```
import pandas as pd
from prophet import Prophet
import matplotlib.pyplot as plt
from sklearn.metrics import mean_squared_error
import numpy as np
from statsmodels.tsa.seasonal import seasonal_decompose
from sklearn.preprocessing import StandardScaler
```

```
In [21]: # df = pd.read_csv('train_clean_v2.csv')
# df.drop([df.columns[0]],inplace=True,axis=1)
# df.head()
```

Out[21]:

•		Timestamp	% Baseline	maxtempC	mintempC	sunHour	uvIndex	DewPointC	FeelsLikeC
	0	2014-01-01 07:00:00	0.0079	-3	-6	8.7	2	-14	-13
	1	2014-01-01 08:00:00	0.1019	-3	-6	8.7	2	-14	-12
	2	2014-01-01 09:00:00	0.3932	-3	-6	8.7	2	-14	-11
	3	2014-01-01 10:00:00	0.5447	-3	-6	8.7	2	-14	-10
	4	2014-01-01 11:00:00	0.5485	-3	-6	8.7	2	-14	-10

5 rows × 25 columns

```
In [22]: ## Load the trained .h5 model
# from tensorflow.keras.models import load_model
# model = load_model('ann_modelv2.h5')

## Identify rows with NaN in % Baseline
# nan_rows = df[df['% Baseline'].isna()]

## Prepare the features (excluding Timestamp and % Baseline)
# features = nan_rows.drop(columns=['Timestamp', '% Baseline'])

## Scale the features (ensure the same scaler used during training is applied here
# from sklearn.preprocessing import StandardScaler
# scaler = StandardScaler()
# scaled_features = scaler.fit_transform(features)

## Make predictions using the model
# predictions = model.predict(scaled_features)

## Fill the NaN values in the original dataframe
# df.loc[df['% Baseline'].isna(), '% Baseline'] = predictions
```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be buil t. `model.compile_metrics` will be empty until you train or evaluate the model.

570/570 ________ 1s 1ms/step

```
In [25]: # # Display the updated dataframe
         # df.head()
         # # Save the updated dataframe to a new CSV file
         # df.to_csv('analysisv2.csv', index=False)
In [60]: # Load the analysis.csv data
         analysis = pd.read csv('analysisv2.csv')
         # Ensure the 'Timestamp' column is in datetime format
         analysis['Timestamp'] = pd.to_datetime(analysis['Timestamp'])
         # Extract hour and month from the 'ds' column
         analysis['hour'] = analysis['Timestamp'].dt.hour
         analysis['month'] = analysis['Timestamp'].dt.month
         # Create the interaction term 'hour_month'
         analysis['hour_month'] = analysis['hour'] * analysis['month']
         # Define the cutoff date
         cutoff_date = pd.Timestamp('2017-10-01 00:00:00')
         analysis_df = analysis[analysis['Timestamp'] < cutoff_date]</pre>
         ttest = analysis[analysis['Timestamp'] >= cutoff_date]
         # Prepare the data for Prophet
         # Rename the columns to 'ds' for the date/time and 'y' for the target value
         analysis_df['ds'] = pd.to_datetime(analysis_df['Timestamp'])
         analysis_df['y'] = analysis_df['% Baseline']
         # Optional: Add cap and floor if using logistic growth
         analysis df['cap'] = 1.0 # Example cap
         analysis_df['floor'] = 0.0 # Example floor
         # Fit and transform the humidity and sunHour columns
         scaler = StandardScaler()
         analysis_df[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit tra
         # Keep only the columns required by Prophet
         prophet_df = analysis_df[['ds', 'y','cap','floor', 'ssunHour', 'shumidity','scloudd
         split_point = int(len(analysis_df) * 0.8)
         # Split the data based on the split point
         train_df = prophet_df[:split_point]
         test_df = prophet_df[split_point:]
         # Extract the target variable '% Baseline'
         #y_train = train_df['% Baseline'].values
         #y_test = test_df['% Baseline'].values
         # Create a DataFrame for prediction, using the 'ds' from test df
         #future_test = test_df[['ds','ssunHour','shumidity']].copy()
```

```
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:17: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser guide/indexing.html#returning-a-view-versus-a-copy
  analysis_df['ds'] = pd.to_datetime(analysis_df['Timestamp'])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:18: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
  analysis_df['y'] = analysis_df['% Baseline']
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:20: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser guide/indexing.html#returning-a-view-versus-a-copy
  analysis_df['cap'] = 1.0 # Example cap
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:21: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
  analysis_df['floor'] = 0.0 # Example floor
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel 28172\3752391712.py:24: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
  analysis_df[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_tr
ansform(analysis_df[['humidity', 'sunHour','cloudcover','hour_month']])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:24: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser guide/indexing.html#returning-a-view-versus-a-copy
  analysis_df[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_tr
ansform(analysis_df[['humidity', 'sunHour','cloudcover','hour_month']])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:24: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
```

```
ser_guide/indexing.html#returning-a-view-versus-a-copy
    analysis_df[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_tr
    ansform(analysis_df[['humidity', 'sunHour','cloudcover','hour_month']])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3752391712.py:24: SettingWithCopy
Warning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
    analysis_df[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_tr
    ansform(analysis_df[['humidity', 'sunHour','cloudcover','hour_month']])
```

In [61]: prophet_df

Out[61]:		ds	у	сар	floor	ssunHour	shumidity	scloudcover	shour_month
	0	2014- 01-01 07:00:00	0.007900	1.0	0.0	-0.422555	-1.681535	-0.795529	-1.037312
	1	2014- 01-01 08:00:00	0.101900	1.0	0.0	-0.422555	-1.888416	-0.852592	-1.021246
	2	2014- 01-01 09:00:00	0.393200	1.0	0.0	-0.422555	-2.095297	-0.909655	-1.005181
	3	2014- 01-01 10:00:00	0.544700	1.0	0.0	-0.422555	-2.095297	-0.738466	-0.989116
	4	2014- 01-01 11:00:00	0.548500	1.0	0.0	-0.422555	-2.164257	-0.567277	-0.973050
	•••								
	32844	2017- 09-30 19:00:00	0.189902	1.0	0.0	-0.826374	0.180396	0.431328	1.597404
	32845	2017- 09-30 20:00:00	0.322137	1.0	0.0	-0.826374	0.111436	0.003354	1.741992
	32846	2017- 09-30 21:00:00	0.251104	1.0	0.0	-0.826374	0.042475	-0.396087	1.886580
	32847	2017- 09-30 22:00:00	0.241324	1.0	0.0	-0.826374	0.111436	-0.595808	2.031169
	32848	2017- 09-30 23:00:00	0.271460	1.0	0.0	-0.826374	0.180396	-0.824060	2.175757

32849 rows × 8 columns

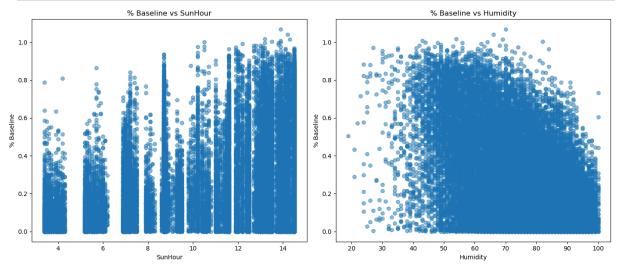
```
In [62]: # Scatter plot between % Baseline and sunHour
plt.figure(figsize=(14, 6))

plt.subplot(1, 2, 1)
plt.scatter(analysis_df['sunHour'], analysis_df['% Baseline'], alpha=0.5)
plt.title('% Baseline vs SunHour')
plt.xlabel('SunHour')
plt.ylabel('% Baseline')

# Scatter plot between % Baseline and humidity
plt.subplot(1, 2, 2)
```

```
plt.scatter(analysis_df['humidity'], analysis_df['% Baseline'], alpha=0.5)
plt.title('% Baseline vs Humidity')
plt.xlabel('Humidity')
plt.ylabel('% Baseline')

plt.tight_layout()
plt.show()
```



```
In [63]: import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         # Assume 'train_df' is your DataFrame and 'Timestamp' is your time column
         #train_df['Timestamp'] = pd.to_datetime(train_df['Timestamp'])
         # Extract time-based features
         train_df['hour'] = train_df['ds'].dt.hour
         train_df['day_of_week'] = train_df['ds'].dt.dayofweek
         train_df['month'] = train_df['ds'].dt.month
         # Plot boxplot for hourly seasonality
         plt.figure(figsize=(14, 7))
         sns.boxplot(x='hour', y='y', data=train_df)
         plt.title('Hourly Seasonality: % Baseline by Hour of the Day')
         plt.show()
         # Plot boxplot for daily seasonality
         plt.figure(figsize=(14, 7))
         sns.boxplot(x='day_of_week', y='y', data=train_df)
         plt.title('Daily Seasonality: % Baseline by Day of the Week')
         plt.show()
         # Plot boxplot for monthly seasonality
         plt.figure(figsize=(14, 7))
         sns.boxplot(x='month', y='y', data=train_df)
         plt.title('Monthly Seasonality: % Baseline by Month')
         plt.show()
```

C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\1788810459.py:9: SettingWithCopyW
arning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

train_df['hour'] = train_df['ds'].dt.hour

C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\1788810459.py:10: SettingWithCopy
Warning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

train_df['day_of_week'] = train_df['ds'].dt.dayofweek

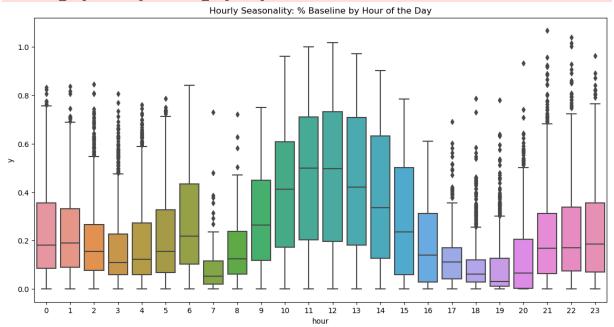
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\1788810459.py:11: SettingWithCopy
Warning:

A value is trying to be set on a copy of a slice from a DataFrame.

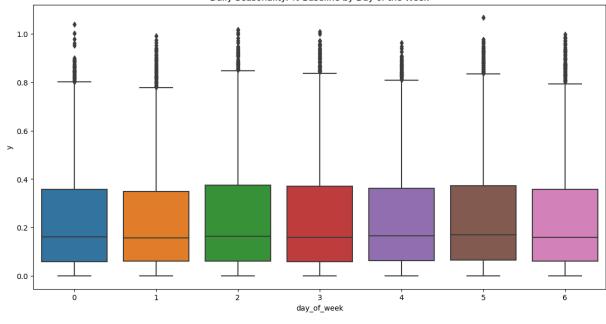
Try using .loc[row_indexer,col_indexer] = value instead

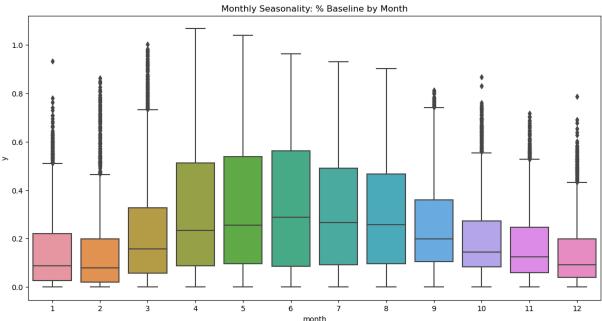
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

train_df['month'] = train_df['ds'].dt.month



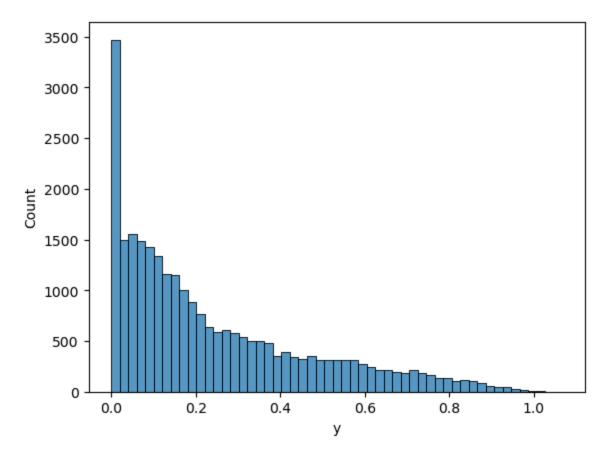






In [64]: sns.histplot(train_df['y'])
 plt.show

Out[64]: <function matplotlib.pyplot.show(close=None, block=None)>



```
In [65]: # Calculate Q1 (25th percentile) and Q3 (75th percentile)
Q1 = train_df['y'].quantile(0.25)
Q3 = train_df['y'].quantile(0.75)
IQR = Q3 - Q1

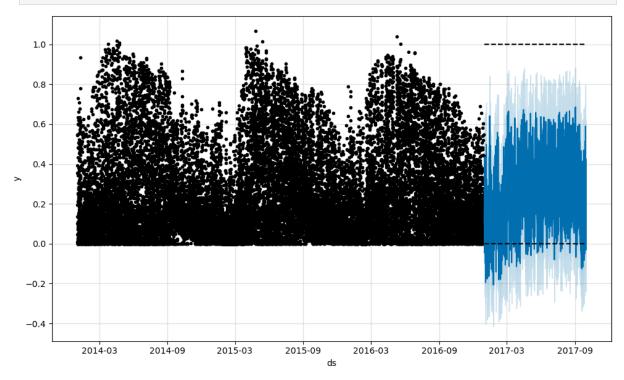
# Define outliers as values below Q1 - 1.5*IQR or above Q3 + 1.5*IQR
tol = 2
outliers = train_df[(train_df['y'] < (Q1 - tol * IQR)) | (train_df['y'] > (Q3 + tol print(f"Number of outliers detected: {len(outliers)}")
```

Number of outliers detected: 25

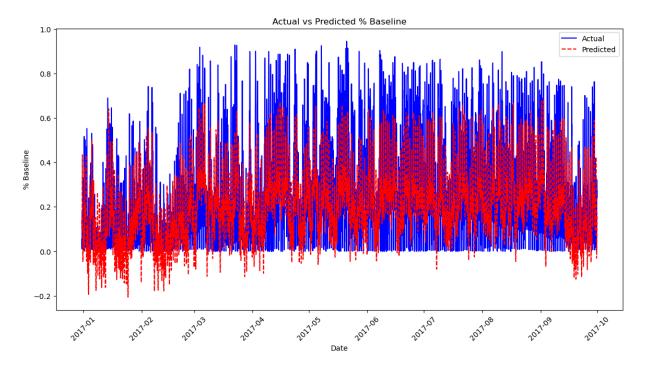
```
In [66]: result = seasonal_decompose(train_df['y'], model='additive',period=24)
    fig = plt.figure()
    fig = result.plot()
    fig.set_size_inches(16, 9)
```

<Figure size 640x480 with 0 Axes>

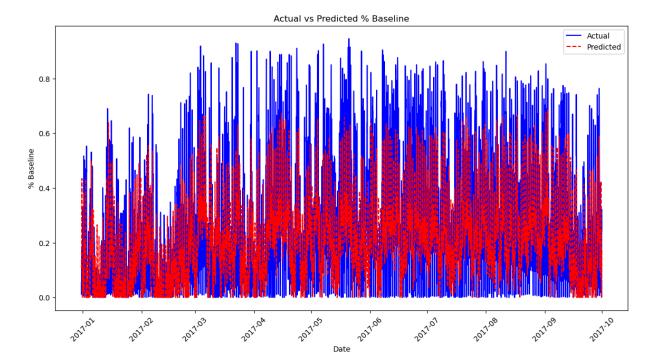
```
In [69]: # Plot the forecast
    model.plot(forecast)
    plt.show()
```



```
In [70]: def rmse(x,y):
             return np.sqrt(mean_squared_error(x,y))
In [71]: # Assuming 'test_predictions' contains 'ds' (timestamps), 'y' (actual values), and
         plt.figure(figsize=(14, 7))
         # Plot the actual values
         plt.plot(test_predictions['ds'], test_predictions['y'], label='Actual', color='blue
         # Plot the predicted values
         plt.plot(test_predictions['ds'], test_predictions['yhat'], label='Predicted', color
         # Add labels and title
         plt.xlabel('Date')
         plt.ylabel('% Baseline')
         plt.title('Actual vs Predicted % Baseline')
         plt.legend()
         # Rotate the x-axis labels for better readability
         plt.xticks(rotation=45)
         # Display the plot
         plt.show()
```



```
In [72]: # Assuming 'forecast' is your DataFrame with the predicted values
         test_predictions['yhat1'] = test_predictions['yhat'].apply(lambda x: max(x, 0))
         # Assuming 'test_predictions' contains 'ds' (timestamps), 'y' (actual values), and
         plt.figure(figsize=(14, 7))
         # Plot the actual values
         plt.plot(test_predictions['ds'], test_predictions['y'], label='Actual', color='blue
         # Plot the predicted values
         plt.plot(test_predictions['ds'], test_predictions['yhat1'], label='Predicted', colo
         # Add labels and title
         plt.xlabel('Date')
         plt.ylabel('% Baseline')
         plt.title('Actual vs Predicted % Baseline')
         plt.legend()
         # Rotate the x-axis labels for better readability
         plt.xticks(rotation=45)
         # Display the plot
         plt.show()
```



```
In [73]: print(rmse(test_predictions['yhat1'],test_predictions['y']))
    print(rmse(test_predictions['yhat'],test_predictions['y']))
```

- 0.17355783250325976
- 0.17648716554008712

PRED

```
In [77]: # Rename the 'Timestamp' column in ttest to 'ds'
    #ttest.rename(columns={'Timestamp': 'ds'}, inplace=True)
    ttest[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_transform
    ttest
```

```
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3652278203.py:3: SettingWithCopyW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
 ttest[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_transfor
m(ttest[['humidity', 'sunHour','cloudcover','hour_month']])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3652278203.py:3: SettingWithCopyW
arning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
 ttest[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_transfor
m(ttest[['humidity', 'sunHour','cloudcover','hour_month']])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3652278203.py:3: SettingWithCopyW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
 ttest[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_transfor
m(ttest[['humidity', 'sunHour','cloudcover','hour_month']])
C:\Users\NAUFAL\AppData\Local\Temp\ipykernel_28172\3652278203.py:3: SettingWithCopyW
arning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/u
ser_guide/indexing.html#returning-a-view-versus-a-copy
 ttest[['shumidity', 'ssunHour','scloudcover','shour_month']] = scaler.fit_transfor
m(ttest[['humidity', 'sunHour','cloudcover','hour_month']])
```

Out[77]:

	ds	% Baseline	maxtempC	mintempC	sunHour	uvlndex	DewPointC	FeelsLike
32849	2017- 10-01 00:00:00	0.237560	21	9	11.6	5	7	
32850	2017- 10-01 01:00:00	0.352637	21	9	11.6	5	6	
32851	2017- 10-01 02:00:00	0.192091	21	9	11.6	5	6	
32852	2017- 10-01 03:00:00	0.118935	21	9	11.6	5	6	
32853	2017- 10-01 04:00:00	0.119415	21	9	11.6	5	7	
•••								
35052	2017- 12-31 19:00:00	0.123297	-9	-13	8.7	1	-17	-2
35053	2017- 12-31 20:00:00	0.147535	-9	-13	8.7	1	-17	-2
35054	2017- 12-31 21:00:00	0.180168	-9	-13	8.7	1	-17	-2
35055	2017- 12-31 22:00:00	0.248257	-9	-13	8.7	1	-17	-í
35056	2017- 12-31 23:00:00	0.216462	-9	-13	8.7	1	-17	-2

2208 rows × 32 columns

In [80]: prophet_df

Out[80]:		ds	у	сар	floor	ssunHour	shumidity	scloudcover	shour_month
	0	2014- 01-01 07:00:00	0.007900	1.0	0.0	-0.422555	-1.681535	-0.795529	-1.037312
	1	2014- 01-01 08:00:00	0.101900	1.0	0.0	-0.422555	-1.888416	-0.852592	-1.021246
	2	2014- 01-01 09:00:00	0.393200	1.0	0.0	-0.422555	-2.095297	-0.909655	-1.005181
	3	2014- 01-01 10:00:00	0.544700	1.0	0.0	-0.422555	-2.095297	-0.738466	-0.989116
	4	2014- 01-01 11:00:00	0.548500	1.0	0.0	-0.422555	-2.164257	-0.567277	-0.973050
	32844	2017- 09-30 19:00:00	0.189902	1.0	0.0	-0.826374	0.180396	0.431328	1.597404
	32845	2017- 09-30 20:00:00	0.322137	1.0	0.0	-0.826374	0.111436	0.003354	1.741992
	32846	2017- 09-30 21:00:00	0.251104	1.0	0.0	-0.826374	0.042475	-0.396087	1.886580
	32847	2017- 09-30 22:00:00	0.241324	1.0	0.0	-0.826374	0.111436	-0.595808	2.031169
	32848	2017- 09-30 23:00:00	0.271460	1.0	0.0	-0.826374	0.180396	-0.824060	2.175757

32849 rows × 8 columns

```
In [79]: # Create a date range from 1st October 2014 00:00 to 31st December 2017 23:00
date_range = pd.date_range(start='2017-10-01 00:00', end='2017-12-31 23:00', freq='
# Create a DataFrame with this date range
date_df = pd.DataFrame(date_range, columns=['ds'])
date_df = pd.merge(date_df, ttest[['ds', 'shumidity', 'ssunHour','scloudcover','sho date_df
```

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υu	L	/ / /	

	ds	shumidity	ssunHour	scloudcover	shour_month
0	2017-10-01 00:00:00	0.644684	1.666754	-0.983866	-1.644198
1	2017-10-01 01:00:00	0.713048	1.666754	-1.011822	-1.514222
2	2017-10-01 02:00:00	0.781411	1.666754	-1.067734	-1.384246
3	2017-10-01 03:00:00	0.781411	1.666754	-1.095690	-1.254270
4	2017-10-01 04:00:00	0.713048	1.666754	-1.095690	-1.124294
•••					
2203	2017-12-31 19:00:00	-0.107313	0.471993	-0.983866	1.319258
2204	2017-12-31 20:00:00	-0.038950	0.471993	-1.039778	1.475229
2205	2017-12-31 21:00:00	0.029414	0.471993	-1.095690	1.631200
2206	2017-12-31 22:00:00	0.097777	0.471993	-1.095690	1.787172
2207	2017-12-31 23:00:00	0.166140	0.471993	-1.095690	1.943143

2208 rows × 5 columns

```
In [83]: # Create a date range from 1st October 2014 00:00 to 31st December 2017 23:00
    date_range = pd.date_range(start='2017-10-01 00:00', end='2017-12-31 23:00', freq='
    # Create a DataFrame with this date range
    date_df = pd.DataFrame(date_range, columns=['ds'])
    # Ensure the future dataframe also has cap and floor values
    date_df['cap'] = 1.0 # Same cap as in the training data
    date_df['floor'] = 0 # Same floor as in the training data
    date_df = pd.merge(date_df, ttest[['ds', 'shumidity', 'ssunHour', 'scloudcover', 'sho
    # Reorder the columns to swap 'ssunHour' and 'shumidity'
    date_df = date_df[['ds', 'cap', 'floor', 'ssunHour', 'shumidity', 'scloudcover', 'sho
    forecast = model.predict(date_df)
    # Merge the forecast with the test set (to include actual values for comparison)
    pred = pd.merge(date_df, forecast[['ds', 'yhat']], on='ds', how='left')
```

```
In [84]: submit = pd.read_csv('dataset/sample_submission.csv')
# Convert 'Timestamp' in submit to datetime format
a = submit.copy()
submit['Timestamp'] = pd.to_datetime(submit['Timestamp'], format='%b %d, %Y %I%p')

# Filter the pred DataFrame to include only the rows with matching timestamps in su
filtered_pred = pred[pred['ds'].isin(submit['Timestamp'])]
filtered_pred
```

		ds	сар	floor	ssunHour	shumidity	scloudcover	shour_month	yhat
		2017-10- 01 06:00:00	1.0	0	1.666754	0.507958	-1.095690	-0.864341	0.117762
		2017-10- 01 07:00:00	1.0	0	1.666754	-0.107313	-1.095690	-0.734365	0.125560
	8	2017-10- 01 08:00:00	1.0	0	1.666754	-0.654221	-1.095690	-0.604389	0.172740
	9	2017-10- 01 09:00:00	1.0	0	1.666754	-1.201128	-1.095690	-0.474413	0.284134
	10	2017-10- 01 10:00:00	1.0	0	1.666754	-1.474582	-1.095690	-0.344437	0.410183
	•••								
2	2196	2017-12- 31 12:00:00	1.0	0	0.471993	-0.927674	-0.480658	0.227458	0.398960
2	2197	2017-12- 31 13:00:00	1.0	0	0.471993	-0.790947	-0.284966	0.383430	0.339246
2	2198	2017-12- 31 14:00:00	1.0	0	0.471993	-0.722584	-0.061318	0.539401	0.247683
2	2199	2017-12- 31 15:00:00	1.0	0	0.471993	-0.585857	0.134374	0.695372	0.150915
2	2200	2017-12- 31 16:00:00	1.0	0	0.471993	-0.449130	-0.201098	0.851344	0.079800

1077 rows × 8 columns

Out[84]:

```
In [85]: # Reset the index of both DataFrames before concatenation
    a_reset = a.reset_index(drop=True)
    filtered_pred_reset = filtered_pred.reset_index(drop=True)

# Concatenate the 'Timestamp' from a and 'yhat' from filtered_pred
    result_df = pd.concat([a_reset['Timestamp'], filtered_pred_reset['yhat']], axis=1)

# Rename the columns if needed
    result_df.columns = a.columns
    result_df
```

Out[85]:		Timestamp	% Baseline
	0	Oct 1, 2017 6am	0.117762
	1	Oct 1, 2017 7am	0.125560
	2	Oct 1, 2017 8am	0.172740
	3	Oct 1, 2017 9am	0.284134
	4	Oct 1, 2017 10am	0.410183
	•••		
	1072	Dec 31, 2017 12pm	0.398960
	1073	Dec 31, 2017 1pm	0.339246
	1074	Dec 31, 2017 2pm	0.247683
	1075	Dec 31, 2017 3pm	0.150915

1077 rows × 2 columns

1076 Dec 31, 2017 4pm

```
In [86]: result_df.to_csv('submit/submission_prophetv3.csv',index=False)
```

0.079800

In []: