Machine Learning Lab program 1



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AIM: Study and implement the Naive Bayes learner on a breast cancer dataset

GitHub link: - https://github.com/pranshuag9/machine-learning-lab/blob/main/lab4/Naive%20Bayes.ipynb

Program Snippets: 1. Loading dataset

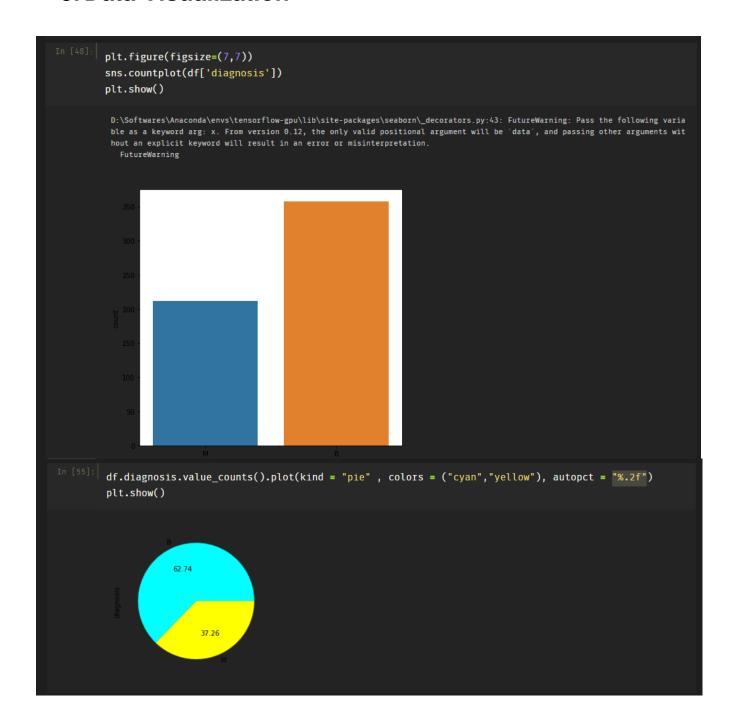
	ļ	Load	ling da	ataset						
In [2]:			ndas <mark>as</mark> po ead_csv("o	d cancer.csv")					
In [3]:	df.	head(1	0)							
		id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	со
	0 8	342302	M	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3
	1 8	342517	М	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0
	2 8	34300903	М	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1
	3 8	34348301	М	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2
	4 8	34358402	М	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1
	5 8	343786	М	12.45	15.70	82.57	477.1	0.12780	0.17000	0.1
	6 8	344359	М	18.25	19.98	119.60	1040.0	0.09463	0.10900	0.1
	7 8	34458202	M	13.71	20.83	90.20	577.9	0.11890	0.16450	0.0
	8 8	344981	M	13.00	21.82	87.50	519.8	0.12730	0.19320	0.1
	9 8	34501001	M	12.46	24.04	83.97	475.9	0.11860	0.23960	0.2
	10 r	ows × 33	columns							
In [4]:	df.	tail(5)							
		id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	со
	564	926424	М	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.2
	565	926682	М	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.1
	566	926954	M	16.60	28.08	108.30	858.1	0.08455	0.10230	0.0
	567	927241	M	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.3
	568	92751	В	7.76	24.54	47.92	181.0	0.05263	0.04362	0.0
_	5 ro	ws × 33 (columns							

2. Data Cleaning/ Preprocessing

```
dropping id and unnamed: 32 column because they are noise are doesnt affect result
            df = df.drop(["id"], axis= 1)
            df.columns.values
              array(['diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',
                       'area_mean', 'smoothness_mean', 'compactness_mean',
                       'concavity_mean', 'concave points_mean', 'symmetry_mean',
                      'fractal_dimension_mean', 'radius_se', 'texture_se',
'perimeter_se', 'area_se', 'smoothness_se', 'compactness_se',
                      'concavity_se', 'concave points_se', 'symmetry_se',
                      'fractal_dimension_se', 'radius_worst', 'texture_worst',
                      'perimeter_worst', 'area_worst', 'smoothness_worst',
                      'compactness_worst', 'concavity_worst', 'concave points_worst', 'symmetry_worst', 'fractal_dimension_worst', 'Unnamed: 32'],
                      dtype=object)
In [42]: df = df.drop(["Unnamed: 32"], axis=1)
In [43]: df.columns.values
              array(['diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',
                       'area_mean', 'smoothness_mean', 'compactness_mean',
                      'concavity_mean', 'concave points_mean', 'symmetry_mean',
'fractal_dimension_mean', 'radius_se', 'texture_se',
                      'perimeter_se', 'area_se', 'smoothness_se', 'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
                      'fractal_dimension_se', 'radius_worst', 'texture_worst',
                      'perimeter_worst', 'area_worst', 'smoothness_worst',
'compactness_worst', 'concavity_worst', 'concave points_worst',
                       'symmetry_worst', 'fractal_dimension_worst'], dtype=object)
```

Х												
	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean					
0	0.521037	0.022658	0.545989	0.363733	0.593753	0.792037	0.703140					
1	0.643144	0.272574	0.615783	0.501591	0.289880	0.181768	0.203608					
2	0.601496	0.390260	0.595743	0.449417	0.514309	0.431017	0.462512					
3	0.210090	0.360839	0.233501	0.102906	0.811321	0.811361	0.565604					
4	0.629893	0.156578	0.630986	0.489290	0.430351	0.347893	0.463918					
564	0.690000	0.428813	0.678668	0.566490	0.526948	0.296055	0.571462					
565	0.622320	0.626987	0.604036	0.474019	0.407782	0.257714	0.337395					
566	0.455251	0.621238	0.445788	0.303118	0.288165	0.254340	0.216753					
567	0.644564	0.663510	0.665538	0.475716	0.588336	0.790197	0.823336					
568	0.036869	0.501522	0.028540	0.015907	0.000000	0.074351	0.000000					

3. Data Visualization



4. Dividing data into train/test splits

```
In [67]: from sklearn.model_selection import train_test_split

#for checking testing results
from sklearn.metrics import classification_report, confusion_matrix

#for visualizing tree
from sklearn.tree import plot_tree

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2, random_state = 0)

print("Training split input- ", x_train.shape)
print("Testing split input- ", x_test.shape)

Training split input- (455, 10)
Testing split input- (114, 10)
```

5. Using Naive Bayes gaussian learning method on breast cancer dataset

```
In [41]: from sklearn.model_selection import train_test_split
        from sklearn.naive_bayes import GaussianNB
        from sklearn.metrics import accuracy_score
         from sklearn import metrics
In [31]: nb = GaussianNB()
        nb.fit(x_train,y_train)
         GaussianNB()
        y_pred = nb.predict(x_test)
        print("Classification report - \n", classification_report(y_test,y_pred))
         Classification report -
                     precision recall f1-score support
                  М
                       0.87
                               0.85 0.86
                                      0.89
            accuracy
                    0.88
0.89
                                0.88
                                        0.88
                                                 114
           macro avg
         weighted avg
                                        0.89
```

6. Confusion matrix

```
cm=confusion_matrix(y_test,y_pred)
cm
array([[61, 6],
[ 7, 40]], dtype=int64)
plt.figure(figsize=(5,5))
sns.heatmap(data=cm,linewidths=1.0, annot=True,square = True, cmap = 'Blues')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
all_sample_title = 'Accuracy Score: {0}'.format(round(nb.score(x_test, y_test),2))
plt.title(all_sample_title, size = 15)
plt.savefig("accuracy_nb.png")
```