Importing required libraries

```
In [1]: import pandas as pd
import numpy as np

from statsmodels.tsa.ar_model import AutoReg
from statsmodels.tsa.arima.model import ARIMA
from statsmodels.tsa.statespace.sarimax import SARIMAX
from statsmodels.tsa.stattools import adfuller

from sklearn.metrics import mean_absolute_error, mean_squared_error
import matplotlib.pyplot as plt

import warnings
warnings.filterwarnings("ignore")
```

Paths for data import and result storage ¶

```
In [2]: dpath = r"/Users/bashirulazambiswas/Documents/Sadia/SUNY courses/2022_
path = r"/Users/bashirulazambiswas/Documents/Sadia/SUNY courses/2022_F
```

Specification Selection

```
In [3]: station_name = 'QUEE'
  target_variable = 'temperature'
  window = 5
  train_method = 'prediction with target only'
  model = "AR"
```

Data preparation function

```
In [4]: def data_prep(dpath, station_name, target_variable):
            data = classical_time_series_data(dpath,station_name,target_variab
            data.time = time coversion(data.time)
            return data
        def classical time series data(path, station name, dependent variable la
            data_preprocessed = preprocess(path,station_name,dependent_variabl
            if dependent variable label == 'temperature':
                dependent_variable = 'temp_2m_avg [degF]'
            elif dependent_variable_label == 'humidity':
                dependent_variable = 'relative_humidity_avg [percent]'
            elif dependent_variable_label == 'precipitation':
                dependent_variable = 'precip_total [inch]'
            data_classical = data_preprocessed[0][['station','time',dependent]
            return data classical
        def time_coversion(time_col): #Converts time from string to datetime
            time list = [time col[i].split()[0] for i in range(len(time col))]
            return pd.to_datetime(time_list)
```

Preprocessing function

```
In [5]: def preprocess(path, station_name, dependent_variable_label):
            data = pd.read csv(path+"/"+station name+".csv").drop("Unnamed: 0"
            correlation = data.corr().round(2)
            index = correlation.index
            for i in correlation.index:
                if i in index:
                    temp = correlation.loc[i]
                    temp = temp[temp==1]
                    for j in temp.index:
                        if j != i:
                            if j not in index:
                                 continue
                             index = index.drop(j)
            preprocessed_data = pd.DataFrame()
            preprocessed_data['station'] = data['station']
            preprocessed data['time'] = data['time']
            preprocessed data[index] = data[index]
            if dependent variable label == 'temperature':
                dependent_variable = 'temp_2m_avg [degF]'
                independent_feature_label = preprocessed_data.columns.drop(['t
            elif dependent_variable_label == 'humidity':
                dependent_variable = 'relative_humidity_avg [percent]'
                independent_feature_label = preprocessed_data.columns.drop(['r
            elif dependent variable label == 'precipitation':
                dependent_variable = 'precip_total [inch]'
                independent_feature_label = preprocessed_data.columns.drop(['p
            x = preprocessed_data[independent_feature_label]
            y = preprocessed_data[dependent_variable]
            return preprocessed_data,x,y
```

Data generated after preprocessing

```
In [6]: data = data_prep(dpath, station_name, target_variable)
data.head(5)
```

Out [6]:

temp_2m_avg [degl	time	station	
59.	2018-09-10	QUEE	0
67.	2018-09-11	QUEE	1
72.	2018-09-12	QUEE	2
70.	2018-09-13	QUEE	3
70	2018-09-14	QUEE	4

Train-test split function

```
In [7]: | def TrainTestSplit(data, target_variable):
            train_year = [2018, 2019, 2020]
            test year = [2021, 2022]
            if target_variable == 'temperature':
                target = 'temp_2m_avg [degF]'
            elif target variable == 'humidity':
                target = 'relative_humidity_avg [percent]'
            elif target_variable == 'precipitation':
                target = 'precip_total [inch]'
            train = pd.DataFrame()
            for year in train_year:
                 train = train.append(data[(data['time'].dt.year == year)])
            train = train.drop(train[train[target].isna()].index)
            test = pd.DataFrame()
            for year in test year:
                 test = test.append(data[(data['time'].dt.year == year)])
            test = test.drop(test[test[target].isna()].index)
            return train,test
```

Train and test set after splitting

```
In [8]: train,test = TrainTestSplit(data,target_variable)
    print("Train:",train.shape)
    print("Test:",test.shape)
Train: (822, 3)
```

SARIMA function

Test: (618, 3)

```
In [9]: | def SARIMA_model(path, train, test, target_variable, p, d, q, station_name, md
            if target variable == 'temperature':
                target = 'temp_2m_avg [degF]'
            elif target variable == 'humidity':
                target = 'relative_humidity_avg [percent]'
            elif target variable == 'precipitation':
                target = 'precip_total [inch]'
            seasonal\_order = (p,d,q,2)
            model = ARIMA(train[target],order=(0,0,0),seasonal_order=seasonal_
            model fit = model.fit(low memory = True, cov type = None)
            y pred = model fit.predict(1, len(test))
            prediction = pd.DataFrame({"time":test['time'].reset_index(drop=Tr
            return prediction
        def SARIMA_compute_d(data, target_variable):
            # Computes the d for ARIMA model
            if target variable == 'temperature':
                target = 'temp_2m_avg [degF]'
            elif target variable == 'humidity':
                target = 'relative_humidity_avg [percent]'
            elif target_variable == 'precipitation':
                target = 'precip_total [inch]'
            temp = data[target].dropna()
            score = adfuller(temp)[1]
            d = 0
            while score > 0.05:
                temp = temp.diff().dropna()
                score = adfuller(temp)[1]
                d = d+1
            return d
```

Prediction using AutoRegression

```
In [10]: p = window
d = SARIMA_compute_d(train, target_variable)
q = window
prediction = SARIMA_model(path, train, test, target_variable, p, d, q, static
prediction.head(10)
```

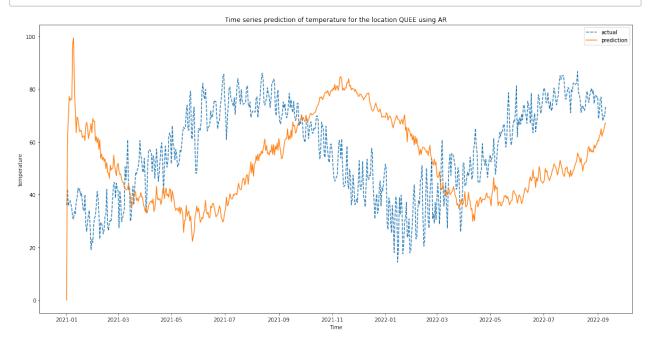
Out [10]:

	time	predicted	actual
0	2021-01-01	0.000000	36.0
1	2021-01-02	60.743908	42.1
2	2021-01-03	69.086232	35.7
3	2021-01-04	77.232233	37.3
4	2021-01-05	75.628392	37.7
5	2021-01-06	76.314170	36.4
6	2021-01-07	76.141697	34.5
7	2021-01-08	97.311764	31.2
8	2021-01-09	99.390809	30.7
9	2021-01-10	79.981243	33.6

Plotting function

```
In [11]: def PlotPrediction(prediction, target_variable, window, station_name, mode
    plt.figure(figsize=(20,10))
    plt.plot(prediction['time'], prediction['actual'], '--', label='actua
    plt.plot(prediction['time'], prediction['predicted'], '-', label='pre
    plt.legend()
    plt.xlabel("Time")
    plt.ylabel(target_variable)
    plt.title("Time series prediction of "+target_variable+" for the l
```

In [12]: PlotPrediction(prediction, target_variable, window, station_name, model, tr



In []: