

```
In [2]: # necessary libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

```
In [3]: # dataset
dataset = pd.read_csv('gdp per capita.csv')

'''here we select only two years to verify for all countries in world'''
dataFrame = dataset[["2010", "2011"]]

'''may some countires skipped bcoz they include null values'''

x = dataFrame.dropna()

# make array
X = x.values
```

```
In [5]: '''count total number of values on each year
mean values for both years for all countries
min and max value is also described
after which values is most in dataframe
is all about the description'''
dataFrame.describe()
```

```
Out[5]:
```

	2010	2011
count	256.000000	258.000000
mean	14749.500789	16198.068908
std	22142.907842	24251.351938
min	234.235539	249.577979
25%	1655.486199	1866.674318
50%	5475.194363	5979.460608
75%	16722.986730	19868.319985
max	150737.892500	169016.196100

```
In [6]: # normalizing the dataset

from sklearn import preprocessing

'''normalize the dataset with builtin library of python sklearn using preprocessing'''
normalized = preprocessing.normalize(X)
normalized
```

```
Out[6]: array([[0.68533719, 0.72822588],
               [0.67826274, 0.73481947],
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```

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```

In [15]:

```
#curve_fit() method

#Import curve fitting package from scipy
from scipy.optimize import curve_fit

xdata = x["2010"]
ydata = x["2011"]

'''guassian function that use in curve fit this is chunk of function that use in future
call this method to make curve fit that is builtin function of scipy'''
def guass(x, A, B):
    y = A*np.exp(-1*B*x**2)
    return y
param, cov = curve_fit(guass,xdata, ydata)

A = param[0]
B = param[1]

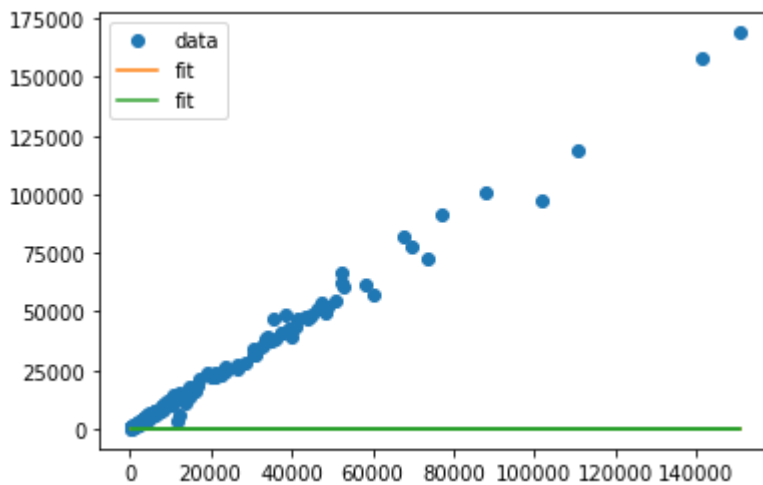
#    fit y
y = guass(x, A, B)

#    plotting
plt.plot(xdata, ydata, 'o', label='data')
plt.plot(xdata, y, '-', label='fit')
plt.legend()
```

E:\Files\lib\site-packages\scipy\optimize\minpack.py:833: OptimizeWarning: Covariance of the parameters could not be estimated

warnings.warn('Covariance of the parameters could not be estimated',
<matplotlib.legend.Legend at 0x2873bdeaca0>

Out[15]:



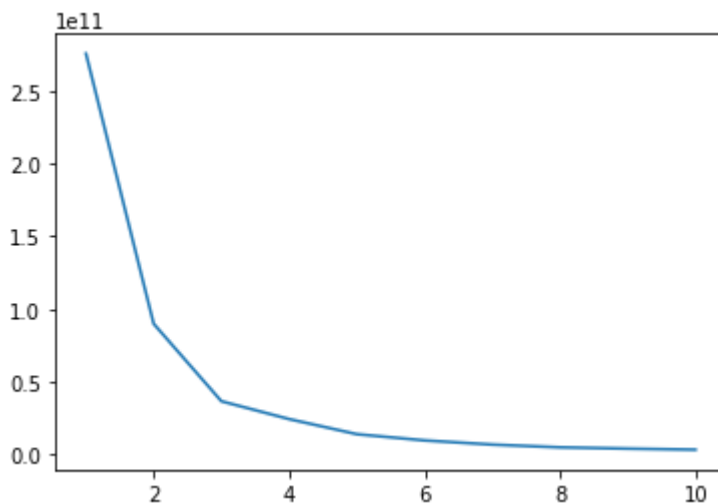
```
In [18]: # prediction for future values
from sklearn.cluster import KMeans

'''how many clusters make from this dataset'''
wcss = []
for i in range(1,11):
    kmeans = KMeans(n_clusters = i, init = "k-means++")
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)
```

