

```
In [1]: import pandas as pd
import pickle
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: a_train=pd.read_csv("C:\\Users\\reshma_koduri\\Downloads\\archive (2)\\train.csv")
```

```
In [3]: a_train.head(10)
```

```
Out[3]:
```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores
0	842	0	2.2	0	1	0	7	0.6	188	2
1	1021	1	0.5	1	0	1	53	0.7	136	3
2	563	1	0.5	1	2	1	41	0.9	145	5
3	615	1	2.5	0	0	0	10	0.8	131	6
4	1821	1	1.2	0	13	1	44	0.6	141	2
5	1859	0	0.5	1	3	0	22	0.7	164	1
6	1821	0	1.7	0	4	1	10	0.8	139	8
7	1954	0	0.5	1	0	0	24	0.8	187	4
8	1445	1	0.5	0	0	0	53	0.7	174	7
9	509	1	0.6	1	2	1	9	0.1	93	5

10 rows × 11 columns



```
In [4]: a_train.tail(10)
```

```
Out[4]:
```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_co
1990	1617	1	2.4	0	8	1	36	0.8	85	
1991	1882	0	2.0	0	11	1	44	0.8	113	
1992	674	1	2.9	1	1	0	21	0.2	198	
1993	1467	1	0.5	0	0	0	18	0.6	122	
1994	858	0	2.2	0	1	0	50	0.1	84	
1995	794	1	0.5	1	0	1	2	0.8	106	
1996	1965	1	2.6	1	0	0	39	0.2	187	
1997	1911	0	0.9	1	1	1	36	0.7	108	
1998	1512	0	0.9	0	4	1	46	0.1	145	
1999	510	1	2.0	1	5	1	45	0.9	168	

10 rows × 11 columns



In [5]: `a_train.describe()`

Out[5]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	
count	2000.000000	2000.0000	2000.000000	2000.000000	2000.000000	2000.000000	2000.000000	2
mean	1238.518500	0.4950	1.522250	0.509500	4.309500	0.521500	32.046500	
std	439.418206	0.5001	0.816004	0.500035	4.341444	0.499662	18.145715	
min	501.000000	0.0000	0.500000	0.000000	0.000000	0.000000	2.000000	
25%	851.750000	0.0000	0.700000	0.000000	1.000000	0.000000	16.000000	
50%	1226.000000	0.0000	1.500000	1.000000	3.000000	1.000000	32.000000	
75%	1615.250000	1.0000	2.200000	1.000000	7.000000	1.000000	48.000000	
max	1998.000000	1.0000	3.000000	1.000000	19.000000	1.000000	64.000000	

8 rows × 21 columns

In [6]: `a_train.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):
#   Column          Non-Null Count  Dtype
---  -
0   battery_power    2000 non-null   int64
1   blue             2000 non-null   int64
2   clock_speed      2000 non-null   float64
3   dual_sim         2000 non-null   int64
4   fc               2000 non-null   int64
5   four_g          2000 non-null   int64
6   int_memory       2000 non-null   int64
7   m_dep           2000 non-null   float64
8   mobile_wt        2000 non-null   int64
9   n_cores          2000 non-null   int64
10  pc               2000 non-null   int64
11  px_height        2000 non-null   int64
12  px_width         2000 non-null   int64
13  ram              2000 non-null   int64
14  sc_h             2000 non-null   int64
15  sc_w             2000 non-null   int64
16  talk_time        2000 non-null   int64
17  three_g          2000 non-null   int64
18  touch_screen     2000 non-null   int64
19  wifi             2000 non-null   int64
20  price_range      2000 non-null   int64
dtypes: float64(2), int64(19)
memory usage: 328.2 KB
```

In [7]: `list(a_train)`

Out[7]:

```
['battery_power',
 'blue',
 'clock_speed',
 'dual_sim',
 'fc',
 'four_g',
```

```
'int_memory',
'm_dep',
'mobile_wt',
'n_cores',
'pc',
'px_height',
'px_width',
'ram',
'sc_h',
'sc_w',
'talk_time',
'three_g',
'touch_screen',
'wifi',
'price_range']
```

```
In [8]: a_train.isna().sum()
```

Out[8]:

battery_power	0
blue	0
clock_speed	0
dual_sim	0
fc	0
four_g	0
int_memory	0
m_dep	0
mobile_wt	0
n_cores	0
pc	0
px_height	0
px_width	0
ram	0
sc_h	0
sc_w	0
talk_time	0
three_g	0
touch_screen	0
wifi	0
price_range	0

dtype: int64

```
In [9]: a_train['price_range']=a_train['price_range'].map({0:0,1:0,2:1,3:1})
a_train
```

Out[9]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_co
0	842	0	2.2	0	1	0	7	0.6	188	
1	1021	1	0.5	1	0	1	53	0.7	136	
2	563	1	0.5	1	2	1	41	0.9	145	
3	615	1	2.5	0	0	0	10	0.8	131	
4	1821	1	1.2	0	13	1	44	0.6	141	
...
1995	794	1	0.5	1	0	1	2	0.8	106	
1996	1965	1	2.6	1	0	0	39	0.2	187	
1997	1911	0	0.9	1	1	1	36	0.7	108	
1998	1512	0	0.9	0	4	1	46	0.1	145	

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_co
1999	510	1	2.0	1	5	1	45	0.9	168	

2000 rows × 21 columns

```
In [10]: b=a_train.drop(['n_cores','sc_h','sc_w','m_dep','clock_speed','mobile_wt','px_width'])
```

```
In [11]: b['price_range'].unique()
```

Out[11]: array([0, 1], dtype=int64)

```
In [12]: b.groupby(['price_range']).count()
```

	battery_power	blue	dual_sim	fc	four_g	int_memory	ram	talk_time	three_g	touch_scre
price_range										
0	1000	1000	1000	1000	1000	1000	1000	1000	1000	1000
1	1000	1000	1000	1000	1000	1000	1000	1000	1000	1000



```
In [13]: y=b['price_range']
y
```

Out[13]: 0 0
1 1
2 1
3 1
4 0
..
1995 0
1996 1
1997 1
1998 0
1999 1
Name: price_range, Length: 2000, dtype: int64

```
In [14]: x=b.drop(['price_range'],axis=1)
x
```

	battery_power	blue	dual_sim	fc	four_g	int_memory	ram	talk_time	three_g	touch_scre
0	842	0	0	1	0	7	2549	19	0	
1	1021	1	1	0	1	53	2631	7	1	
2	563	1	1	2	1	41	2603	9	1	
3	615	1	0	0	0	10	2769	11	1	
4	1821	1	0	13	1	44	1411	15	1	
...
1995	794	1	1	0	1	2	668	19	1	

	battery_power	blue	dual_sim	fc	four_g	int_memory	ram	talk_time	three_g	touch_screen
1996	1965	1	1	0	0	39	2032	16	1	
1997	1911	0	1	1	1	36	3057	5	1	
1998	1512	0	0	4	1	46	869	19	1	
1999	510	1	1	5	1	45	3919	2	1	

2000 rows × 11 columns

```
In [15]: 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=42)
```

```
In [16]: from sklearn.linear_model import LogisticRegression
reg=LogisticRegression()
reg.fit(x_train,y_train)
```

```
Out[16]: LogisticRegression()
```

```
In [17]: ypred=reg.predict(x_test)
ypred
```

```
Out[17]: array([0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0,
1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0,
1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0,
1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1,
1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0,
0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0,
0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0,
1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0,
1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0,
1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0,
1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1,
1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1,
1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1,
0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1,
1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1,
1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0,
0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1,
1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1,
1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0,
0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0,
1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0,
1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1,
0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0], dtype=int64)
```

```
In [18]: from sklearn.metrics import confusion_matrix
confusion_matrix(y_test,ypred)
```

```
Out[18]: array([[213,  37],
[ 20, 230]], dtype=int64)
```

```
In [19]: from sklearn.metrics import accuracy_score
accuracy_score(ypred,y_test)
```

Out[19]: 0.886

```
In [20]: from sklearn.model_selection import GridSearchCV
from sklearn.ensemble import RandomForestClassifier
reg=RandomForestClassifier()
n_estimators=[25,50,75,100,125,150,175,200]
criterion=['gini','entropy']
max_depth=[3,5,10]
parameters={'n_estimators': n_estimators,'criterion':criterion,'max_depth':max_depth}
rfc_reg = GridSearchCV(reg, parameters)
rfc_reg.fit(x_train,y_train)
```

Out[20]: GridSearchCV(estimator=RandomForestClassifier(),
param_grid={'criterion': ['gini', 'entropy'],
'max_depth': [3, 5, 10],
'n_estimators': [25, 50, 75, 100, 125, 150, 175, 200]})

```
In [25]: rfc_reg.best_params_
```

Out[25]: {'criterion': 'gini', 'max_depth': 10, 'n_estimators': 100}

```
In [26]: reg=RandomForestClassifier(n_estimators=100,criterion='gini',max_depth=10)
reg.fit(x_train,y_train)
```

Out[26]: RandomForestClassifier(max_depth=10)

```
In [27]: ypred=reg.predict(x_test)
ypred
```

Out[27]: array([0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0,
0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0,
1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0,
1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1,
1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0,
0, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0,
1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0,
1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0,
1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0,
1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0,
1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1,
1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1,
0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1,
1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1,
1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0,
0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1,
0, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1,
0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1,
0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0,
1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0,
1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1,
0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0], dtype=int64)

```
In [28]: from sklearn.metrics import accuracy_score
accuracy_score(y_test,ypred)
```

Out[28]: 0.936

In []: