Importing required libraries

```
In [11]: 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 [12]: dpath = r"/Users/bashirulazambiswas/Documents/Sadia/SUNY courses/2022_
path = r"/Users/bashirulazambiswas/Documents/Sadia/SUNY courses/2022_F
```

Specification Selection

```
In [13]: station_name = 'QUEE'
target_variable = 'temperature'
window = 5
train_method = 'prediction with target only'
model = "AR"
```

Data preparation function

```
In [14]: 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 [15]: 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 [16]: data = data_prep(dpath, station_name, target_variable)
    data.head(5)
```

Out[16]:

	station	time	temp_2m_avg [degF]
0	QUEE	2018-09-10	59.7
1	QUEE	2018-09-11	67.9
2	QUEE	2018-09-12	72.6
3	QUEE	2018-09-13	70.8
4	QUEE	2018-09-14	70.7

Train-test split function

```
In [17]: | 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 [18]: train,test = TrainTestSplit(data,target_variable)
    print("Train:",train.shape)
    print("Test:",test.shape)

Train: (822, 3)
    Test: (618, 3)
```

ARIMA function

```
In [19]: | def ARIMA_model(path,train,test,target_variable,p,d,q,station_name,mod
             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]'
             model = ARIMA(train[target], order=(p,d,q))
             model fit = model.fit()
             y pred = model fit.predict(1, len(test))
             prediction = pd.DataFrame({"time":test['time'].reset index(drop=Tr
             return prediction
         def ARIMA_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 [20]: p = window
d = ARIMA_compute_d(train, target_variable)
q = window
prediction = ARIMA_model(path, train, test, target_variable, p, d, q, station
prediction.head(10)
```

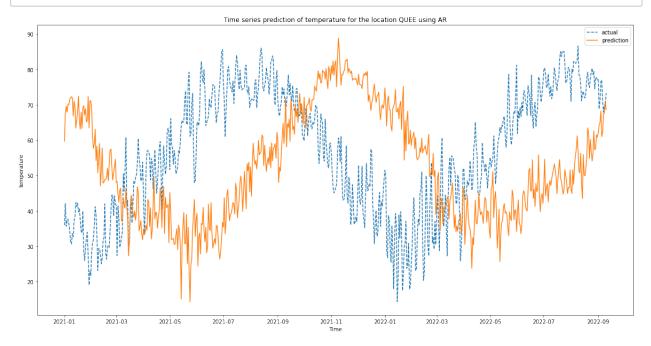
Out [20]:

	time	predicted	actual
0	2021-01-01	59.699943	36.0
1	2021-01-02	67.664962	42.1
2	2021-01-03	69.612503	35.7
3	2021-01-04	67.932879	37.3
4	2021-01-05	70.116839	37.7
5	2021-01-06	70.020964	36.4
6	2021-01-07	71.575969	34.5
7	2021-01-08	71.848404	31.2
8	2021-01-09	72.296489	30.7
9	2021-01-10	71.966349	33.6

Plotting function

```
In [21]: 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 [22]: PlotPrediction(prediction, target_variable, window, station_name, model, tr



In []: