lab1

September 9, 2020

```
[1]: import numpy as np
   import pandas as pd
   from matplotlib import pyplot as plt
   from sklearn.datasets import load_breast_cancer
   from sklearn.metrics import confusion_matrix
   from sklearn.naive_bayes import GaussianNB
   from sklearn.model_selection import train_test_split
   from sklearn.metrics import accuracy_score
   from sklearn.metrics import classification_report

import seaborn as sns
   sns.set()
   breast_cancer = load_breast_cancer()

X = pd.DataFrame(breast_cancer.data, columns=breast_cancer.feature_names)
```

[2]: print (X)

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	\
0	17.99	10.38	122.80	1001.0	0.11840	
1	20.57	17.77	132.90	1326.0	0.08474	
2	19.69	21.25	130.00	1203.0	0.10960	
3	11.42	20.38	77.58	386.1	0.14250	
4	20.29	14.34	135.10	1297.0	0.10030	
	•••	•••	•••	•••	***	
564	21.56	22.39	142.00	1479.0	0.11100	
565	20.13	28.25	131.20	1261.0	0.09780	
566	16.60	28.08	108.30	858.1	0.08455	
567	20.60	29.33	140.10	1265.0	0.11780	
568	7.76	24.54	47.92	181.0	0.05263	
	mean compactn	ess mean con	cavity mean com	ncave points	mean symmetry	\
0	0.27	760 0	.30010	0.14710	0.2419	
1	0.07	7864 0	.08690	0.07017	0.1812	
2	0.15	5990 0	.19740	0.12790	0.2069	
3	0.28	390 0	.24140	0.10520	0.2597	
4	0.13	3280 0	.19800	0.10430	0.1809	
		•••		•••	•••	

```
564
               0.11590
                                0.24390
                                                      0.13890
                                                                        0.1726
565
               0.10340
                                0.14400
                                                      0.09791
                                                                        0.1752
                                0.09251
                                                                        0.1590
566
               0.10230
                                                      0.05302
567
               0.27700
                                0.35140
                                                      0.15200
                                                                        0.2397
               0.04362
                                0.00000
                                                      0.00000
568
                                                                        0.1587
     mean fractal dimension ... worst radius worst texture \
                     0.07871
                                         25.380
                                                          17.33
0
1
                     0.05667
                                         24.990
                                                          23.41
2
                     0.05999
                                                          25.53
                                         23.570
3
                     0.09744 ...
                                                          26.50
                                         14.910
4
                     0.05883 ...
                                         22.540
                                                          16.67
. .
                                         •••
564
                     0.05623 ...
                                         25.450
                                                          26.40
565
                     0.05533 ...
                                                          38.25
                                         23.690
                                                          34.12
566
                     0.05648 ...
                                         18.980
567
                     0.07016
                                         25.740
                                                          39.42
                     0.05884 ...
                                                          30.37
568
                                          9.456
     worst perimeter worst area worst smoothness
                                                      worst compactness
                                                                  0.66560
0
               184.60
                            2019.0
                                              0.16220
1
               158.80
                           1956.0
                                              0.12380
                                                                  0.18660
2
               152.50
                           1709.0
                                              0.14440
                                                                  0.42450
3
                98.87
                            567.7
                                              0.20980
                                                                  0.86630
4
               152.20
                           1575.0
                                              0.13740
                                                                  0.20500
564
               166.10
                            2027.0
                                              0.14100
                                                                  0.21130
565
                                                                  0.19220
               155.00
                           1731.0
                                              0.11660
566
               126.70
                           1124.0
                                              0.11390
                                                                  0.30940
567
               184.60
                            1821.0
                                              0.16500
                                                                  0.86810
568
                59.16
                             268.6
                                              0.08996
                                                                  0.06444
     worst concavity worst concave points worst symmetry \
0
               0.7119
                                      0.2654
                                                       0.4601
1
               0.2416
                                      0.1860
                                                       0.2750
2
               0.4504
                                      0.2430
                                                       0.3613
3
               0.6869
                                      0.2575
                                                       0.6638
4
               0.4000
                                      0.1625
                                                       0.2364
. .
                  •••
                                       •••
                                                       0.2060
564
               0.4107
                                      0.2216
565
               0.3215
                                      0.1628
                                                       0.2572
566
               0.3403
                                      0.1418
                                                       0.2218
567
               0.9387
                                      0.2650
                                                       0.4087
568
               0.0000
                                      0.0000
                                                       0.2871
     worst fractal dimension
0
                      0.11890
1
                      0.08902
```

```
3
                          0.17300
    4
                          0.07678
    . .
                               •••
    564
                          0.07115
    565
                          0.06637
    566
                          0.07820
    567
                          0.12400
    568
                          0.07039
    [569 rows x 30 columns]
    dir(breast_cancer)
[3]:
[3]: ['DESCR',
      'data',
      'feature_names',
      'filename',
      'frame',
      'target',
      'target_names']
[4]: y=pd.Categorical.from_codes(breast_cancer.target,breast_cancer.target_names)
     print(y)
     [malignant, malignant, malignant, malignant, malignant, ..., malignant,
    malignant, malignant, malignant, benign]
    Length: 569
    Categories (2, object): [malignant, benign]
[5]: X.describe()
[5]:
            mean radius
                                         mean perimeter
                                                            mean area
                          mean texture
             569.000000
                                             569.000000
     count
                            569.000000
                                                           569.000000
              14.127292
                             19.289649
                                              91.969033
                                                           654.889104
     mean
     std
               3.524049
                              4.301036
                                              24.298981
                                                           351.914129
     min
               6.981000
                              9.710000
                                              43.790000
                                                           143.500000
     25%
                                              75.170000
              11.700000
                             16.170000
                                                           420.300000
     50%
                             18.840000
                                              86.240000
                                                           551.100000
              13.370000
     75%
              15.780000
                                                           782.700000
                             21.800000
                                             104.100000
     max
              28.110000
                             39.280000
                                             188.500000
                                                          2501.000000
            mean smoothness
                              mean compactness
                                                 mean concavity
                                                                  mean concave points
                                                                            569.000000
     count
                  569.000000
                                     569.000000
                                                      569.000000
                                       0.104341
                                                                              0.048919
     mean
                    0.096360
                                                        0.088799
                    0.014064
                                       0.052813
                                                        0.079720
                                                                              0.038803
     std
                                                                              0.00000
     min
                    0.052630
                                       0.019380
                                                        0.000000
     25%
                    0.086370
                                       0.064920
                                                        0.029560
                                                                              0.020310
```

