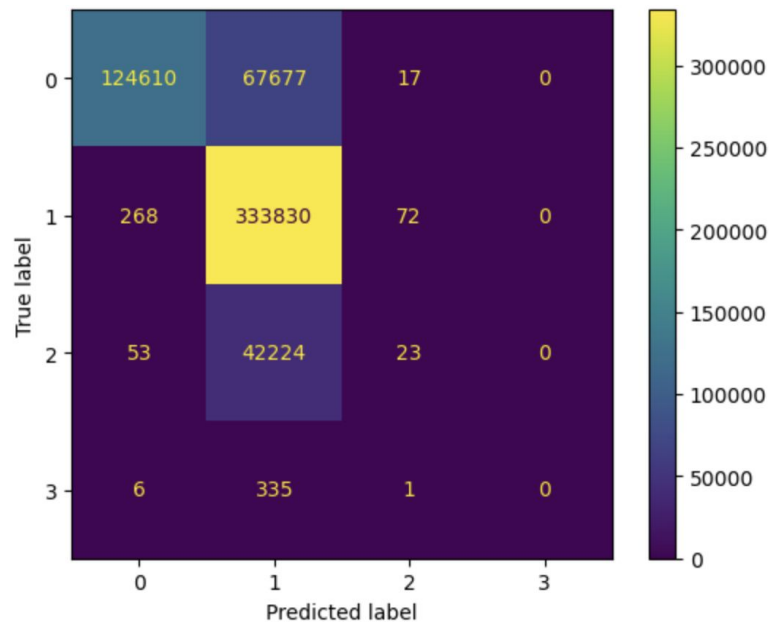
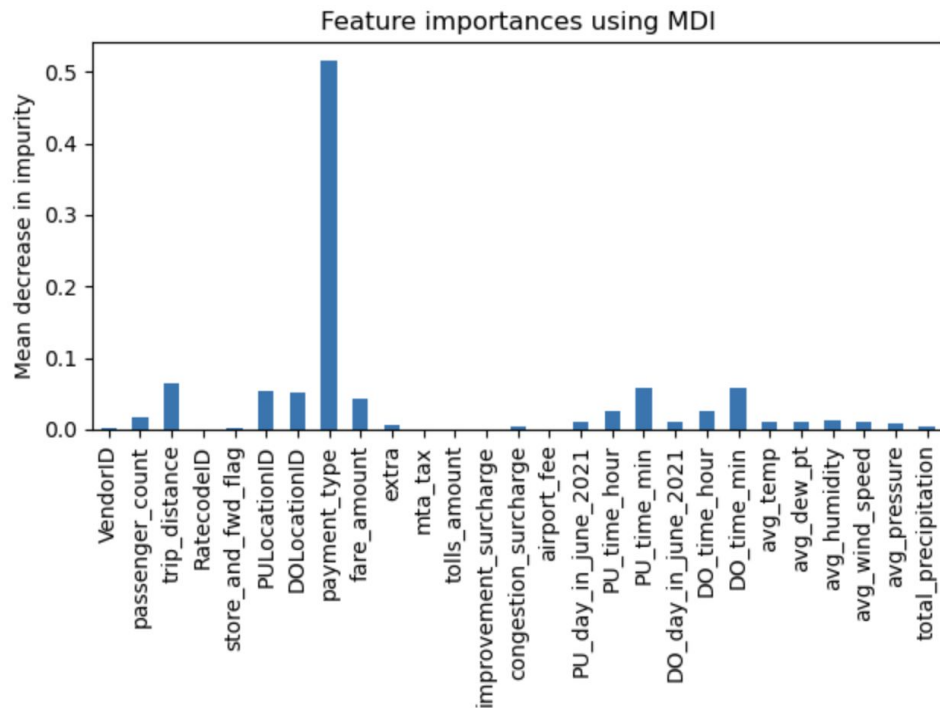
Training vs Validation Loss; ADAM Optimizer



# Decision Tree (Test $R^2 = 0.8058$ )

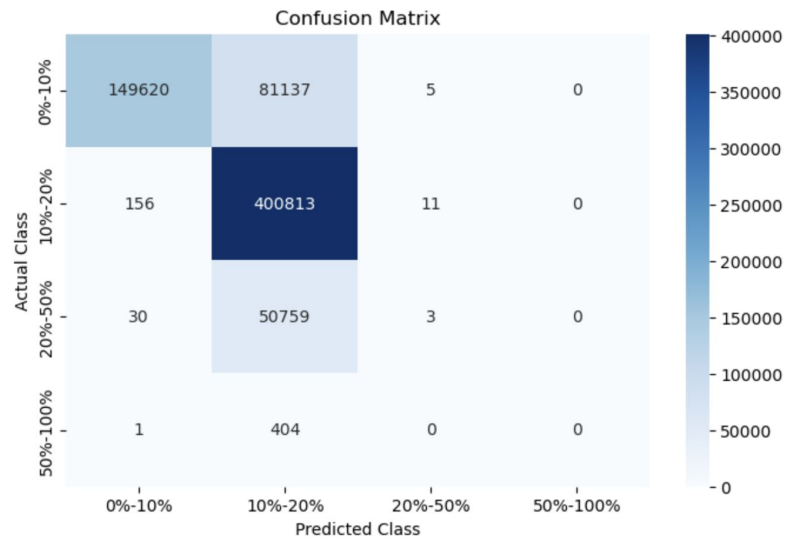
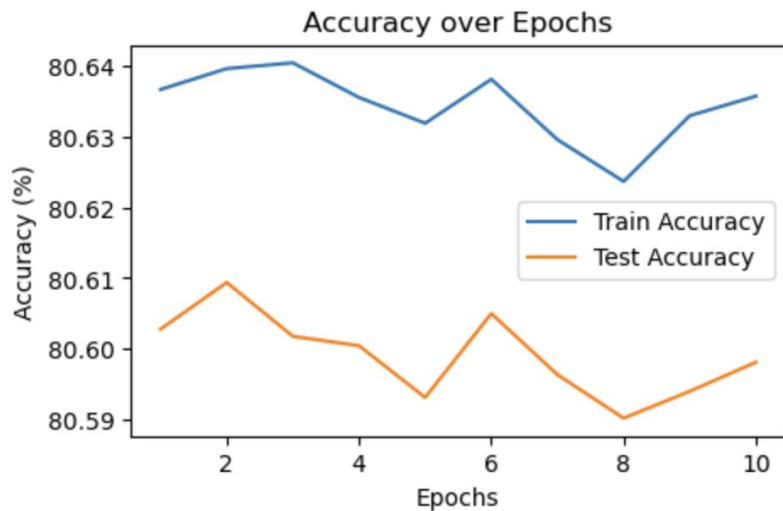


# Random Forest (Test $R^2 = 0.8056$ )





# Softmax Regression

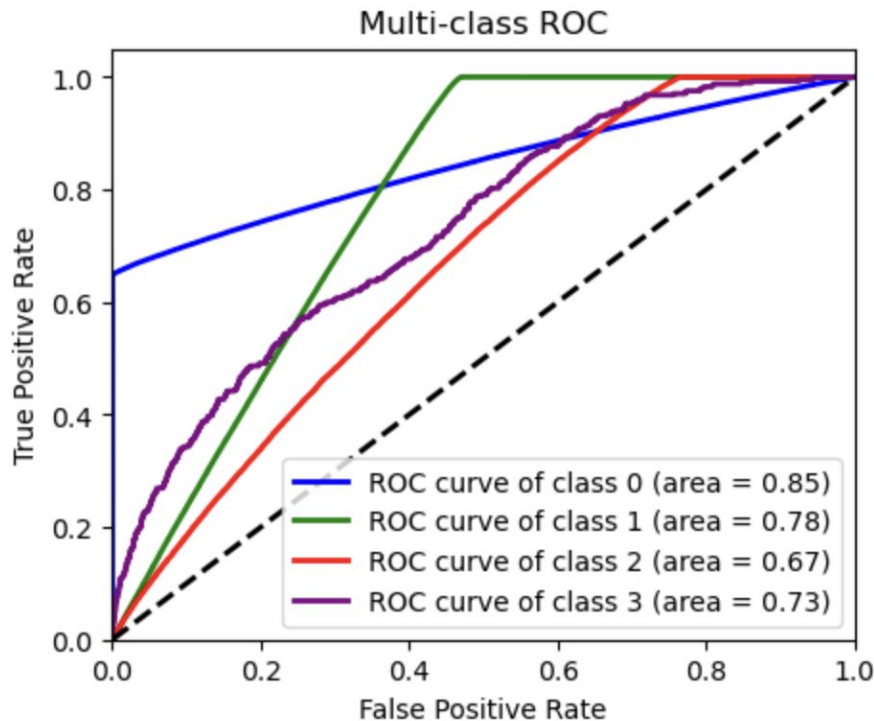




# Softmax Regression

## Advanced Method(Receiver Operating Characteristic, ROC):

**Definition:** The ROC curve is a graphical plot that illustrates the diagnostic ability of a binary classifier system as its discrimination threshold is varied. It is created by plotting the true positive rate (TPR) against the false positive rate (FPR) at various threshold settings.



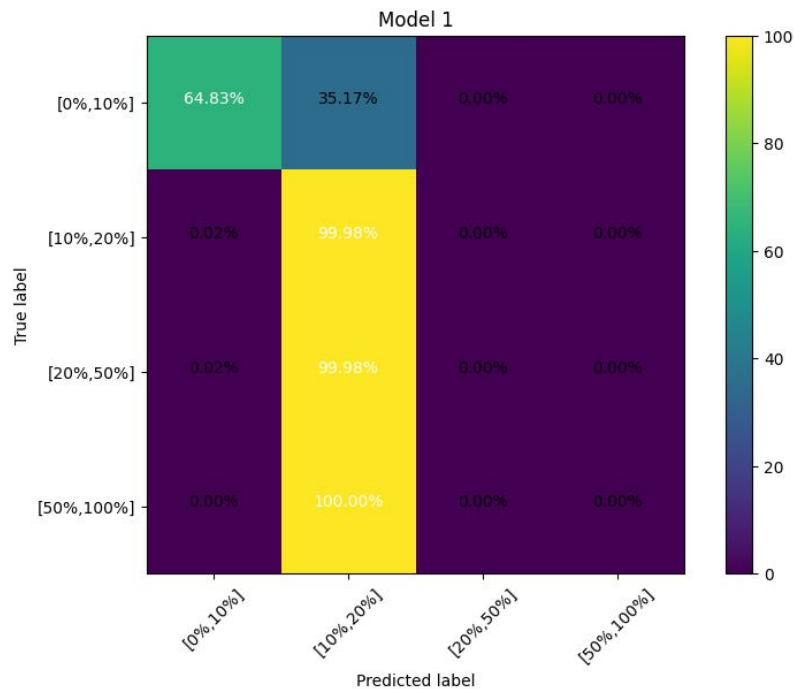


# Multinomial Logistic Regression

Model Identifier	Solver	Penalty	Training Time	Accuracy
Model 1	lbfgs	l2	~6 mins	80.61%
Model 2	lbfgs	None	~9 mins	80.61%
Model 3	newton-cg	None	~18 mins	80.61%
Model 4	newton-cg	l2	~16 mins	80.61%
Model 5	sag	None	~85 mins	80.61%
Model 6	saga	None	~93 mins	80.61%



# Multinomial Logistic Regression Confusion Matrix for Model 1





## Conclusion/ Recommendations

Traditional ML works well for tabular data (NN not as much)

Apply the models to more data (i.e. for hire trip reviews, driver ratings, etc.)

Do more models (TabPFN, TP-BERTa, etc.)

Group tips into more buckets





**Thank you!**

