```
In [2]: 1 import pandas as pd
2 import numpy as np
3 import seaborn as sns
4 import matplotlib.pyplot as plt

In [3]: 1 mydata=pd.read_csv("mobile_price.csv")

In [4]: 1 mydata
```

Out[4]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	 px_height	px_width	ram
0	842	0	2.2	0	1	0	7	0.6	188	2	 20	756	2549
1	1021	1	0.5	1	0	1	53	0.7	136	3	 905	1988	2631
2	563	1	0.5	1	2	1	41	0.9	145	5	 1263	1716	2603
3	615	1	2.5	0	0	0	10	0.8	131	6	 1216	1786	2769
4	1821	1	1.2	0	13	1	44	0.6	141	2	 1208	1212	1411
1995	794	1	0.5	1	0	1	2	8.0	106	6	 1222	1890	668
1996	1965	1	2.6	1	0	0	39	0.2	187	4	 915	1965	2032
1997	1911	0	0.9	1	1	1	36	0.7	108	8	 868	1632	3057
1998	1512	0	0.9	0	4	1	46	0.1	145	5	 336	670	869
1999	510	1	2.0	1	5	1	45	0.9	168	6	 483	754	3919

2000 rows × 21 columns

4

```
In [5]: 1 mydata.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):

Data	cordinis (cocar	ZI COIUMIIS).	
#	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 <u>g</u>	2000 non-null	int64
6	int_memory	2000 non-null	int64
7	m_dep	2000 non-null	float64
8	<pre>mobile_wt</pre>	2000 non-null	int64
9	n_cores	2000 non-null	int64
10	рс	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
1 5	SC_W	2000 non-null	int64
16	talk_time	2000 non-null	int64
17	three <u></u> 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
dtype	es: float64(2),	int64(19)	
memor	ry usage: 328.2	KB	

In [6]:

1 mydata.describe()

Out[6]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_core
count	2000.000000	2000.0000	2000.000000	2000.000000	2000.000000	2000.000000	2000.000000	2000.000000	2000.000000	2000.00000
mean	1238.518500	0.4950	1.522250	0.509500	4.309500	0.521500	32.046500	0.501750	140.249000	4.52050
std	439.418206	0.5001	0.816004	0.500035	4.341444	0.499662	18.145715	0.288416	35.399655	2.28783
min	501.000000	0.0000	0.500000	0.000000	0.000000	0.000000	2.000000	0.100000	80.000000	1.00000
25%	851.750000	0.0000	0.700000	0.000000	1.000000	0.000000	16.000000	0.200000	109.000000	3.00000
50%	1226.000000	0.0000	1.500000	1.000000	3.000000	1.000000	32.000000	0.500000	141.000000	4.00000
75%	1615.250000	1.0000	2.200000	1.000000	7.000000	1.000000	48.000000	0.800000	170.000000	7.00000
max	1998.000000	1.0000	3.000000	1.000000	19.000000	1.000000	64.000000	1.000000	200.000000	8.00000

8 rows × 21 columns

- -

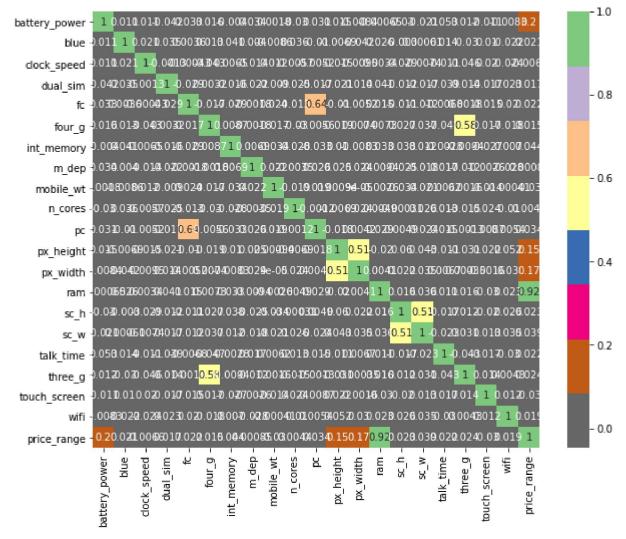
```
1 mydata.isnull().sum()
In [7]:
Out[7]: battery_power
                         0
        blue
                         0
        clock_speed
                         0
        dual_sim
                         0
        fc
                         0
        four_g
                         0
        int_memory
                         0
        m_dep
                         0
        mobile_wt
        n_cores
        рс
                         0
        px_height
        px_width
        ram
        sc_h
                         0
        SC_W
                         0
        talk_time
        three_g
        touch_screen
                         0
        wifi
                         0
        price_range
                         0
        dtype: int64
          1 mydata_corr=mydata.corr()
In [8]:
```

In [9]: 1 mydata_corr

Out[9]:

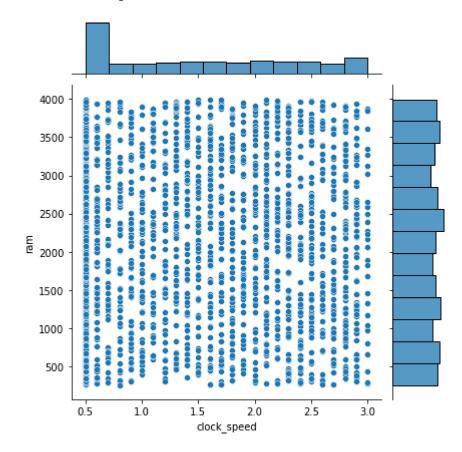
	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	1
battery_power	1.000000	0.011252	0.011482	-0.041847	0.033334	0.015665	-0.004004	0.034085	0.001844	-0.029727	
blue	0.011252	1.000000	0.021419	0.035198	0.003593	0.013443	0.041177	0.004049	-0.008605	0.036161	
clock_speed	0.011482	0.021419	1.000000	-0.001315	-0.000434	-0.043073	0.006545	-0.014364	0.012350	-0.005724	
dual_sim	-0.041847	0.035198	-0.001315	1.000000	-0.029123	0.003187	-0.015679	-0.022142	-0.008979	-0.024658	
fc	0.033334	0.003593	-0.000434	-0.029123	1.000000	-0.016560	-0.029133	-0.001791	0.023618	-0.013356	
four_g	0.015665	0.013443	-0.043073	0.003187	-0.016560	1.000000	0.008690	-0.001823	-0.016537	-0.029706	
int_memory	-0.004004	0.041177	0.006545	-0.015679	-0.029133	0.008690	1.000000	0.006886	-0.034214	-0.028310	
m_dep	0.034085	0.004049	-0.014364	-0.022142	-0.001791	-0.001823	0.006886	1.000000	0.021756	-0.003504	
mobile_wt	0.001844	-0.008605	0.012350	-0.008979	0.023618	-0.016537	-0.034214	0.021756	1.000000	-0.018989	
n_cores	-0.029727	0.036161	-0.005724	-0.024658	-0.013356	-0.029706	-0.028310	-0.003504	-0.018989	1.000000	
рс	0.031441	-0.009952	-0.005245	-0.017143	0.644595	-0.005598	-0.033273	0.026282	0.018844	-0.001193	
px_height	0.014901	-0.006872	-0.014523	-0.020875	-0.009990	-0.019236	0.010441	0.025263	0.000939	-0.006872	
px_width	-0.008402	-0.041533	-0.009476	0.014291	-0.005176	0.007448	-0.008335	0.023566	0.000090	0.024480	
ram	-0.000653	0.026351	0.003443	0.041072	0.015099	0.007313	0.032813	-0.009434	-0.002581	0.004868	
sc_h	-0.029959	-0.002952	-0.029078	-0.011949	-0.011014	0.027166	0.037771	-0.025348	-0.033855	-0.000315	
sc_w	-0.021421	0.000613	-0.007378	-0.016666	-0.012373	0.037005	0.011731	-0.018388	-0.020761	0.025826	
talk_time	0.052510	0.013934	-0.011432	-0.039404	-0.006829	-0.046628	-0.002790	0.017003	0.006209	0.013148	
three_g	0.011522	-0.030236	-0.046433	-0.014008	0.001793	0.584246	-0.009366	-0.012065	0.001551	-0.014733	
touch_screen	-0.010516	0.010061	0.019756	-0.017117	-0.014828	0.016758	-0.026999	-0.002638	-0.014368	0.023774	
wifi	-0.008343	-0.021863	-0.024471	0.022740	0.020085	-0.017620	0.006993	-0.028353	-0.000409	-0.009964	
price_range	0.200723	0.020573	-0.006606	0.017444	0.021998	0.014772	0.044435	0.000853	-0.030302	0.004399	

21 rows × 21 columns



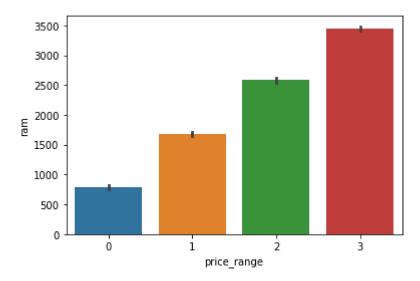
In [11]: 1 sns.jointplot(x='clock_speed', y='ram', data=mydata)

Out[11]: <seaborn.axisgrid.JointGrid at 0x20608862580>

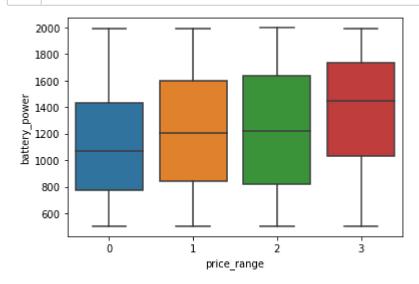


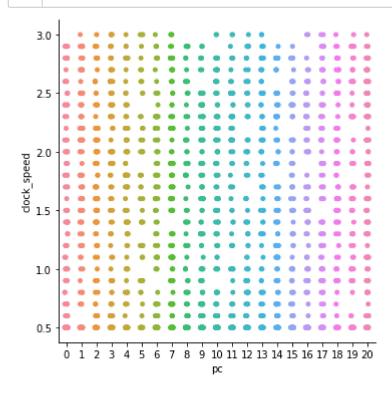
In [12]: 1 sns.barplot(x='price_range', y='ram', data=mydata)

Out[12]: <AxesSubplot:xlabel='price_range', ylabel='ram'>



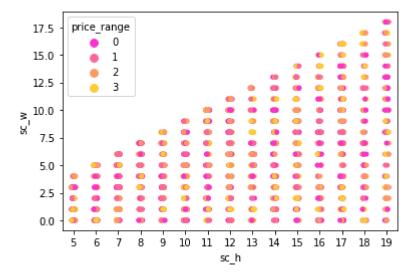
In [13]: 1 sns.boxplot(x='price_range', y='battery_power', data=mydata);





```
In [17]: 1 sns.stripplot(x='sc_h',y='sc_w', data=mydata,palette='spring',hue='price_range')
```

Out[17]: <AxesSubplot:xlabel='sc_h', ylabel='sc_w'>



```
In [18]: 1 x_ind=mydata.drop('price_range', axis=1)
```

In [19]: 1 x_ind

Out[19]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	рс	px_height	px_width	ram
0	842	0	2.2	0	1	0	7	0.6	188	2	2	20	756	2549
1	1021	1	0.5	1	0	1	53	0.7	136	3	6	905	1988	2631
2	563	1	0.5	1	2	1	41	0.9	145	5	6	1263	1716	2603
3	615	1	2.5	0	0	0	10	0.8	131	6	9	1216	1786	2769
4	1821	1	1.2	0	13	1	44	0.6	141	2	14	1208	1212	1411
1995	794	1	0.5	1	0	1	2	0.8	106	6	14	1222	1890	668
1996	1965	1	2.6	1	0	0	39	0.2	187	4	3	915	1965	2032
1997	1911	0	0.9	1	1	1	36	0.7	108	8	3	868	1632	3057
1998	1512	0	0.9	0	4	1	46	0.1	145	5	5	336	670	869
1999	510	1	2.0	1	5	1	45	0.9	168	6	16	483	754	3919

2000 rows × 20 columns

In [20]: 1 y_dep=mydata.price_range

```
In [21]:
           1 y_dep
Out[21]: 0
                 1
                 2
         2
                 2
         3
                 2
         4
                 1
         1995
                 0
         1996
                 2
         1997
                 3
         1998
                 0
         1999
         Name: price_range, Length: 2000, dtype: int64
           1 from sklearn.model_selection import train_test_split
In [22]:
In [23]:
           1 x_train,x_test,y_train,y_test=train_test_split(x_ind,y_dep,train_size=0.8,random_state=2)
In [24]:
           1 from sklearn.naive_bayes import GaussianNB
           1 model=GaussianNB()
In [25]:
           1 model.fit(x_train,y_train)
In [27]:
Out[27]:
         GaussianNB()
           1 y_pred=model.predict(x_test)
In [28]:
In [29]:
           1 from sklearn.metrics import confusion_matrix, accuracy_score
```