2

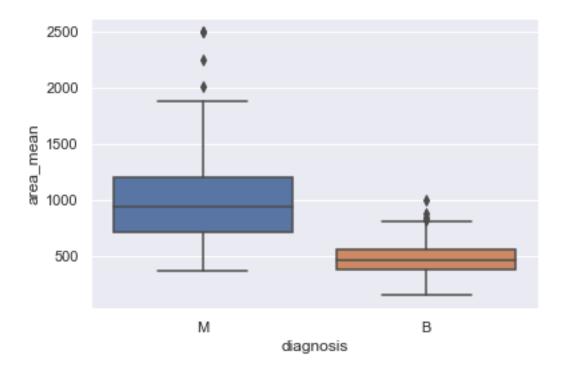
0.08758

50% 75% max	0.095870 0.105300 0.163400	0.092630 0.130400 0.345400	0.061540 0.130700 0.426800	0.033500 0.074000 0.201200
count mean std min 25% 50% 75% max	mean symmetry mean 569.000000 0.181162 0.027414 0.106000 0.161900 0.179200 0.195700 0.304000	fractal dimension 569.000000 0.062798 0.007060 0.049960 0.057700 0.061540 0.066120 0.097440	worst radius \ 569.000000 16.269190 4.833242 7.930000 13.010000 14.970000 18.790000 36.040000	
count mean std min 25% 50% 75% max	worst texture worst 569.000000 25.677223 6.146258 12.020000 21.080000 25.410000 29.720000 49.540000	t perimeter worst 569.000000 569.00 107.261213 880.58 33.602542 569.38 50.410000 185.20 84.110000 515.30 97.660000 686.50 125.400000 1084.00 251.200000 4254.00	00000 569.00000 83128 0.132369 56993 0.022832 00000 0.071170 00000 0.116600 00000 0.131300 00000 0.146000)))))
count mean std min 25% 50% 75% max	worst compactness 569.000000 0.254265 0.157336 0.027290 0.147200 0.211900 0.339100 1.058000	worst concavity worst 569.000000 0.272188 0.208624 0.000000 0.114500 0.226700 0.382900 1.252000	569.000000 0.114606 0.065732 0.000000 0.064930 0.099930 0.161400 0.291000	
count mean std min 25% 50% 75% max	worst symmetry worst 569.000000 0.290076 0.061867 0.156500 0.250400 0.282200 0.317900 0.663800	569.000000 0.083946 0.01806: 0.055046 0.071466 0.080046 0.092086 0.207506	0 6 1 0 0 0	

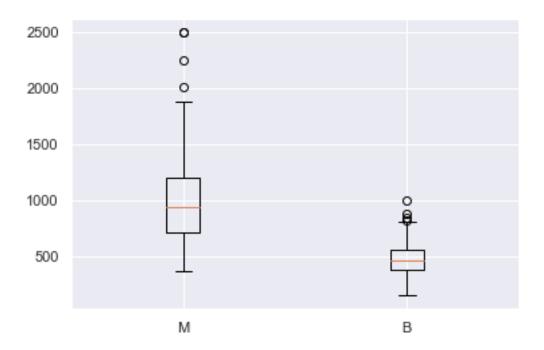
[8 rows x 30 columns]

```
[6]: #We will do this using SciKit-Learn library in Python using the
      \hookrightarrow train\_test\_split method.
      from sklearn.model_selection import train_test_split
      X_train, X_test, Y_train, Y_test = train_test_split(X, y, test_size = 0.25, __
      →random_state = 0)
      #Feature Scaling to bring attribute to one range (say 0-100 or 0-1)
      from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      X_train = sc.fit_transform(X_train)
      X_test = sc.transform(X_test)
 [7]: print(X_train)
      print(X test)
      \begin{bmatrix} [-0.65079907 & -0.43057322 & -0.68024847 & \dots & -0.36433881 & 0.32349851 \end{bmatrix} 
       -0.7578486 ]
      [-0.82835341 0.15226547 -0.82773762 ... -1.45036679 0.62563098
       -1.03071387]
      -0.96601386]
      [-1.33114223 -0.22172269 -1.3242844 \dots -0.98806491 -0.69995543
       -0.12266325]
      [-1.25110186 -0.24600763 -1.28700242 ... -1.75887319 -1.56206114
       -1.00989735]
      0.2126516 ]]
      \begin{bmatrix} [-0.21395901 & 0.3125461 & -0.14355187 \ ... & 1.37043754 & 1.08911166 \end{bmatrix} 
        1.53928319]
      [-0.26750714 \quad 1.461224 \quad -0.32955207 \dots \quad -0.84266106 \quad -0.71577388
       -0.88105993]
      [-0.03922298 -0.86770223 -0.10463112 ... -0.505318 -1.20298225
       -0.92494342
       \begin{bmatrix} -0.51270124 & -1.69096186 & -0.54095317 & \dots & -0.12632201 & 0.33773512 \end{bmatrix} 
       -0.428722441
      [-0.17732081 -2.01395163 -0.17345939 ... -0.62875108 -0.29500302
       -0.65432858]
      [ 1.5305829 -0.26300709 1.57961296 ... 1.6694843
                                                            1.18085869
        0.48889253]]
[16]: print("Cancer data set dimensions : {}".format(X.shape,y.shape))
     Cancer data set dimensions: (569, 30)
[17]: df=pd.read_csv(r"D:\msc3\machine learning\lab1\data.csv")
      sns.boxplot(x='diagnosis', y='area_mean', data=df)
```

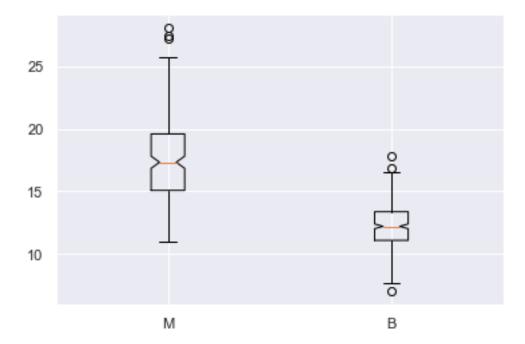
[17]: <matplotlib.axes._subplots.AxesSubplot at 0x23766372388>



```
[18]: malignant = df[df['diagnosis']=='M']['area_mean']
      benign = df[df['diagnosis']=='B']['area_mean']
      fig = plt.figure()
      ax = fig.add_subplot(111)
      ax.boxplot([malignant,benign], labels=['M', 'B'])
[18]: {'whiskers': [<matplotlib.lines.Line2D at 0x23771870288>,
        <matplotlib.lines.Line2D at 0x23771884a08>,
        <matplotlib.lines.Line2D at 0x237718918c8>,
        <matplotlib.lines.Line2D at 0x2377188a848>],
       'caps': [<matplotlib.lines.Line2D at 0x23771884ec8>,
        <matplotlib.lines.Line2D at 0x23771884bc8>,
        <matplotlib.lines.Line2D at 0x23771891f48>,
        <matplotlib.lines.Line2D at 0x2377189a908>],
       'boxes': [<matplotlib.lines.Line2D at 0x23771879fc8>,
        <matplotlib.lines.Line2D at 0x2377188aec8>],
       'medians': [<matplotlib.lines.Line2D at 0x2377188a8c8>,
        <matplotlib.lines.Line2D at 0x2377189ae88>],
       'fliers': [<matplotlib.lines.Line2D at 0x2377188ad88>,
        <matplotlib.lines.Line2D at 0x2377189ab88>],
       'means': []}
```

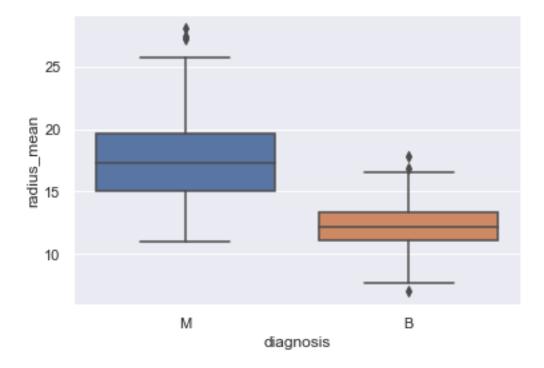


```
[19]: malignant = df[df['diagnosis']=='M']['radius_mean']
  benign = df[df['diagnosis']=='B']['radius_mean']
  fig = plt.figure()
  ax = fig.add_subplot(111)
  ax.boxplot([malignant,benign], notch = True, labels=['M', 'B']);
```

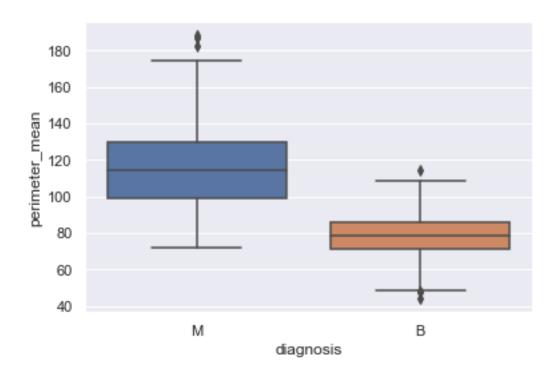


```
[20]: sns.boxplot(x='diagnosis', y='radius_mean', data=df)
```

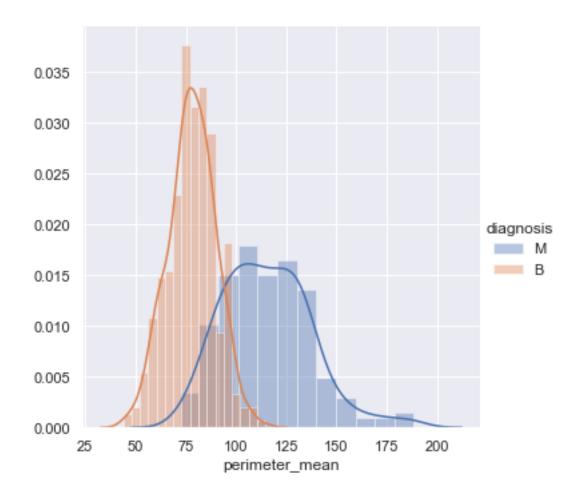
[20]: <matplotlib.axes._subplots.AxesSubplot at 0x23771926048>



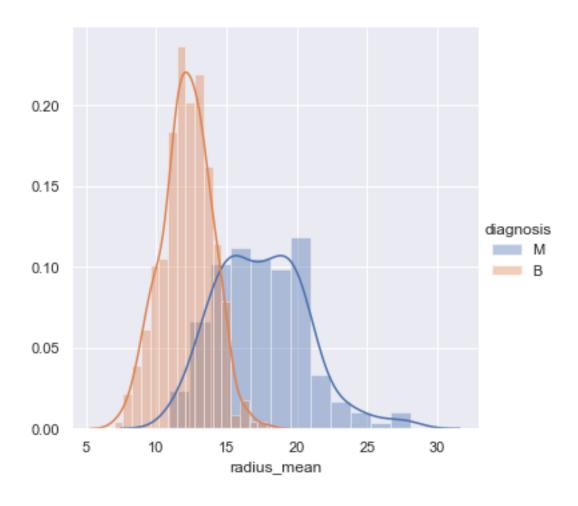
[21]: <matplotlib.axes._subplots.AxesSubplot at 0x23771997c88>



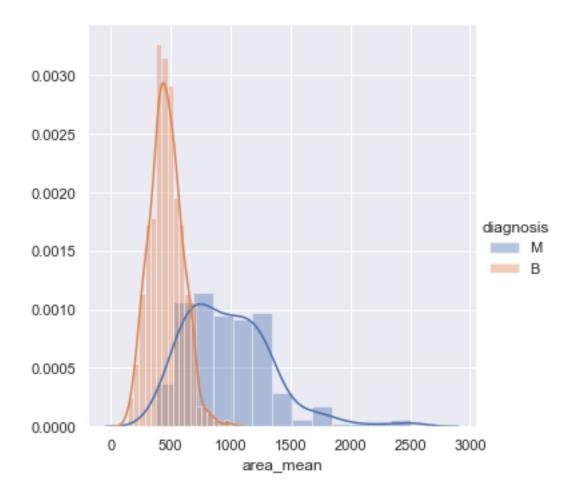
```
[22]: sns.FacetGrid(df, hue='diagnosis', height=5) \
    .map(sns.distplot, 'perimeter_mean') \
     .add_legend()
    plt.show()
```



```
[23]: sns.FacetGrid(df, hue='diagnosis', height=5) \
   .map(sns.distplot, 'radius_mean') \
   .add_legend()
  plt.show()
```



```
[24]: sns.FacetGrid(df, hue='diagnosis', height=5) \
   .map(sns.distplot, 'area_mean') \
   .add_legend()
  plt.show()
```



```
[25]: df.isnull().sum() df.isna().sum()
```

```
[25]: id
                                  0
      diagnosis
                                  0
      radius_mean
                                  0
      texture_mean
                                  0
      perimeter_mean
                                  0
      area_mean
                                  0
      smoothness_mean
                                  0
      compactness_mean
                                  0
      concavity_mean
                                  0
      concave points_mean
                                  0
      symmetry_mean
                                  0
      fractal_dimension_mean
                                  0
      radius_se
                                  0
      texture_se
                                  0
                                  0
      perimeter_se
```

area_se	0			
smoothness_se				
compactness_se				
concavity_se	0			
concave points_se	0			
symmetry_se	0			
fractal_dimension_se	0			
radius_worst	0			
texture_worst	0			
perimeter_worst	0			
area_worst	0			
smoothness_worst				
compactness_worst	0			
concavity_worst	0			
concave points_worst	0			
symmetry_worst				
fractal_dimension_worst				
dtype: int64				

[]:

[]: