

Building Serverless ML Platform

Data Analysis And Visualization Of Green Taxi Records

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Building a serverless Machine Learning platform that analyses, visualizes Green-taxi user data and predicts future customer base.





SERVERLESS ARCHITECTURE

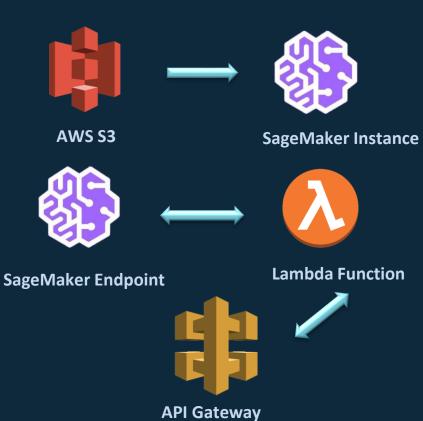
- Serverless is the native architecture of the cloud that enables you to shift more of your operational responsibilities to AWS.
- Serverless enables us to build modern applications with increased agility and lower total cost of ownership.



METHODOLOGY



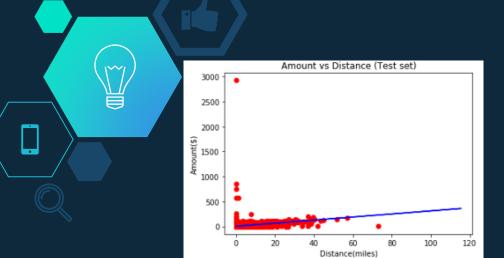
- Amazon S3
- Lambda function
- Sagemaker Instance
- Sagemaker Endpoint
- AWS API

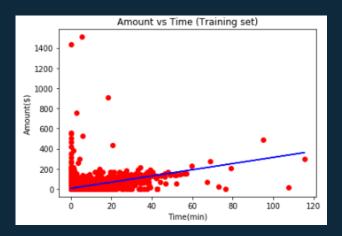


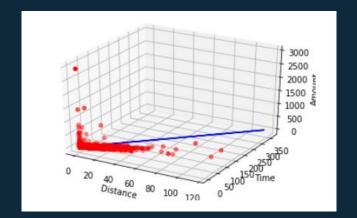


- ♦ 6 years Green taxi Trip records for the state of New York have been collected.
- Monthly Raw data files are stored in S3.
- We have implemented Python to organize and analyze the Raw data.
- ML Algorithms Multiple Linear Regression.

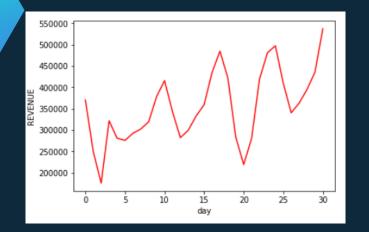
```
In [158]: data = pd.read csv('https://fianlprojectcc.s3-us-west-1.amazonaws.com/2014-01.csv')
          data.columns = ['lpep pickup datetime', 'Lpep dropoff datetime', 'Store and fwd flag', 'RateCodeID', 'Pickup longitude', 'Pickup
          data.to csv("data.csv", sep=',', index=False)
          columns = ['lpep_pickup_datetime', 'Lpep_dropoff_datetime', 'Total_amount', 'Trip_distance']
          df = pd.read csv("data.csv", usecols = columns)
          df=df.dropna()
          #print(df)
          df['lpep_pickup_datetime'] = pd.to_datetime(df['lpep_pickup_datetime'],format='%m/%d/%Y %H:%M')
          df['Lpep_dropoff_datetime'] = pd.to_datetime(df['Lpep_dropoff_datetime'],format='%m/%d/%Y %H:%M')
          df['trip duration']= df['Lpep dropoff datetime'] - df['lpep pickup datetime']
          data = df[~(df['lpep pickup datetime'] < '2014-01-01')]</pre>
          data = data[~(df['lpep pickup datetime'] > '2014-02-01')]
          data['count']=1
          #print(data)
          #print(data)
          regcolumns = ['Trip distance', 'Total amount']
          regcolumns2 = ['trip duration', 'Total amount']
          data distance=data[regcolumns]
          data time = data[regcolumns2]
          #print(data time)
          by hour=data.groupby(Grouper(key='lpep pickup datetime', freq='H')).sum()
          by days=data.groupby(Grouper(key='lpep pickup datetime', freq='d')).sum()
          print(by days)
          # data atribute has non group data
```

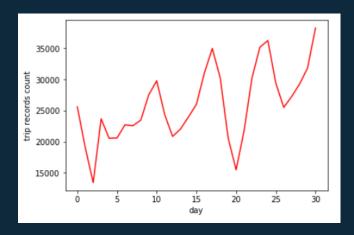


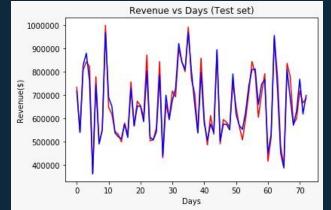












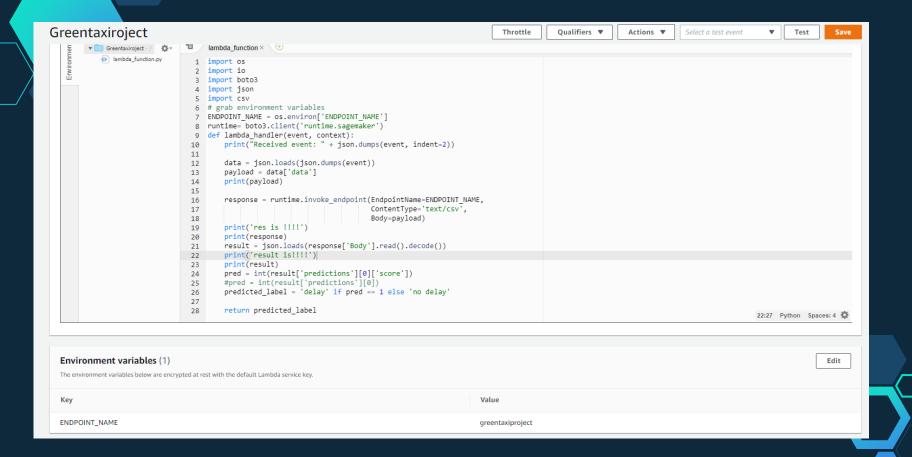




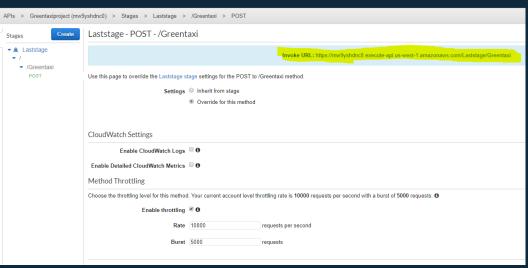
Prediction Accuracy for trip pricing – 92.4%

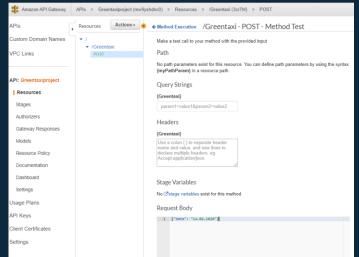
Prediction Accuracy for revenue and customer base – 86.8%













CHALLENGES

- Column names and data types were different for different monthly reports so preprocessing the data for analysis was a challenging job
- We were unable to integrate Location specific data in our predictions.



FUTURE WORK

- We would like to use API collaboration platforms like Postman and provide two user interfaces. We would also integrate location data into our analysis.
- User-side Application.
- Company Interface.



Thank You!

Any questions?

