```
1.1.1
In [44]:
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          ECGR 4105
          Homework 2
          Problem 4
          '\nPatrick Ballou\nID: 801130521\nECGR 4105\nHomework 2\nProblem 4\n'
Out[44]:
          import numpy as np
In [45]:
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
          from sklearn.model selection import KFold
          from sklearn.linear model import LogisticRegression
          from sklearn import metrics
          from sklearn.model_selection import cross_val_score
          from sklearn.model selection import train test split
          from sklearn.datasets import load breast cancer
          from sklearn.preprocessing import MinMaxScaler, StandardScaler
In [46]:
          breast = load_breast_cancer()
          breast data = breast.data
          breast data.shape
          breast input = pd.DataFrame(breast data)
In [47]:
          breast_labels = breast.target
          breast_labels.shape
          labels = np.reshape(breast labels,(569,1))
          final breast data = np.concatenate([breast data, labels],axis=1)
          final_breast_data.shape
          (569, 31)
Out[47]:
          breast dataset = pd.DataFrame(final breast data)
In [48]:
          features = breast.feature names
          features_labels = np.append(features, 'label')
          breast dataset.columns = features labels
          breast dataset.head()
Out[48]:
                                                                                 mean
             mean
                                                  mean
                                                              mean
                                                                        mean
                     mean
                               mean
                                      mean
                                                                                           mean
                                                                               concave
                                                                                       symmetry
             radius texture perimeter
                                       area smoothness compactness concavity
                                                                                points
                                                                                                  dim
             17.99
                               122.80 1001.0
          0
                      10.38
                                                0.11840
                                                             0.27760
                                                                        0.3001
                                                                               0.14710
                                                                                           0.2419
                                                                                                    (
              20.57
          1
                      17.77
                               132.90 1326.0
                                                0.08474
                                                             0.07864
                                                                        0.0869
                                                                               0.07017
                                                                                           0.1812
          2
             19.69
                               130.00 1203.0
                                                0.10960
                                                             0.15990
                                                                        0.1974
                                                                               0.12790
                                                                                           0.2069
                     21.25
```

20.29 14.34 135.10 1297.0 5 rows × 31 columns

77.58

386.1

0.14250

0.10030

0.28390

0.13280

0.2414

0.1980

0.10520

0.10430

0.2597

0.1809

11.42

20.38

3

(

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In [49]: x = breast_dataset[features]
         Y = breast dataset['label']
In [50]: #standard scaler is better here too
         scaler = StandardScaler()
         #scaler = MinMaxScaler()
         X = scaler.fit transform(x)
         \#k-fold with k = 5 and k = 10
In [51]:
          kfold1 = KFold(n splits=5, random state=7, shuffle=True)
          kfold2 = KFold(n_splits=10, random_state=7, shuffle=True)
         model = LogisticRegression()
          results 5 = cross val score(model, X, Y, cv=kfold1)
          results 10 = cross val score(model, X, Y, cv=kfold2)
         #basically the same accuracy
In [52]:
          print("Accuracy for K = 5: %.3f% (%.3f%)" % (results_5.mean()*100, results_5.std()*1
         print("Accuracy for K = 10: %.3f%% (%.3f%%)" % (results 10.mean()*100, results 10.std()
         Accuracy for K = 5: 97.539% (1.024%)
         Accuracy for K = 10: 97.716\% (1.578\%)
         #4b: add penalty
In [53]:
          kfold1 = KFold(n splits=5, random state=7, shuffle=True)
          kfold2 = KFold(n splits=10, random state=7, shuffle=True)
         model = LogisticRegression(penalty='12')
          results 5 = cross val score(model, X, Y, cv=kfold1)
          results_10 = cross_val_score(model, X, Y, cv=kfold2)
         #very similar results
In [54]:
         print("Accuracy for K = 5: %.3f%% (%.3f%%)" % (results 5.mean()*100, results 5.std()*1
         print("Accuracy for K = 10: %.3f%% (%.3f%%)" % (results 10.mean()*100, results 10.std()
         Accuracy for K = 5: 97.539% (1.024%)
         Accuracy for K = 10: 97.716\% (1.578\%)
```