CSE519_HW2

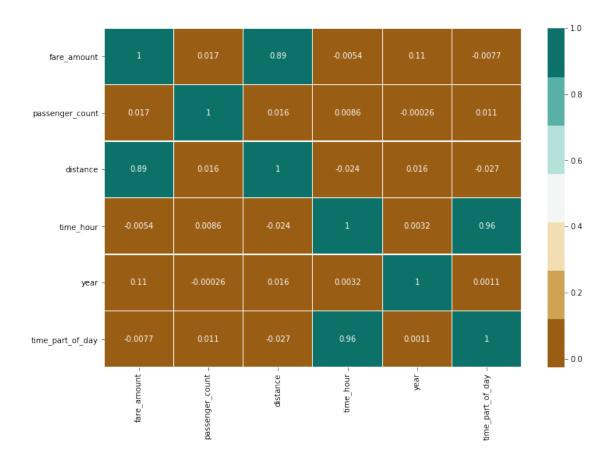
September 25, 2018

```
In [243]: import pandas
          import datetime
          import numpy as np
          import geopy.distance
          import os
          from IPython.display import display, HTML
          import random
          import seaborn
          from matplotlib import pyplot as plt
          from matplotlib import cm
          from sklearn import datasets, linear_model
          from sklearn.model_selection import train_test_split
          from sklearn.metrics import mean_squared_error
          from sklearn import linear_model
          from sklearn import svm
          from sklearn.ensemble import RandomForestRegressor
          from sklearn.datasets import make_regression
          from sklearn.ensemble import GradientBoostingRegressor
In [9]: def get_distance(a, b):
            return geopy.distance.distance(a, b).miles
        #https://stackoverflow.com/questions/40807225/how-to-call-data-from-a-dataframe-into-h
        def haversine(lon1, lat1, lon2, lat2):
            Calculate the great circle distance between two points
            on the earth (specified in decimal degrees)
            All args must be of equal length.
            11 11 11
            lon1, lat1, lon2, lat2 = map(np.radians, [lon1, lat1, lon2, lat2])
            dlon = lon2 - lon1
            dlat = lat2 - lat1
            a = np.sin(dlat/2.0)**2 + np.cos(lat1) * np.cos(lat2) * np.sin(dlon/2.0)**2
```

```
c = 2 * np.arcsin(np.sqrt(a))
            km = 6367 * c
            return km
In [10]: def pre_process(chunk):
             chunk['distance'] = chunk[['pickup_latitude',
                                          'pickup_longitude',
                                          'dropoff_latitude',
                                          'dropoff_longitude']
                                        ].apply(lambda x: get_distance(x[0], x[1], x[2], x[3]),
             chunk.drop(chunk[ chunk['distance'] == 0 ].index, inplace = True)
             write_chunk(chunk, "cleaned_data.csv")
In [11]: def clean(chunk):
             newyork_lat = 40.730610
             newyork_long = -73.935242
             range_var = 0.5
             chunk.dropna(inplace=True)
             chunk.drop(chunk[chunk['fare_amount'] <= 0.0].index, inplace = True)</pre>
              chunk.drop(chunk[chunk['passenger_count'] == 0].index, inplace = True)
              chunk.drop(chunk[(chunk['pickup_latitude'] < newyork_lat-range_var) | (chunk['pic
              chunk.drop(chunk[(chunk['dropoff_latitude'] < newyork_lat-range_var) | (chunk['dropoff_latitude'] |
              chunk.drop(chunk[(chunk['pickup_longitude'] < newyork_long-range_var) | (chunk['pickup_longitude']</pre>
             chunk.drop(chunk[(chunk['dropoff_longitude'] < newyork_long-range_var) | (chunk['dropoff_longitude']</pre>
In [12]: def write_chunk(chunk, filename):
             chunk.to_csv(filename, columns = ['fare_amount', 'passenger_count', 'distance', 't
In [13]: chunk_size = 10**7
         count = 0
         if os.path.exists('cleaned_data.csv'):
             os.remove('cleaned_data.csv')
         chunk = pandas.read_csv('all/train.csv', nrows = chunk_size)
         count += 1
         #display(chunk.describe())
         print("\rProcessing count: {}".format(count * chunk_size), end="")
Processing count: 10000000
In [14]: clean(chunk)
In [15]: pre_process(chunk)
In [18]: chunk['time_hour'] = pandas.to_datetime(
                  chunk['pickup_datetime'].apply(lambda x: x.split()[1]),
                  format='%H:%M:%S').dt.hour
```

```
In [19]: chunk['year'] = pandas.to_datetime(
                 chunk['pickup_datetime'].apply(lambda x: x.split()[0]),
                 format='%Y-%m-%d'
             ).dt.year.astype(int)
In [20]: chunk['month'] = pandas.to_datetime(
                 chunk['pickup_datetime'].apply(lambda x: x.split()[0]),
                 format='%Y-%m-%d'
             ).dt.month.astype(int)
  Pearsons Coefficient
In [119]: percentage = 0.001
          random_rows = pandas.read_csv('cleaned_data.csv',
                                       dtype={
                                            'fare_amount': np.float32,
                                            'distance': np.float32,
                                            'passenger_count': int,
                                            'time_hour': int,
                                           'year' : int
                                       },
                                         skiprows=lambda i: i>0 and random.random() > percentage
          random_rows['time_part_of_day'] = pandas.cut(random_rows['time_hour'],
                                                        [-1, 6, 12, 16, 24],
                                                        labels = [1, 2, 3,4]).astype(int)
          random_rows = random_rows.filter(['fare_amount', 'passenger_count', 'distance', 'time
          display(random_rows.head())
          random_rows.describe()
  fare_amount passenger_count
                                 distance time_hour year time_part_of_day
0
           7.7
                                                    4 2012
                              1
                                 1.737223
                                                                             1
           6.5
                              1 1.109777
                                                   11 2013
                                                                             2
1
2
           4.5
                              1 0.811588
                                                   17 2012
                                                                             4
3
                                                                             4
          21.5
                              1 4.734918
                                                   23 2013
4
          16.9
                              2 3.950192
                                                    3 2012
                                                                             1
Out[119]:
                 fare_amount
                              passenger_count
                                                   distance
                                                               time_hour
                                                                                  year
                 9646.000000
                                  9646.000000
                                                9646.000000
                                                             9646.000000
                                                                          9646.000000
          count
                                                   2.090501
                                                               13.427016
          mean
                   11.321514
                                     1.699150
                                                                          2011.724238
          std
                    9.420152
                                     1.312497
                                                   2.215263
                                                                6.494773
                                                                              1.869692
          min
                    2.500000
                                      1.000000
                                                   0.000148
                                                                0.000000
                                                                          2009.000000
          25%
                                                                9.000000
                                                                          2010.000000
                    6.100000
                                      1.000000
                                                   0.808063
          50%
                    8.500000
                                      1.000000
                                                   1.356755
                                                               14.000000
                                                                          2012.000000
          75%
                   12.500000
                                      2.000000
                                                   2.484034
                                                               19.000000
                                                                          2013.000000
          max
                  124.500000
                                     6.000000
                                                  17.819082
                                                               23.000000 2015.000000
                 time_part_of_day
```

```
9646.000000
          count
                         2.814120
          mean
          std
                         1.102884
          min
                         1.000000
          25%
                         2.000000
          50%
                         3.000000
          75%
                         4.000000
          max
                         4.000000
In [120]: # Reference: https://stackoverflow.com/questions/22258491/read-a-small-random-sample
In [121]: pearson_corr = random_rows[['fare_amount', 'passenger_count', 'distance', 'time_hour
          display(pearson_corr)
          labels = ['fare_amount', 'passenger_count', 'distance', 'time_hour',
                          'year', 'time_part_of_day']
          seaborn.heatmap(pearson_corr,
                          cmap=seaborn.color_palette("BrBG", 7),
                      linewidths=.1,
                      annot=True,
                      xticklabels=labels,
                      yticklabels=labels)
          plt.gcf().set_size_inches(12,8)
                  fare_amount passenger_count distance time_hour
                                                                          year
                                      0.017381 0.890199 -0.005405 0.113453
fare_amount
                     1.000000
                                                            0.008638 -0.000265
                                      1.000000 0.016091
passenger_count
                     0.017381
distance
                     0.890199
                                      0.016091 1.000000 -0.023727 0.016486
time_hour
                                      0.008638 -0.023727
                    -0.005405
                                                            1.000000 0.003175
year
                     0.113453
                                     -0.000265 0.016486
                                                            0.003175 1.000000
time_part_of_day
                    -0.007684
                                      0.010713 -0.026650
                                                            0.958932 0.001134
                  time_part_of_day
                         -0.007684
fare_amount
passenger_count
                          0.010713
distance
                         -0.026650
time_hour
                          0.958932
                          0.001134
year
time_part_of_day
                          1.000000
```

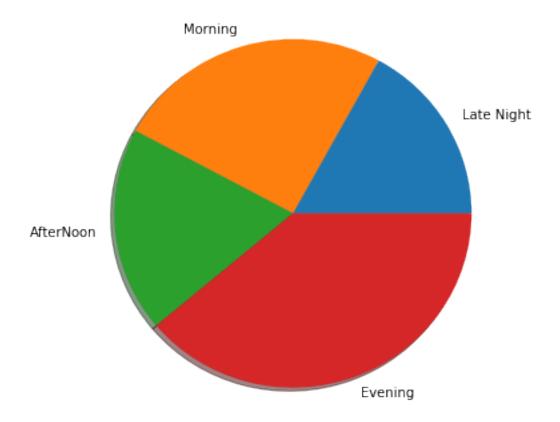


Visualization



Reference: https://medium.com/@kvnamipara/a-better-visualisation-of-pie-charts-by-matplotlib-935b7667d77f

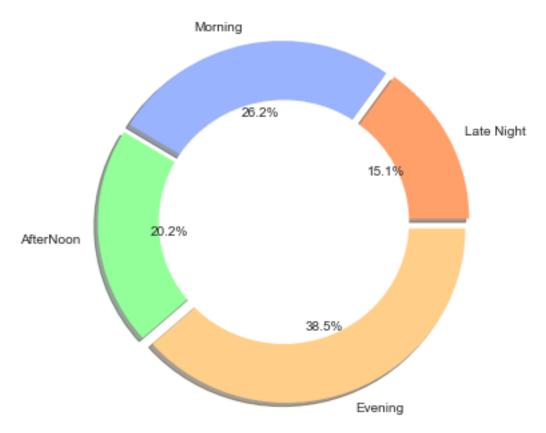
Total Distance traveled in Different times of the day

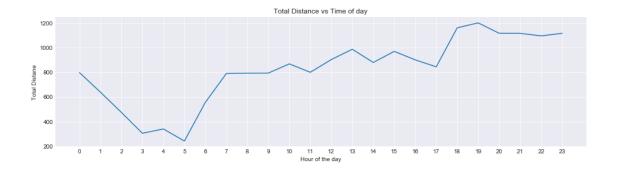


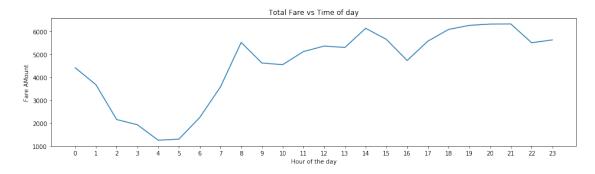
```
In [7]: sum_fareamount = random_rows.groupby('time_part_of_day')['fare_amount'].sum()
       plot = sum_fareamount.plot(kind ='pie',
                            x = 'time_hour',
                            y = 'fare_amount',
                            title='Earning Share in Different times of the day',
                            legend = False,
                            figsize=(6,6),
                            shadow=True,
                            labels=["Late Night", 'Morning', 'AfterNoon', 'Evening'],
                            colors = ['#ff9f69','#99b3ff','#92ff99','#ffcf89'],
                            explode = (0.04, 0.04, 0.04, .04),
                            autopct='%1.1f%%')
        centre_circle = plt.Circle((0,0),0.70, fc = 'white')
       plt.gcf().gca().add_artist(centre_circle)
       plot.set_xlabel('')
       plot.set_ylabel('')
```

Out[7]: Text(0,0.5,'')

Earning Share in Different times of the day









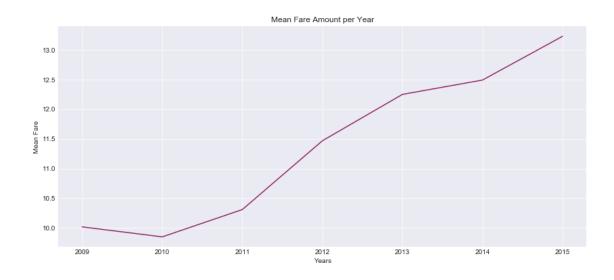
```
 \  \, \hbox{In [10]: \#Reference : $https://stackoverflow.com/questions/11927715/how-to-give-a-pandas-matple} \, \, \\
In [11]: year_fare_mean = pandas.DataFrame(random_rows.groupby('year')['fare_amount'].mean().a
                                            pandas.Series(range(2009,2016), name="year"))
         display(year_fare_mean)
         plot = year_fare_mean.plot(kind = 'line',
                          stacked=True,
                          color = cm.inferno_r(np.linspace(.6,.9, 24)),
                          legend=False,
                          figsize=(14,6),
                          title="Mean Fare Amount per Year")
         plot.set_xlabel("Years")
         _ = plot.set_ylabel("Mean Fare")
      fare_amount
year
2009
        10.017088
2010
         9.848972
2011
        10.309377
2012
        11.468890
2013
        12.251040
```

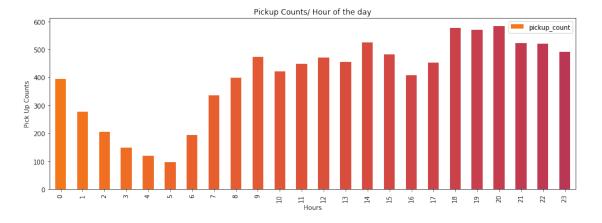
2014

2015

12.492615

13.230117



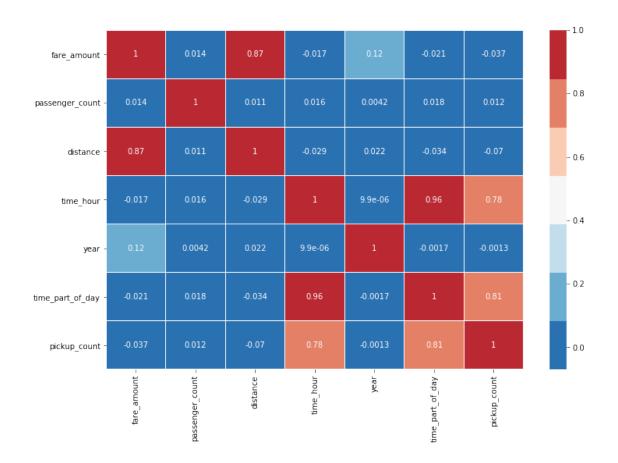


New generated Features

```
In [291]: cleaned_csv = pandas.read_csv('cleaned_data.csv')
          cleaned_csv['time_part_of_day'] = pandas.cut(cleaned_csv['time_hour'],
                                                        [-1, 6, 12, 16, 24],
                                                        labels = [1, 2, 3, 4]).astype(int)
          feature_csv = pandas.merge(cleaned_csv,
                                     sum_pickup[['time_hour','pickup_count']],
                                     on='time hour')
          feature_csv = feature_csv.filter(['fare_amount',
                                      'passenger_count',
                                      'distance',
                                      'time_hour',
                                      'year',
                                      'time_part_of_day',
                                      'pickup_count',
                                      ], axis=1)
          display(feature_csv.head())
   fare_amount passenger_count distance
                                           time_hour year time_part_of_day
0
          4.5
                                 0.639764
                                                   17 2009
          16.5
                                                  17 2012
                                                                            4
1
                              1 2.584861
           7.7
2
                              2 1.037366
                                                  17 2011
                                                                            4
3
                                                                            4
          7.5
                             1 1.007130
                                                  17 2012
4
          13.5
                              1 1.997993
                                                  17 2013
                                                                            4
  pickup_count
0
            452
            452
1
2
            452
3
            452
            452
4
  Correlation Heatmap of the new features
In [292]: pearson_corr = feature_csv.corr(method='pearson')
          display(feature_csv.head())
          labels = [[ 'fare_amount', 'passenger_count', 'distance', 'time_hour',
                          'year', 'time_part_of_day', 'pickup_count']]
          seaborn.heatmap(pearson_corr,
                          cmap=seaborn.color_palette("RdBu_r", 7),
                      linewidths=.1.
                      annot=True)
          plt.gcf().set_size_inches(12,8)
  fare_amount passenger_count distance time_hour year time_part_of_day \
0
           4.5
                              1 0.639764
                                                  17 2009
```

1	16.5	1	2.584861	17	2012	4
2	7.7	2	1.037366	17	2011	4
3	7.5	1	1.007130	17	2012	4
4	13.5	1	1.997993	17	2013	4

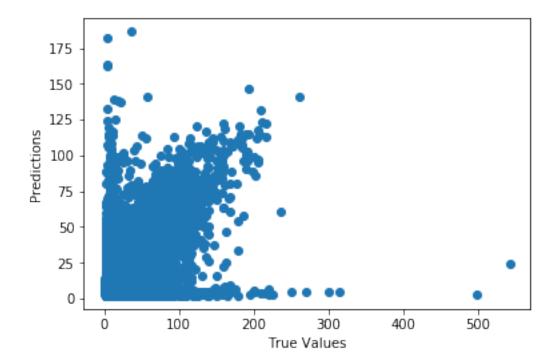
pickup_count
0 452
1 452
2 452
3 452
4 452



Baseline model: Linear Regression

Generate Baseline Model Scores

Out[296]: Text(0,0.5,'Predictions')



Accuracy: 77.0% RMS Error: 20.936110128423508

Advanced model: Using Random Forest Regressor Reference: https://stackoverflow.com/questions/41925157/logisticregression-unknown-label-type-continuous-using-sklearn-in-python