## Virtual bidding in NYISO's markets

Deadline: Nov 27 2023

In this lab, we will implement a simple virtual trading strategy in New York ISO's electricity markets. The goal is to maximize profits. We shall train our model on price data from one year, and implement the strategy on the data from the next year. How much can you earn with a certain daily budget? Say \$250K?

We will present a trading strategy. You are welcome to try other strategies and compare the gains over multiple runs.

Let's start with customary imports.

```
In [2]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   from numpy.random import choice
   from sklearn.preprocessing import StandardScaler
   import seaborn as sns
   import pickle

from sklearn.svm import SVC
   from sklearn.linear_model import LogisticRegression
   from sklearn.neural_network import MLPClassifier
```

# Load the day-ahead and real-time prices from 11 zones in New York.

The day-ahead prices are defined every hour. The real-time market runs every 5 minutes. For each zone, an average of these prices over an hour is published.

Store the list of zones in the variable 'listOfZones'. Also, store the number of options as the number of zones times the 24 hours available for trading. Finally create another list containing the option names (zone + hour).

# Parse the files with DA and RT prices along with DA load forecast.

Define a function that parses three files containing DA and RT prices, along with DA load predictions from a year from all different zones in the list defined before. This function will be used for to both load the data for training the classifiers and testing them. This function has 3 outputs: they are DA prices,

difference between DA and RT prices, and finally DA load predictions. The outputs are pandas data frames whose columns are the options, and rows are the days in the year.

```
In [4]: def loadNYISOData(year):
            # Open the relevant files for DA prices, RT prices, DA load.
            dfPriceDA = pd.read_csv("DAM_NYISO_Zonal_LBMP_" + str(year) + ".csv")
            dfPriceRT = pd.read_csv("RTM_NYISO_Zonal_LBMP_" + str(year) + ".csv")
            dfLoadDA = pd.read csv("DAM NYISO LoadForecast " + str(year) + ".csv")
            # Collect the DA and RT prices from each zone from each hour and create a pandas
            # The data should have prices and loads from all days of a year, where each day
            # contributes 24 rows, corresponding to each hour.
            priceDA = pd.DataFrame({zone: (dfPriceDA.loc[dfPriceDA['Zone Name'] == zone,
                                                          'DAM Zonal LBMP']).values
                                     for zone in listOfZones})
            priceRT = pd.DataFrame({zone: (dfPriceRT.loc[dfPriceRT['Zone Name'] == zone,
                                                          'TWI Zonal LBMP']).values
                                     for zone in listOfZones})
            loadDA = pd.DataFrame({zone: (dfLoadDA.loc[dfLoadDA['Zone Name'] == zone,
                                                        'DAM Forecast Load']).values
                                    for zone in listOfZones})
            numberOfDays = int(len(priceDA.index)/24)
            # Compute the price differences between DA and RT prices for all options on
            # all days of the year. Store it as a pandas data frame where the 24 rows for
            # each day is flattened into one row. This operation essentially allows us to
            # independently think of each zone in each hour as a separate option. Also,
            # reshape the prices for the DA market in the same manner.
            priceDART = pd.DataFrame([priceRT.sub(priceDA).loc[day * 24:
                                                                       (day + 1) * 24 - 1,
                                      listOfZones].values.flatten()
                                      for day in range(numberOfDays)],
                                     columns=optionNames)
            priceDA = pd.DataFrame([priceDA.loc[day * 24: (day + 1) * 24 - 1,
                                    listOfZones].values.flatten()
                                    for day in range(numberOfDays)],
                                   columns=optionNames)
            return priceDA, priceDART, loadDA
```

# Create a function that creates the inputs for training a classifier

Create a function that takes the price and load data and creates two arrays 'X' and 'Y'. Essentially, the rows of 'X' contains all information relevant to predicting the sign of the price difference on the various options on the next day. It takes as an input, three pandas frames corresponding to the DA prices,

price differences, and the DA load predictions, and produces three outputs: the arrays 'X', 'Y', and the

```
In [5]: def createClassifierIO(priceDA, priceDART, loadDA):
            # Define how many past days of prices to use for classification.
            pastPrices = range(1, 3)
            # Define how many past days of load predictions to use for classification.
            pastLoad = range(1, 3)
            # Define a date range within the year to create the arrays 'X' and 'Y' in a way
            # that past price and load data for the first day is within the date range in t\!\!\!\!/
            # pandas frames passed as inputs.
            rangeOfDays = range(3, len(priceDA.index))
            # 'X' will contain three sets of variables:
               1. the DA prices from past days in the list 'pastDays',
               2. the differences between DA and RT prices from the same past days,
                3. the load predictions from past days in the list 'pastLoad'
            X = [np.concatenate((
                priceDA.loc[[(day - h) for h in pastPrices]].values.flatten(),
                priceDART.loc[[(day - h) for h in pastPrices]].values.flatten(),
                loadDA.loc[[(day - h) for h in pastLoad]].values.flatten()
            )) for day in rangeOfDays]
            # Scale the array 'X' to make its data zero mean and unit variance.
            X = StandardScaler().fit_transform(X)
            # 'Y' will contain zeros and ones, where a one indicates that the price in DA is
            # higher than in RT for a particular option. Recall that an option corresponds a
            # a zone at a particular hour of the day.
            Y = np.array([(priceDART.loc[day].values > 0).astype(int)
                          for day in rangeOfDays])
            # Return the arrays 'X' and 'Y', and finally the range of days from the year the
            # will be utilized for training or testing the classifier.
            return X, Y, rangeOfDays
```

## Design the training module.

The training module utilizes a year's worth of data to determine the following for each option, i.e., for each zone for each hour of the day:

- 1. Classifiers that predict the sign of the difference between DA and RT prices.
- 2. Statistics of the mean of the price difference.
- 3. A quantile of the day-ahead prices that we will use as our bid for each option. You will either train the classifiers here or load them from the folder './Classifiers'. Storing the classifiers from time to time allows you to only vary the bidding strategy and observe the annual reward rather than having to train the classifiers every time.

Define and train the classifiers or load pre-trained classifiers.

```
In [6]: classifiers = []
        # We have two options here. Use previous training experience, or learn anew.
        useSavedClassifiers = False
        if not useSavedClassifiers:
            print("Starting training module...\n")
            trainPriceDA, trainPriceDART, trainLoadDA = loadNYISOData(2015)
            numberOfDays = int(len(trainPriceDA.index))
            print("Loaded hourly prices from 2015 for %d days." % numberOfDays)
            # We will implement a trading strategy, where we bid a particular quantile of t\!\!\!/
            # DA prices for an option. If you do not know what a quantile means, refer to t\!\!\!/
            # article on it. Essentially, a 95% quantile of the DA prices equals that value
            # 95% of the DA prices are below it. Store all quantiles starting from 50% in st
            # 5% in a dictionary. Store them in a pickle file.
            quantilesToStore = [0.70, 0.75, 0.80, 0.85, 0.90, 0.95]
            offerPrices = trainPriceDA.quantile(q=quantilesToStore).transpose().to dict()
            pickle.dump(offerPrices, open("./Training/OfferPrices", 'wb'))
            # Calculate the average price spread for each option over the entire year. This
            # us in choosing our portfolio. Store it as a dictionary. Our bid will choose th
            # options that our classifier indicates that they will be profitable and histori
            # have higher average price differences, indicating that they have higher rate \epsilon
            # Store them using pickle.
            averagePriceSpread = trainPriceDART.mean(axis=0).transpose().to dict()
            pickle.dump(averagePriceSpread, open("./Training/AveragePriceSpread", 'wb'))
            # Create the training dataset using the function 'createClassifierIO' on the pr
            # Loads, and store them in 'trainX', and 'trainY'.
            trainX, trainY, = createClassifierIO(trainPriceDA, trainPriceDART, trainLoadD
            # Define a collection of classifiers, one for each option. You can try different
            # as that based on an SVM, logistic regression, multilayer perceptron based, etc
            # measure training accuracy to indicate how well the classifier works on the \mathsf{tr}_{\mathsf{c}}
            # However, good training accuracy does not always indicate good test performance
            # Avoid over-fitting.
            classifiers = [MLPClassifier(hidden layer sizes=(20, 10), max iter=200)
                           for _ in range(nOptions)]
            trainingAccuracy = 0
            for ii in range(nOptions):
                classifiers[ii].fit(trainX, trainY[:, ii])
                print("Classifier trained for option " + optionNames[ii])
                trainingAccuracy += classifiers[ii].score(trainX, trainY[:, ii])
                # Store the classifier.
                pickle.dump(classifiers[ii], open("./Training/Classifier_" + optionNames[ii]
            print("\noverall training accuracy = %1.2f percent." % (100 * trainingAccuracy/)
```

```
del numberOfDays, trainPriceDA, trainLoadDA, trainPriceDART, trainX, trainY
else:
    # Load the classifiers, the offer prices at various quantiles, and the average [
    print("Loading previously trained variables...\n")
    classifiers = [pickle.load(open("./Training/Classifier_" + optionNames[ii], 'rb
                   for ii in range(nOptions)]
    offerPrices = pickle.load(open("./Training/OfferPrices", 'rb'))
    averagePriceSpread = pickle.load(open("./Training/AveragePriceSpread", 'rb'))
    print("All training variables were loaded successfully...\n")
Starting training module...
Loaded hourly prices from 2015 for 365 days.
Classifier trained for option CAPITL Hour 0
Classifier trained for option CAPITL_Hour_1
Classifier trained for option CAPITL_Hour_2
Classifier trained for option CAPITL Hour 3
Classifier trained for option CAPITL_Hour_4
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option CAPITL Hour 5
Classifier trained for option CAPITL_Hour_6
Classifier trained for option CAPITL_Hour_7
Classifier trained for option CAPITL Hour 8
Classifier trained for option CAPITL_Hour_9
Classifier trained for option CAPITL Hour 10
Classifier trained for option CAPITL_Hour_11
Classifier trained for option CAPITL_Hour 12
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option CAPITL_Hour_13
Classifier trained for option CAPITL_Hour_14
Classifier trained for option CAPITL_Hour 15
Classifier trained for option CAPITL_Hour_16
Classifier trained for option CAPITL_Hour_17
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
```

```
Classifier trained for option CAPITL Hour 18
Classifier trained for option CAPITL Hour 19
Classifier trained for option CAPITL Hour 20
Classifier trained for option CAPITL Hour 21
Classifier trained for option CAPITL Hour 22
Classifier trained for option CAPITL Hour 23
Classifier trained for option CENTRL Hour 0
Classifier trained for option CENTRL_Hour_1
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option CENTRL Hour 2
Classifier trained for option CENTRL_Hour_3
Classifier trained for option CENTRL Hour 4
Classifier trained for option CENTRL Hour 5
Classifier trained for option CENTRL Hour 6
Classifier trained for option CENTRL_Hour_7
Classifier trained for option CENTRL Hour 8
Classifier trained for option CENTRL Hour 9
Classifier trained for option CENTRL Hour 10
Classifier trained for option CENTRL_Hour_11
Classifier trained for option CENTRL Hour 12
Classifier trained for option CENTRL Hour 13
Classifier trained for option CENTRL Hour 14
Classifier trained for option CENTRL Hour 15
Classifier trained for option CENTRL Hour 16
Classifier trained for option CENTRL Hour 17
Classifier trained for option CENTRL Hour 18
Classifier trained for option CENTRL_Hour_19
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option CENTRL Hour 20
Classifier trained for option CENTRL_Hour_21
Classifier trained for option CENTRL Hour 22
Classifier trained for option CENTRL Hour 23
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option DUNWOD Hour 0
Classifier trained for option DUNWOD_Hour_1
Classifier trained for option DUNWOD_Hour_2
Classifier trained for option DUNWOD Hour 3
Classifier trained for option DUNWOD Hour 4
Classifier trained for option DUNWOD Hour 5
Classifier trained for option DUNWOD Hour 6
Classifier trained for option DUNWOD_Hour_7
Classifier trained for option DUNWOD Hour 8
Classifier trained for option DUNWOD_Hour_9
Classifier trained for option DUNWOD Hour 10
```

Classifier trained for option DUNWOD\_Hour\_11 Classifier trained for option DUNWOD\_Hour\_12

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eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option DUNWOD\_Hour\_13 Classifier trained for option DUNWOD Hour 14

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eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option DUNWOD\_Hour\_15 Classifier trained for option DUNWOD\_Hour\_16

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eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option DUNWOD\_Hour\_17 Classifier trained for option DUNWOD Hour 18

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eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option DUNWOD\_Hour\_19 Classifier trained for option DUNWOD\_Hour\_20 Classifier trained for option DUNWOD\_Hour\_21 Classifier trained for option DUNWOD\_Hour\_22

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option DUNWOD\_Hour\_23 Classifier trained for option GENESE Hour 0

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option GENESE\_Hour\_1

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option GENESE Hour 2
Classifier trained for option GENESE Hour 3
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option GENESE_Hour_4
Classifier trained for option GENESE Hour 5
Classifier trained for option GENESE_Hour 6
Classifier trained for option GENESE_Hour_7
Classifier trained for option GENESE_Hour_8
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option GENESE_Hour_9
Classifier trained for option GENESE_Hour_10
Classifier trained for option GENESE Hour 11
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option GENESE Hour 12
Classifier trained for option GENESE Hour 13
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option GENESE Hour 14
Classifier trained for option GENESE Hour 15
Classifier trained for option GENESE_Hour_16
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option GENESE_Hour_17
Classifier trained for option GENESE Hour 18
Classifier trained for option GENESE_Hour_19
Classifier trained for option GENESE Hour 20
Classifier trained for option GENESE_Hour_21
Classifier trained for option GENESE Hour 22
Classifier trained for option GENESE_Hour_23
Classifier trained for option HUD VL Hour 0
Classifier trained for option HUD VL Hour 1
Classifier trained for option HUD VL Hour 2
Classifier trained for option HUD VL_Hour_3
```

Classifier trained for option HUD VL\_Hour\_4

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option HUD VL\_Hour\_5 Classifier trained for option HUD VL\_Hour\_6

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option HUD VL\_Hour\_7 Classifier trained for option HUD VL\_Hour\_8 Classifier trained for option HUD VL\_Hour\_9 Classifier trained for option HUD VL Hour 10

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eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option HUD VL\_Hour\_11 Classifier trained for option HUD VL\_Hour\_12 Classifier trained for option HUD VL\_Hour\_13 Classifier trained for option HUD VL\_Hour\_14 Classifier trained for option HUD VL Hour 15

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option HUD VL\_Hour\_16 Classifier trained for option HUD VL\_Hour\_17

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option HUD VL\_Hour\_18 Classifier trained for option HUD VL\_Hour\_19 Classifier trained for option HUD VL\_Hour\_20 Classifier trained for option HUD VL\_Hour\_21 Classifier trained for option HUD VL\_Hour\_22

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option HUD VL\_Hour\_23

Classifier trained for option LONGIL\_Hour\_0

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option LONGIL\_Hour\_1 Classifier trained for option LONGIL\_Hour\_2 Classifier trained for option LONGIL Hour 3

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option LONGIL\_Hour\_4 Classifier trained for option LONGIL\_Hour\_5 Classifier trained for option LONGIL\_Hour\_6 Classifier trained for option LONGIL\_Hour\_7 Classifier trained for option LONGIL\_Hour\_8 Classifier trained for option LONGIL Hour\_9

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option LONGIL\_Hour\_10

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option LONGIL\_Hour\_11 Classifier trained for option LONGIL\_Hour\_12 Classifier trained for option LONGIL\_Hour\_13 Classifier trained for option LONGIL Hour 14

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option LONGIL Hour 15
Classifier trained for option LONGIL Hour 16
Classifier trained for option LONGIL Hour 17
Classifier trained for option LONGIL Hour 18
Classifier trained for option LONGIL Hour 19
Classifier trained for option LONGIL Hour 20
Classifier trained for option LONGIL Hour 21
Classifier trained for option LONGIL_Hour_22
Classifier trained for option LONGIL Hour 23
Classifier trained for option MHK VL Hour 0
Classifier trained for option MHK VL Hour 1
Classifier trained for option MHK VL Hour 2
Classifier trained for option MHK VL Hour 3
Classifier trained for option MHK VL Hour 4
Classifier trained for option MHK VL_Hour_5
Classifier trained for option MHK VL_Hour_6
Classifier trained for option MHK VL Hour 7
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option MHK VL_Hour_8
Classifier trained for option MHK VL Hour 9
Classifier trained for option MHK VL Hour 10
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option MHK VL Hour 11
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option MHK VL_Hour_12
Classifier trained for option MHK VL Hour 13
Classifier trained for option MHK VL Hour 14
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option MHK VL Hour 15
Classifier trained for option MHK VL Hour 16
Classifier trained for option MHK VL_Hour_17
Classifier trained for option MHK VL_Hour_18
Classifier trained for option MHK VL Hour 19
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural_network\_multilayer_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
```

Classifier trained for option MHK VL\_Hour\_20

Classifier trained for option MHK VL\_Hour\_21 Classifier trained for option MHK VL\_Hour\_22 Classifier trained for option MHK VL\_Hour\_23 Classifier trained for option MILLWD\_Hour\_0

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option MILLWD\_Hour\_1

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option MILLWD Hour 2

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option MILLWD\_Hour\_3

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option MILLWD\_Hour\_4

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option MILLWD\_Hour\_5

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option MILLWD\_Hour\_6
Classifier trained for option MILLWD\_Hour\_7
Classifier trained for option MILLWD\_Hour\_8
Classifier trained for option MILLWD\_Hour\_9
Classifier trained for option MILLWD\_Hour\_10
Classifier trained for option MILLWD\_Hour\_11
Classifier trained for option MILLWD\_Hour\_12
Classifier trained for option MILLWD\_Hour\_13
Classifier trained for option MILLWD\_Hour\_14
Classifier trained for option MILLWD\_Hour\_15
Classifier trained for option MILLWD\_Hour\_16

Classifier trained for option MILLWD\_Hour\_17

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option MILLWD_Hour_18 Classifier trained for option MILLWD_Hour_19 Classifier trained for option MILLWD_Hour_20 Classifier trained for option MILLWD_Hour_21 Classifier trained for option MILLWD_Hour_22 Classifier trained for option MILLWD_Hour_23 Classifier trained for option N.Y.C._Hour_0 Classifier trained for option N.Y.C. Hour 1
```

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option N.Y.C._Hour_2
Classifier trained for option N.Y.C._Hour_3
Classifier trained for option N.Y.C._Hour_4
Classifier trained for option N.Y.C. Hour_5
```

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option N.Y.C._Hour_6 Classifier trained for option N.Y.C._Hour_7 Classifier trained for option N.Y.C. Hour 8
```

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option N.Y.C.\_Hour\_9

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option N.Y.C. Hour 10
Classifier trained for option N.Y.C. Hour 11
Classifier trained for option N.Y.C. Hour 12
Classifier trained for option N.Y.C. Hour 13
Classifier trained for option N.Y.C. Hour 14
Classifier trained for option N.Y.C. Hour 15
Classifier trained for option N.Y.C._Hour_16
Classifier trained for option N.Y.C._Hour_17
Classifier trained for option N.Y.C. Hour 18
Classifier trained for option N.Y.C. Hour 19
Classifier trained for option N.Y.C. Hour 20
Classifier trained for option N.Y.C. Hour 21
Classifier trained for option N.Y.C. Hour 22
Classifier trained for option N.Y.C. Hour 23
Classifier trained for option NORTH_Hour_0
Classifier trained for option NORTH Hour 1
Classifier trained for option NORTH Hour 2
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option NORTH_Hour_3
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option NORTH_Hour_4
Classifier trained for option NORTH Hour 5
Classifier trained for option NORTH_Hour_6
Classifier trained for option NORTH Hour 7
Classifier trained for option NORTH Hour 8
Classifier trained for option NORTH Hour 9
Classifier trained for option NORTH Hour 10
Classifier trained for option NORTH_Hour_11
Classifier trained for option NORTH Hour 12
Classifier trained for option NORTH_Hour_13
Classifier trained for option NORTH Hour 14
Classifier trained for option NORTH Hour 15
Classifier trained for option NORTH Hour 16
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
Classifier trained for option NORTH Hour 17
Classifier trained for option NORTH_Hour_18
Classifier trained for option NORTH Hour 19
Classifier trained for option NORTH Hour 20
C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural network\ multilayer perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
```

Classifier trained for option NORTH\_Hour\_21

warnings.warn(

Classifier trained for option NORTH\_Hour\_22 Classifier trained for option NORTH\_Hour\_23 Classifier trained for option WEST\_Hour\_0 Classifier trained for option WEST\_Hour\_1 Classifier trained for option WEST\_Hour\_2 Classifier trained for option WEST\_Hour\_3

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option WEST\_Hour\_4

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option WEST\_Hour\_5

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option WEST\_Hour\_6 Classifier trained for option WEST\_Hour\_7 Classifier trained for option WEST\_Hour\_8

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option WEST\_Hour\_9 Classifier trained for option WEST\_Hour\_10

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option WEST Hour 11

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option WEST_Hour_12
Classifier trained for option WEST_Hour_13
Classifier trained for option WEST_Hour_14
Classifier trained for option WEST_Hour_15
Classifier trained for option WEST_Hour_16
Classifier trained for option WEST_Hour_17
Classifier trained for option WEST_Hour_18
Classifier trained for option WEST_Hour_19
```

Classifier trained for option WEST\_Hour\_20

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

Classifier trained for option WEST Hour 21

C:\Users\matt3\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\n
eural\_network\\_multilayer\_perceptron.py:679: ConvergenceWarning: Stochastic Optimi
zer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(

```
Classifier trained for option WEST_Hour_22 Classifier trained for option WEST_Hour_23
```

Overall training accuracy = 99.96 percent.

#### Test the classifier's accuracy on test data.

```
In [7]:
        # First, load the test data from NYISO for the year 2016. Again, utilize our function
        # named 'LoadNYISOData'.
        print("Starting the testing module...\n")
        testPriceDA, testPriceDART, testLoadDA = loadNYISOData(2016)
        # Create the data for the classifier using the function 'createClassifierIO'.
        testX, testY, rangeOfDays = createClassifierIO(testPriceDA, testPriceDART, testLoad
        # The next step is not useful for implementing the trading strategy, but quickly che
        # your trained classifiers are for the test data. Training accuracy is not always in
        # test accuracy.
        testingAccuracy = [classifiers[ii].score(testX, testY[:, ii]) for ii in range(nOptic
        print("Test Accuracy Stats: Min = %.2f%%, Avg = %.2f%%, Max = %.2f%%" %
              (100 * np.min(testingAccuracy),
               100 * np.mean(testingAccuracy),
               100 * np.max(testingAccuracy)))
        # Utilize the classifiers to predict the sign of DA - RT prices for each day in 'ran
        # the test data. Store the result in a pandas data frame with columns as the option
        # day in year as index.
        predictedY = pd.DataFrame(np.column stack([classifiers[ii].predict(testX) for ii in
                                  columns=optionNames, index=rangeOfDays)
```

Starting the testing module...

Test Accuracy Stats: Min = 49.04%, Avg = 58.94%, Max = 69.70%

### Design and implement the trading strategy.

We define a fairly simple trading strategy. Define a total budget that you are willing to spend in the DA market. Recall that we only invest in options where we buy at the DA market and sell at the RT market. When your bid for one unit of an option clears in the DA market, you have to pay the DA price for that option. There are two possibilities:

- 1. Your bid clears: Therefore, your offer price was higher than the DA price.
- 2. Your bid does not clear: Then, the DA price was higher than your bid. In both these cases, the maximum you have to pay is your bid. Therefore, we will enforce that your bids across all options in a day does not exceed your total budget. Keep track of how rewards grow (or fall) through the year as you utilize your strategy. Also, keep track of how much rewards you get from each option. We shall visualize these results after implementing the trading strategy over the NYISO data for 2016.

Choose the bid prices as a suitable quantile of the historical DA prices. The higher the quantile, the better your chances are that your bid will be cleared. However, a higher quantile also indicates that you are budgeting more money for each option, and hence, you will buy fewer options.

```
In [11]:
         dailyBudget = 250000
         quantileOffer =0.8
         # Keep track of your rewards in each day, a cumulative reward over the year, and a t
         # year. Also, keep track of the total reward from each option. Store them as diction
         # indexed by day of the year.
         reward = {}
         cumulativeReward = {}
         totalReward = 0
         optionReturn = dict((option, 0) for option in optionNames)
         # Implement the trading strategy on each day!
         for day in rangeOfDays:
             reward[day] = 0
             # Find the options that your classifier says that should be profitable. Store th
             # names in chosenOptions.
             chosenOptionNumbers = np.ravel(list(np.nonzero(predictedY.loc[day].values)))
             if np.size(chosenOptionNumbers) == 0:
                 continue
             chosenOptions = [optionNames[i] for i in chosenOptionNumbers]
             # Design the portfolio based on average price spreads. Our strategy is that as
             # exceeded your daily budget, pick an option from the list of 'chosenOptions' p
             # the probability of choosing it is proportional to exponential(historical rewal
             # is, a historically profitable option is chosen more often than one that is not
             # decreasing your budget with each bid.
             # Start with an empty portfolio.
             portfolio = dict((option, 0) for option in chosenOptions)
             # Calculate the probabilities of choosing each option among the list 'chosenOpt'
             # 'chosenOptions' contains the options that your classifier indicates as being |
             priceSpreads = [1.0 * averagePriceSpread[option] for option in chosenOptions]
             probabilityOptions = [np.exp(p) for p in priceSpreads]
             probabilityOptions /= np.sum(probabilityOptions)
             # Start with your daily budget.
             budget = dailyBudget
             # Sampling among the profitable options and bid based on them.
             while budget > np.median([offerPrices[quantileOffer][option] for option in chose
                 optionToBuy = choice(chosenOptions, p=probabilityOptions)
                 if budget >= offerPrices[quantileOffer][optionToBuy]:
                     portfolio[optionToBuy] += 1
                     budget -= offerPrices[quantileOffer][optionToBuy]
             # Compute the reward from the day. Go through each of the options you have decid
             # If the DA price is lower than the bid price, then your bid is cleared. For eac
             # have bought, you get a reward equal to the DA - RT price.
             for option in chosenOptions:
                 if testPriceDA.at[day, option] < offerPrices[quantileOffer][option]:</pre>
```

Day 3: Reward (in \$) = -77,997 Day 4: Reward (in \$) = 1,414,047 Day 5: Reward (in \$) = 25,178 Day 6: Reward (in \$) = -28,369 Day 7: Reward (in \$) = -12,575 Day 8: Reward (in \$) = -39,926 Day 9: Reward (in \$) = -42,636 Day 10: Reward (in \$) = -104,316 Day 11: Reward (in \$) = -43,041 Day 12: Reward (in \$) = -56,447 Day 13: Reward (in \$) = -39,753 Day 14: Reward (in \$) = -36,639 Day 15: Reward (in \$) = 8,801 Day 16: Reward (in \$) = 43,385 Day 17: Reward (in \$) = -67,580 Day 18: Reward (in \$) = -39,822 Day 19: Reward (in \$) = 30,899 Day 20: Reward (in \$) = -5,094 Day 21: Reward (in \$) = 35,761 Day 22: Reward (in \$) = 85,031 Day 23: Reward (in \$) = 34,459 Day 24: Reward (in \$) = -86,349 Day 25: Reward (in \$) = -56,894 Day 26: Reward (in \$) = -27,164 Day 27: Reward (in \$) = -6,392 Day 28: Reward (in \$) = -13,114 Day 29: Reward (in \$) = -22,813 Day 30: Reward (in \$) = -28,635 Day 31: Reward (in \$) = -31,701 Day 32: Reward (in \$) = -3,263 Day 33: Reward (in \$) = -211,493 Day 34: Reward (in \$) = -15,485 Day 35: Reward (in \$) = 42,196 Day 36: Reward (in \$) = -36,750 Day 37: Reward (in \$) = -40,318 Day 38: Reward (in \$) = -351 Day 39: Reward (in \$) = 12,944 Day 40: Reward (in \$) = -32,792 Day 41: Reward (in \$) = -38,097 Day 42: Reward (in \$) = -15,285 Day 43: Reward (in \$) = -55,504 Day 44: Reward (in \$) = -33,368 Day 45: Reward (in \$) = -17,148 Day 46: Reward (in \$) = 91,170 Day 47: Reward (in \$) = 54,701 Day 48: Reward (in \$) = -54,520 Day 49: Reward (in \$) = -51,205 Day 50: Reward (in \$) = -55,337 Day 51: Reward (in \$) = 5,147 Day 52: Reward (in \$) = -19,548 Day 53: Reward (in \$) = -7,132 Day 54: Reward (in \$) = 5,048 Day 55: Reward (in \$) = -3,940 Day 56: Reward (in \$) = -43,593 Day 57: Reward (in \$) = -28,529 Day 58: Reward (in \$) = -19,809 Day 59: Reward (in \$) = -16,370 Day 60: Reward (in \$) = -68,574

Day 61: Reward (in \$) = -40,195 Day 62: Reward (in \$) = 1,123 Day 63: Reward (in \$) = -7,826 Day 64: Reward (in \$) = -31,865 Day 65: Reward (in \$) = 13,179 Day 66: Reward (in \$) = 62,808 Day 67: Reward (in \$) = -15,213 Day 68: Reward (in \$) = -44,821 Day 69: Reward (in \$) = -38,966 Day 70: Reward (in \$) = 83,903 Day 71: Reward (in \$) = -9,202 Day 72: Reward (in \$) = -9,939 Day 73: Reward (in \$) = -95,805 Day 74: Reward (in \$) = 768,102 Day 75: Reward (in \$) = -25,953 Day 76: Reward (in \$) = -111,352 Day 77: Reward (in \$) = -47,624 Day 78: Reward (in \$) = -27,026 Day 79: Reward (in \$) = 10,793 Day 80: Reward (in \$) = -40,065 Day 81: Reward (in \$) = 70,832 Day 82: Reward (in \$) = 76,873 Day 83: Reward (in \$) = -107,978 Day 84: Reward (in \$) = 518 Day 85: Reward (in \$) = 67,328 Day 86: Reward (in \$) = -33,391 Day 87: Reward (in \$) = 3,018,976 Day 88: Reward (in \$) = 632,168 Day 89: Reward (in \$) = -41,623 Day 90: Reward (in \$) = -29,384 Day 91: Reward (in \$) = -27,713 Day 92: Reward (in \$) = -28,636 Day 93: Reward (in \$) = 1,835 Day 94: Reward (in \$) = 35,415 Day 95: Reward (in \$) = -7,868 Day 96: Reward (in \$) = 252,318 Day 97: Reward (in \$) = -25,361 Day 98: Reward (in \$) = -2,589 Day 99: Reward (in \$) = -23,776 Day 100: Reward (in \$) = 17,028 Day 101: Reward (in \$) = -24,852 Day 102: Reward (in \$) = -178 Day 103: Reward (in \$) = 158,058 Day 104: Reward (in \$) = -70,129 Day 105: Reward (in \$) = -11,185 Day 106: Reward (in \$) = 3,242 Day 107: Reward (in \$) = 1,191 Day 108: Reward (in \$) = 50,158 Day 109: Reward (in \$) = -4,834 Day 110: Reward (in \$) = -45,891 Day 111: Reward (in \$) = -29,466 Day 112: Reward (in \$) = 7,748 Day 113: Reward (in \$) = -44,023 Day 114: Reward (in \$) = -58,149 Day 115: Reward (in \$) = -121,665 Day 116: Reward (in \$) = 11,151 Day 117: Reward (in \$) = -24,559 Day 118: Reward (in \$) = 8,410 Day 119: Reward (in \$) = -75,064

Day 120: Reward (in \$) = 2,139 Day 121: Reward (in \$) = 39,661 Day 122: Reward (in \$) = 30 Day 123: Reward (in \$) = -62,588 Day 124: Reward (in \$) = 2,813 Day 125: Reward (in \$) = -9,448 Day 126: Reward (in \$) = -11,220 Day 127: Reward (in \$) = -38,888 Day 128: Reward (in \$) = -41,880 Day 129: Reward (in \$) = -93,371 Day 130: Reward (in \$) = -36,004 Day 131: Reward (in \$) = 23,653 Day 132: Reward (in \$) = 157,660 Day 133: Reward (in \$) = 72,988 Day 134: Reward (in \$) = -46,338 Day 135: Reward (in \$) = -13,832 Day 136: Reward (in \$) = -51,746 Day 137: Reward (in \$) = -47,771 Day 138: Reward (in \$) = -87,170 Day 139: Reward (in \$) = 4,327 Day 140: Reward (in \$) = -44,837 Day 141: Reward (in \$) = 18,545 Day 142: Reward (in \$) = 23,988 Day 143: Reward (in \$) = -99,713 Day 144: Reward (in \$) = -77,452 Day 145: Reward (in \$) = -53,809 Day 146: Reward (in \$) = -19,545 Day 147: Reward (in \$) = -25,641 Day 148: Reward (in \$) = 224,376 Day 149: Reward (in \$) = -7,101 Day 150: Reward (in \$) = 50,411 Day 151: Reward (in \$) = 405,633 Day 152: Reward (in \$) = 370,844 Day 153: Reward (in \$) = 281,098 Day 154: Reward (in \$) = -732 Day 155: Reward (in \$) = -40,357 Day 156: Reward (in \$) = -41,118 Day 157: Reward (in \$) = -130,491 Day 158: Reward (in \$) = 17,031 Day 159: Reward (in \$) = -46,387 Day 160: Reward (in \$) = -139,293 Day 161: Reward (in \$) = 5,078 Day 162: Reward (in \$) = 164,122 Day 163: Reward (in \$) = -14,154 Day 164: Reward (in \$) = -6,555 Day 165: Reward (in \$) = 263,544 Day 166: Reward (in \$) = -28,050 Day 167: Reward (in \$) = 24,915 Day 168: Reward (in \$) = -1,148 Day 169: Reward (in \$) = -2,785 Day 170: Reward (in \$) = -49,354 Day 171: Reward (in \$) = -3,990 Day 172: Reward (in \$) = -4,698 Day 173: Reward (in \$) = 178,984 Day 174: Reward (in \$) = -16,519 Day 175: Reward (in \$) = 17,003 Day 176: Reward (in \$) = -16,356 Day 177: Reward (in \$) = -27,640 Day 178: Reward (in \$) = -10,544

Day 179: Reward (in \$) = -45,741 Day 180: Reward (in \$) = -22,836 Day 181: Reward (in \$) = -22,746 Day 182: Reward (in \$) = -35,162 Day 183: Reward (in \$) = -85,605 Day 184: Reward (in \$) = -62,407 Day 185: Reward (in \$) = -20,279 Day 186: Reward (in \$) = 85,492 Day 187: Reward (in \$) = 16,244 Day 188: Reward (in \$) = 5,554 Day 189: Reward (in \$) = -3,008 Day 190: Reward (in \$) = -101,760 Day 191: Reward (in \$) = -24,506 Day 192: Reward (in \$) = -23,310 Day 193: Reward (in \$) = -64,756 Day 194: Reward (in \$) = -239 Day 195: Reward (in \$) = 1,080 Day 196: Reward (in \$) = 6,780 Day 197: Reward (in \$) = 34,721 Day 198: Reward (in \$) = 58,016 Day 199: Reward (in \$) = -1,742 Day 200: Reward (in \$) = -51,952 Day 201: Reward (in \$) = -27,929 Day 202: Reward (in \$) = 1,822 Day 203: Reward (in \$) = 5,709 Day 204: Reward (in \$) = 3,917 Day 205: Reward (in \$) = -72,237 Day 206: Reward (in \$) = -4,342 Day 207: Reward (in \$) = -5,268 Day 208: Reward (in \$) = 187 Day 209: Reward (in \$) = 5,652 Day 210: Reward (in \$) = 78,185 Day 211: Reward (in \$) = 47,243 Day 212: Reward (in \$) = 49,293 Day 213: Reward (in \$) = -1,333 Day 214: Reward (in \$) = -6,153 Day 215: Reward (in \$) = -68,275 Day 216: Reward (in \$) = 7,499 Day 217: Reward (in \$) = 1,578 Day 218: Reward (in \$) = 34,380 Day 219: Reward (in \$) = -25,161 Day 220: Reward (in \$) = -1,783 Day 221: Reward (in \$) = -32,249 Day 222: Reward (in \$) = 15,327 Day 223: Reward (in \$) = 3,433 Day 224: Reward (in \$) = 211 Day 225: Reward (in \$) = 701 Day 226: Reward (in \$) = -23,685 Day 227: Reward (in \$) = -601 Day 228: Reward (in \$) = 1,644 Day 229: Reward (in \$) = -1,446 Day 230: Reward (in \$) = -2,441 Day 231: Reward (in \$) = 2,529 Day 232: Reward (in \$) = -38,296 Day 233: Reward (in \$) = -52,423 Day 234: Reward (in \$) = -103,445 Day 235: Reward (in \$) = -107,855 Day 236: Reward (in \$) = 3,696 Day 237: Reward (in \$) = 51,077

Day 238: Reward (in \$) = -2,805 Day 239: Reward (in \$) = -14,784 Day 240: Reward (in \$) = -13,335 Day 241: Reward (in \$) = -1,161 Day 242: Reward (in \$) = -43,271 Day 243: Reward (in \$) = -3,556 Day 244: Reward (in \$) = -920 Day 245: Reward (in \$) = -87,449 Day 246: Reward (in \$) = -46,785 Day 247: Reward (in \$) = -10,087 Day 248: Reward (in \$) = -32,282 Day 249: Reward (in \$) = -10,263 Day 250: Reward (in \$) = -19,060 Day 251: Reward (in \$) = 1,197 Day 252: Reward (in \$) = 6,249 Day 253: Reward (in \$) = 174,116 Day 254: Reward (in \$) = -13,399 Day 255: Reward (in \$) = -48,373 Day 256: Reward (in \$) = -4,606 Day 257: Reward (in \$) = -6,530 Day 258: Reward (in \$) = -20,819 Day 259: Reward (in \$) = -30,373 Day 260: Reward (in \$) = -32,660 Day 261: Reward (in \$) = 23,488 Day 262: Reward (in \$) = 6,580 Day 263: Reward (in \$) = -641 Day 264: Reward (in \$) = -60,472 Day 265: Reward (in \$) = -25,220 Day 266: Reward (in \$) = -71,096 Day 267: Reward (in \$) = -17,166 Day 268: Reward (in \$) = -22,876 Day 269: Reward (in \$) = -67,001 Day 270: Reward (in \$) = -70,989 Day 271: Reward (in \$) = -31,167 Day 272: Reward (in \$) = -60,764 Day 273: Reward (in \$) = -54,737 Day 274: Reward (in \$) = 12,477 Day 275: Reward (in \$) = 23,891 Day 276: Reward (in \$) = -32,245 Day 277: Reward (in \$) = -6,498 Day 278: Reward (in \$) = 187,018 Day 279: Reward (in \$) = -74,542 Day 280: Reward (in \$) = 13,300 Day 281: Reward (in \$) = -12,646 Day 282: Reward (in \$) = 14,699 Day 283: Reward (in \$) = -2,246 Day 284: Reward (in \$) = 7,632 Day 285: Reward (in \$) = -64,832 Day 286: Reward (in \$) = -57,023 Day 287: Reward (in \$) = -68,967 Day 288: Reward (in \$) = 11,157 Day 289: Reward (in \$) = 16,366 Day 290: Reward (in \$) = 6,418 Day 291: Reward (in \$) = -6,988 Day 292: Reward (in \$) = -3,303 Day 293: Reward (in \$) = 125,354 Day 294: Reward (in \$) = 15,819 Day 295: Reward (in \$) = -13,737 Day 296: Reward (in \$) = -61,193 Day 297: Reward (in \$) = -40,019 Day 298: Reward (in \$) = -12,452 Day 299: Reward (in \$) = -61,745 Day 300: Reward (in \$) = -30,474 Day 301: Reward (in \$) = -11,293 Day 302: Reward (in \$) = -8,657 Day 303: Reward (in \$) = 15,668 Day 304: Reward (in \$) = -51,900 Day 305: Reward (in \$) = -15,861 Day 306: Reward (in \$) = -8,025 Day 307: Reward (in \$) = 45,511 Day 308: Reward (in \$) = -41,108 Day 309: Reward (in \$) = -12,742 Day 310: Reward (in \$) = 2,402 Day 311: Reward (in \$) = -30,469 Day 312: Reward (in \$) = 46,825 Day 313: Reward (in \$) = -8,138 Day 314: Reward (in \$) = -31,158 Day 315: Reward (in \$) = -33,025 Day 316: Reward (in \$) = -11,931 Day 317: Reward (in \$) = -30,702

```
Day 318: Reward (in \$) = -14,243
Day 319: Reward (in $) = 2,940
Day 320: Reward (in $) = 14,970
Day 321: Reward (in $) = 18,877
Day 322: Reward (in \$) = 1,378
Day 323: Reward (in $) = 7,113
Day 324: Reward (in $) = 16,003
Day 325: Reward (in $) = -22,554
Day 326: Reward (in \$) = 864
Day 327: Reward (in $) = 20,204
Day 328: Reward (in $) = 114,144
Day 329: Reward (in $) = 36,110
Day 330: Reward (in $) = 6,351
Day 331: Reward (in $) = -10,353
Day 332: Reward (in $) = -56,458
Day 333: Reward (in $) = -69,144
Day 334: Reward (in $) = -2,620
Day 335: Reward (in \$) = -21,536
Day 336: Reward (in $) = -17,588
Day 337: Reward (in $) = 9,219
Day 338: Reward (in $) = -9,027
Day 339: Reward (in $) = -32,691
Day 340: Reward (in $) = -36,962
Day 341: Reward (in $) = -41,449
Day 342: Reward (in $) = -91,573
Day 343: Reward (in $) = -40,680
Day 344: Reward (in $) = 70,904
Day 345: Reward (in $) = 74,547
Day 346: Reward (in $) = -38,321
Day 347: Reward (in \$) = 9,290
Day 348: Reward (in $) = -43,442
Day 349: Reward (in $) = 40,498
Day 350: Reward (in $) = -375
Day 351: Reward (in $) = 177,879
Day 352: Reward (in $) = 9,244
Day 353: Reward (in $) = -10,948
Day 354: Reward (in $) = -41,961
Day 355: Reward (in \$) = -20,750
Day 356: Reward (in $) = 1,395
Day 357: Reward (in $) = -82,409
Day 358: Reward (in \$) = -48,321
Day 359: Reward (in $) = -10,042
Day 360: Reward (in $) = -12,383
Day 361: Reward (in \$) = -82,146
Day 362: Reward (in $) = 25,445
Day 363: Reward (in $) = 23,829
Day 364: Reward (in $) = -3,940
Day 365: Reward (in \$) = -27,416
Total money earned over the year (in $) = 3,182,875
```

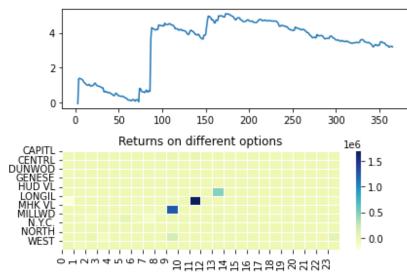
### Task1: Visualize the rewards (25 points)

We would like to plot the cumulative reward over the year 2016. By cumulative reward on a particular date, we mean the total reward from the start of the year till that date.

Also, plot a heat map of the returns from each option.

#### Fill in the missing lines below.

```
In [10]:
         # Plot the cumulative reward over the year 2016. Also, plot a heat map of the return
         # each option.
         fig, axs = plt.subplots(2, 1, tight_layout=True)
         axs = axs.ravel()
         ## Enter code here where you plot in axs[0].
         axs[0].plot(*zip(*sorted(cumulativeReward.items())))
         axs[1] = sns.heatmap(np.reshape(list(optionReturn.values()), (len(listOfZones), 24))
                             linewidth=0.5,
                             cmap="YlGnBu")
         axs[1].set_yticks(range(len(listOfZones)))
         axs[1].set yticklabels(listOfZones, rotation=0)
         axs[1].set_xticks(range(24))
         axs[1].set_xticklabels(range(24), rotation=90)
         axs[1].set_title('Returns on different options')
         plt.show()
```



### Task 2: Choosing a classifier (30 points)

We used a multilayer perceptron classifier. Your task is to try out SVM and logistic regression classifiers, and explore which one leads to more profits. Use the relevant functions from 'sklearn'.

(comments here, add a new code cell below)

```
In [*]: def trading(dailyBudget, quantileOffer):
            # Keep track of your rewards in each day, a cumulative reward over the year, and
            # year. Also, keep track of the total reward from each option. Store them as di\epsilon
            # indexed by day of the year.
            global reward
            global cumulativeReward
            global totalReward
            global optionReturn
            reward = \{\}
            cumulativeReward = {}
            totalReward = 0
            optionReturn = dict((option, 0) for option in optionNames)
            # Implement the trading strategy on each day!
            for day in rangeOfDays:
                reward[day] = 0
                # Find the options that your classifier says that should be profitable. Stol
                # names in chosenOptions.
                chosenOptionNumbers = np.ravel(list(np.nonzero(predictedY.loc[day].values)))
                if np.size(chosenOptionNumbers) == 0:
                    continue
                chosenOptions = [optionNames[i] for i in chosenOptionNumbers]
                # Design the portfolio based on average price spreads. Our strategy is that
                # exceeded your daily budget, pick an option from the list of 'chosenOption's
                # the probability of choosing it is proportional to exponential(historical |
                # is, a historically profitable option is chosen more often than one that is
                # decreasing your budget with each bid.
                # Start with an empty portfolio.
                portfolio = dict((option, 0) for option in chosenOptions)
                # Calculate the probabilities of choosing each option among the list 'chose'
                # 'chosenOptions' contains the options that your classifier indicates as be
                priceSpreads = [1.0 * averagePriceSpread[option] for option in chosenOption
                probabilityOptions = [np.exp(p) for p in priceSpreads]
                probabilityOptions /= np.sum(probabilityOptions)
                # Start with your daily budget.
                budget = dailyBudget
                # Sampling among the profitable options and bid based on them.
                while budget > np.median([offerPrices[quantileOffer][option] for option in
                    optionToBuy = choice(chosenOptions, p=probabilityOptions)
                    if budget >= offerPrices[quantileOffer][optionToBuy]:
                        portfolio[optionToBuy] += 1
                        budget -= offerPrices[quantileOffer][optionToBuy]
                # Compute the reward from the day. Go through each of the options you have \epsilon
                # If the DA price is lower than the bid price, then your bid is cleared. For
                # have bought, you get a reward equal to the DA - RT price.
```

```
for option in chosenOptions:
    if testPriceDA.at[day, option] < offerPrices[quantileOffer][option]:
        rewardOptionDay = testPriceDART.at[day, option] * portfolio[option]
        optionReturn[option] += rewardOptionDay
        reward[day] += rewardOptionDay

totalReward += reward[day]

# Calculate the cumulative reward in millions of dollars.
    cumulativeReward[day] = totalReward/1000000

print("Total money earned over the year (in $) = " + "{0:,.0f}".format(totalReward)</pre>
```

SVC with the kernel set to linear makes the most profit out of all the classifiers.

#### Task 3: Quantile (25 points)

For the best classifier, try different quantile choices from the following list: 0.70, 0.75, 0.80, 0.85, 0.90, 0.95. Rank them in terms of profits (low to high).

How do you expect the quantile to affect the portfolio?

#### (comments here, add a new code cell below)

```
In [*]: # Code for task 3
qlist = [0.70, 0.75, 0.80, 0.85, 0.90, 0.95]
profit = dict((i, 0) for i in qlist)
for q in qlist:
    trading(dailyBudget=250000, quantileOffer=q)
    profit[q] = totalReward
print("\nRank of quantiles:")
print(dict(sorted(profit.items(), key=lambda item: item[1])).keys())
```

Rank in terms of profits:

```
0.7, 0.95, 0.9, 0.85, 0.8, 0.75
```

If the quantile is too small, then the bid price would have a larger probability to be lower than the DA price which would result in lower profit. In turn if the quantile is too high, the budget would be used up faster which would also lead to a smaller chance to make a profit.

### Task 4: Daily budget (25 points)

We used a daily budget of 250,000. Try different budgets, let's say, 100,000, 150,000, 200,000, and 300,000. Rank them in terms of profits (low to high). Fix the quantile to the one that led to maximum profits in the previous task.

(comments here, add a new code cell below)

```
In []: # Code for task 4
blist = [100000, 150000, 200000, 300000]
profit = dict((i, 0) for i in blist)
for b in blist:
    trading(dailyBudget=b, quantileOffer=0.75)
    profit[b] = totalReward
print("\nRank of budgets:")
print(dict(sorted(profit.items(), key=lambda item: item[1])).keys())
```

Rank of daily budget:

100,000, 150,000, 200,000, 300,000

### Task 5: Beat the algorithm! (40 points)

Make changes to increase the profits further! Try implementing a different trading strategy. Explain what did for improvement and how much profits you made. Do not exceed the 250,000 budget!

(comments here, add a new code cell below)

```
In [*]:
        # Code for task 5
        dailyBudget = 250000
        quantileOffer = 0.75
        tune = 0.85
        # Keep track of your rewards in each day, a cumulative reward over the year, and a t
        # year. Also, keep track of the total reward from each option. Store them as diction
        # indexed by day of the year.
        reward = \{\}
        cumulativeReward = {}
        totalReward = 0
        optionReturn = dict((option, 0) for option in optionNames)
        # Implement the trading strategy on each day!
        for day in rangeOfDays:
            reward[day] = 0
            # Find the options that your classifier says that should be profitable. Store the
            # names in chosenOptions.
            chosenOptionNumbers = np.ravel(list(np.nonzero(predictedY.loc[day].values)))
            if np.size(chosenOptionNumbers) == 0:
                continue
            chosenOptions = [optionNames[i] for i in chosenOptionNumbers]
            # Design the portfolio based on average price spreads. Our strategy is that as
            # exceeded your daily budget, pick an option from the list of 'chosenOptions' p
            # the probability of choosing it is proportional to exponential(historical rewal
            # is, a historically profitable option is chosen more often than one that is not
            # decreasing your budget with each bid.
            # Start with an empty portfolio.
            portfolio = dict((option, 0) for option in chosenOptions)
            # Calculate the probabilities of choosing each option among the list 'chosenOpt'
            # 'chosenOptions' contains the options that your classifier indicates as being |
            priceSpreads = [1.0 * averagePriceSpread[option] for option in chosenOptions]
            if tune:
                k_least = int(len(priceSpreads) * tune)
                idx = np.argpartition(np.array(priceSpreads), k_least)[:k_least]
                chosenOptions = np.delete(np.array(chosenOptions), idx)
                priceSpreads = np.delete(np.array(priceSpreads), idx)
            probabilityOptions = [np.exp(p) for p in priceSpreads]
            probabilityOptions /= np.sum(probabilityOptions)
            # Start with your daily budget.
            budget = dailyBudget
            # Sampling among the profitable options and bid based on them.
            while budget > np.median([offerPrices[quantileOffer][option] for option in chose
                optionToBuy = choice(chosenOptions, p=probabilityOptions)
                if budget >= offerPrices[quantileOffer][optionToBuy]:
```

```
portfolio[optionToBuy] += 1
    budget -= offerPrices[quantileOffer][optionToBuy]

# Compute the reward from the day. Go through each of the options you have decid
# If the DA price is lower than the bid price, then your bid is cleared. For ead
# have bought, you get a reward equal to the DA - RT price.

for option in chosenOptions:
    if testPriceDA.at[day, option] < offerPrices[quantileOffer][option]:
        rewardOptionDay = testPriceDART.at[day, option] * portfolio[option]
        optionReturn[option] += rewardOptionDay
        reward[day] += rewardOptionDay

totalReward += reward[day]

# Calculate the cumulative reward in millions of dollars.
cumulativeReward[day] = totalReward/1000000

print("Day " + str(day) + ": Reward (in $) = " + "{0:,.0f}".format(reward[day])

print("Total money earned over the year (in $) = " + "{0:,.0f}".format(totalReward)</pre>
```

Type *Markdown* and LaTeX:  $\alpha^2$ 

For improvement, I added a tuning option to remove some of the options in the data with lower price spreads, increasing the profits compared to the original implementation.