Question1 import pandas as pd import numpy as np # Read file and add column 'TrueLabel' of SPY dfSPY = pd.read csv("C:/Users/Lee/iCloudDrive/Document/Boston University/CS677 DS with Python/Homework/Assignme tLabel = np.zeros like(dfSPY.shape[0]) dfSPY['TrueLabel'] = tLabel # Read file and add column 'TrueLabel' of Google dfGOO = pd.read csv("C:/Users/Lee/iCloudDrive/Document/Boston University/CS677 DS with Python/Homework/Assignme tLabel = np.zeros like(dfSPY.shape[0]) dfGOO['TrueLabel'] = tLabel # Assigning '-' or '+' based on column 'Return' # I personally define Return = 0 as '-' # SPY profitReturn = (dfSPY.Return > 0) lossReturn = (dfSPY.Return <= 0)</pre> dfSPY.loc[profitReturn, 'TrueLabel'] = '+' dfSPY.loc[lossReturn, 'TrueLabel'] = '-' # GOOGL profitReturn = (dfG00.Return > 0) lossReturn = (dfG00.Return <=0)</pre> dfG00.loc[profitReturn, 'TrueLabel'] = '+' dfGOO.loc[lossReturn, 'TrueLabel'] = '-' In [4]: # Locate the index of the last date of the third year temp = dfSPY[dfSPY.Year == 2019].index.values[0] # Creat sub dataframe of first 3 years and sub dataframe of last 2 years # For SPY dfSPY3 = dfSPY.loc[:temp-1, ['Date', 'Return', 'TrueLabel']] dfSPY2 = dfSPY.loc[temp:, ['Date', 'Return', 'TrueLabel']] dfSPY2 = dfSPY2.reset index(drop=True) # Reset the index of testing data # For GOOGL dfG003 = dfG00.loc[:temp-1, ['Date', 'Return', 'TrueLabel']] dfGOO2 = dfGOO.loc[temp:, ['Date', 'Return', 'TrueLabel']] dfG002 = dfG002.reset index(drop=True) # Reset the index of testing data Lplus = sum((dfSPY3.Return > 0)) Lminus = sum((dfSPY3.Return <= 0))</pre> totalDay = dfSPY3.shape[0] print('### Default probability p* = {:.1f}% that the next day is a "up" day'.format(Lplus*100/totalDay)) ### Default probability $p^* = 54.8\%$ that the next day is a "up" day In [6]: def labelSlicing(key, labelArray): k = key predictLabel = ["NA" for i in range(k)] index = klastIndex = len(label) - 1while index <= lastIndex:</pre> i = index sub = labelArray[i-k : i-k+k] index = index + 1temp = '' for e in sub: temp = temp + epredictLabel.append(temp) return predictLabel if __name__ == "__main__": # Creat label with k = 1, 2, 3, 4## For APY label = np.array(dfSPY3.TrueLabel) ### K = 1, 2, 3, 4k1 = labelSlicing(key = 1 , labelArray = label) k2 = labelSlicing(key = 2 , labelArray = label) k3 = labelSlicing(key = 3, labelArray = label)k4 = labelSlicing(key = 4, labelArray = label)### Add three columns into the dfSPY3 dataframe dfSPY3['K=1'] = k1dfSPY3['K=2'] = k2dfSPY3['K=3'] = k3dfSPY3['K=4'] = k4## For GOOGL label = np.array(dfG003.TrueLabel) ### K = 1, 2, 3, 4k1 = labelSlicing(key = 1 , labelArray = label) k2 = labelSlicing(key = 2, labelArray = label)k3 = labelSlicing(key = 3, labelArray = label)k4 = labelSlicing(key = 4, labelArray = label)### Add three columns into the dfSPY3 dataframe dfGOO3['K=1'] = k1dfGOO3['K=2'] = k2dfGOO3['K=3'] = k3dfGOO3['K=4'] = k4# Creat label on last 2 years data with K = 1,2,3,4## For SPY label = np.array(dfSPY2.TrueLabel) ### K = 1, 2, 3, 4k1 = labelSlicing(key = 1 , labelArray = label) k2 = labelSlicing(key = 2 , labelArray = label) k3 = labelSlicing(key = 3 , labelArray = label) k4 = labelSlicing(key = 4, labelArray = label)### Add three columns into the dfSPY3 dataframe dfSPY2['K=1'] = k1dfSPY2['K=2'] = k2dfSPY2['K=3'] = k3dfSPY2['K=4'] = k4## For GOOGL ### K = 1, 2, 3, 4label = np.array(dfG002.TrueLabel) k1 = labelSlicing(key = 1, labelArray = label)k2 = labelSlicing(key = 2 , labelArray = label) k3 = labelSlicing(key = 3, labelArray = label)k4 = labelSlicing(key = 4, labelArray = label)### Add three columns into the dfSPY3 dataframe dfGOO2['K=1'] = k1dfGOO2['K=2'] = k2dfGOO2['K=3'] = k3dfGOO2['K=4'] = k4def keyMatch(preKey, posKey, preLable, posLable): matchCount = 0 for (pre, pos) in zip(preLable, posLable): if (pre, pos) == (preKey, posKey): matchCount = matchCount + 1 return matchCount if name == " main ": # For SPY ### For K consecutive 'down days' tLabel = np.array(dfSPY3['TrueLabel']) ### Count the probability of '- +' as opposed to '- -' preLabel = np.array(dfSPY3['K=1']) upCount = keyMatch(preKey= '-', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '-', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP.append(('K=1', temp)) ### Count the probability of '-- +' as opposed to '-- -' preLabel = np.array(dfSPY3['K=2']) upCount = keyMatch(preKey='--', posKey='+', preLable=preLabel, posLable=tLabel) downCount = keyMatch(preKey='--', posKey='-', preLable=preLabel, posLable=tLabel) temp = upCount / (downCount + upCount) kP.append(('K=2', temp)) ### Count the probability of '--- +' as opposed to '--- -' preLabel = np.array(dfSPY3['K=3']) upCount = keyMatch(preKey='---', posKey='+', preLable=preLabel, posLable=tLabel) downCount = keyMatch(preKey='---', posKey='-', preLable=preLabel, posLable=tLabel) temp = upCount / (downCount + upCount) kP.append(('K=3', temp)) ### Print print('### For SPY, k cosnsecutive "down days"') for p in kP: print('### for {}, the probability of the next day is a ''up day'' is {:.1f}%'.format(p[0], p[1]*100)) print() # For GOO ### For K consecutive 'down days' tLabel = np.array(dfG003['TrueLabel']) ### Count the probability of '- +' as opposed to '- -' preLabel = np.array(dfG003['K=1']) upCount = keyMatch(preKey= '-', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '-', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP.append(('K=1', temp)) ### Count the probability of '-- +' as opposed to '-- -' preLabel = np.array(dfG003['K=2']) upCount = keyMatch(preKey='--', posKey='+', preLable=preLabel, posLable=tLabel) downCount = keyMatch(preKey='--', posKey='-', preLable=preLabel, posLable=tLabel) temp = upCount / (downCount + upCount) kP.append(('K=2', temp)) ### Count the probability of '--- +' as opposed to '--- -' preLabel = np.array(dfG003['K=3']) upCount = keyMatch(preKey='---', posKey='+', preLable=preLabel, posLable=tLabel) downCount = keyMatch(preKey='---', posKey='-', preLable=preLabel, posLable=tLabel) temp = upCount / (downCount + upCount) kP.append(('K=3', temp)) ### Print print('### For GOOGL, k cosnsecutive "down days"') for p in kP: print('### for {} , the probability of the next day is a ''up day'' is {:.1f}%'.format(p[0], p[1]*100)) print() # For SPY ### For consecutive 'up days' tLabel = np.array(dfSPY3['TrueLabel']) kP1 = []### Count the probability of '+ +' as opposed to'+ -' preLabel = np.array(dfSPY3['K=1']) upCount = keyMatch(preKey= '+', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '+', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP1.append(('K=1', temp)) ### Count the probability of '++ +' as opposed to '++ -' preLabel = np.array(dfSPY3['K=2']) upCount = keyMatch(preKey= '++', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '++', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP1.append(('K=2', temp)) ### Count the probability of '+++ +' as opposed to'+++ -' preLabel = np.array(dfSPY3['K=3']) upCount = keyMatch(preKey= '+++', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '+++', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP1.append(('K=3', temp)) ### Print print('### For k cosnsecutive "up days"') print('### for {} , the probability of the next day is a ''up day'' is {:.1f}%'.format(p[0], p[1]*100)) print() # For GOOGL ### For consecutive 'up days' tLabel = np.array(dfG003['TrueLabel']) ### Count the probability of '+ +' as opposed to '+ -' preLabel = np.array(dfG003['K=1']) upCount = keyMatch(preKey= '+', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '+', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP1.append(('K=1', temp)) ### Count the probability of '++ +' as opposed to '++ -' preLabel = np.array(dfG003['K=2']) upCount = keyMatch(preKey= '++', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '++', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP1.append(('K=2', temp)) ### Count the probability of '+++ +' as opposed to'+++ -' preLabel = np.array(dfG003['K=3']) upCount = keyMatch(preKey= '+++', posKey= '+', preLable= preLabel, posLable= tLabel) downCount = keyMatch(preKey= '+++', posKey= '-', preLable= preLabel, posLable= tLabel) temp = upCount / (downCount + upCount) kP1.append(('K=3', temp)) ### Print print('### For GOOGL, k cosnsecutive "up days"') for p in kP1: print('### for $\{\}$, the probability of the next day is a ''up day'' is $\{:.1f\}$ %'.format(p[0], p[1]*100)) print() ### For SPY, k cosnsecutive "down days" ### for K=1 , the probability of the next day is a up day is 58.7%### for K=2 , the probability of the next day is a up day is 58.2% ### for K=3 , the probability of the next day is a up day is 61.0% ### For GOOGL, k cosnsecutive "down days" ### for K=1 , the probability of the next day is a up day is 51.4% ### for K=2 , the probability of the next day is a up day is 47.1% ### for K=3 , the probability of the next day is a up day is 54.9% ### For k cosnsecutive "up days" ### for K=1 , the probability of the next day is a up day is 51.7%### for K=2 , the probability of the next day is a up day is 49.8%### for K=3 , the probability of the next day is a up day is 45.3% ### For GOOGL, k cosnsecutive "up days" ### for K=1 , the probability of the next day is a up day is 53.9% ### for K=2 , the probability of the next day is a up day is 53.7%### for K=3 , the probability of the next day is a up day is 53.9% Question2 In [9]: # Make a prediction dictionary based on the training data (first 3 years data) from itertools import permutations def possiblePreKey(w): # Creat possible preKey of w # Eg. if w = 2, then return ['-+', '++', '--', '+-'] pool = '+' * w + '-' * w preSet = list(set(permutations(pool, w))) for i in range(len(preSet)): preSet[i] = ''.join(preSet[i]) return preSet def predictModel(wValue, preKeyPool, preLabel, posLabel): # Provide the prediction outcome based on w# Eg. for w = 2, return {'-+':'+', '+-':'+', '++': '-', '--':'+'} w = wValuepoolList = possiblePreKey(w) modelDic = dict.fromkeys(poolList,'NA') for p in poolList: upCount = keyMatch(preKey=p, posKey='+', preLable=preLabel, posLable=posLabel) downCount = keyMatch(preKey=p, posKey='-', preLable=preLabel, posLable=posLabel) if upCount > downCount: modelDic[p] = '+'else: modelDic[p] = '-'return modelDic if name == " main ": # For SPY ## W = 2w = 2posLabel = np.array(dfSPY3['TrueLabel']) preKeyPool = possiblePreKey(w) preLabel = np.array(dfSPY3['K=2']) w2Model = predictModel(wValue= w, preKeyPool= preKeyPool, preLabel= preLabel= posLabel= posLabel ##W = 3w = 3posLabel = np.array(dfSPY3['TrueLabel']) preKeyPool = possiblePreKey(w) preLabel = np.array(dfSPY3['K=3']) w3Model = predictModel(wValue= w, preKeyPool= preKeyPool, preLabel= preLabel= posLabel= posLabel ##W = 4w = 4posLabel = np.array(dfSPY3['TrueLabel']) preKeyPool = possiblePreKey(w) preLabel = np.array(dfSPY3['K=4']) w4Model = predictModel(wValue= w, preKeyPool= preKeyPool, preLabel= preLabel= posLabel=) # For GOOGL ## W = 2w = 2 posLabel = np.array(dfG003['TrueLabel']) preKeyPool = possiblePreKey(w) preLabel = np.array(dfG003['K=2']) w2ModelG = predictModel(wValue= w, preKeyPool= preKeyPool, preLabel= preLabel= posLabel=) ##W = 3w = 3posLabel = np.array(dfG003['TrueLabel']) preKeyPool = possiblePreKey(w) preLabel = np.array(dfG003['K=3']) w3ModelG = predictModel(wValue= w, preKeyPool= preKeyPool, preLabel= preLabel= posLabel=) ##W = 4w = 3posLabel = np.array(dfG003['TrueLabel']) preKeyPool = possiblePreKey(w) preLabel = np.array(dfG003['K=4']) w4ModelG = predictModel(wValue= w, preKeyPool= preKeyPool, preLabel= preLabel= posLabel=) # Creat prediction lable based on W = 2,3,4def makePrediction(testData, wValue, trainModel): preDictLable = ['na' for i in range(wValue)] preKey = testData for pk in preKey[wValue:]: preDictLable.append(trainModel[pk]) return preDictLable if name == " main ": # For SPY ## W = 2w2PredictLabel = makePrediction(testData = np.array(dfSPY2['K=2']), wValue= 2, trainModel= w2Model) dfSPY2['w2PredictLabel'] = w2PredictLabel ## W = 3w3PredictLabel = makePrediction(testData= np.array(dfSPY2['K=3']), wValue= 3, trainModel= w3Model) dfSPY2['w3PredictLabel'] = w3PredictLabel w4PredictLabel = makePrediction(testData= np.array(dfSPY2['K=4']), wValue= 4, trainModel= w4Model) dfSPY2['w4PredictLabel'] = w4PredictLabel # For GOOGI ## W = 2w2PredictLabel = makePrediction(testData = np.array(dfG002['K=2']), wValue= 2, trainModel= w2Model) dfG002['w2PredictLabel'] = w2PredictLabel w3PredictLabel = makePrediction(testData= np.array(dfGOO2['K=3']), wValue= 3, trainModel= w3Model) dfGOO2['w3PredictLabel'] = w3PredictLabel w4PredictLabel = makePrediction(testData= np.array(dfGOO2['K=4']), wValue= 4, trainModel= w4Model) dfG002['w4PredictLabel'] = w4PredictLabel Question3 def ensemLabel(data): eLab = ['na' for i in range(4)] temp = data.loc[4:, ['w2PredictLabel', 'w3PredictLabel', 'w4PredictLabel']] for (a, b, c) in zip(temp.w2PredictLabel, temp.w3PredictLabel, temp.w4PredictLabel): value = (a == '+') + (b == '+') + (c == '+') if value >= 2: p = ' + 'else: p = ' - 'eLab.append(p) return eLab if name == " main ": # For SPY data = dfSPY2 eLab = ensemLabel(data) dfSPY2['ensembleLabel'] = eLab # For GOOGL data = dfG002eLab = ensemLabel(data) dfG002['ensembleLabel'] = eLab # Compute the accuracy of ensembleLabel # For SPY size = dfSPY2.shape[0] e = sum(dfSPY2['TrueLabel'] == dfSPY2['ensembleLabel']) / size print("### For SPY") print("### The accuracy of ensembleLable is {:.1f}%".format(e*100)) print() # For GOOGL size = dfSPY2.shape[0] e = sum(dfG002['TrueLabel'] == dfG002['ensembleLabel']) / size print("### For GOOGL") print("### The accuracy of ensembleLable is {:.1f}%".format(e*100)) ### For SPY ### The accuracy of ensembleLable is 57.9% ### For GOOGL ### The accuracy of ensembleLable is 53.2% # Prediction '-' labels ## For SPY size = sum(dfSPY2['TrueLabel'] == '-') w2 = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w2PredictLabel == '-')) / size w3 = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w3PredictLabel == '-')) / size w4 = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w4PredictLabel == '-')) / size eL = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.ensembleLabel == '-')) / size print("For SPY") print("The accuracy on predicting '-' of W = 2 is $\{:.1f\}$ %".format(w2*100)) print("The accuracy on predicting '-' of W = 3 is {:.1f}%".format(w3*100)) print("The accuracy on predicting '-' of W = 4 is $\{:.1f\}$ %".format(w4*100)) print("The accuracy on predicting '-' of ensembleLabel is {:.1f}%".format(eL*100)) print() ## GOOGL size = sum(dfGOO2['TrueLabel'] == '-') w2 = sum((dfG002.TrueLabel == '-') & (dfG002.w2PredictLabel == '-')) / size w3 = sum((dfG002.TrueLabel == '-') & (dfG002.w3PredictLabel == '-')) / size w4 = sum((dfG002.TrueLabel == '-') & (dfG002.w4PredictLabel == '-')) / size eL = sum((dfGOO2.TrueLabel == '-') & (dfGOO2.ensembleLabel == '-')) / size print("For GOOGL") print("The accuracy on predicting '-' of W = 2 is {:.1f}%".format(w2*100)) print("The accuracy on predicting '-' of W = 3 is $\{:.1f\}$ %".format(w3*100)) print("The accuracy on predicting '-' of W = 4 is $\{:.1f\}$ %".format(w4*100)) print("The accuracy on predicting '-' of ensembleLabel is {:.1f}%".format(eL*100)) print() # Prediction '+' labels ## For SPY size = sum(dfSPY2['TrueLabel'] == '+') w2 = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w2PredictLabel == '+')) / size w3 = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w3PredictLabel == '+')) / size w4 = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w4PredictLabel == '+')) / size eL = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.ensembleLabel == '+')) / size print("For SPY") print("The accuracy on predicting '+' of W = 2 is $\{:.1f\}$ %".format(w2*100)) print("The accuracy on predicting '+' of W = 3 is $\{:.1f\}$ %".format(w3*100)) print("The accuracy on predicting '+' of W = 4 is $\{:.1f\}\%$ ".format(w4*100)) print("The accuracy on predicting '+' of ensembleLabel is {:.1f}%".format(eL*100)) print() ## For GOOGL size = sum(dfG002['TrueLabel'] == '+') w2 = sum((dfG002.TrueLabel == '+') & (dfG002.w2PredictLabel == '+')) / size w3 = sum((dfG002.TrueLabel == '+') & (dfG002.w3PredictLabel == '+')) / size w4 = sum((dfG002.TrueLabel == '+') & (dfG002.w4PredictLabel == '+')) / size eL = sum((dfGOO2.TrueLabel == '+') & (dfGOO2.ensembleLabel == '+')) / size print("For GOO") print("The accuracy on predicting '+' of W = 2 is $\{:.1f\}$ %".format(w2*100)) print("The accuracy on predicting '+' of W = 3 is $\{:.1f\}$ %".format(w3*100)) print("The accuracy on predicting '+' of W = 4 is $\{:.1f\}\%$ ".format(w4*100)) print("The accuracy on predicting '+' of ensembleLabel is {:.1f}%".format(eL*100)) print() For SPY The accuracy on predicting '-' of W = 2 is 33.0% The accuracy on predicting '-' of W = 3 is 23.0%The accuracy on predicting '-' of W = 4 is 36.8% The accuracy on predicting '-' of ensembleLabel is 23.0% For GOOGL The accuracy on predicting '-' of W = 2 is 31.0% The accuracy on predicting '-' of W = 3 is 16.4%The accuracy on predicting '-' of W = 4 is 27.9%The accuracy on predicting '-' of ensembleLabel is 16.4%For SPY The accuracy on predicting '+' of W = 2 is 67.1%The accuracy on predicting '+' of W = 3 is 83.1% The accuracy on predicting '+' of W = 4 is 71.2% The accuracy on predicting '+' of ensembleLabel is 82.7% For GOO The accuracy on predicting '+' of W = 2 is 70.1%The accuracy on predicting '+' of W = 3 is 83.1% The accuracy on predicting '+' of W = 4 is 72.3% The accuracy on predicting '+' of ensembleLabel is 83.1% **Question 4** In [14]: # True Positive ## For SPY w2TPSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w2PredictLabel == '+')) w3TPSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w3PredictLabel == '+')) w4TPSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w4PredictLabel == '+')) eLTPSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.ensembleLabel == '+')) print("For SPY") print('True Positives for W = 2 is {}'.format(w2TPSPY)) print('True Positives for W = 3 is {}'.format(w3TPSPY)) print('True Positives for W = 4 is {}'.format(w4TPSPY)) print('True Positives for ensembleLabel is {}'.format(eLTPSPY)) print() ## For GOOGL w2TPG = sum((dfG002.TrueLabel == '+') & (dfG002.w2PredictLabel == '+')) w3TPG = sum((dfG002.TrueLabel == '+') & (dfG002.w3PredictLabel == '+')) w4TPG = sum((dfG002.TrueLabel == '+') & (dfG002.w4PredictLabel == '+')) eLTPG = sum((dfGOO2.TrueLabel == '+') & (dfGOO2.ensembleLabel == '+')) print("For GOOGL") print('True Positives for W = 2 is {}'.format(w2TPG)) print('True Positives for W = 3 is {}'.format(w3TPG)) print('True Positives for W = 4 is {}'.format(w4TPG)) print('True Positives for ensembleLabel is {}'.format(eLTPG)) print() For SPY True Positives for W = 2 is 198 True Positives for W = 3 is 245 True Positives for W = 4 is 210 True Positives for ensembleLabel is 244 For GOOGL True Positives for W = 2 is 195 True Positives for W = 3 is 231 True Positives for W = 4 is 201 True Positives for ensembleLabel is 231 # False Positive w2FPSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w2PredictLabel == '+')) w3FPSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w3PredictLabel == '+')) w4FPSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w4PredictLabel == '+')) eLFPSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.ensembleLabel == '+')) print("For SPY") print('False Positives for W = 2 is {}'.format(w2FPSPY)) print('False Positives for W = 3 is {}'.format(w3FPSPY)) print('False Positives for W = 4 is {}'.format(w4FPSPY)) print('False Positives for ensembleLabel is {}'.format(eLFPSPY)) print() ## For GOOGL w2FPG = sum((dfG002.TrueLabel == '-') & (dfG002.w2PredictLabel == '+')) w3FPG = sum((dfG002.TrueLabel == '-') & (dfG002.w3PredictLabel == '+')) w4FPG = sum((dfG002.TrueLabel == '-') & (dfG002.w4PredictLabel == '+')) eLFPG = sum((dfG002.TrueLabel == '-') & (dfG002.ensembleLabel == '+')) print("For GOOGL") print('False Positives for W = 2 is {}'.format(w2FPG)) print('False Positives for W = 3 is {}'.format(w3FPG)) print('False Positives for W = 4 is {}'.format(w4FPG)) print('False Positives for ensembleLabel is {}'.format(eLFPG)) print() For SPY False Positives for W = 2 is 139 False Positives for W = 3 is 160 False Positives for W = 4 is 131 False Positives for ensembleLabel is 160 For GOOGL False Positives for W = 2 is 155 False Positives for W = 3 is 188 False Positives for W = 4 is 161 False Positives for ensembleLabel is 187 # True Negetive w2TNSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w2PredictLabel == '-')) w3TNSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w3PredictLabel == '-')) w4TNSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.w4PredictLabel == '-')) eLTNSPY = sum((dfSPY2.TrueLabel == '-') & (dfSPY2.ensembleLabel == '-')) print("For SPY") print('True Negative for W = 2 is {}'.format(w2TNSPY)) print('True Negative for W = 3 is {}'.format(w3TNSPY)) print('True Negative for W = 4 is {}'.format(w4TNSPY)) print('True Negative for ensembleLabel is {}'.format(eLTNSPY)) print() ## For GOOGL w2TNG = sum((dfG002.TrueLabel == '-') & (dfG002.w2PredictLabel == '-')) w3TNG = sum((dfG002.TrueLabel == '-') & (dfG002.w3PredictLabel == '-')) w4TNG = sum((dfG002.TrueLabel == '-') & (dfG002.w4PredictLabel == '-')) eLTNG = sum((dfG002.TrueLabel == '-') & (dfG002.ensembleLabel == '-')) print("For GOOGL") print('True Negative for W = 2 is {}'.format(w2TNG)) print('True Negative for W = 3 is {}'.format(w3TNG)) print('True Negative for W = 4 is {}'.format(w4TNG)) print('True Negative for ensembleLabel is {}'.format(eLTNG)) For SPY True Negative for W = 2 is 69 True Negative for W = 3 is 48 True Negative for W = 4 is 77 True Negative for ensembleLabel is 48 For GOOGL True Negative for W = 2 is 70 True Negative for W = 3 is 37 True Negative for W = 4 is 63 True Negative for ensembleLabel is 37 # False Negative ## For SPY w2FNSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w2PredictLabel == '-')) w3FNSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w3PredictLabel == '-')) w4FNSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.w4PredictLabel == '-')) eLFNSPY = sum((dfSPY2.TrueLabel == '+') & (dfSPY2.ensembleLabel == '-')) print("For SPY") print('False Negative for W = 2 is {}'.format(w2FNSPY)) print('False Negative for W = 3 is {}'.format(w3FNSPY)) print('False Negative for W = 4 is {}'.format(w4FNSPY)) print('False Negative for ensembleLabel is {}'.format(eLFNSPY)) print() ## For GOOGL w2FNG = sum((dfGOO2.TrueLabel == '+') & (dfGOO2.w2PredictLabel == '-'))w3FNG = sum((dfG002.TrueLabel == '+') & (dfG002.w3PredictLabel == '-')) w4FNG = sum((dfGOO2.TrueLabel == '+') & (dfGOO2.w4PredictLabel == '-'))eLFNG = sum((dfG002.TrueLabel == '+') & (dfG002.ensembleLabel == '-'))print("For GOOGL") print('False Negative for W = 2 is {}'.format(w2FNG)) print('False Negative for W = 3 is {}'.format(w3FNG)) print('False Negative for W = 4 is {}'.format(w4FNG)) print('False Negative for ensembleLabel is {}'.format(eLFNG)) print() For SPY False Negative for W = 2 is 96 False Negative for W = 3 is 48 False Negative for W = 4 is 82 False Negative for ensembleLabel is 48 For GOOGL False Negative for W = 2 is 82 False Negative for W = 3 is 45 False Negative for W = 4 is 75 False Negative for ensembleLabel is 45 # TPR = TP/(TP+FN)## For SPY w2TPR SPY = w2TPSPY/(w2TPSPY + w2FNSPY) w3TPR_SPY = w3TPSPY/(w3TPSPY + w3FNSPY) $w4TPR_SPY = w4TPSPY/(w4TPSPY + w4FNSPY)$ eLTPR_SPY = eLTPSPY/(eLTPSPY + eLFNSPY) print("TPR for SPY") $print("w2 = {:.1f}%".format(w2TPR_SPY*100))$ print("w3 = {:.1f}%".format(w3TPR SPY*100)) $print("w4 = {:.1f}%".format(w4TPR_SPY*100))$ print("eL = {:.1f}%".format(eLTPR_SPY*100)) ## For GOOGL w2TPR G = w2TPG/(w2TPG + w2FNG)w3TPR G = w3TPG/(w3TPG + w3FNG) $w4TPR_G = w4TPG/(w4TPG + w4FNG)$ eLTPR_G = eLTPG/(eLTPG + eLFNG) print("TPR for GOOGL") $print("w2 = {:.1f}%".format(w2TPR_G*100))$ print("w3 = $\{:.1f\}$ %".format(w3TPR G*100)) $print("w4 = {:.1f}%".format(w4TPR_G*100))$ print("eL = {:.1f}%".format(eLTPR_G*100)) TPR for SPY w2 = 67.3% w3 = 83.6% w4 = 71.9% eL = 83.6%TPR for GOOGL w2 = 70.4%w3 = 83.7%w4 = 72.8% eL = 83.7%# TNR = TN/(TN+FP)## For SPY w2TNR_SPY = w2TNSPY/(w2TNSPY + w2FPSPY) w3TNR_SPY = w3TNSPY/(w3TNSPY + w3FPSPY) $w4TNR_SPY = w4TNSPY/(w4TNSPY + w4FPSPY)$ eLTNR SPY = eLTNSPY/(eLTNSPY + eLFPSPY) print("TPR for SPY") print("w2 = $\{:.1f\}$ %".format(w2TNR_SPY*100)) print("w3 = $\{:.1f\}$ %".format(w3TNR_SPY*100)) print(" $w4 = {:.1f}$ %".format($w4TNR_SPY*100$)) print("eL = {:.1f}%".format(eLTNR SPY*100)) print() ## For GOOGL $w2TNR_G = w2TNG/(w2TNG + w2FPG)$ w3TNR G = w3TNG/(w3TNG + w3FPG)w4TNR G = w4TNG/(w4TNG + w4FPG)eLTNR_G = eLTNG/(eLTNG + eLFPG) print("TPR for SPY") print(" $w2 = {:.1f}$ %".format($w2TNR_G*100$)) $print("w3 = {:.1f}%".format(w3TNR_G*100))$ $print("w4 = {:.1f}%".format(w4TNR_G*100))$ $print("eL = {:.1f}%".format(eLTNR_G*100))$ TPR for SPY w2 = 33.2%w3 = 23.1%w4 = 37.0%eL = 23.1%TPR for SPY w2 = 31.1% w3 = 16.4%w4 = 28.1% eL = 16.5%