

STEP OF MACHINE LEARNING

1. PREPROCESSING + EDA + FEATURE SELECTION
2. EXTRACT INPUT AND OUTPUT COLUMNS
3. TRAIN TEST SPLIT
4. TRAIN THE MODEL
5. EVALUATE THE MODEL/MODEL SELECTION
6. DEPLOY THE MODEL

#IMPORTING THE NECESSARY LIBRARIES

```
import numpy as np
import pandas as pd
```

#Reading the data

```
data = pd.read_csv('/content/placement.csv')
data
```

	Unnamed: 0	cgpa	iq	placement
0	0	6.8	123	1
1	1	5.9	106	0
2	2	5.3	121	0
3	3	7.4	132	1
4	4	5.8	142	0
...
95	95	4.3	200	0
96	96	4.4	42	0
97	97	6.7	182	1
98	98	6.3	103	1
99	99	6.2	113	1

100 rows × 4 columns

data.head()

	Unnamed: 0	cgpa	iq	placement
0	0	6.8	123	1
1	1	5.9	106	0
2	2	5.3	121	0
3	3	7.4	132	1
4	4	5.8	142	0

data.tail()

	Unnamed: 0	cgpa	iq	placement
95	95	4.3	200	0
96	96	4.4	42	0
97	97	6.7	182	1
98	98	6.3	103	1
99	99	6.2	113	1

data.shape

```
(100, 4)
```

```
data.shape[0]
```

```
100
```

```
#removing missing values
```

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Unnamed: 0    100 non-null    int64
1   cgpa          100 non-null    float64
2   iq            100 non-null    int64
3   placement     100 non-null    int64
dtypes: float64(1), int64(3)
memory usage: 3.2 KB
```

here as you can see there are no null values. so no data preprocessing required

```
data.head()
```

	Unnamed: 0	cgpa	iq	placement
0	0	6.8	123	1
1	1	5.9	106	0
2	2	5.3	121	0
3	3	7.4	132	1
4	4	5.8	142	0

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Unnamed: 0    100 non-null    int64
1   cgpa          100 non-null    float64
2   iq            100 non-null    int64
3   placement     100 non-null    int64
dtypes: float64(1), int64(3)
memory usage: 3.2 KB
```

```
data = data.iloc[:,1:]
```

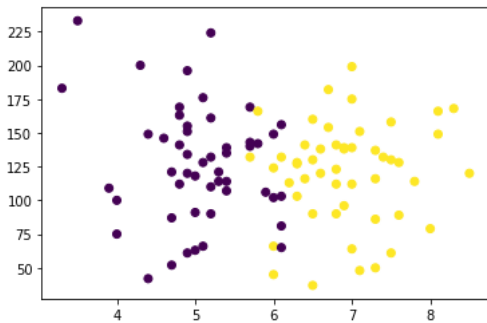
```
data
```

cgpa iq placement 

```
import matplotlib.pyplot as plt
```

```
plt.scatter(data['cgpa'],data['iq'],c = data['placement'])
#xaxis = cgpa , y axis = iq # colored the placement coloumn
```

<matplotlib.collections.PathCollection at 0x7f9f1463fbe0>



**Using LOGISTIC REGRESSION NAMED LOGISTIC REGRESSION. **

#LOGISTIC REGRESSION BECAUSE THIS IS CLASSIFICATION PROBLEM

Independent Variable = Input = cgpa & iq dependent variable = output = Placement

```
x = data.iloc[:, 0:2]
y = data.iloc[:, -1]
```

X

	cgpa	iq
0	6.8	123
1	5.9	106
2	5.3	121
3	7.4	132
4	5.8	142
...
95	4.3	200
96	4.4	42
97	6.7	182
98	6.3	103
99	6.2	113

100 rows × 2 columns

y

0	1
1	0
2	0
3	1
4	0
...	...
95	0

```

96 0
97 1
98 1
99 1
Name: placement, Length: 100, dtype: int64

```

```

#train test split first
# then scale the values

```

```

from sklearn.model_selection import train_test_split
#train_test_split(inpendant variable , dependent variable, give t
x_train,x_test, y_train, y_test = train_test_split(x,y,test_size = 0

```

x_train

	cgpa	iq
54	6.4	141
79	6.5	90
78	6.1	81
40	4.9	134
61	7.3	137
...
84	5.7	169
35	6.8	90
75	4.8	169
15	5.1	176
87	5.7	132

90 rows × 2 columns

x_test

	cgpa	iq
25	5.0	91
24	4.7	121
77	7.3	50
90	7.3	86
72	7.3	116
44	7.5	61
33	6.0	149
89	4.9	151
46	5.3	114
81	5.4	107

y_train

```

54 1
79 1
78 0
40 0
61 1
..
84 0
35 1
75 0
15 0
87 1
Name: placement, Length: 90, dtype: int64

```

y_test

```

25    0
24    0
77    1
90    1
72    1
44    1
33    0
89    0
46    0
81    0
Name: placement, dtype: int64

```

#scaling all values between 0 and 1
 #it is not necessary but you should do it

from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()

x_train = scaler.fit_transform(x_train)
 x_train

```

array([[ 0.36580376,  0.38287711],
       [ 0.45336317, -0.89245377],
       [ 0.10312553, -1.11751216],
       [-0.94758739,  0.2078317 ],
       [ 1.15383845,  0.28285116],
       [ 0.45336317, -2.21779762],
       [ 0.89116022, -1.54262245],
       [ 0.54092258, -0.14225913],
       [-0.86002798, -1.56762894],
       [ 2.20455136, -0.14225913],
       [-0.68490916,  2.45841561],
       [ 0.01556612, -0.04223318],
       [-2.34853795,  1.4331496 ],
       [ 0.27824435,  0.05779277],
       [ 0.10312553, -0.56736942],
       [-0.1595527 ,  1.00803931],
       [ 0.54092258,  0.30785765],
       [-0.59734975, -0.11725264],
       [ 0.7160414 ,  0.38287711],
       [ 0.97871963, -1.94272626],
       [ 0.62848199,  1.40814312],
       [-0.77246857, -1.49260948],
       [-1.12270621, -0.96747323],
       [ 0.27824435,  0.03278628],
       [ 0.36580376, -0.24228508],
       [-0.94758739, -1.61764191],
       [ 1.85431373,  1.00803931],
       [-0.50979034,  0.33286414],
       [ 1.41651668,  0.05779277],
       [-1.12270621, -1.84270031],
       [ 0.89116022, -0.34231103],
       [ 0.80360081, -0.74241484],
       [-0.24711211,  0.43289009],
       [-0.50979034,  0.23283819],
       [-1.0351468 ,  0.38287711],
       [ 0.97871963,  0.63294199],
       [ 1.41651668, -0.91746025],
       [-1.73562208, -1.26755108],
       [-1.47294385,  1.8582599 ],
       [-1.0351468 ,  0.93301985],
       [-0.77246857,  0.05779277],
       [-0.1595527 ,  0.4078836 ],
       [ 0.45336317,  0.10780575],
       [ 1.85431373,  0.58292902],
       [-0.68490916, -0.39232401],
       [ 0.89116022,  1.2330977 ],
       [ 1.32895727,  0.10780575],
       [-0.50979034, -0.29229806],
       [-1.73562208, -0.64238889],
       [ 0.80360081,  0.33286414],
       [-0.94758739,  0.73296794],
       [-0.07199329, -0.49234996],
       [ 0.89116022,  1.83325341],
       [ 0.19068494, -0.31730455],

```

```
[ -0.24711211,  0.35787063],
[  0.7160414 , -0.06723967],
[ -0.68490916,  0.15781872],
[  1.5016355 , -0.70770806]
```

```
x_test = scaler.fit_transform(x_test)
x_test
```

```
array([[ -0.97591834, -0.42941754],
       [-1.2495403 ,  0.51782704],
       [ 1.12185005, -1.72398513],
       [ 1.12185005, -0.58729164],
       [ 1.12185005,  0.35995294],
       [ 1.3042647 , -1.37666212],
       [-0.06384512,  1.40192198],
       [-1.06712566,  1.46507162],
       [-0.70229637,  0.2968033 ],
       [-0.61108905,  0.07577957]])
```

```
*TRAIN THE CLASSIFIER *
```

```
from sklearn.linear_model import LogisticRegression
```

```
clf = LogisticRegression()
```

```
clf.fit(x_train, y_train)
```

```
LogisticRegression()
```

```
y_pred = clf.predict(x_test)
y_pred
```

```
array([0, 0, 1, 1, 1, 1, 0, 0, 0, 0])
```

```
y_test
```

```
25  0
24  0
77  1
90  1
72  1
44  1
33  0
89  0
46  0
81  0
Name: placement, dtype: int64
```

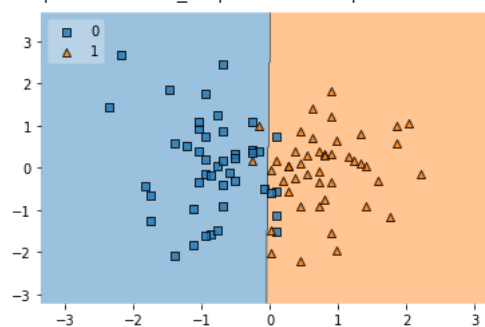
```
from sklearn.metrics import accuracy_score
```

```
accuracy_score(y_test,y_pred)
```

```
1.0
```

```
from mlxtend.plotting import plot_decision_regions
plot_decision_regions(x_train, y_train.values, clf = clf, legend=2)
```

```
/usr/local/lib/python3.8/dist-packages/mlxtend/plotting/decision_regions.py:244: MatplotlibDeprecationWarning:
  ax.axis(xmin=xx.min(), xmax=xx.max(), y_min=yy.min(), y_max=yy.max())
<matplotlib.axes._subplots.AxesSubplot at 0x7f9f1456efd0>
```



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