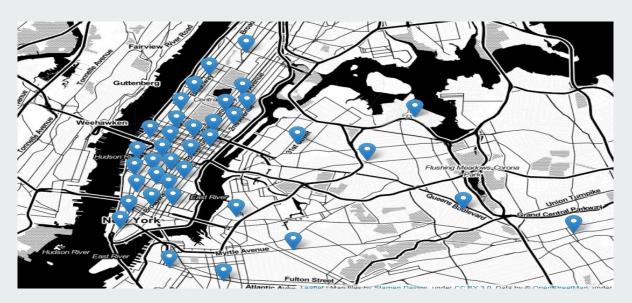
# **Taxi Demand Prediction**



**By Eddie Amaitum** 

### **Demand in taxi/car share industry**



**Drivers (Supply) Vs Customers (Demand)** 

Companies need to forecast demand



### Solution: Predict taxi demand per hour by location

Operations team can adjust the distribution of drivers



+ \$3 for pick-ups around the arena



**Operations team** 

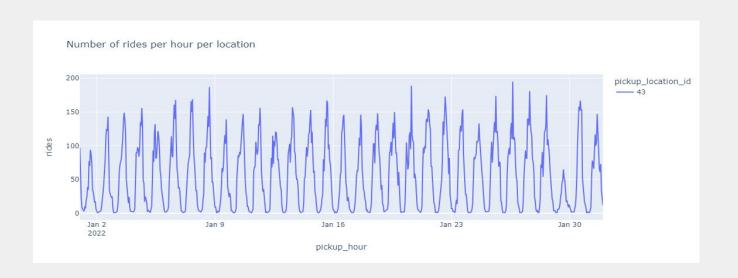
### **Impact**

- Predictive models shown to cut wait times by up to 20%
- Improved efficiency in driver deployment hence companies generate more revenue
- Increased customer satisfaction



- NYC Taxi & Limousine Commission (TLC) Trip Records
- The data is relatively clean
- Feature engineering needed

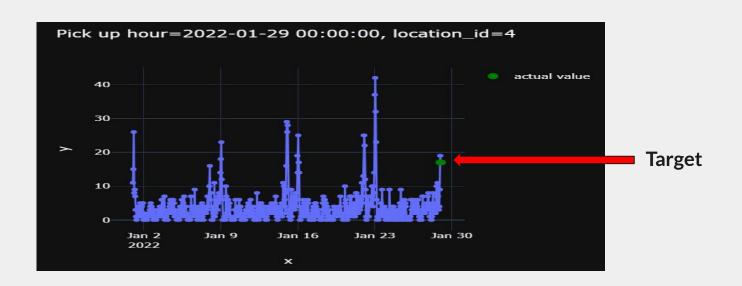
## **Sample Time Series Data**



### **Data transformation into (features, targets)**

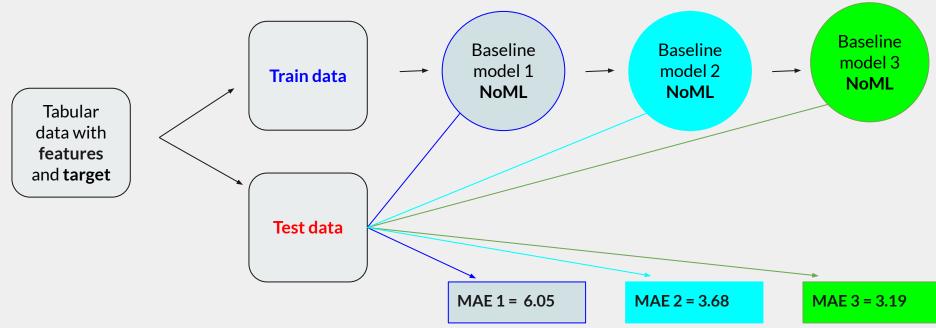


### Sample Features & Target Variables



# Split the data **Train data** Tabular data with Split by date e.g Aug Data preparation features and **target Test data**

### **Baseline models**



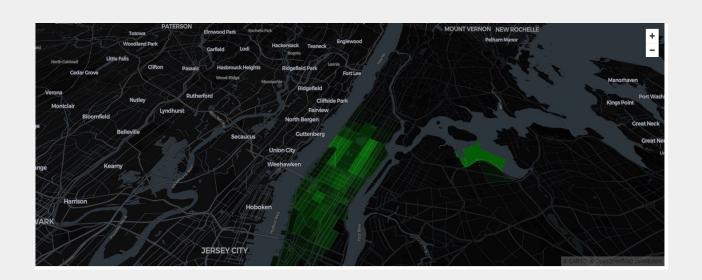
# Model comparison

Model	Mean Absolute Error (MAE)	Notes
Ad Hoc model 1	6.05	Baseline model
Ad Hoc model 2	3.68	Baseline model
Ad Hoc model 3	3.19	Baseline model
XGBoost	2.70	Models improved
Lightgbm	2.57	Models improved
Lightgbm + feature engineering	2.59	Added average rides per month
Lightgbm + hyperparameter tuning	2.54(num_leaves,min_child_samples,etc)	Best model for production



- → Further improve model performance by adding more features
- → Build pipelines to automate processes
- → Complete model operationalization

# **Application dashboard**





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# THANK YOU

