

In [42]:

```

1 #import libraries
2 import xgboost as xgb
3 import itertools
4 from prophet import Prophet
5 import pandas as pd
6 import numpy as np

```

In [2]:

```

1 data = pd.read_csv("D:/Forecasting Models and Time Series for Business in Python/Dai
2 data.head()

```

Out[2]:

	instant	dteday	season	yr	mnth	holiday	weekday	workingday	weathersit	temp
0	1	2011-01-01	1	0	1	0	6	0	2	0.344167
1	2	2011-01-02	1	0	1	0	0	0	2	0.363478
2	3	2011-01-03	1	0	1	0	1	1	1	0.196364
3	4	2011-01-04	1	0	1	0	2	1	1	0.200000
4	5	2011-01-05	1	0	1	0	3	1	1	0.226957

In [32]:

```

1 #select variables
2 dataset = data.loc[:, ["dteday", "cnt", "holiday", "workingday", "weathersit",
3 "temp", "atemp", "hum", "windspeed"]]
4 dataset.head()

```

Out[32]:

	dteday	cnt	holiday	workingday	weathersit	temp	atemp	hum	windspeed
0	2011-01-01	985	0	0	2	0.344167	0.363625	0.805833	0.160446
1	2011-01-02	801	0	0	2	0.363478	0.353739	0.696087	0.248539
2	2011-01-03	1349	0	1	1	0.196364	0.189405	0.437273	0.248309
3	2011-01-04	1562	0	1	1	0.200000	0.212122	0.590435	0.160296
4	2011-01-05	1600	0	1	1	0.226957	0.229270	0.436957	0.186900

In [33]:

```

1 #Date variable
2 dataset.dteday = pd.to_datetime(dataset.dteday,
3                                 format = "%Y-%m-%d")
4 dataset.dteday

```

Out[33]:

```

0      2011-01-01
1      2011-01-02
2      2011-01-03
3      2011-01-04
4      2011-01-05
...
726    2012-12-27
727    2012-12-28
728    2012-12-29
729    2012-12-30
730    2012-12-31
Name: dteday, Length: 731, dtype: datetime64[ns]

```

In [34]:

```

1 #renaming variable
2 dataset = dataset.rename(columns = {'cnt' : 'y'})
3 dataset = dataset.rename(columns = {'dteday' : 'ds'})
4 dataset.head()

```

Out[34]:

	ds	y	holiday	workingday	weathersit	temp	atemp	hum	windspeed
0	2011-01-01	985	0	0	2	0.344167	0.363625	0.805833	0.160446
1	2011-01-02	801	0	0	2	0.363478	0.353739	0.696087	0.248539
2	2011-01-03	1349	0	1	1	0.196364	0.189405	0.437273	0.248309
3	2011-01-04	1562	0	1	1	0.200000	0.212122	0.590435	0.160296
4	2011-01-05	1600	0	1	1	0.226957	0.229270	0.436957	0.186900

In [35]:

```

1 holiday_dates = dataset[dataset.holiday == 1].ds
2 holidays = pd.DataFrame({'holiday' : 'holi',
3                           'ds': pd.to_datetime(holiday_dates),
4                           'lower_window': -3,
5                           'upper_window': 1})
6 holidays.head()

```

Out[35]:

	holiday	ds	lower_window	upper_window
16	holi	2011-01-17	-3	1
51	holi	2011-02-21	-3	1
104	holi	2011-04-15	-3	1
149	holi	2011-05-30	-3	1
184	holi	2011-07-04	-3	1

In [36]:

```

1 #removing holiday column
2 dataset = dataset.drop(columns = "holiday")
3 dataset.head(1)

```

Out[36]:

	ds	y	workingday	weathersit	temp	atemp	hum	windspeed
0	2011-01-01	985	0	2	0.344167	0.363625	0.805833	0.160446

In [37]:

```

1 #Training and test set
2 test_days = 31
3 training_set = dataset.iloc[:-test_days, :]
4 test_set = dataset.iloc[-test_days:, :]
5 test_set.tail()

```

Out[37]:

	ds	y	workingday	weathersit	temp	atemp	hum	windspeed
726	2012-12-27	2114	1	2	0.254167	0.226642	0.652917	0.350133
727	2012-12-28	3095	1	2	0.253333	0.255046	0.590000	0.155471
728	2012-12-29	1341	0	2	0.253333	0.242400	0.752917	0.124383
729	2012-12-30	1796	0	1	0.255833	0.231700	0.483333	0.350754
730	2012-12-31	2729	1	2	0.215833	0.223487	0.577500	0.154846

In [38]:

```

1  #Facebook Prophet model
2  m = Prophet(growth = "linear",
3              yearly_seasonality = True,
4              weekly_seasonality = True,
5              daily_seasonality = False,
6              holidays = holidays,
7              seasonality_mode = "multiplicative",
8              seasonality_prior_scale = 20,
9              holidays_prior_scale = 20,
10             changepoint_prior_scale = 0.01)
11 m.add_regressor('workingday')
12 m.add_regressor('weathersit')
13 m.add_regressor('temp')
14 m.add_regressor('atemp')
15 m.add_regressor('hum')
16 m.add_regressor('windspeed')
17 m.fit(training_set)

```

18:12:20 - cmdstanpy - INFO - Chain [1] start processing

18:12:20 - cmdstanpy - INFO - Chain [1] done processing

Out[38]:

&lt;prophet.forecaster.Prophet at 0x2a7f2ef3640&gt;

In [39]:

```

1  #Create Future Dataframe
2  future = m.make_future_dataframe(periods = len(test_set),
3                                  freq = "D")
4  future.tail()

```

Out[39]:

	ds
726	2012-12-27
727	2012-12-28
728	2012-12-29
729	2012-12-30
730	2012-12-31

In [40]:

```
1 #merge regressors
2 future = pd.concat([future, dataset.iloc[:,2:]],
3                     axis = 1)
4 future.tail()
```

Out[40]:

	ds	workingday	weathersit	temp	atemp	hum	windspeed
726	2012-12-27	1	2	0.254167	0.226642	0.652917	0.350133
727	2012-12-28	1	2	0.253333	0.255046	0.590000	0.155471
728	2012-12-29	0	2	0.253333	0.242400	0.752917	0.124383
729	2012-12-30	0	1	0.255833	0.231700	0.483333	0.350754
730	2012-12-31	1	2	0.215833	0.223487	0.577500	0.154846

In [41]:

```
1 #forecast
2 forecast = m.predict(future)
3 forecast.head()
```

Out[41]:

	ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	atemp	atemp_
0	2011-01-01	2429.776904	274.639879	2164.623680	2429.776904	2429.776904	-0.055503	-0.0
1	2011-01-02	2435.398244	52.781019	1982.814006	2435.398244	2435.398244	-0.060175	-0.0
2	2011-01-03	2441.019585	370.166907	2254.347723	2441.019585	2441.019585	-0.137830	-0.1
3	2011-01-04	2446.640925	450.453198	2305.364375	2446.640925	2446.640925	-0.127095	-0.1
4	2011-01-05	2452.262265	639.018538	2532.923198	2452.262265	2452.262265	-0.118992	-0.1

5 rows × 46 columns



In [43]:

```
1 #XGBoost
2
3 prophet_variables = forecast.loc[:, ["trend", "holi", "weekly", "yearly"]]
4 df_xgb = pd.concat([dataset, prophet_variables], axis = 1)
5 df_xgb.head()
```

Out[43]:

	ds	y	workingday	weathersit	temp	atemp	hum	windspeed	tren
0	2011-01-01	985	0	2	0.344167	0.363625	0.805833	0.160446	2429.7769C
1	2011-01-02	801	0	2	0.363478	0.353739	0.696087	0.248539	2435.39824
2	2011-01-03	1349	1	1	0.196364	0.189405	0.437273	0.248309	2441.0195E
3	2011-01-04	1562	1	1	0.200000	0.212122	0.590435	0.160296	2446.6409E
4	2011-01-05	1600	1	1	0.226957	0.229270	0.436957	0.186900	2452.2622E

In [44]:

```
1 #Training and test set
2 test_days = 31
3 training_set = df_xgb.iloc[:-test_days, :]
4 test_set = df_xgb.iloc[-test_days:, :]
5 test_set.tail()
```

Out[44]:

	ds	y	workingday	weathersit	temp	atemp	hum	windspeed	tr
726	2012-12-27	2114	1	2	0.254167	0.226642	0.652917	0.350133	6403.27C
727	2012-12-28	3095	1	2	0.253333	0.255046	0.590000	0.155471	6408.71E
728	2012-12-29	1341	0	2	0.253333	0.242400	0.752917	0.124383	6414.15E
729	2012-12-30	1796	0	1	0.255833	0.231700	0.483333	0.350754	6419.59E
730	2012-12-31	2729	1	2	0.215833	0.223487	0.577500	0.154846	6425.03E

In [45]:

```
1 #isolate X and Y
2 y_train = training_set.y
3 y_test = test_set.y
4 X_train = training_set.iloc[:, 2:]
5 X_test = test_set.iloc[:, 2:]
```

In [46]:

```
1 #create XGBoost Matrices
2 Train = xgb.DMatrix(data = X_train, label = y_train)
3 Test = xgb.DMatrix(data = X_test, label = y_test)
```

In [50]:

```
1 #Set the parameters
2 parameters = {'learning_rate': 0.1,
3               'max_depth': 3,
4               'colsample_bytree': 1,
5               'subsample': 1,
6               'min_child_weight': 1,
7               'gamma': 1,
8               'random_state': 1502,
9               'eval_metric': "rmse",
10              'objective': "reg:squarederror"}
```

In [52]:

```
1 #XGBoost Model
2 model = xgb.train(params = parameters,
3                   dtrain = Train,
4                   num_boost_round = 100,
5                   evals = [(Test, "y")])
```



```
[0]    y-rmse:3891.58894
[1]    y-rmse:3563.69059
[2]    y-rmse:3274.75734
[3]    y-rmse:2950.62380
[4]    y-rmse:2716.06266
[5]    y-rmse:2522.39072
[6]    y-rmse:2317.96280
[7]    y-rmse:2156.40260
[8]    y-rmse:2024.56200
[9]    y-rmse:1934.05555
[10]   y-rmse:1866.22580
[11]   y-rmse:1791.30150
[12]   y-rmse:1740.37022
[13]   y-rmse:1686.21001
[14]   y-rmse:1645.00477
[15]   y-rmse:1607.49185
[16]   y-rmse:1550.18027
[17]   y-rmse:1505.37005
[18]   y-rmse:1497.57975
[19]   y-rmse:1468.93298
[20]   y-rmse:1447.49851
[21]   y-rmse:1470.55166
[22]   y-rmse:1458.53921
[23]   y-rmse:1450.04251
[24]   y-rmse:1436.79017
[25]   y-rmse:1437.45231
[26]   y-rmse:1434.34955
[27]   y-rmse:1426.79344
[28]   y-rmse:1405.47873
[29]   y-rmse:1382.48373
[30]   y-rmse:1368.36629
[31]   y-rmse:1372.42708
[32]   y-rmse:1364.42167
[33]   y-rmse:1354.23317
[34]   y-rmse:1351.62577
[35]   y-rmse:1345.77694
[36]   y-rmse:1345.84280
[37]   y-rmse:1341.13245
[38]   y-rmse:1338.39152
[39]   y-rmse:1326.23727
[40]   y-rmse:1326.54284
[41]   y-rmse:1325.68060
[42]   y-rmse:1313.05545
[43]   y-rmse:1304.07389
[44]   y-rmse:1299.94952
[45]   y-rmse:1290.71035
[46]   y-rmse:1293.13980
[47]   y-rmse:1293.57000
[48]   y-rmse:1279.87623
[49]   y-rmse:1279.88202
[50]   y-rmse:1282.58339
[51]   y-rmse:1282.58685
[52]   y-rmse:1279.46683
[53]   y-rmse:1268.14083
[54]   y-rmse:1262.12109
[55]   y-rmse:1257.84226
[56]   y-rmse:1258.36123
[57]   y-rmse:1261.15983
[58]   y-rmse:1261.33869
[59]   y-rmse:1260.90076
[60]   y-rmse:1257.12425
```

```
[61] y-rmse:1237.10839
[62] y-rmse:1236.71193
[63] y-rmse:1235.84508
[64] y-rmse:1238.61690
[65] y-rmse:1235.14405
[66] y-rmse:1235.04261
[67] y-rmse:1223.69949
[68] y-rmse:1221.92818
[69] y-rmse:1216.91122
[70] y-rmse:1216.01084
[71] y-rmse:1218.76650
[72] y-rmse:1218.01949
[73] y-rmse:1217.88212
[74] y-rmse:1207.76128
[75] y-rmse:1205.51613
[76] y-rmse:1202.63937
[77] y-rmse:1202.52231
[78] y-rmse:1203.09201
[79] y-rmse:1202.82743
[80] y-rmse:1201.42101
[81] y-rmse:1192.39748
[82] y-rmse:1187.60552
[83] y-rmse:1187.49280
[84] y-rmse:1188.94596
[85] y-rmse:1188.47885
[86] y-rmse:1186.20477
[87] y-rmse:1186.19375
[88] y-rmse:1186.19311
[89] y-rmse:1184.95346
[90] y-rmse:1184.79903
[91] y-rmse:1183.38123
[92] y-rmse:1183.37107
[93] y-rmse:1178.39280
[94] y-rmse:1178.29195
[95] y-rmse:1181.48770
[96] y-rmse:1182.13371
[97] y-rmse:1181.01612
[98] y-rmse:1180.26977
[99] y-rmse:1174.05661
```

