

```

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

training=pd.read_csv(r'mobile_price_range_data.csv')
test=pd.read_csv(r'test.csv')

pd.set_option('display.max_rows',None)
pd.set_option('display.max_columns',None)

training.head()

```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory
m_dep \							
0	842	0	2.2	0	1	0	7
0.6							
1	1021	1	0.5	1	0	1	53
0.7							
2	563	1	0.5	1	2	1	41
0.9							
3	615	1	2.5	0	0	0	10
0.8							
4	1821	1	1.2	0	13	1	44
0.6							

	mobile_wt	n_cores	pc	px_height	px_width	ram	sc_h	sc_w
talk_time \								
0	188	2	2	20	756	2549	9	7
19								
1	136	3	6	905	1988	2631	17	3
7								
2	145	5	6	1263	1716	2603	11	2
9								
3	131	6	9	1216	1786	2769	16	8
11								
4	141	2	14	1208	1212	1411	8	2
15								

	three_g	touch_screen	wifi	price_range
0	0		0	1
1	1		1	0
2	1		1	0
3	1		0	0
4	1		1	0

```
test.head()
```

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g
int_memory \							
0	1	1043	1	1.8	1	14	0
5							

1	2	841	1	0.5	1	4	1
61							
2	3	1807	1	2.8	0	1	0
27							
3	4	1546	0	0.5	1	18	1
25							
4	5	1434	0	1.4	0	11	1
49							

	m_dep	mobile_wt	n_cores	pc	px_height	px_width	ram	sc_h
sc_w \								
0	0.1	193	3	16	226	1412	3476	12
7								
1	0.8	191	5	12	746	857	3895	6
0								
2	0.9	186	3	4	1270	1366	2396	17
10								
3	0.5	96	8	20	295	1752	3893	10
0								
4	0.5	108	6	18	749	810	1773	15
8								

	talk_time	three_g	touch_screen	wifi
0	2	0	1	0
1	7	1	0	0
2	10	0	1	1
3	7	1	1	0
4	7	1	0	1

```
test.drop('id',axis=1,inplace=True)
```

```
test.head()
```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory
m_dep \							
0	1043	1	1.8	1	14	0	5
0.1							
1	841	1	0.5	1	4	1	61
0.8							
2	1807	1	2.8	0	1	0	27
0.9							
3	1546	0	0.5	1	18	1	25
0.5							
4	1434	0	1.4	0	11	1	49
0.5							

	mobile_wt	n_cores	pc	px_height	px_width	ram	sc_h	sc_w
talk_time \								
0	193	3	16	226	1412	3476	12	7
2								

1	191	5	12	746	857	3895	6	0
7								
2	186	3	4	1270	1366	2396	17	10
10								
3	96	8	20	295	1752	3893	10	0
7								
4	108	6	18	749	810	1773	15	8
7								

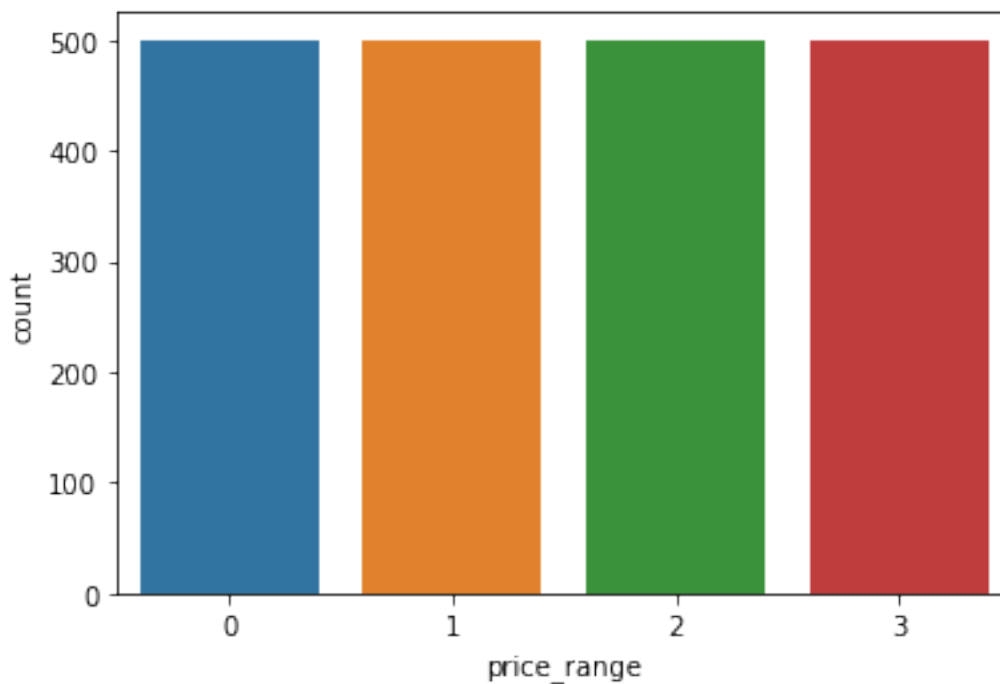
	three_g	touch_screen	wifi
0	0	1	0
1	1	0	0
2	0	1	1
3	1	1	0
4	1	0	1

```
sns.countplot(training['price_range'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
<AxesSubplot:xlabel='price_range', ylabel='count'>
```



```
training.shape, test.shape
```

```
((2000, 21), (1000, 20))
```

```
training.isnull().sum()
```

```
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
```

```
training.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

```

```
test.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   battery_power         1000 non-null   int64
1   blue                  1000 non-null   int64
2   clock_speed           1000 non-null   float64
3   dual_sim              1000 non-null   int64
4   fc                    1000 non-null   int64
5   four_g               1000 non-null   int64
6   int_memory            1000 non-null   int64
7   m_dep                 1000 non-null   float64
8   mobile_wt             1000 non-null   int64
9   n_cores               1000 non-null   int64
10  pc                    1000 non-null   int64
11  px_height             1000 non-null   int64
12  px_width              1000 non-null   int64
13  ram                   1000 non-null   int64
14  sc_h                  1000 non-null   int64
15  sc_w                  1000 non-null   int64
16  talk_time             1000 non-null   int64
17  three_g               1000 non-null   int64
18  touch_screen          1000 non-null   int64
19  wifi                  1000 non-null   int64
dtypes: float64(2), int64(18)
memory usage: 156.4 KB

```

```
training.describe()
```

	battery_power	blue	clock_speed	dual_sim	fc
\count	2000.000000	2000.0000	2000.000000	2000.000000	2000.000000
mean	1238.518500	0.4950	1.522250	0.509500	4.309500
std	439.418206	0.5001	0.816004	0.500035	4.341444
min	501.000000	0.0000	0.500000	0.000000	0.000000
25%	851.750000	0.0000	0.700000	0.000000	1.000000
50%	1226.000000	0.0000	1.500000	1.000000	3.000000

75%	1615.250000	1.0000	2.200000	1.000000	7.000000
max	1998.000000	1.0000	3.000000	1.000000	19.000000

	four_g	int_memory	m_dep	mobile_wt	n_cores
\					
count	2000.000000	2000.000000	2000.000000	2000.000000	2000.000000
mean	0.521500	32.046500	0.501750	140.249000	4.520500
std	0.499662	18.145715	0.288416	35.399655	2.287837
min	0.000000	2.000000	0.100000	80.000000	1.000000
25%	0.000000	16.000000	0.200000	109.000000	3.000000
50%	1.000000	32.000000	0.500000	141.000000	4.000000
75%	1.000000	48.000000	0.800000	170.000000	7.000000
max	1.000000	64.000000	1.000000	200.000000	8.000000

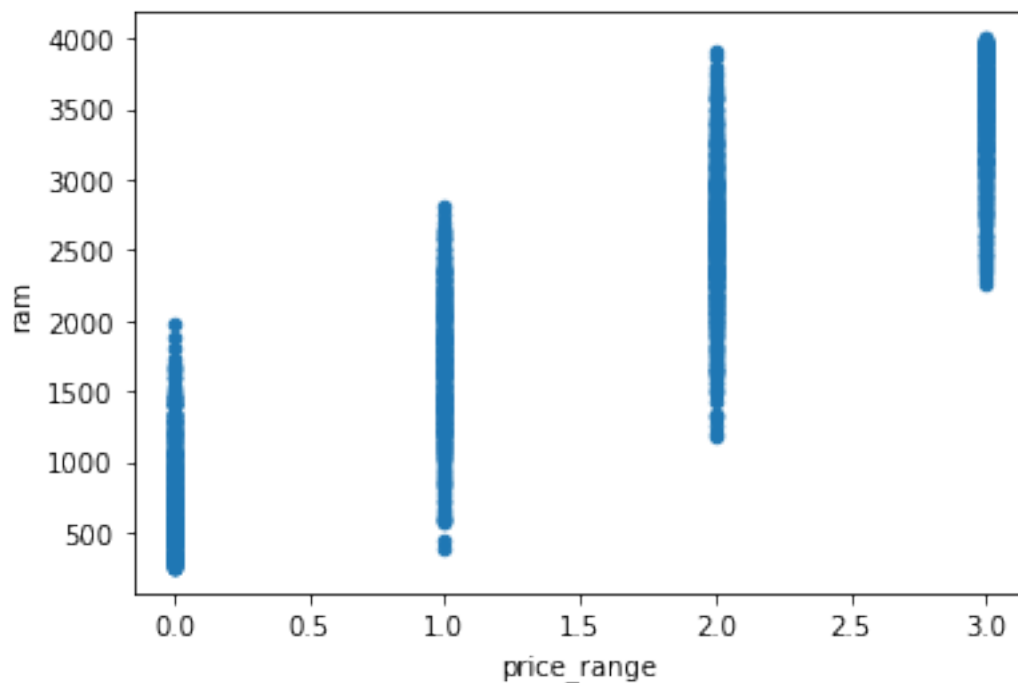
	pc	px_height	px_width	ram	sc_h
\					
count	2000.000000	2000.000000	2000.000000	2000.000000	2000.000000
mean	9.916500	645.108000	1251.515500	2124.213000	12.306500
std	6.064315	443.780811	432.199447	1084.732044	4.213245
min	0.000000	0.000000	500.000000	256.000000	5.000000
25%	5.000000	282.750000	874.750000	1207.500000	9.000000
50%	10.000000	564.000000	1247.000000	2146.500000	12.000000
75%	15.000000	947.250000	1633.000000	3064.500000	16.000000
max	20.000000	1960.000000	1998.000000	3998.000000	19.000000

	sc_w	talk_time	three_g	touch_screen
wifi \				
count	2000.000000	2000.000000	2000.000000	2000.000000
2000.000000				

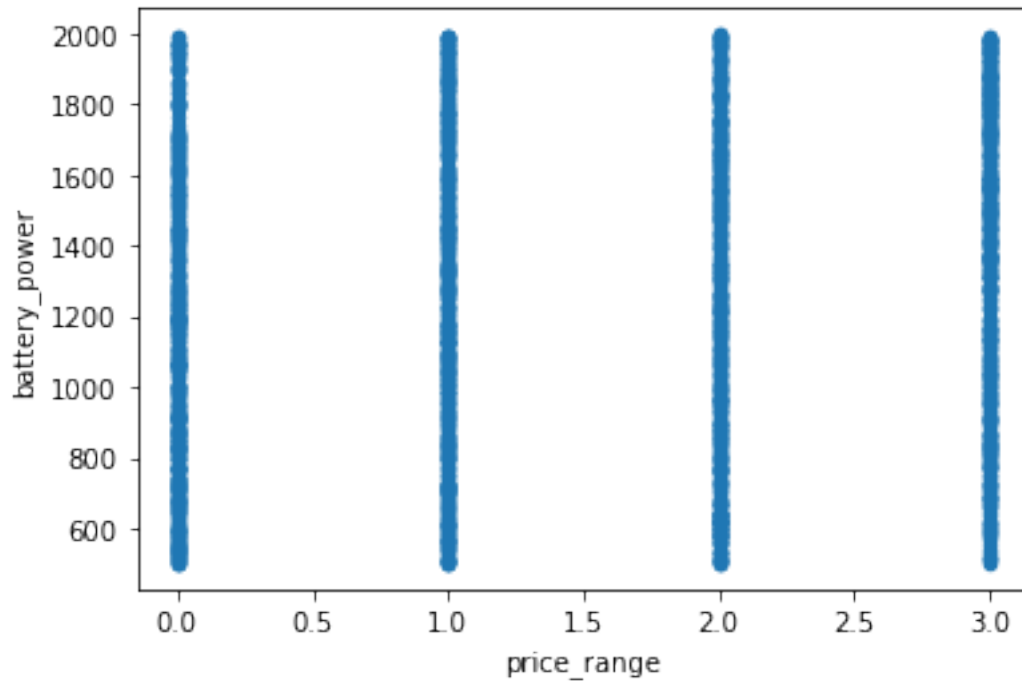
mean	5.767000	11.011000	0.761500	0.503000
std	4.356398	5.463955	0.426273	0.500116
min	0.000000	2.000000	0.000000	0.000000
25%	2.000000	6.000000	1.000000	0.000000
50%	5.000000	11.000000	1.000000	1.000000
75%	9.000000	16.000000	1.000000	1.000000
max	18.000000	20.000000	1.000000	1.000000

	price_range
count	2000.000000
mean	1.500000
std	1.118314
min	0.000000
25%	0.750000
50%	1.500000
75%	2.250000
max	3.000000

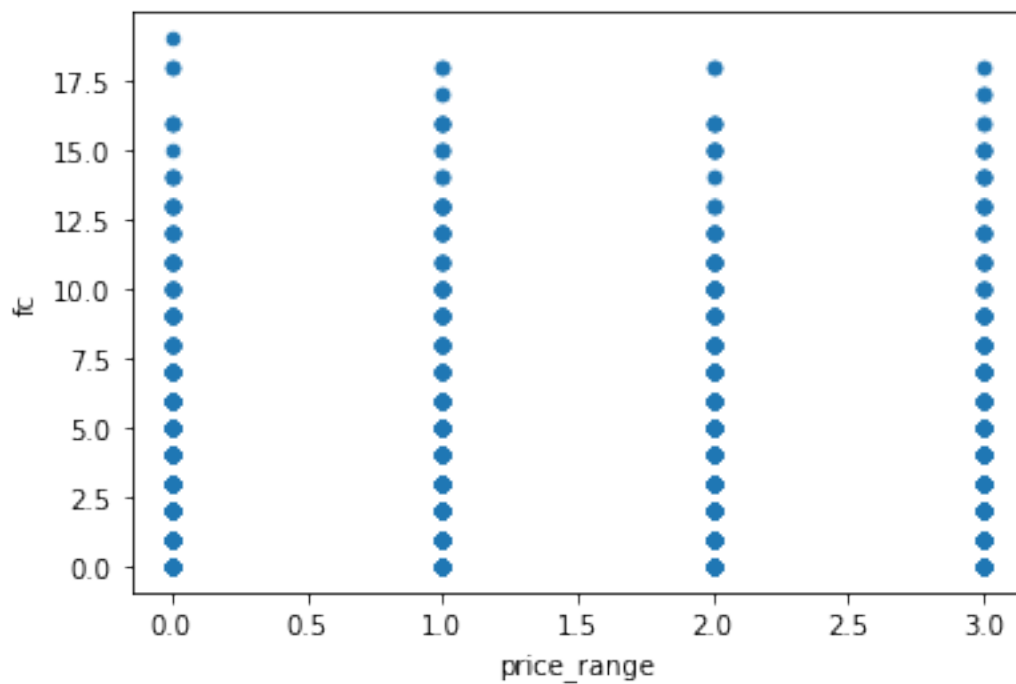
```
training.plot(x='price_range',y='ram',kind='scatter')
plt.show()
```



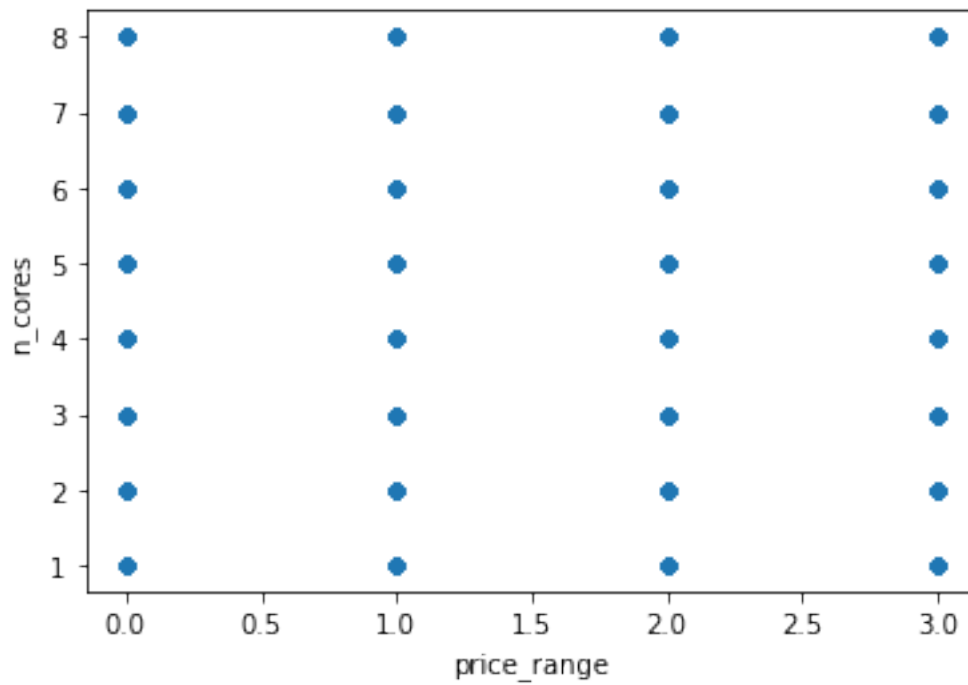
```
training.plot(x='price_range',y='battery_power',kind='scatter')  
plt.show()
```



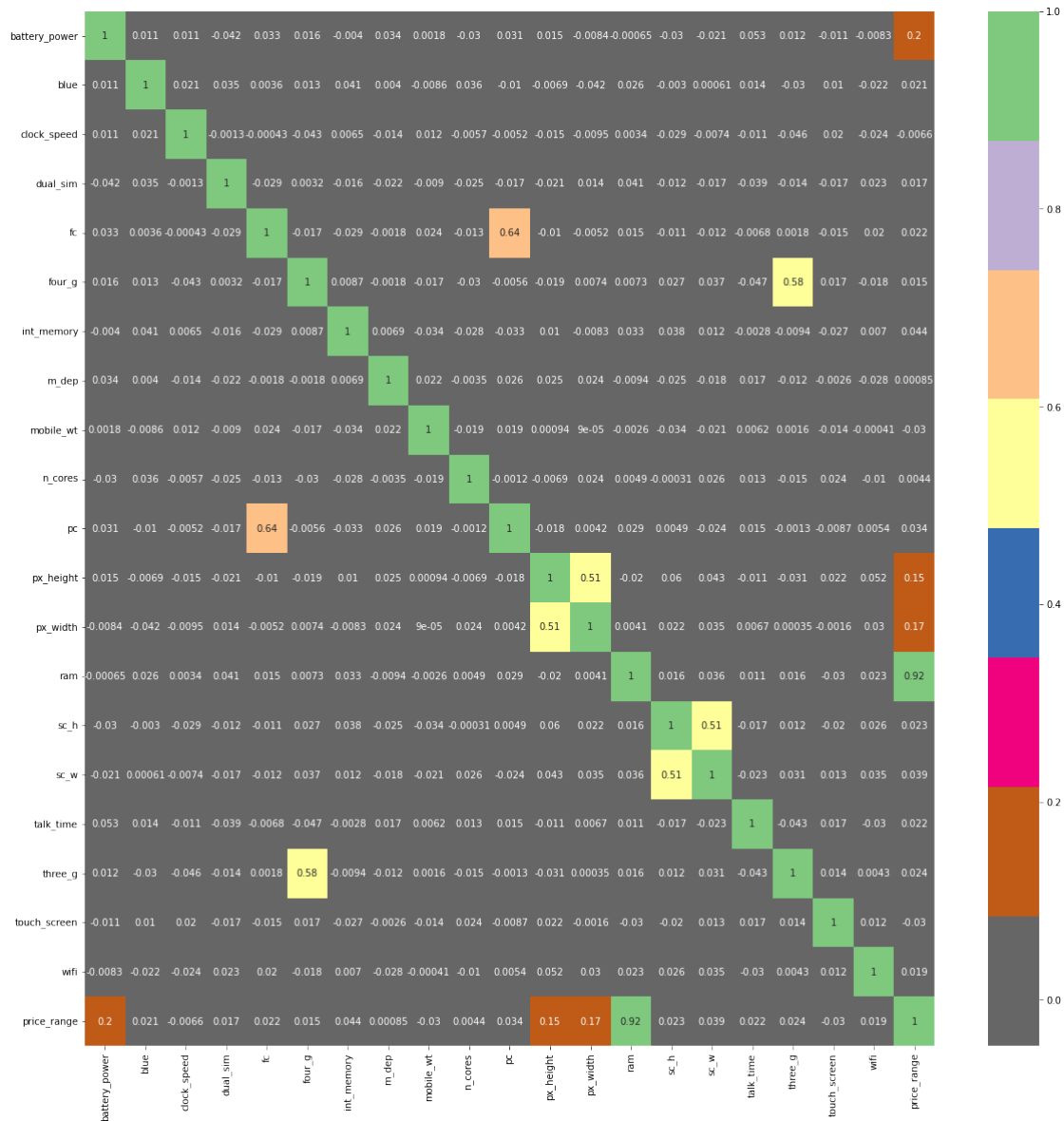
```
training.plot(x='price_range',y='fc',kind='scatter')  
plt.show()
```



```
training.plot(x='price_range',y='n_cores',kind='scatter')  
plt.show()
```

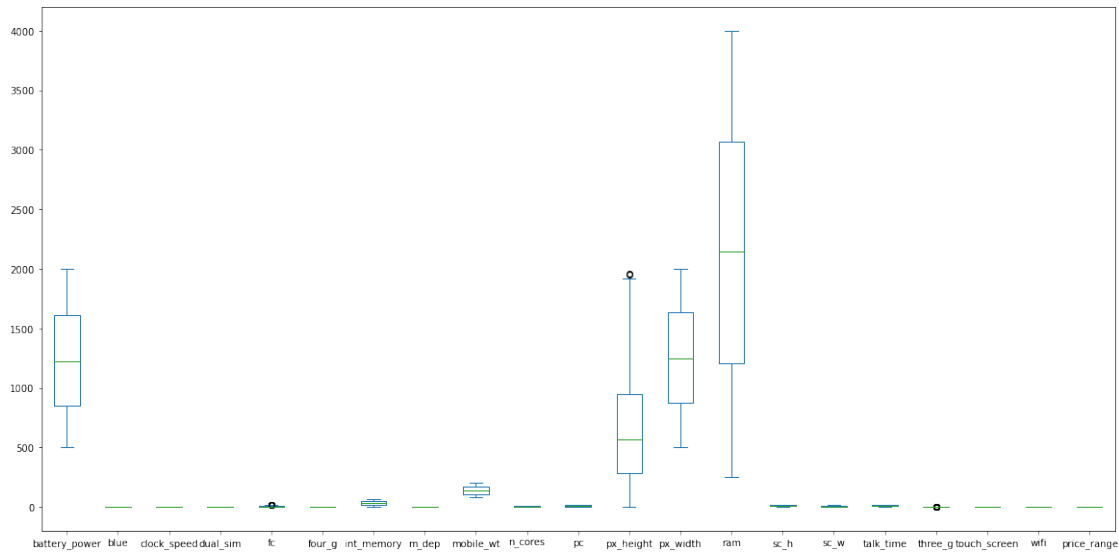



```
import seaborn as sns
plt.figure(figsize=(20,20))
sns.heatmap(training.corr(),annot=True,cmap=plt.cm.Accent_r)
plt.show()
```



```
training.plot(kind='box',figsize=(20,10))
```

<AxesSubplot:>



```
X = training.drop('price_range',axis=1)
y = training['price_range']
```

```
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test =
train_test_split(X,y,test_size=0.1,random_state=101)
```

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
test = sc.transform(test)
```

X_train

```
array([[ -1.62737257,  -0.98675438,  -1.01271559, ...,  -1.78222729,
        -1.00892875,  -0.99888951],
       [ -0.75199354,   1.01342342,   0.58093235, ...,  -1.78222729,
         0.99115027,  -0.99888951],
       [ -0.20630271,   1.01342342,   0.70352065, ...,   0.56109566,
        -1.00892875,   1.00111173],
       ...,
       [  0.69636086,   1.01342342,  -0.03200917, ...,   0.56109566,
        -1.00892875,  -0.99888951],
       [  0.83733099,  -0.98675438,  -1.2578922 , ...,   0.56109566,
        -1.00892875,   1.00111173],
       [  0.4144206 ,  -0.98675438,  -0.39977408, ...,   0.56109566,
         0.99115027,   1.00111173]])
```

X_test

```
array([[ 0.28481903,  -0.98675438,  -1.2578922 , ...,   0.56109566,
        -1.00892875,  -0.99888951],
       [-1.44092821,  -0.98675438,  -1.2578922 , ...,   0.56109566,
         0.99115027,   1.00111173],
```

```

[-1.49322358, -0.98675438, -0.15459747, ..., 0.56109566,
 -1.00892875, 1.00111173],
...,
[-0.55418061, 1.01342342, 0.33575574, ..., 0.56109566,
 -1.00892875, -0.99888951],
[ 0.09610095, -0.98675438, -0.89012729, ..., 0.56109566,
 0.99115027, 1.00111173],
[-1.60690917, -0.98675438, 1.07128556, ..., 0.56109566,
 0.99115027, -0.99888951]])

test
array([[ -0.4541373,  1.01342342,  0.33575574, ..., -1.78222729,
         0.99115027, -0.99888951],
       [ -0.91342707,  1.01342342, -1.2578922 , ...,  0.56109566,
        -1.00892875, -0.99888951],
       [  1.2829785 ,  1.01342342,  1.56163877, ..., -1.78222729,
         0.99115027,  1.00111173],
       ...,
       [ -0.13127022, -0.98675438, -0.15459747, ...,  0.56109566,
        -1.00892875, -0.99888951],
       [  0.65998148,  1.01342342, -1.2578922 , ..., -1.78222729,
         0.99115027, -0.99888951],
       [  0.06199528,  1.01342342, -1.2578922 , ...,  0.56109566,
        -1.00892875,  1.00111173]])

from sklearn.tree import DecisionTreeClassifier
dtc = DecisionTreeClassifier()
dtc.fit(X_train , Y_train)

DecisionTreeClassifier()

p = dtc.predict(X_test)
p
array([1, 1, 2, 1, 0, 1, 2, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 3,
1,
      2, 3, 2, 2, 2, 1, 0, 0, 2, 3, 0, 0, 3, 0, 0, 0, 1, 1, 1, 2, 3,
2,
      2, 1, 1, 3, 3, 1, 0, 0, 2, 3, 3, 2, 0, 3, 3, 3, 2, 2, 3, 1, 3,
2,
      0, 1, 0, 2, 1, 2, 3, 2, 2, 3, 3, 2, 0, 2, 0, 0, 2, 1, 2, 2, 2,
2,
      0, 0, 3, 3, 0, 2, 0, 3, 2, 0, 2, 3, 0, 1, 2, 3, 0, 2, 0, 0, 2,
0,
      1, 0, 3, 2, 2, 2, 1, 3, 2, 0, 3, 3, 1, 3, 1, 3, 3, 2, 1, 1, 1,
0,
      1, 1, 0, 2, 3, 0, 2, 3, 2, 3, 0, 1, 0, 0, 1, 3, 2, 0, 2, 1, 3,
2,
      3, 2, 2, 0, 3, 1, 2, 2, 2, 2, 1, 2, 1, 1, 3, 3, 1, 2, 0, 3, 1,
3,

```

```

1, 2, 3, 1, 2, 1, 0, 1, 3, 2, 1, 2, 1, 3, 1, 0, 2, 2, 0, 3, 0,
0,
3, 0], dtype=int64)

```

```

from sklearn.metrics import accuracy_score, confusion_matrix
dtc_acc = accuracy_score(p,Y_test)
print(dtc_acc)
print(confusion_matrix(p,Y_test))

```

```

0.845
[[43  3  0  0]
 [ 7 38  7  0]
 [ 0  5 49  3]
 [ 0  0  6 39]]

```

```

from sklearn.svm import SVC
knn=SVC()
knn.fit(X_train,Y_train)

```

```

SVC()

```

```

pred1 = knn.predict(X_test)
pred1

```

```

array([1, 1, 2, 1, 1, 1, 2, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 3,
1,
2, 3, 2, 2, 2, 2, 0, 0, 2, 3, 0, 0, 3, 0, 0, 0, 1, 1, 1, 1, 3,
2,
3, 0, 2, 3, 3, 1, 0, 1, 2, 3, 2, 2, 0, 3, 2, 3, 2, 2, 3, 1, 3,
1,
0, 1, 0, 2, 1, 2, 3, 2, 1, 3, 3, 2, 1, 2, 0, 0, 2, 2, 2, 2, 2,
1,
0, 0, 3, 2, 0, 2, 0, 3, 2, 0, 2, 3, 0, 1, 3, 3, 0, 3, 0, 0, 2,
0,
1, 0, 3, 2, 1, 1, 1, 3, 1, 0, 3, 2, 2, 3, 1, 2, 3, 2, 1, 1, 1,
0,
0, 1, 0, 1, 3, 0, 2, 3, 1, 3, 0, 0, 0, 1, 1, 3, 2, 0, 2, 0, 2,
2,
3, 2, 2, 0, 3, 2, 2, 2, 1, 2, 1, 2, 1, 0, 3, 3, 1, 2, 0, 3, 1,
3,
2, 2, 3, 2, 1, 1, 0, 1, 2, 2, 2, 2, 0, 3, 1, 0, 2, 2, 0, 2, 0,
0,
3, 0], dtype=int64)

```

```

from sklearn.metrics import accuracy_score
svc_acc = accuracy_score(pred1,Y_test)
print(svc_acc)
print(confusion_matrix(pred1,Y_test))

```

```

0.88
[[46  3  0  0]
 [ 4 40  8  0]

```

```

[ 0  3 52  4]
[ 0  0  2 38]]

from sklearn.linear_model import LogisticRegression # its a
classification
lr=LogisticRegression()
lr.fit(X_train,Y_train)

LogisticRegression()

pred2 = lr.predict(X_test)
pred2

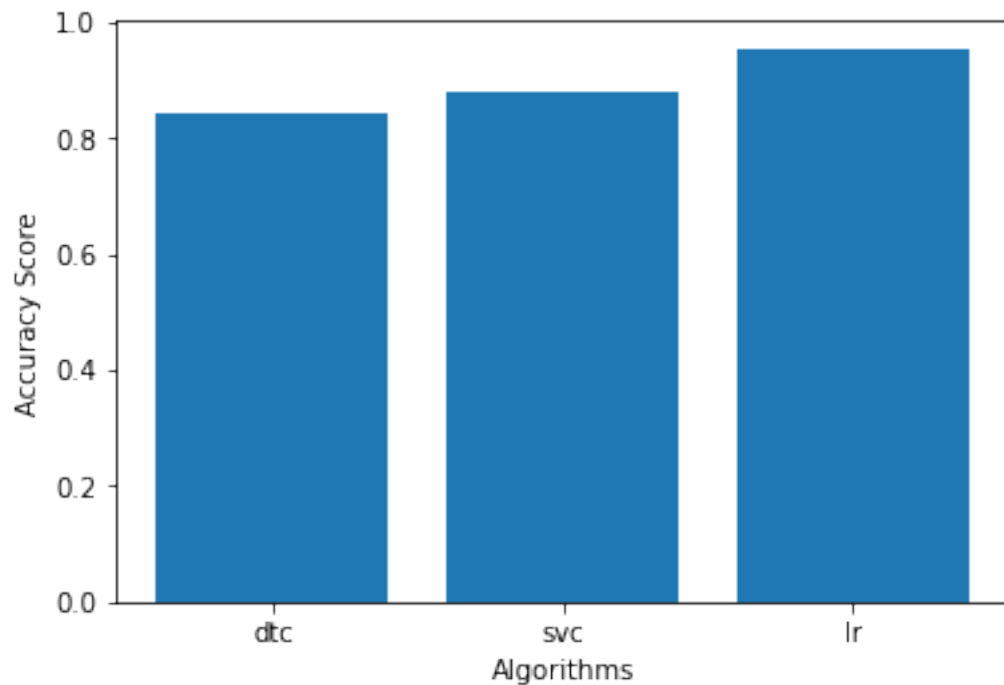
array([1, 1, 2, 1, 1, 1, 2, 1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 3,
1,
      2, 3, 2, 2, 2, 2, 0, 0, 2, 3, 0, 0, 3, 0, 0, 0, 1, 1, 1, 2, 3,
2,
      3, 0, 1, 3, 3, 1, 0, 0, 3, 3, 3, 3, 1, 3, 2, 3, 2, 2, 3, 1, 3,
1,
      0, 0, 0, 2, 1, 2, 3, 2, 1, 3, 3, 2, 0, 2, 0, 0, 2, 1, 2, 2, 2,
1,
      0, 0, 3, 2, 0, 2, 0, 3, 2, 0, 2, 3, 0, 1, 3, 3, 0, 3, 0, 0, 2,
0,
      1, 0, 3, 2, 2, 1, 1, 3, 1, 0, 3, 2, 2, 3, 1, 2, 3, 2, 1, 1, 1,
0,
      0, 1, 0, 2, 3, 0, 2, 3, 1, 3, 0, 0, 0, 1, 1, 2, 2, 0, 3, 1, 2,
2,
      3, 2, 2, 0, 3, 2, 2, 2, 2, 2, 1, 2, 1, 1, 3, 3, 1, 2, 0, 3, 1,
3,
      2, 2, 3, 2, 2, 1, 0, 1, 3, 2, 1, 2, 0, 3, 1, 0, 2, 2, 0, 2, 0,
0,
      3, 0], dtype=int64)

from sklearn.metrics import accuracy_score
lr_acc = accuracy_score(pred2,Y_test)
print(lr_acc)
print(confusion_matrix(pred2,Y_test))

0.955
[[49  1  0  0]
 [ 1 45  3  0]
 [ 0  0 56  1]
 [ 0  0  3 41]]

plt.bar(x=['dtc','svc','lr'],height=[dtc_acc,svc_acc,lr_acc])
plt.xlabel("Algorithms")
plt.ylabel("Accuracy Score")
plt.show()

```



```
lr.predict(test)
```

```
array([3, 3, 2, 3, 1, 3, 3, 1, 3, 0, 3, 3, 0, 0, 2, 0, 2, 1, 3, 2, 1,
3,
      1, 1, 3, 0, 2, 0, 3, 0, 2, 0, 3, 0, 1, 1, 3, 1, 2, 1, 1, 2, 0,
0,
      0, 1, 0, 3, 1, 2, 1, 0, 3, 0, 3, 0, 3, 1, 1, 3, 3, 3, 0, 1, 1,
1,
      2, 3, 1, 2, 1, 2, 2, 3, 3, 0, 2, 0, 1, 3, 0, 3, 3, 0, 3, 0, 3,
1,
      3, 0, 1, 2, 2, 1, 2, 2, 0, 2, 1, 2, 1, 0, 0, 3, 0, 2, 1, 1, 2,
3,
      3, 3, 1, 3, 3, 3, 3, 2, 3, 0, 0, 3, 2, 1, 2, 0, 3, 2, 2, 2, 0,
2,
      2, 1, 3, 1, 1, 0, 3, 2, 1, 2, 1, 3, 2, 3, 3, 3, 2, 3, 2, 3, 1,
0,
      3, 2, 3, 3, 3, 3, 3, 2, 3, 3, 3, 3, 1, 0, 3, 0, 0, 0, 2, 1, 0,
1,
      0, 0, 1, 2, 1, 0, 0, 1, 1, 2, 2, 1, 0, 0, 0, 1, 0, 3, 1, 0, 2,
2,
      3, 3, 1, 2, 3, 2, 3, 2, 2, 1, 0, 0, 1, 3, 0, 2, 3, 3, 0, 2, 0,
3,
      2, 3, 3, 1, 0, 1, 0, 3, 0, 1, 0, 2, 2, 1, 3, 1, 3, 0, 3, 1, 2,
0,
      0, 2, 1, 3, 3, 3, 1, 1, 3, 0, 0, 2, 3, 3, 1, 3, 1, 1, 3, 2, 1,
2,
      3, 3, 3, 1, 0, 0, 2, 3, 1, 1, 3, 2, 1, 3, 0, 0, 3, 0, 0, 3, 2,
3,
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