```
In [131]: import numpy as np
          import pandas as pd
          import os
          from dateutil.parser import parse
          import matplotlib.pyplot as plt
          from math import radians, cos, sin, asin, sqrt
          from scipy.stats import pearsonr
          from sklearn.linear_model import LinearRegression as LR
          import re
          import random
          from datetime import datetime
          from datetime import timedelta
          from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error
          import logging
          from sklearn.ensemble import RandomForestRegressor as RF
          import warnings
         warnings.filterwarnings('ignore')
In [111]: data_path = 'data/Year1/'
         TIME_BEGIN = datetime(2014,12,1)
          TIME_END = datetime(2015,12,1)
         TIME_DELTA = timedelta(0,0,0,0,15) #15 minutes delta
          MIN_INPUT = 96 #entries in a day
         MAX INPUT = 34944 #max_minutes - 1 day
         DAY_MINUTES = 96 #entry count in a day
In [134]: def get_data(path, count=None):
             train_df = pd.read_csv(path, nrows=count, header=None)
             return train_df
In [135]: home dfs = []
          for home_path in home_paths:
             home dfs.append (get data(data path + home path))
In [136]: #building time features
         curtime = TIME_BEGIN
          month_list = []
         day_list = [] #day of month
weekday_list = []
          minute_list = []
          while curtime < TIME END:
             month_list.append(curtime.month)
             day_list.append(curtime.day)
             weekday_list.append(curtime.isoweekday())
             minute_list.append(curtime.hour*60 + curtime.minute)
             curtime += TIME DELTA
In [101]: #enrich data with time features
          #features considered : month, day_of_month, day_of_week, minute
          date_df = pd.DataFrame()
          date df['month'] = month list
          date_df['day'] = day_list
          date_df['weekday'] = weekday_list
         date df['minute'] = minute list
In [102]: X = date_df
```

```
In [112]: timeslot = int(input('Input Timeslot in entry index: '))
          if timeslot > MAX INPUT:
              timeslot = MAX_INPUT
          if timeslot < MIN_INPUT:</pre>
              timeslot = MIN INPUT
          print ('Timeslot considered is: ' + str(timeslot))
          Input Timeslot in entry index: 1232354235
          Timeslot considered is: 34944
 In [ ]: X train = X[:timeslot]
          X_test = X[timeslot:timeslot+DAY_MINUTES]
In [129]: # Linear Regression
          print ('MODEL : LINEAR REGRESSION')
          linReg = LR (normalize=True, n_jobs=-1) #n_jobs=-1 uses all the cpus
          i = 1
          for home in home_dfs:
              Y_train = home[0][:timeslot]
              Y_test = home[0][timeslot:timeslot+DAY_MINUTES]
              linReg.fit(X_train,Y_train)
              Y_pred_linear = linReg.predict(X_test)
              print ('MAE: House[' + str(i) + '], ' + str(mean_absolute_error(Y_pred_linear, Y_test)))
              i += 1
          MODEL : LINEAR REGRESSION
          MAE: House[1], 1.3842771912426108
          MAE: House[2], 1.6575772741759216
          MAE: House[3], 1.757195698276184
          MAE: House[4], 0.8508289271089308
          MAE: House[5], 1.6159036257161234
          MAE: House[6], 1.0995928273446747
          MAE: House[7], 1.7165909204053094
          MAE: House[8], 1.6850882502435782
          MAE: House[9], 1.264611838218349
          MAE: House[10], 1.3748836072516915
In [132]: print ('MODEL : RANDOM FOREST')
          rand_forest_reg = RF(random_state=0, n_jobs = -1, n_estimators = 10, oob_score = True)
          i = 1
          for home in home_dfs:
              Y_train = home[0][:timeslot]
              Y_test = home[0][timeslot:timeslot+DAY_MINUTES]
              rand_forest_reg.fit (X_train, Y_train)
              Y_pred_rf = rand_forest_reg.predict(X_test)
              print ('MAE: House[' + str(i) + '], ' + str(mean_absolute_error(Y_pred_rf, Y_test)))
              i += 1
          MODEL : RANDOM FOREST
          MAE: House[1], 1.1018142677083334
          MAE: House[2], 1.4142444208333333
          MAE: House[3], 1.0788536979166667
          MAE: House[4], 0.8294385572916667
          MAE: House[5], 1.0180408385416666
          MAE: House[6], 1.0075472694444445
          MAE: House[7], 1.0564943020833333
          MAE: House[8], 1.2987328552083335
          MAE: House[9], 0.7396080083333333
          MAE: House[10], 1.1589714322916664
```