In [53]:

```
1 #Forecasting
2 predictions_xgb = pd.Series(model.predict(Test), name = "XGBoost")
3 predictions_xgb.index = test_set.ds
4 predictions_xgb[:2]
```

Out[53]:

```
ds
2012-12-01    3996.192383
2012-12-02    3068.570312
Name: XGBoost, dtype: float32
```

In [56]:

```
1 #set up index
2 training_set.index = training_set.ds
3 test_set.index = test_set.ds
```

Out[56]:

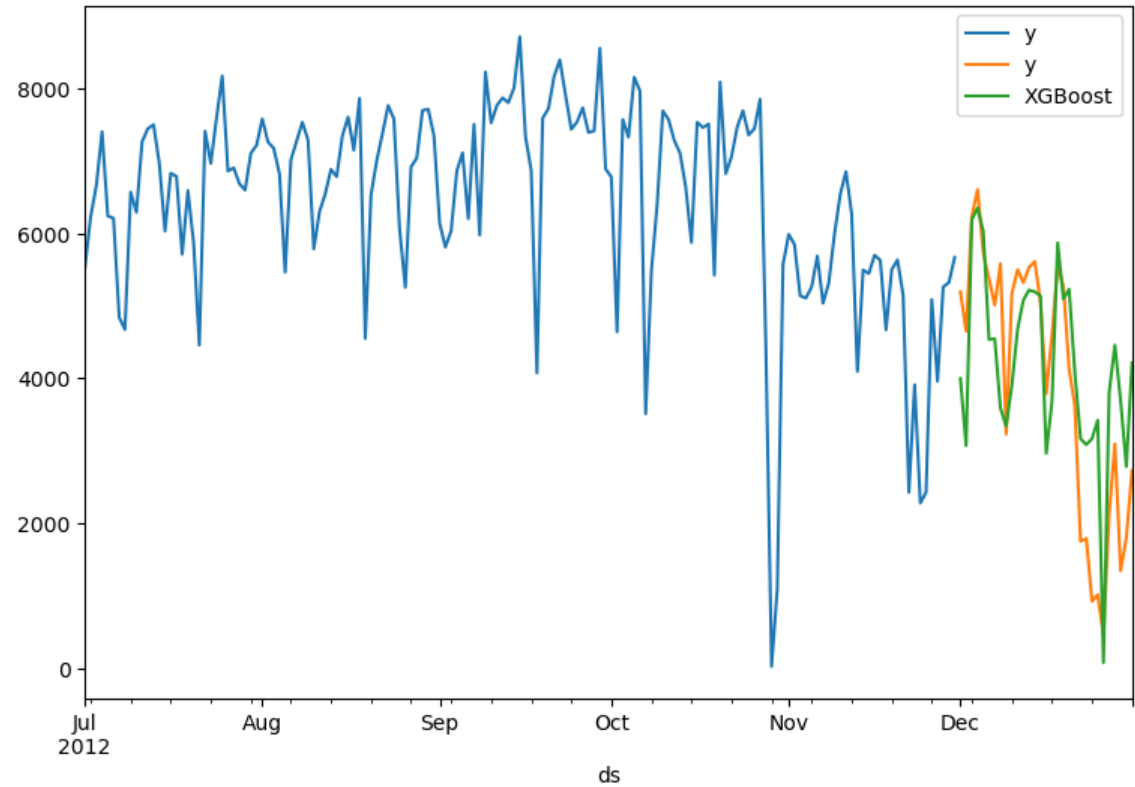
	ds	y	workingday	weathersit	temp	atemp	hum	windspeed	
	ds								
2012-12-01	2012-12-01	5191	0	2	0.298333	0.316904	0.806667	0.059704	6261.8
2012-12-02	2012-12-02	4649	0	2	0.347500	0.359208	0.823333	0.124379	6267.2
2012-12-03	2012-12-03	6234	1	1	0.452500	0.455796	0.767500	0.082721	6272.7
2012-12-04	2012-12-04	6606	1	1	0.475833	0.469054	0.733750	0.174129	6278.1
2012-12-05	2012-12-05	5729	1	1	0.438333	0.428012	0.485000	0.324021	6283.6
2012-12-06	2012-12-06	5375	1	1	0.255833	0.258204	0.508750	0.174754	6289.0
2012-12-07	2012-12-07	5008	1	2	0.320833	0.321958	0.764167	0.130600	6294.4
2012-12-08	2012-12-08	5582	0	2	0.381667	0.389508	0.911250	0.101379	6299.9
2012-12-09	2012-12-09	3228	0	2	0.384167	0.390146	0.905417	0.157975	6305.3
2012-12-10	2012-12-10	5170	1	2	0.435833	0.435575	0.925000	0.190308	6310.8
2012-12-11	2012-12-11	5501	1	2	0.353333	0.338363	0.596667	0.296037	6316.2
2012-12-12	2012-12-12	5319	1	2	0.297500	0.297338	0.538333	0.162937	6321.6
2012-12-13	2012-12-13	5532	1	1	0.295833	0.294188	0.485833	0.174129	6327.1
2012-12-14	2012-12-14	5611	1	1	0.281667	0.294192	0.642917	0.131229	6332.5
2012-12-15	2012-12-15	5047	0	1	0.324167	0.338383	0.650417	0.106350	6337.9
2012-12-16	2012-12-16	3786	0	2	0.362500	0.369938	0.838750	0.100742	6343.4
2012-12-17	2012-12-17	4585	1	2	0.393333	0.401500	0.907083	0.098258	6348.8
2012-12-18	2012-12-18	5557	1	1	0.410833	0.409708	0.666250	0.221404	6354.3
2012-12-19	2012-12-19	5267	1	1	0.332500	0.342162	0.625417	0.184092	6359.7
2012-12-20	2012-12-20	4128	1	2	0.330000	0.335217	0.667917	0.132463	6365.1
2012-12-21	2012-12-21	3623	1	2	0.326667	0.301767	0.556667	0.374383	6370.6
2012-12-22	2012-12-22	1749	0	1	0.265833	0.236113	0.441250	0.407346	6376.0
2012-12-23	2012-12-23	1787	0	1	0.245833	0.259471	0.515417	0.133083	6381.5

	ds	y	workingday	weathersit	temp	atemp	hum	windspeed	
	ds								
2012-12-24	2012-12-24	920	1	2	0.231304	0.258900	0.791304	0.077230	6386.9
2012-12-25	2012-12-25	1013	0	2	0.291304	0.294465	0.734783	0.168726	6392.3
2012-12-26	2012-12-26	441	1	3	0.243333	0.220333	0.823333	0.316546	6397.8
2012-12-27	2012-12-27	2114	1	2	0.254167	0.226642	0.652917	0.350133	6403.2
2012-12-28	2012-12-28	3095	1	2	0.253333	0.255046	0.590000	0.155471	6408.7
2012-12-29	2012-12-29	1341	0	2	0.253333	0.242400	0.752917	0.124383	6414.1
2012-12-30	2012-12-30	1796	0	1	0.255833	0.231700	0.483333	0.350754	6419.5
2012-12-31	2012-12-31	2729	1	2	0.215833	0.223487	0.577500	0.154846	6425.0

```
In [57]:
1 #viz
2 training_set.y['2012-07-01:'].plot(figsize = (9,6), legend = True)
3 test_set.y.plot(legend = True)
4 predictions_xgb.plot(legend = True)
```

Out[57]:

<AxesSubplot: xlabel='ds'>



In [58]:

```
1 #MAE and RMSE
2 from sklearn.metrics import mean_squared_error, mean_absolute_error
3 print(round(mean_absolute_error(test_set['y'], predictions_xgb),0))
4 print(round(np.sqrt(mean_squared_error(test_set['y'], predictions_xgb)), 0))
```

946.0

1174.0

In [59]:

```
1 #MAPE function
2 def MAPE(y_true, y_pred):
3     y_true, y_pred = np.array(y_true), np.array(y_pred)
4     return np.mean(np.abs((y_true - y_pred) / y_true)) * 100
5 MAPE(test_set['y'], predictions_xgb)
```

Out[59]:

45.377004733275925

In [ ]:

1