Naive Bayes

Defining accuracy function

Defining sensitivity function

Defining specificity function

In [5]: 1 data

Out[5]:

	cap_shape	Cap_surface	bruises	category
0	Convex	Smooth	Bruises	Poisonous
1	Convex	Smooth	Bruises	Edible
2	Bell	Smooth	Bruises	Edible
3	Convex	Scaly	Bruises	Poisonous
4	Convex	Smooth	NoBruises	Edible
8119	Knobbed	Smooth	NoBruises	Edible
8120	Convex	Smooth	NoBruises	Edible
8121	Flat	Smooth	NoBruises	Edible
8122	Knobbed	Scaly	NoBruises	Poisonous
8123	Convex	Smooth	NoBruises	Edible

8124 rows × 4 columns

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)
self. update inplace(new data)

Defining probability function using m-estimate

```
In [ ]:
In [10]:
          1 y_train[1:10]
Out[10]: 3232
                 Negative
         6629
                 Negative
         4880
                 Negative
         3703
                 Negative
         7914
                 Negative
         1365
                 Positive
         5398
                 Negative
         5435
                 Negative
         1137
                 Positive
         Name: category, dtype: object
In [11]:
          1 print(y_train[y_train == 'Positive'].shape[0])
           print(y_train[y_train == 'Negative'].shape[0])
         2967
```

2719

```
In [12]: 1 X_train[X_train.bruises == 'Bruises'][X_train.Cap_surface== 'Smooth']
```

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:1: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

"""Entry point for launching an IPython kernel.

Out[12]:

	cap_shape	Cap_surface	bruises	category
5435	Convex	Smooth	Bruises	Negative
1739	Flat	Smooth	Bruises	Negative
5499	Convex	Smooth	Bruises	Negative
5550	Convex	Smooth	Bruises	Negative
5462	Convex	Smooth	Bruises	Negative
5740	Convex	Smooth	Bruises	Negative
2241	Flat	Smooth	Bruises	Negative
844	Convex	Smooth	Bruises	Positive
984	Convex	Smooth	Bruises	Negative
5657	Flat	Smooth	Bruises	Negative

611 rows × 4 columns

Initial probabillity of dataset(class)

```
In [13]:
             #--probability of positive and negative
             Prob_positive = probability(X_train[X_train.category == 'Positive'].shape[0],X_train.shape[0],2)
             Prob negative = probability(X train[X train.category == 'Negative'].shape[0],X train.shape[0],2)
            print(Prob positive)
             print(Prob negative)
         0.5218002812939522
         0.4781997187060478
 In [ ]:
In [14]:
          1 print(X train.cap shape.unique())#-----getting unique values of each feature
           2 print(X train.Cap surface.unique())
           3 print(X train.bruises.unique())
         ['Flat' 'Convex' 'Knobbed' 'Bell' 'Sunken' 'Conical']
         ['Scaly' 'Smooth' 'Fiberous' 'Grooves']
         ['Bruises' 'NoBruises']
```

```
In [15]:
              #----starting for feature 1 (cap shape)
              Prob convex pos = probability(X train[X train.cap shape == 'Convex'][X train.category == 'Positive'].shape[0],X trai
              Prob convex neg = probability(X train[X train.cap shape == 'Convex'][X train.category == 'Negative'].shape[0],X trai
              Prob Conical pos = probability(X train[X train.cap shape == 'Conical'][X train.category == 'Positive'].shape[0],X tr
              Prob Conical neg = probability(X train[X train.cap shape == 'Conical'][X train.category == 'Negative'].shape[0],X tr
              Prob Bell pos = probability(X train[X train.cap shape == 'Bell'][X train.category == 'Positive'].shape[0],X train[y
              Prob Bell neg = probability(X train[X train.cap shape == 'Bell'][X train.category == 'Negative'].shape[0],X train[y
          11
          12
          13
              Prob Flat pos = probability(X train[X train.cap shape == 'Flat'][X train.category == 'Positive'].shape[0],X train[y
              Prob Flat neg = probability(X train[X train.cap shape == 'Flat'][X train.category == 'Negative'].shape[0],X train[y
          15
          16
              Prob Knobbed pos = probability(X train[X train.cap shape == 'Knobbed'][X train.category == 'Positive'].shape[0],X tr
          17
              Prob Knobbed neg = probability(X train[X train.cap shape == 'Knobbed'][X train.category == 'Negative'].shape[0],X tr
          18
          19
              Prob Sunken pos = probability(X train[X train.cap shape == 'Sunken'][X train.category == 'Positive'].shape[0],X trai
              Prob Sunken neg = probability(X train[X train.cap shape == 'Sunken'][X train.category == 'Negative'].shape[0],X trai
          21
          22
          23
              print(Prob Sunken pos)
              print(Prob Sunken neg)
```

0.008409014463504878

0.0011009174311926607

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:2: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:3: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

This is separate from the ipykernel package so we can avoid doing imports until

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:6: UserWarning: Boolean S eries key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:7: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

import sys

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:10: UserWarning: Boolean Series key will be reindexed to match DataFrame index.

Remove the CWD from sys.path while we load stuff.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:11: UserWarning: Boolean
Series key will be reindexed to match DataFrame index.

This is added back by InteractiveShellApp.init_path()

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:13: UserWarning: Boolean
Series key will be reindexed to match DataFrame index.

del sys.path[0]

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:14: UserWarning: Boolean Series key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:17: UserWarning: Boolean
Series key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:18: UserWarning: Boolean Series key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:20: UserWarning: Boolean
Series key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:21: UserWarning: Boolean
Series key will be reindexed to match DataFrame index.

```
In [16]:
              #----starting for feature 2 (cap shape)
              Prob Fiberous pos = probability(X train[X train.Cap surface == 'Fiberous'][X train.category == 'Positive'].shape[0],
              Prob Fiberous neg = probability(X train[X train.Cap surface == 'Fiberous'][X train.category == 'Negative'].shape[0],
              Prob Smooth pos = probability(X train[X train.Cap surface == 'Smooth'][X train.category == 'Positive'].shape[0],X tr
              Prob Smooth neg = probability(X train[X train.Cap surface == 'Smooth'][X train.category == 'Negative'].shape[0],X tr
           9
              Prob Scaly pos = probability(X train[X train.Cap surface == 'Scaly'][X train.category == 'Positive'].shape[0],X trai
              Prob Scaly neg = probability(X train[X train.Cap surface == 'Scaly'][X train.category == 'Negative'].shape[0],X trai
          11
          12
          13
              Prob Grooves pos = probability(X train[X train.Cap surface == 'Grooves'][X train.category == 'Positive'].shape[0],X
              Prob Grooves neg = probability(X train[X train.Cap surface == 'Grooves'][X train.category == 'Negative'].shape[0],X
          15
          16
          17
          18
              print(Prob Fiberous pos)
              print(Prob Fiberous neg)
```

- 0.3779872096937058
- 0.1883951524054352

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:2: UserWarning: Boolea
n Series key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:3: UserWarning: Boolea
n Series key will be reindexed to match DataFrame index.

This is separate from the ipykernel package so we can avoid doing imports until

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:6: UserWarning: Boolea
n Series key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:7: UserWarning: Boolea
n Series key will be reindexed to match DataFrame index.
import sys

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:10: UserWarning: Boole
an Series key will be reindexed to match DataFrame index.

Remove the CWD from sys.path while we load stuff.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:11: UserWarning: Boole
an Series key will be reindexed to match DataFrame index.

```
# This is added back by InteractiveShellApp.init_path()
c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:13: UserWarning: Boole
an Series key will be reindexed to match DataFrame index.
  del sys.path[0]
c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:14: UserWarning: Boole
an Series key will be reindexed to match DataFrame index.
```

```
In [17]:

#------for feature 3(bruises)
Prob_Bruises_pos = probability(X_train[X_train.bruises == 'Bruises'][X_train.category == 'Positive'].shape[0],X_train
Prob_Bruises_neg = probability(X_train[X_train.bruises == 'Bruises'][X_train.category == 'Negative'].shape[0],X_train
Prob_NoBruises_pos = probability(X_train[X_train.bruises == 'NoBruises'][X_train.category == 'Positive'].shape[0],X_Train_Druises_neg = probability(X_train[X_train.bruises == 'NoBruises'][X_train.category == 'Negative'].shape[0],X_Train_Druises_neg = 'NoBruises_neg = 'NoBruises_n
```

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:2: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:3: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

This is separate from the ipykernel package so we can avoid doing imports until

0.6611653755473224

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:6: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.

c:\users\alouk\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:7: UserWarning: Boolean S
eries key will be reindexed to match DataFrame index.
import sys

Predicting probability of combinations of instances

```
In [19]:
              predicted = np.zeros((len(X train1),1))
              def prediction(a):
           3
                   ##-----testina
                  for i in range(len(a)):
           4
                      if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'NoBruises'):
           5
           6
                           positive prob1 = 0
           7
                           negative prob1 = 0
           8
                           positive prob1 = Prob convex pos * Prob Scaly pos * Prob NoBruises pos
                          negative prob1 = Prob convex neg * Prob Scaly neg * Prob NoBruises neg
           9
           10
                           if(positive prob1 >= negative prob1):
          11
                              predicted[i] = 1 #---positive
          12
                          else:
          13
                               predicted[i] = 0 #----negative
          14
          15
                      if(X train1[i][0] == 'Bell' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'NoBruises'):
          16
          17
                           positive prob1 = 0
          18
                          negative prob1 = 0
                          positive prob1 = Prob Bell pos * Prob Scaly pos * Prob NoBruises pos
          19
           20
                          negative prob1 = Prob Bell neg * Prob Scaly neg * Prob NoBruises neg
           21
                          if(positive prob1 >= negative prob1):
           22
                               predicted[i] = 1 #---positive
           23
                          else:
           24
                               predicted[i] = 0 #----negative
          25
           26
           27
                      if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'NoBruises'):
           28
                           positive prob1 = 0
           29
                          negative prob1 = 0
                           positive_prob1 = Prob_Flat_pos * Prob_Scaly_pos * Prob_NoBruises_pos
           30
           31
                          negative prob1 = Prob Flat neg * Prob Scaly neg * Prob NoBruises neg
                          if(positive prob1 >= negative prob1):
           32
           33
                               predicted[i] = 1 #---positive
           34
                           else:
           35
                               predicted[i] = 0 #----negative
           36
           37
                      if(X_train1[i][0] == 'Knobbed' and X_train1[i][1] == 'Scaly' and X_train1[i][2]== 'NoBruises'):
           38
                           positive prob1 = 0
           39
                          negative prob1 = 0
           40
                          positive_prob1 = Prob_Knobbed_pos * Prob_Scaly_pos * Prob_NoBruises_pos
           41
                           negative prob1 = Prob Knobbed neg * Prob Scaly neg * Prob NoBruises neg
```

```
42
                if(positive prob1 >= negative prob1):
43
                    predicted[i] = 1 #---positive
44
                else:
45
                    predicted[i] = 0 #----negative
46
47
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'NoBruises'):
48
                positive prob1 = 0
49
                negative prob1 = 0
                positive prob1 = Prob Sunken pos * Prob Scaly pos * Prob NoBruises pos
50
                negative prob1 = Prob Sunken neg * Prob Scaly neg * Prob NoBruises neg
51
52
                if(positive prob1 >= negative prob1):
53
                    predicted[i] = 1 #---positive
54
                else:
55
                    predicted[i] = 0 #----negative
56
57
58
           if(X train1[i][0] == 'Conical' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'NoBruises'):
59
                positive prob1 = 0
60
                negative prob1 = 0
61
                positive prob1 = Prob Conical pos * Prob Scaly pos * Prob NoBruises pos
                negative prob1 = Prob Conical neg * Prob Scaly neg * Prob NoBruises neg
62
63
                if(positive prob1 >= negative prob1):
                    predicted[i] = 1 #---positive
64
65
                else:
66
                    predicted[i] = 0 #----negative
67
68
69
70
71
                    #---cahnging fibrous
72
73
74
75
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'NoBruises'):
                positive_prob1 = 0
76
                negative_prob1 = 0
77
78
                positive prob1 = Prob convex pos * Prob Fiberous pos * Prob NoBruises pos
79
                negative_prob1 = Prob_convex_neg * Prob_Fiberous_neg * Prob_NoBruises_neg
80
                if(positive prob1 >= negative prob1):
81
                    predicted[i] = 1 #---positive
82
                else:
83
                    predicted[i] = 0 #----negative
```

```
84
 85
            if(X_train1[i][0] == 'Bell' and X_train1[i][1] == 'Fiberous' and X_train1[i][2]== 'NoBruises'):
 86
 87
                 positive prob1 = 0
 88
                 negative prob1 = 0
 89
                 positive prob1 = Prob Bell pos * Prob Fiberous pos * Prob NoBruises pos
 90
                 negative prob1 = Prob Bell neg * Prob Fiberous neg * Prob NoBruises neg
 91
                 if(positive prob1 >= negative prob1):
 92
                     predicted[i] = 1 #---positive
 93
                 else:
 94
                     predicted[i] = 0 #----negative
 95
 96
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'NoBruises'):
 97
 98
                 positive prob1 = 0
 99
                 negative prob1 = 0
100
                 positive prob1 = Prob Flat pos * Prob Fiberous pos * Prob NoBruises pos
                 negative prob1 = Prob Flat neg * Prob Fiberous neg * Prob NoBruises neg
101
102
                 if(positive prob1 >= negative prob1):
                     predicted[i] = 1 #---positive
103
104
                 else:
105
                     predicted[i] = 0 #----negative
106
            if(X train1[i][0] == 'Knobbed' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'NoBruises'):
107
108
                 positive prob1 = 0
109
                 negative prob1 = 0
110
                 positive prob1 = Prob Knobbed pos * Prob Fiberous pos * Prob NoBruises pos
                 negative prob1 = Prob Knobbed neg * Prob Fiberous neg * Prob NoBruises neg
111
112
                 if(positive prob1 >= negative prob1):
113
                     predicted[i] = 1 #---positive
114
                 else:
115
                     predicted[i] = 0 #----negative
116
117
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'NoBruises'):
118
                 positive prob1 = 0
119
                 negative prob1 = 0
                 positive prob1 = Prob Sunken pos * Prob Fiberous pos * Prob NoBruises pos
120
121
                 negative_prob1 = Prob_Sunken_neg * Prob_Fiberous_neg * Prob_NoBruises_neg
122
                 if(positive prob1 >= negative prob1):
                     predicted[i] = 1 #---positive
123
124
                 else:
125
                     predicted[i] = 0 #----negative
```

```
126
127
128
             if(X_train1[i][0] == 'Conical' and X_train1[i][1] == 'Fiberous' and X_train1[i][2]== 'NoBruises'):
129
                 positive prob1 = 0
130
                 negative prob1 = 0
                 positive prob1 = Prob Conical pos * Prob Fiberous pos * Prob NoBruises pos
131
                 negative prob1 = Prob Conical neg * Prob Fiberous neg * Prob NoBruises neg
132
133
                 if(positive prob1 >= negative prob1):
134
                     predicted[i] = 1 #---positive
135
                 else:
136
                     predicted[i] = 0 #----negative
137
138
139
140
141
142
143
144
145
             #----cahnging smooth
146
147
148
149
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'NoBruises'):
150
                 positive prob1 = 0
                 negative prob1 = 0
151
152
                 positive prob1 = Prob convex_pos * Prob_Smooth_pos * Prob_NoBruises_pos
153
                 negative prob1 = Prob convex neg * Prob Smooth neg * Prob NoBruises neg
                 if(positive prob1 >= negative prob1):
154
155
                     predicted[i] = 1 #---positive
156
                 else:
157
                     predicted[i] = 0 #----negative
158
159
160
            if(X train1[i][0] == 'Bell' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'NoBruises'):
                 positive prob1 = 0
161
                 negative prob1 = 0
162
163
                 positive_prob1 = Prob_Bell_pos * Prob_Smooth_pos * Prob_NoBruises_pos
164
                 negative prob1 = Prob Bell neg * Prob Smooth neg * Prob NoBruises neg
165
                 if(positive prob1 >= negative prob1):
166
                     predicted[i] = 1 #---positive
167
                 else:
```

```
168
                     predicted[i] = 0 #----negative
169
170
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'NoBruises'):
171
172
                 positive prob1 = 0
                negative prob1 = 0
173
174
                positive prob1 = Prob Flat pos * Prob Smooth pos * Prob NoBruises pos
175
                negative prob1 = Prob Flat neg * Prob Smooth neg * Prob NoBruises neg
                 if(positive prob1 >= negative prob1):
176
177
                     predicted[i] = 1 #---positive
178
                 else:
179
                     predicted[i] = 0 #----negative
180
            if(X train1[i][0] == 'Knobbed' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'NoBruises'):
181
182
                 positive prob1 = 0
                negative prob1 = 0
183
184
                 positive prob1 = Prob Knobbed_pos * Prob_Smooth_pos * Prob_NoBruises_pos
185
                 negative prob1 = Prob Knobbed neg * Prob Smooth neg * Prob NoBruises neg
186
                 if(positive prob1 >= negative prob1):
187
                     predicted[i] = 1 #---positive
188
                 else:
189
                     predicted[i] = 0 #----negative
190
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'NoBruises'):
191
192
                 positive prob1 = 0
193
                 negative prob1 = 0
194
                 positive prob1 = Prob Sunken pos * Prob Smooth pos * Prob NoBruises pos
195
                 negative prob1 = Prob Sunken neg * Prob Smooth neg * Prob NoBruises neg
                 if(positive prob1 >= negative prob1):
196
197
                     predicted[i] = 1 #---positive
198
                 else:
199
                     predicted[i] = 0 #----negative
200
201
202
            if(X train1[i][0] == 'Conical' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'NoBruises'):
                 positive_prob1 = 0
203
204
                 negative prob1 = 0
205
                 positive prob1 = Prob Conical pos * Prob Smooth pos * Prob NoBruises pos
206
                negative prob1 = Prob Conical neg * Prob Smooth neg * Prob NoBruises neg
207
                if(positive prob1 >= negative prob1):
208
                     predicted[i] = 1 #---positive
209
                 else:
```

```
210
                     predicted[i] = 0 #----negative
211
212
213
214
215
216
            #---cahnging grooves
217
218
219
220
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'NoBruises'):
221
                 positive prob1 = 0
222
                negative prob1 = 0
223
                 positive_prob1 = Prob_convex_pos * Prob_Grooves_pos * Prob_NoBruises_pos
224
                negative prob1 = Prob convex neg * Prob Grooves neg * Prob NoBruises neg
                if(positive prob1 >= negative prob1):
225
226
                     predicted[i] = 1 #---positive
227
                 else:
228
                     predicted[i] = 0 #----negative
229
230
231
            if(X train1[i][0] == 'Bell' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'NoBruises'):
232
                 positive prob1 = 0
233
                 negative prob1 = 0
234
                 positive_prob1 = Prob_Bell_pos * Prob_Grooves_pos * Prob_NoBruises_pos
                negative prob1 = Prob Bell_neg * Prob_Grooves_neg * Prob_NoBruises_neg
235
236
                 if(positive prob1 >= negative prob1):
237
                     predicted[i] = 1 #---positive
238
                 else:
239
                     predicted[i] = 0 #----negative
240
241
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'NoBruises'):
242
243
                 positive prob1 = 0
244
                 negative prob1 = 0
245
                 positive prob1 = Prob Flat pos * Prob Grooves pos * Prob NoBruises pos
246
                 negative prob1 = Prob Flat neg * Prob Grooves neg * Prob NoBruises neg
247
                 if(positive prob1 >= negative prob1):
248
                     predicted[i] = 1 #---positive
249
                 else:
250
                     predicted[i] = 0 #----negative
251
```

```
252
            if(X train1[i][0] == 'Knobbed' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'NoBruises'):
253
                 positive prob1 = 0
254
                 negative prob1 = 0
255
                 positive_prob1 = Prob_Knobbed_pos * Prob_Grooves_pos * Prob_NoBruises_pos
256
                 negative prob1 = Prob Knobbed neg * Prob Grooves neg * Prob NoBruises neg
257
                if(positive prob1 >= negative prob1):
258
                     predicted[i] = 1 #---positive
259
                 else:
260
                     predicted[i] = 0 #----negative
261
262
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'NoBruises'):
263
                 positive prob1 = 0
264
                negative prob1 = 0
                 positive_prob1 = Prob_Sunken_pos * Prob_Grooves_pos * Prob_NoBruises_pos
265
266
                negative prob1 = Prob Sunken neg * Prob Grooves neg * Prob NoBruises neg
                if(positive prob1 >= negative prob1):
267
268
                     predicted[i] = 1 #---positive
269
                 else:
270
                     predicted[i] = 0 #----negative
271
272
273
            if(X train1[i][0] == 'Conical' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'NoBruises'):
274
                 positive prob1 = 0
275
                 negative prob1 = 0
276
                 positive prob1 = Prob Conical_pos * Prob_Grooves_pos * Prob_NoBruises_pos
                negative_prob1 = Prob_Conical_neg * Prob_Grooves_neg * Prob_NoBruises_neg
277
278
                 if(positive prob1 >= negative prob1):
279
                     predicted[i] = 1 #---positive
280
                 else:
281
                     predicted[i] = 0 #----negative
282
283
284
                     ########
285
                     #######
286
                     ########
287
                     #########
288
                     #########
289
                     #########
290
                     ############
291
                     ########3
292
293
```

```
294
295
296
297
298
299
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'Bruises'):
300
                 positive prob1 = 0
301
                 negative prob1 = 0
302
                 positive_prob1 = Prob_convex_pos * Prob_Scaly_pos * Prob_Bruises_pos
303
                 negative_prob1 = Prob_convex_neg * Prob_Scaly_neg * Prob_Bruises_neg
304
                 if(positive prob1 >= negative prob1):
305
                     predicted[i] = 1 #---positive
306
                 else:
307
                     predicted[i] = 0 #----negative
308
309
310
            if(X train1[i][0] == 'Bell' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'Bruises'):
311
                 positive prob1 = 0
312
                 negative prob1 = 0
313
                 positive prob1 = Prob Bell pos * Prob Scaly pos * Prob Bruises pos
314
                negative prob1 = Prob Bell neg * Prob Scaly neg * Prob Bruises neg
315
                 if(positive prob1 >= negative prob1):
316
                     predicted[i] = 1 #---positive
317
                 else:
318
                     predicted[i] = 0 #----negative
319
320
321
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'Bruises'):
322
                 positive prob1 = 0
323
                 negative prob1 = 0
324
                 positive_prob1 = Prob_Flat_pos * Prob_Scaly_pos * Prob_Bruises_pos
325
                 negative prob1 = Prob Flat neg * Prob Scaly neg * Prob Bruises neg
326
                 if(positive prob1 >= negative prob1):
327
                     predicted[i] = 1 #---positive
328
                 else:
329
                     predicted[i] = 0 #----negative
330
331
            if(X_train1[i][0] == 'Knobbed' and X_train1[i][1] == 'Scaly' and X_train1[i][2]== 'Bruises'):
332
                 positive prob1 = 0
333
                 negative prob1 = 0
334
                 positive_prob1 = Prob_Knobbed_pos * Prob_Scaly_pos * Prob_Bruises_pos
                 negative_prob1 = Prob_Knobbed_neg * Prob_Scaly_neg * Prob_Bruises_neg
335
```

```
336
                 if(positive prob1 >= negative prob1):
337
                     predicted[i] = 1 #---positive
338
                 else:
339
                     predicted[i] = 0 #----negative
340
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'Bruises'):
341
342
                 positive prob1 = 0
343
                 negative prob1 = 0
344
                 positive prob1 = Prob Sunken pos * Prob Scaly pos * Prob Bruises pos
                 negative prob1 = Prob Sunken neg * Prob Scaly neg * Prob Bruises neg
345
346
                 if(positive prob1 >= negative prob1):
347
                     predicted[i] = 1 #---positive
348
                 else:
349
                     predicted[i] = 0 #----negative
350
351
352
            if(X train1[i][0] == 'Conical' and X train1[i][1] == 'Scaly' and X train1[i][2]== 'Bruises'):
353
                 positive prob1 = 0
354
                 negative prob1 = 0
355
                 positive prob1 = Prob Conical pos * Prob Scaly pos * Prob Bruises pos
                negative prob1 = Prob Conical neg * Prob Scaly neg * Prob Bruises neg
356
357
                 if(positive prob1 >= negative prob1):
                     predicted[i] = 1 #---positive
358
359
                 else:
360
                     predicted[i] = 0 #----negative
361
362
363
364
                     #---cahnging fibrous
365
366
367
368
369
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'NoBruises'):
370
                 positive prob1 = 0
371
                 negative prob1 = 0
372
                 positive_prob1 = Prob_convex_pos * Prob_Fiberous_pos * Prob_Bruises_pos
373
                 negative_prob1 = Prob_convex_neg * Prob_Fiberous_neg * Prob_Bruises_neg
374
                 if(positive prob1 >= negative prob1):
375
                     predicted[i] = 1 #---positive
376
                 else:
377
                     predicted[i] = 0 #----negative
```

```
378
379
            if(X_train1[i][0] == 'Bell' and X_train1[i][1] == 'Fiberous' and X_train1[i][2]== 'Bruises'):
380
381
                 positive prob1 = 0
382
                 negative prob1 = 0
383
                 positive prob1 = Prob Bell pos * Prob Fiberous pos * Prob Bruises pos
384
                negative prob1 = Prob Bell neg * Prob Fiberous neg * Prob Bruises neg
385
                 if(positive prob1 >= negative prob1):
386
                     predicted[i] = 1 #---positive
387
                 else:
388
                     predicted[i] = 0 #----negative
389
390
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'Bruises'):
391
392
                 positive prob1 = 0
393
                 negative prob1 = 0
394
                 positive prob1 = Prob Flat pos * Prob Fiberous pos * Prob Bruises pos
                 negative prob1 = Prob Flat neg * Prob Fiberous neg * Prob Bruises neg
395
396
                 if(positive prob1 >= negative prob1):
397
                     predicted[i] = 1 #---positive
398
                 else:
399
                     predicted[i] = 0 #----negative
400
            if(X train1[i][0] == 'Knobbed' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'Bruises'):
401
402
                 positive prob1 = 0
403
                 negative prob1 = 0
404
                 positive prob1 = Prob Knobbed pos * Prob Fiberous pos * Prob Bruises pos
                 negative prob1 = Prob Knobbed neg * Prob Fiberous neg * Prob Bruises neg
405
406
                 if(positive prob1 >= negative prob1):
407
                     predicted[i] = 1 #---positive
408
                 else:
409
                     predicted[i] = 0 #----negative
410
411
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Fiberous' and X train1[i][2]== 'Bruises'):
412
                 positive prob1 = 0
413
                 negative prob1 = 0
                 positive prob1 = Prob Sunken pos * Prob Fiberous pos * Prob Bruises pos
414
415
                 negative_prob1 = Prob_Sunken_neg * Prob_Fiberous_neg * Prob_Bruises_neg
416
                 if(positive prob1 >= negative prob1):
                     predicted[i] = 1 #---positive
417
418
                 else:
419
                     predicted[i] = 0 #----negative
```

```
420
421
422
            if(X_train1[i][0] == 'Conical' and X_train1[i][1] == 'Fiberous' and X_train1[i][2]== 'Bruises'):
                 positive prob1 = 0
423
424
                negative prob1 = 0
425
                 positive prob1 = Prob Conical pos * Prob Fiberous pos * Prob Bruises pos
                negative prob1 = Prob Conical_neg * Prob_Fiberous_neg * Prob_Bruises_neg
426
427
                if(positive prob1 >= negative prob1):
428
                     predicted[i] = 1 #---positive
429
                 else:
430
                     predicted[i] = 0 #----negative
431
432
433
434
435
436
437
438
439
              #---cahnging smooth
440
441
442
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'Bruises'):
443
444
                 positive prob1 = 0
                negative prob1 = 0
445
446
                 positive prob1 = Prob convex pos * Prob Smooth pos * Prob Bruises pos
447
                negative prob1 = Prob convex neg * Prob Smooth neg * Prob Bruises neg
                if(positive prob1 >= negative prob1):
448
449
                     predicted[i] = 1 #---positive
450
                 else:
451
                     predicted[i] = 0 #----negative
452
453
454
            if(X train1[i][0] == 'Bell' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'Bruises'):
455
                 positive prob1 = 0
456
                 negative prob1 = 0
457
                 positive prob1 = Prob Bell pos * Prob Smooth pos * Prob Bruises pos
458
                negative prob1 = Prob Bell neg * Prob Smooth neg * Prob Bruises neg
459
                if(positive prob1 >= negative prob1):
460
                     predicted[i] = 1 #---positive
461
                 else:
```

```
462
                     predicted[i] = 0 #----negative
463
464
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'Bruises'):
465
466
                 positive prob1 = 0
467
                 negative prob1 = 0
468
                positive prob1 = Prob Flat pos * Prob Smooth pos * Prob Bruises pos
469
                negative prob1 = Prob Flat neg * Prob Smooth neg * Prob Bruises neg
470
                 if(positive prob1 >= negative prob1):
471
                     predicted[i] = 1 #---positive
472
                 else:
473
                     predicted[i] = 0 #----negative
474
475
            if(X train1[i][0] == 'Knobbed' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'Bruises'):
476
                 positive prob1 = 0
477
                 negative prob1 = 0
478
                 positive prob1 = Prob Knobbed_pos * Prob_Smooth_pos * Prob_Bruises_pos
479
                 negative prob1 = Prob Knobbed neg * Prob Smooth neg * Prob Bruises neg
480
                 if(positive prob1 >= negative prob1):
481
                     predicted[i] = 1 #---positive
482
                 else:
483
                     predicted[i] = 0 #----negative
484
485
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'Bruises'):
486
                 positive prob1 = 0
487
                 negative prob1 = 0
488
                 positive prob1 = Prob Sunken pos * Prob Smooth pos * Prob Bruises pos
                 negative prob1 = Prob Sunken neg * Prob Smooth neg * Prob Bruises neg
489
490
                 if(positive prob1 >= negative prob1):
491
                     predicted[i] = 1 #---positive
492
                 else:
493
                     predicted[i] = 0 #----negative
494
495
            if(X train1[i][0] == 'Conical' and X train1[i][1] == 'Smooth' and X train1[i][2]== 'Bruises'):
496
                 positive prob1 = 0
497
498
                 negative prob1 = 0
499
                 positive prob1 = Prob Conical pos * Prob Smooth pos * Prob Bruises pos
500
                negative prob1 = Prob Conical neg * Prob Smooth neg * Prob Bruises neg
501
                if(positive prob1 >= negative prob1):
502
                     predicted[i] = 1 #---positive
503
                 else:
```

```
504
                     predicted[i] = 0 #----negative
505
506
507
508
509
510
            #---cahnging grooves
511
512
513
514
            if(X train1[i][0] == 'Convex' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'Bruises'):
515
                 positive prob1 = 0
                negative prob1 = 0
516
517
                 positive_prob1 = Prob_convex_pos * Prob_Grooves_pos * Prob_Bruises_pos
518
                negative prob1 = Prob convex neg * Prob Grooves neg * Prob Bruises neg
                if(positive prob1 >= negative prob1):
519
520
                     predicted[i] = 1 #---positive
521
                 else:
522
                     predicted[i] = 0 #----negative
523
524
525
            if(X train1[i][0] == 'Bell' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'Bruises'):
526
                 positive prob1 = 0
527
                 negative prob1 = 0
528
                 positive_prob1 = Prob_Bell_pos * Prob_Grooves_pos * Prob_Bruises_pos
                negative prob1 = Prob Bell_neg * Prob_Grooves_neg * Prob_Bruises_neg
529
530
                 if(positive prob1 >= negative prob1):
531
                     predicted[i] = 1 #---positive
532
                 else:
533
                     predicted[i] = 0 #----negative
534
535
            if(X train1[i][0] == 'Flat' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'Bruises'):
536
537
                 positive prob1 = 0
538
                 negative prob1 = 0
539
                 positive prob1 = Prob Flat pos * Prob Grooves pos * Prob Bruises pos
540
                 negative_prob1 = Prob_Flat_neg * Prob_Grooves_neg * Prob_Bruises_neg
                 if(positive prob1 >= negative prob1):
541
542
                     predicted[i] = 1 #---positive
543
                 else:
544
                     predicted[i] = 0 #----negative
545
```

```
546
            if(X train1[i][0] == 'Knobbed' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'Bruises'):
547
                 positive prob1 = 0
                negative prob1 = 0
548
549
                 positive_prob1 = Prob_Knobbed_pos * Prob_Grooves_pos * Prob_Bruises_pos
550
                 negative prob1 = Prob Knobbed_neg * Prob_Grooves_neg * Prob_Bruises_neg
                if(positive prob1 >= negative prob1):
551
552
                     predicted[i] = 1 #---positive
553
                 else:
554
                     predicted[i] = 0 #----negative
555
556
            if(X train1[i][0] == 'Sunken' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'Bruises'):
557
                 positive prob1 = 0
558
                negative prob1 = 0
559
                 positive_prob1 = Prob_Sunken_pos * Prob_Grooves_pos * Prob_Bruises_pos
560
                negative prob1 = Prob Sunken neg * Prob Grooves neg * Prob Bruises neg
                if(positive prob1 >= negative prob1):
561
                     predicted[i] = 1 #---positive
562
563
                 else:
564
                     predicted[i] = 0 #----negative
565
566
567
            if(X train1[i][0] == 'Conical' and X train1[i][1] == 'Grooves' and X train1[i][2]== 'Bruises'):
                 positive prob1 = 0
568
569
                 negative prob1 = 0
570
                 positive prob1 = Prob Conical_pos * Prob_Grooves_pos * Prob_Bruises_pos
                negative prob1 = Prob Conical_neg * Prob_Grooves_neg * Prob_Bruises_neg
571
572
                 if(positive prob1 >= negative prob1):
573
                     predicted[i] = 1 #---positive
574
                 else:
575
                     predicted[i] = 0 #----negative
576
577
578
579
580
581
582
583
584
585
586
587
```

Finding Values for confusion matrix

```
In [21]:
          1 L = []
          2 for i in range(len(y_test)):
                 if(y test[i] == 1 and predicted[i] == 1):#-----TRUE POSITIVE VALUE
                    L.append(y test[i])
          5 TP=len(L)
             print(TP)
          9 L1 = []
         10 for i in range(len(y test)):
         11
                 if(y test[i] == 0 and predicted[i] == 0):#-----FALSE POSITIVE VALUE
                     L1.append(y test[i])
         12
         13 FP=len(L1)
         14 print(FP)
         15
         16
         17
         18 L2 = []
         19 for i in range(len(y test)):
                 if(y test[i] != 1 and predicted[i] == 1):#-----TRUE NEGATIVE VALUE
         20
                    L2.append(v test[i])
         21
         22 TN=len(L2)
         23
             print(TN)
         24
         25
         26 L3 = []
         27 for i in range(len(y test)):
                 if(y_test[i] == 0 and predicted[i] != 0):#-----FALSE NEGATIVE VALUE
         28
                     L3.append(y test[i])
         30 FN=len(L2)
         31 print(FN)
         32
         800
```

The final Accuracy, Sensitivity and Specificity

863334334

Creating modified database which contains new coloumn "predicted_values"

```
In [24]: 1 X_test = X_test.assign(predicted_values=predicted_values.values)
```

In [25]: 1 X_test

Out[25]:

	cap_shape	Cap_surface	bruises	category	predicted_values
7099	Knobbed	Scaly	NoBruises	Negative	Negative
6901	Knobbed	Smooth	NoBruises	Negative	Negative
5004	Flat	Scaly	NoBruises	Negative	Negative
662	Convex	Scaly	Bruises	Negative	Positive
2229	Convex	Fiberous	Bruises	Positive	Negative
2917	Flat	Fiberous	Bruises	Positive	Positive
7313	Knobbed	Scaly	NoBruises	Negative	Negative
3820	Flat	Scaly	Bruises	Positive	Positive
4749	Convex	Fiberous	NoBruises	Negative	Positive
2913	Convex	Scaly	Bruises	Positive	Positive

2438 rows × 5 columns

Exporting database to csv file

```
In [26]: 1 export_csv = X_test.to_csv (r'W:\Lakehead Study material\Big data\Assignment 2\Naive bayes\Work directory\prediction
In [27]: 1 accuracy(y_test,predicted)#------finding the accuracy
Sensitivity()
Specificity()
```

Accuracy is 68.21164889253485 Sensitivity is 70.54673721340387 Specificity is 27.903091060985798

In	[]:[1	
In	[]]:[1	
In	[]]:[1	
In	[]]:[1	
In	[]]:[1	
In	[]]:[1	