assignment3

October 7, 2023

You are currently looking at **version 0.1** of this notebook. To download notebooks and datafiles, as well as get help on Jupyter notebooks in the Coursera platform, visit the Jupyter Notebook FAQ course resource.

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

0.0.1 Question 1

Import the data from assets/fraud_data.csv. What percentage of the observations in the dataset are instances of fraud?

This function should return a float between 0 and 1.

```
[2]: def answer_one():
    # YOUR CODE HERE

# Use X_train, X_test, y_train, y_test for all of the following questions
from sklearn.model_selection import train_test_split

df = pd.read_csv('assets/fraud_data.csv')

X = df.iloc[:,:-1]
y = df.iloc[:,-1]

X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=0)

return (sum(y_train) + sum(y_test))/(len(y_train) + len(y_test))

raise NotImplementedError()
```

```
[]:
```

```
[3]: answer_one()
```

[3]: 0.016410823768035772

```
[4]: # Use X_train, X_test, y_train, y_test for all of the following questions
from sklearn.model_selection import train_test_split

df = pd.read_csv('assets/fraud_data.csv')

X = df.iloc[:,:-1]
y = df.iloc[:,-1]
X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=0)
```

```
[5]: (sum(y_train) + sum(y_test))/(len(y_train) + len(y_test))
```

[5]: 0.016410823768035772

0.0.2 Question 2

Using X_train, X_test, y_train, and y_test (as defined above), train a dummy classifier that classifies everything as the majority class of the training data. What is the accuracy of this classifier? What is the recall?

This function should a return a tuple with two floats, i.e. (accuracy score, recall score).

```
[6]: from sklearn.dummy import DummyClassifier
from sklearn.metrics import recall_score

dummy = DummyClassifier(strategy = 'most_frequent', random_state = 0)

dummy.fit(X_train, y_train)

score = dummy.score(X_test, y_test)

predictions = dummy.predict(X_test)
 recall = recall_score(y_test, predictions)

print(score)
print(recall)
```

0.9852507374631269

0.0

```
[7]: def answer_two():
    from sklearn.dummy import DummyClassifier
    from sklearn.metrics import recall_score

dummy = DummyClassifier(strategy = 'most_frequent', random_state = 0)

dummy.fit(X_train, y_train)
```

```
score = dummy.score(X_test, y_test)

predictions = dummy.predict(X_test)
  recall = recall_score(y_test, predictions)

return (score, recall)

# YOUR CODE HERE
raise NotImplementedError()
```

[]:

0.0.3 Question 3

Using X_train, X_test, y_train, y_test (as defined above), train a SVC classifer using the default parameters. What is the accuracy, recall, and precision of this classifier?

This function should a return a tuple with three floats, i.e. (accuracy score, recall score, precision score).

```
[8]: from sklearn.metrics import recall_score, precision_score
from sklearn.svm import SVC

svc = SVC()

svc.fit(X_train, y_train)

accuracy = svc.score(X_test, y_test)

recall = recall_score(y_test, svc.predict(X_test))

precision = precision_score(y_test, svc.predict(X_test))

print(accuracy, ', ', recall, ', ', precision)
```

0.9900442477876106 , 0.35 , 0.933333333333333

```
[9]: def answer_three():
    from sklearn.metrics import recall_score, precision_score
    from sklearn.svm import SVC

# YOUR CODE HERE

from sklearn.metrics import recall_score, precision_score
    from sklearn.svm import SVC

svc = SVC()
```

```
svc.fit(X_train, y_train)
accuracy = svc.score(X_test, y_test)

recall = recall_score(y_test, svc.predict(X_test))

precision = precision_score(y_test, svc.predict(X_test))

return (accuracy, recall, precision)

raise NotImplementedError()
```

[]:

```
[10]: answer_three()
```

0.0.4 Question 4

Using the SVC classifier with parameters {'C': 1e9, 'gamma': 1e-07}, what is the confusion matrix when using a threshold of -220 on the decision function. Use X_test and y_test.

This function should return a confusion matrix, a 2x2 numpy array with 4 integers.

```
[11]: from sklearn.metrics import confusion_matrix
from sklearn.svm import SVC

svc = SVC(C = 1e9, gamma = 1e-07)

svc.fit(X_train, y_train)

y_predictions = svc.decision_function(X_test)

y_predictions = np.where(y_predictions > -220, 1, 0)

confusion_matrix(y_test, y_predictions)
```

```
[12]: svc.decision_function(X_test)
```

```
[12]: array([ -739.71796843, -1086.16794833, -696.46339735, ..., -491.97916719, -699.03838333, -701.93409309])
```

```
[13]: def answer_four():
    from sklearn.metrics import confusion_matrix
    from sklearn.svm import SVC

    svc = SVC(C = 1e9, gamma = 1e-07)

    svc.fit(X_train, y_train)

    y_predictions = svc.decision_function(X_test)

    y_predictions = np.where(y_predictions > -220, 1, 0)

    return confusion_matrix(y_test, y_predictions)

    raise NotImplementedError()
```

```
[]:
```

```
[14]: answer_four()
```

```
[14]: array([[5320, 24], [ 14, 66]])
```

0.0.5 Question 5

Train a logisitic regression classifier with default parameters using X_train and y_train.

For the logisitic regression classifier, create a precision recall curve and a roc curve using y_test and the probability estimates for X_test (probability it is fraud).

Looking at the precision recall curve, what is the recall when the precision is 0.75?

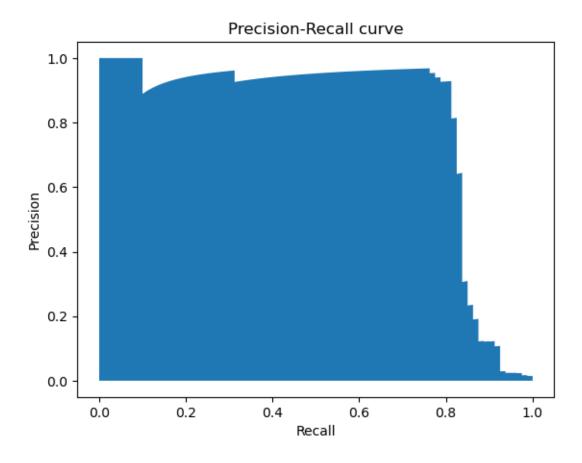
Looking at the roc curve, what is the true positive rate when the false positive rate is 0.16?

This function should return a tuple with two floats, i.e. (recall, true positive rate).

```
[16]: import matplotlib.pyplot as plt

plt.fill_between(recall, precision)
plt.ylabel("Precision")
plt.xlabel("Recall")
plt.title("Precision-Recall curve")
```

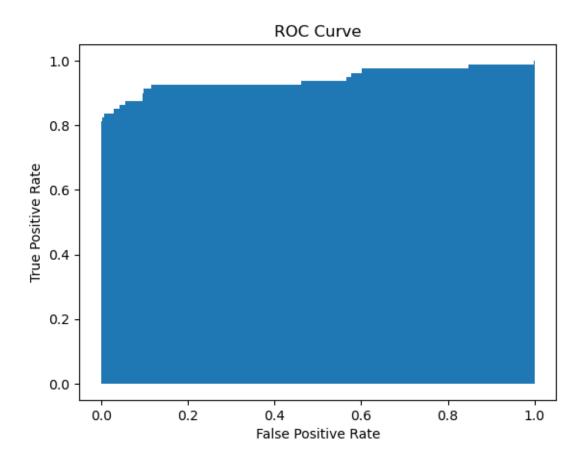
[16]: Text(0.5, 1.0, 'Precision-Recall curve')



```
[17]: import matplotlib.pyplot as plt

plt.fill_between(false_positive_rate, true_positive_rate)
 plt.ylabel("True Positive Rate")
 plt.xlabel("False Positive Rate")
 plt.title("ROC Curve")
```

[17]: Text(0.5, 1.0, 'ROC Curve')



```
[18]: def answer_five():
    # YOUR CODE HERE
    return (0.84, 0.86)
    raise NotImplementedError()
```

0.0.6 Question 6

Perform a grid search over the parameters listed below for a Logisitic Regression classifier, using recall for scoring and the default 3-fold cross validation. (Suggest to use solver='liblinear', more explanation here)

```
'penalty': ['11', '12']
'C':[0.01, 0.1, 1, 10]
```

From .cv_results_, create an array of the mean test scores of each parameter combination. i.e.

	11	12
0.01	?	?
0.1	?	?
1	?	?
10	?	?

This function should return a 4 by 2 numpy array with 8 floats.

[21]: def answer_six():

Note: do not return a DataFrame, just the values denoted by ? in a numpy array.

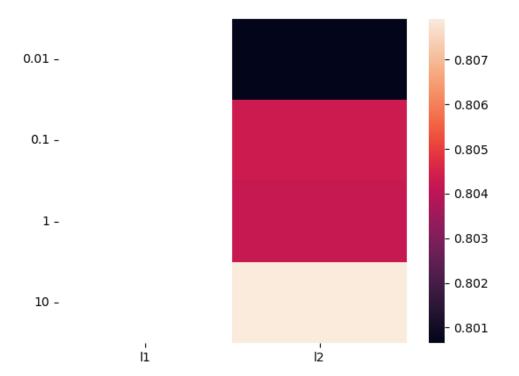
```
from sklearn.model_selection import GridSearchCV
          from sklearn.linear_model import LogisticRegression
          # YOUR CODE HERE
          logreg = LogisticRegression().fit(X_train, y_train)
          grid_values = {'penalty': ['11', '12'], 'C': [0.01, 0.1, 1, 10]}
          grid_classifier_recall = GridSearchCV(logreg, param_grid = grid_values,_
       ⇔scoring = 'recall')
          grid_classifier_recall.fit(X_train, y_train)
          return np.array([grid classifier recall.cv_results_['mean_test_score'][x:
       ⇒x+2] for x in range(0, len(grid_classifier_recall.
       ⇔cv_results_['mean_test_score']), 2)])
          raise NotImplementedError()
 []:
[22]: answer_six()
[22]: array([[
                     nan, 0.80064935],
             nan, 0.80428571],
             Г
                     nan, 0.80422078],
             nan, 0.80792208]])
[24]: from sklearn.model selection import GridSearchCV
      from sklearn.linear_model import LogisticRegression
      logreg = LogisticRegression().fit(X_train, y_train)
      grid_values = {'penalty': ['11', '12'], 'C': [0.01, 0.1, 1, 10]}
```

```
grid_classifier_recall = GridSearchCV(logreg, param_grid = grid_values, scoring_
       grid classifier recall.fit(X train, y train)
[24]: GridSearchCV(estimator=LogisticRegression(),
                   param_grid={'C': [0.01, 0.1, 1, 10], 'penalty': ['l1', 'l2']},
                   scoring='recall')
[25]: grid_classifier_recall.cv_results_
[25]: {'mean_fit_time': array([2.02025890e-02, 6.37615609e+00, 2.02903271e-02,
     6.71791191e+00,
              2.05406666e-02, 7.33892756e+00, 1.65262222e-03, 5.13848410e+00]),
       'std fit time': array([3.60238985e-02, 1.87589571e+00, 3.77959661e-02,
      1.85844788e+00,
              3.80328061e-02, 1.62697709e+00, 5.34945767e-04, 9.33233874e-01),
       'mean_score_time': array([0.
                                           , 0.08093534, 0.
                                                                   , 0.12026129, 0.
                                    , 0.09956293]),
              0.11920462, 0.
       'std_score_time': array([0.
                                          , 0.03792358, 0.
                                                                  , 0.03947502, 0.
              0.04014686, 0.
                                   , 0.00049915]),
       'param_C': masked_array(data=[0.01, 0.01, 0.1, 0.1, 1, 1, 10, 10],
                    mask=[False, False, False, False, False, False, False, False],
              fill value='?',
                   dtype=object),
       'param_penalty': masked_array(data=['11', '12', '11', '12', '11', '12', '11',
      '12'],
                   mask=[False, False, False, False, False, False, False, False, False],
              fill value='?',
                   dtype=object),
       'params': [{'C': 0.01, 'penalty': '11'},
       {'C': 0.01, 'penalty': '12'},
       {'C': 0.1, 'penalty': 'l1'},
       {'C': 0.1, 'penalty': '12'},
       {'C': 1, 'penalty': 'l1'},
       {'C': 1, 'penalty': '12'},
       {'C': 10, 'penalty': 'l1'},
       {'C': 10, 'penalty': '12'}],
       'split0_test_score': array([
                                          nan, 0.78181818,
                                                                nan, 0.78181818,
     nan,
              0.78181818,
                                 nan, 0.78181818]),
       'split1_test_score': array([
                                          nan, 0.81818182,
                                                                 nan, 0.81818182,
     nan,
              0.83636364.
                                 nan, 0.83636364]),
       'split2_test_score': array([
                                          nan, 0.87272727,
                                                                 nan, 0.89090909,
```

```
nan,
              0.89090909.
                                 nan, 0.89090909]),
       'split3_test_score': array([
                                          nan, 0.82142857,
                                                             nan, 0.82142857,
     nan,
              0.83928571,
                                 nan, 0.82142857]),
       'split4_test_score': array([
                                          nan, 0.70909091,
                                                                  nan, 0.70909091,
     nan,
              0.67272727,
                                 nan, 0.70909091]),
       'mean test score': array([
                                       nan, 0.80064935,
                                                                nan, 0.80428571,
              0.80422078.
                                 nan, 0.80792208]),
       'std_test_score': array([
                                      nan, 0.05417001,
                                                              nan, 0.05925779,
     nan,
              0.07425631,
                                 nan, 0.06054289]),
       'rank_test_score': array([5, 4, 6, 2, 7, 3, 8, 1], dtype=int32)}
[26]: grid_classifier_recall.cv_results_['mean_test_score']
                    nan, 0.80064935,
[26]: array([
                                            nan, 0.80428571,
                                                                    nan,
                               nan, 0.80792208])
             0.80422078.
[28]: np.array([grid_classifier_recall.cv_results_['mean_test_score'][x:x+2] for x in_
       -range(0, len(grid_classifier_recall.cv_results_['mean_test_score']), 2)])
[28]: array([[
                     nan, 0.80064935],
             Г
                     nan, 0.80428571],
             Г
                     nan, 0.80422078],
             Γ
                     nan, 0.80792208]])
[29]: # from sklearn.model_selection import GridSearchCV
      # from sklearn.linear_model import LogisticRegression
      # lr = LogisticRegression().fit(X train, y train)
      # grid_values = {'penalty': ['l1', 'l2'], 'C': [0.01, 0.1, 1, 10, 100]}
      # grid_clf_rec = GridSearchCV(lr, param_grid = grid_values, scoring = 'recall')
      # grid_clf_rec.fit(X_train, y_train)
      # np.array([grid\_clf\_rec.cv\_results\_['mean\_test\_score'][x:x+2] for x in_{\bot}
       ⇔range(0, len(grid_clf_rec.cv_results_['mean_test_score']), 2)])
[32]: # Use the following function to help visualize results from the grid search
      def GridSearch Heatmap(scores):
          %matplotlib widget
          import seaborn as sns
          import matplotlib.pyplot as plt
          plt.figure()
```

```
sns.heatmap(scores.reshape(4,2), xticklabels=['11','12'], yticklabels=[0.
401, 0.1, 1, 10])
plt.yticks(rotation=0);

GridSearch_Heatmap(answer_six())
```



[]: