

Machine Learning Assignment – 04
Praveena Goli(700743010)

```
#importing the required libraries to work with Tabular data and also to implement algorithms

import warnings
import numpy as np
import pandas as pd
import seaborn as sns
from sklearn import preprocessing
import matplotlib.pyplot as plt
from scipy.stats.stats import pearsonr
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, recall_score, precision_score,
classification_report, confusion_matrix
warnings.filterwarnings("ignore")
```

Question: 2

Titanic Dataset

1. Find the correlation between 'survived' (target column) and 'sex' column for the Titanic use case in class.
 - a. Do you think we should keep this feature?
2. Do at least two visualizations to describe or show correlations.
3. Implement Naïve Bayes method using scikit-learn library and report the accuracy

Question 3

(Glass Dataset)

1. Implement Naïve Bayes method using scikit-learn library.
 - a. Use the glass dataset available in Link also provided in your assignment.
 - b. Use train_test_split to create training and testing part.
2. Evaluate the model on testing part using score and classification_report(y_true, y_pred)

1. Implement linear SVM method using scikit library
 - a. Use the glass dataset available in Link also provided in your assignment.
 - b. Use train_test_split to create training and testing part.
2. Evaluate the model on testing part using score and

	Duration	Pulse	Maxpulse	Calories
0	60	110	130	409.1
1	60	117	145	479.0
2	60	103	135	340.0
3	45	109	175	282.4
4	45	117	148	406.0

	Duration	Pulse	Maxpulse	Calories
count	169.000000	169.000000	169.000000	164.000000
mean	63.846154	107.461538	134.047337	375.790244
std	42.299949	14.510259	16.450434	266.379919
min	15.000000	80.000000	100.000000	50.300000
25%	45.000000	100.000000	124.000000	250.925000
50%	60.000000	105.000000	131.000000	318.600000
75%	60.000000	111.000000	141.000000	387.600000
max	300.000000	159.000000	184.000000	1860.400000

Duration False
 Pulse False
 Maxpulse False
 Calories True
 dtype: bool
 Duration False
 Pulse False
 Maxpulse False
 Calories False
 dtype: bool

	Maxpulse	Calories
min	100.000000	50.300000
max	184.000000	1860.400000
count	169.000000	169.000000
mean	134.047337	375.790244

	Maxpulse		Calories	
	Duration	Pulse	Maxpulse	Calories
51	80	123	146	643.1
62	160	109	135	853.0
65	180	90	130	800.4
66	150	105	135	873.4
67	150	107	130	816.0
72	90	100	127	700.0
73	150	97	127	953.2
75	90	98	125	563.2
78	120	100	130	500.4
90	180	101	127	600.1
99	90	93	124	604.1
103	90	90	100	500.4
106	180	90	120	800.3
108	90	90	120	500.3

	Duration	Pulse	Maxpulse	Calories
65	180	90	130	800.4
70	150	97	129	1115.0
73	150	97	127	953.2
75	90	98	125	563.2
99	90	93	124	604.1
103	90	90	100	500.4
106	180	90	120	800.3
108	90	90	120	500.3

	Duration	Pulse	Calories
0	60	110	409.1
1	60	117	479.0
2	60	103	340.0
3	45	109	282.4
4	45	117	406.0

	Duration	Pulse	Calories
0	60	110	409.1
1	60	117	479.0
2	60	103	340.0
3	45	109	282.4
4	45	117	406.0

Duration int64
 Pulse int64
 Calories float64
 dtype: object
 Duration int64
 Pulse int64
 Calories int64
 dtype: object
 <Axes: xlabel='Duration', ylabel='Calories'>

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	ParCh	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumin, Mrs. John Bradley (Florence Briggs Th...)	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrell, Mrs. Jacques Heath	female	35.0	1	0	113803	53.1000	C123	S

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
				(Lily May Peel)								
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

-0.5433513806577547

	PassengerId	Survived	Pclass	Sex	Age	SibSp \	
PassengerId	1.000000	-0.005007	-0.035144	0.042939	0.036847	-0.057527	
Survived	-0.005007	1.000000	-0.338481	-0.543351	-0.077221	-0.035322	
Pclass	-0.035144	-0.338481	1.000000	0.131900	-0.369226	0.083081	
Sex	0.042939	-0.543351	0.131900	1.000000	0.093254	-0.114631	
Age	0.036847	-0.077221	-0.369226	0.093254	1.000000	-0.308247	
SibSp	-0.057527	-0.035322	0.083081	-0.114631	-0.308247	1.000000	
Parch	-0.001652	0.081629	0.018443	-0.245489	-0.189119	0.414838	
Fare	0.012658	0.257307	-0.549500	-0.182333	0.096067	0.159651	

	Parch	Fare
PassengerId	-0.001652	0.012658
Survived	0.081629	0.257307
Pclass	0.018443	-0.549500
Sex	-0.245489	-0.182333
Age	-0.189119	0.096067
SibSp	0.414838	0.159651
Parch	1.000000	0.216225
Fare	0.216225	1.000000

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare
PassengerId	1.000000	-0.005007	-0.035144	0.042939	0.036847	-0.057527	-0.001652	0.012658
Survived	-0.005007	1.000000	-0.338481	-0.543351	-0.077221	-0.035322	0.081629	0.257307
Pclass	-0.035144	-0.338481	1.000000	0.131900	-0.369226	0.083081	0.018443	-0.549500
Sex	0.042939	-0.543351	0.131900	1.000000	0.093254	-0.114631	-0.245489	-0.182333

	Passenger Id	Surviv ed	Pclass	Sex	Age	SibSp	Parch	Fare
Age	0.036847	- 0.0772 21	- 0.3692 26	0.0932 54	1.0000 00	- 0.3082 47	- 0.1891 19	0.0960 67
SibSp	-0.057527	- 0.0353 22	0.0830 81	- 0.1146 31	- 0.3082 47	1.0000 00	0.4148 38	0.1596 51
Parch	-0.001652	0.0816 29	0.0184 43	- 0.2454 89	- 0.1891 19	0.4148 38	1.0000 00	0.2162 25
Fare	0.012658	0.2573 07	- 0.5495 00	- 0.1823 33	0.0960 67	0.1596 51	0.2162 25	1.0000 00

GaussianNB(priors=None, var_smoothing=1e-09)

precision recall f1-score support

0.0 0.79 0.80 0.80 85

1.0 0.70 0.69 0.70 58

accuracy 0.76 143

macro avg 0.75 0.74 0.75 143

weighted avg 0.75 0.76 0.75 143

[[68 17]

[18 40]]

accuracy is 0.7552447552447552

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
0	1.521 01	13.64	4.49	1.10	71.78	0. 06	8.75	0.0	0.0	1
1	1.517 61	13.89	3.60	1.36	72.73	0. 48	7.83	0.0	0.0	1
2	1.516 18	13.53	3.55	1.54	72.99	0. 39	7.78	0.0	0.0	1
3	1.517 66	13.21	3.69	1.29	72.61	0. 57	8.22	0.0	0.0	1
4	1.517 42	13.27	3.62	1.24	73.08	0. 55	8.07	0.0	0.0	1

RI Na Mg Al Si K Ca Ba Fe Type

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type	
RI	1.000 000	- 0.191 885	- 0.122 274	- 0.407 326	- 0.54205 2	- 0.289 833	- 0.8104 03	- 0.00 0386	- 0.14 3010	- 0.16 4237	
Na	- 0.191 885	1.000 000	- 0.273 732	0.156 794	- 0.06980 9	- 0.266 087	- 0.2754 42	0.32 6603	- 0.24 1346	0.50 2898	
Mg	- 0.122 274	- 0.273 732	1.000 000	- 0.481 799	- 0.16592 7	0.005 396	- 0.4437 50	- 0.49 2262	0.08 3060	- 0.74 4993	
Al	- 0.407 326	0.156 794	- 0.481 799	1.000 000	- 0.00552 4	0.325 958	- 0.2595 92	0.47 9404	- 0.07 4402	0.59 8829	
Si	- 0.542 052	- 0.069 809	- 0.165 927	- 0.005 524	1.00000 0	- 0.193 331	- 0.2087 32	- 0.10 2151	- 0.09 4201	0.15 1565	
K	- 0.289 833	- 0.266 087	0.005 396	0.325 958	- 0.19333 1	1.000 000	- 0.3178 36	- 0.04 2618	- 0.00 7719	- 0.01 0054	
Ca	0.810 403	- 0.275 442	- 0.443 750	- 0.259 592	- 0.20873 2	- 0.317 836	1.0000 00	- 0.11 2841	0.12 4968	0.00 0952	
Ba	- 0.000 386	0.326 603	- 0.492 262	0.479 404	- 0.10215 1	- 0.042 618	- 0.1128 41	1.00 0000	- 0.05 8692	0.57 5161	
Fe	0.143 010	- 0.241 346	0.083 060	- 0.074 402	- 0.09420 1	- 0.007 719	0.1249 68	- 0.05 8692	1.00 0000	- 0.18 8278	
Type	- 0.164 237	0.502 898	- 0.744 993	0.598 829	0.15156 5	- 0.010 054	0.0009 52	0.57 5161	- 0.18 8278	1.00 0000	

precision recall f1-score support

1	0.90	0.95	0.92	19
2	0.92	0.92	0.92	12
3	1.00	0.50	0.67	6
5	0.00	0.00	0.00	1
6	1.00	1.00	1.00	1
7	0.75	0.75	0.75	4

accuracy		0.84	43
macro avg	0.76	0.69	0.71
weighted avg	0.89	0.84	0.85

```
[[18 1 0 0 0 0]
 [ 1 11 0 0 0 0]
 [ 1 0 3 2 0 0]
 [ 0 0 0 0 0 1]
 [ 0 0 0 0 1 0]
 [ 0 0 0 1 0 3]]
```

accuracy is 0.8372093023255814

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

1	1.00	0.89	0.94	19
2	0.46	1.00	0.63	12
3	0.00	0.00	0.00	6
5	0.00	0.00	0.00	1
6	0.00	0.00	0.00	1
7	0.00	0.00	0.00	4

accuracy		0.67	43
macro avg	0.24	0.32	0.26
weighted avg	0.57	0.67	0.59

```
[[17 2 0 0 0 0]
 [ 0 12 0 0 0 0]
 [ 0 6 0 0 0 0]
 [ 0 1 0 0 0 0]
 [ 0 1 0 0 0 0]
 [ 0 4 0 0 0 0]]
```

accuracy is 0.6744186046511628

GITHUB LINK : https://github.com/Goli18/Machine-Learning_09.git

Video Link :

https://github.com/Goli18/Machine-Learning_09/blob/main/Machine_learning_Assignment-4.ipynb%20-%20Visual%20Studio%20Code%202023-04-05%2022-28-41.mp4