

# MSBX5420 Team Project - Design Document

Team : Team Blanca Peak Date: 04/25/2020

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## **Document History**

| Version | Date      | Author           | Comments                |
|---------|-----------|------------------|-------------------------|
| 1.0     | 4/19/2020 | Team Blanca Peak | Created initial version |

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#### 1. Introduction

This document provides a comprehensive architectural overview of MSBX-5420 project of team blanca peak. This design document is prepared as per the requirement specification defined for our project.

### 2. Application / System Design Specification

#### **Ingest Dataset in S3:**

Download csv files from https://www1.nyc.gov/site/tlc/about/tlc-trip-record-data.page and ingest in Leeds S3 bucket using the following commands:

ssh to the EMR cluster and copy csv files.

sudo ssh -i ./Leed\_HadoopKeypair.pem hadoop@ec2-52-13-183-139.us-west-2.compute.amazonaws.com

Prerequisite: Download the .pem and .ppk files for Leeds AWS

#### Read CSV and Save Dataset as Parquet:

Create a group s3 bucket and copy the 2019 yellow cab files from amazon's open data repository

```
aws s3 mb team-blanca-peak
aws s3 cp s3://nyc-tlc s3://team-blanca-peak --recursive --exclude "*" --include
"yellow_tripdata_2019*"
```

Used a python script and pyarrow to read the csv and transform into parquet

```
import pandas as pd
import pyarrow
for i in range(1,13):
    if i<10:
        tmp_df=pd.read_csv(f'yellow_tripdata_2019-0{i}.csv')
        tmp_df.to_parquet(f'yellow_tripdata_2019-0{i}.parquet')
    else:
        tmp_df=pd.read_csv(f'yellow_tripdata_2019-{i}.csv')
        tmp_df.to_parquet(f'yellow_tripdata_2019-{i}.parquet')</pre>
```

#### Pseudo Code:

- import SparkSession and create SparkContext
- SparkContext.textFile("csv path")
- 3. data frame.write.parquet("s3:team-blanca-peak")



#### Read parquet file and display records:

#### Pseudo Code:

data frame = sqlContext.read.parquet("s3:team-blanca-peak")
data frame.show(5)

#### **Data Analysis:**

Top pickup/dropoff locations
Average trip distance (total and by hour)
Converted dates into datetime
Average trip time by minutes
Most popular months for pickups
Most popular hours for pickups

#### Example:

```
[9] print('The average trip time is ' + str(Q1Q2yellowcab['triptime(min)'].mean()) + ' minutes')

The average trip time is 17.71576354332651 minutes

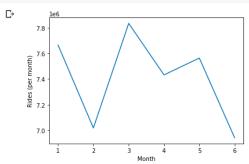
tripagg = Q1Q2yellowcab.groupby(["PULocationID"])["tpep_pickup_datetime"].count()
tripagg = tripagg.sort_values(axis=0, ascending = False)
print('The three pickup locations with the highest total trips are ' + str(totalagg.index[1]) + ' ' +
str(totalagg.index[2]) + " " + str(totalagg.index[3]))
The three pickup locations with the highest total trips are 138 161 230
```

#### **Data Visualization:**

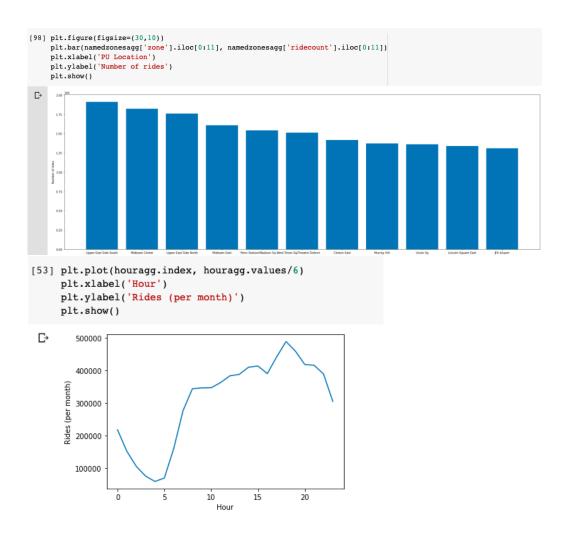
Rides per month
Rides per hour
Average Trip distance per month
Average Trip distance per hour
Top 10 Pickup locations

#### Examples:

```
[52] plt.plot(monthlyagg.iloc[0:6].index, monthlyagg.iloc[0:6].values)
   plt.xlabel('Month')
   plt.ylabel('Rides (per month)')
   plt.show()
```







#### **Machine Learning Model:**

1. Classification model using pyspark.ml.classification RandomForest determine the tip amount based on various factors.

#### Steps:

- 1. Read data from parquet and load in the pyspark dataframe.
- 2. Data cleansing by filling the null values with 0 for float and Unknown for the string fields and drop unnecessary columns.
- 3. Create the boolean Y column based on the tip amount > 0 then 1 else 0
- 4. Converting the categorical columns to vectors using onehotencoder
- 5. Set the Y and X features and labels
- 6. Divide the dataset to train, test and validation based on 80:10:10
- 7. Fit the model to the train dataset
- 8. Test the model using test dataset
- 9. Tune the parameters and re-test
- 10. Fit the cross validation model to the validation dataset
- 11. Draw confusion matrix and evaluate the matrix



#### Pseudo Code:

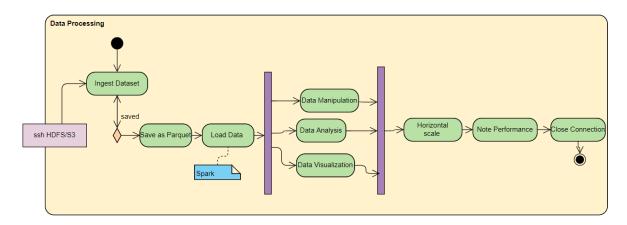
data frame = sqlContext.read.parquet("s3:team-blanca-peak")
data frame = df.na.fill(0)
data frame.withColumn 'tip' when 'tip\_amount' > 0 then 1 otherwise 0
OneHotEncode[input columns, output columns]
pipeline.fit(model\_dataset)
crossvalidation.fit(data set)

2. Any Other ML model to fill here:

## 2.1 Application Components

Need to include Class/method, third party pip install etc here:

#### 2.2 Activity



### 2.3 Schema Definition



| [VendorID - float]                 |  |
|------------------------------------|--|
| [tpep_pickup_datetime - datetime]  |  |
| [tpep_dropoff_datetime - datetime] |  |
| [passenger_count - int]            |  |
| [trip_distance - float]            |  |
| [RatecodeID - int]                 |  |
| [store_and_fwd_flag - bool]        |  |
| [PULocationID - string]            |  |
| [DOLocationID - string]            |  |
| [payment_type - string]            |  |
| [fare_amount - float]              |  |
| [extra - float]                    |  |
| [mta_tax - float]                  |  |
| [tip_amount - float]               |  |
| [tolls_amount - float]             |  |
| [improvement_surcharge - float]    |  |
| [total_amount - float]             |  |
| [congestion_surcharge - float]     |  |

# 3. References, Abbreviations/Acronyms

https://www1.nyc.gov/site/tlc/about/tlc-trip-record-data.page https://github.com/MingChen0919/learning-apache-spark/blob/master/notebooks/06-machine-learning/classification/random-forest-classification.ipynb