E:\Files\lib\site-packages\sklearn\cluster_kmeans.py:881: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

```
In [28]: plt.plot(range(1,11),wcss)
plt.show()
```



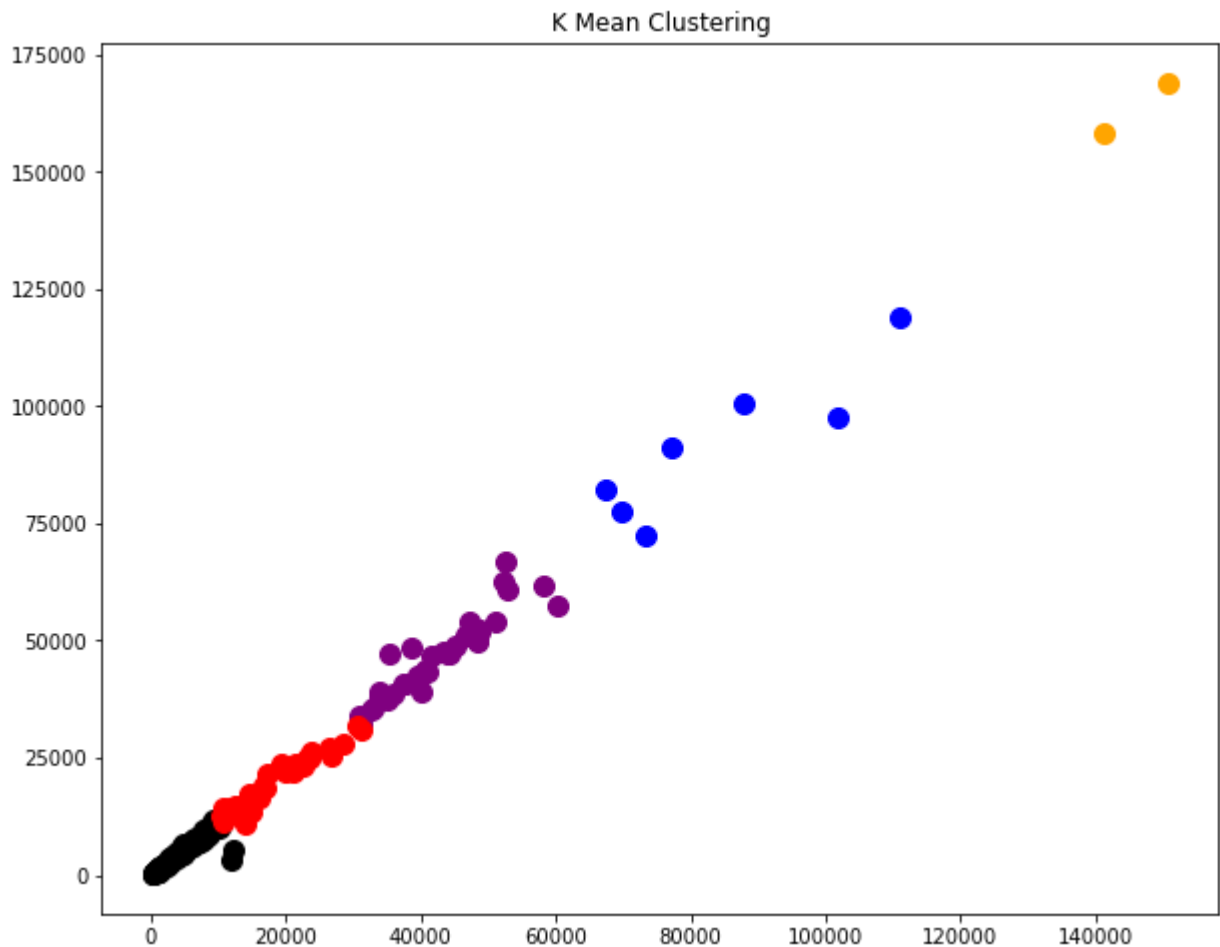
```
In [21]: '''from above elbow chart we easily make 5 different clusters'''

kmeans = KMeans(n_clusters = 5, init = "k-means++")
y_kmeans = kmeans.fit_predict(X)
y_kmeans
```

```
Out[21]: array([3, 0, 0, 0, 0, 0, 1, 0, 1, 3, 0, 0, 3, 1, 1, 0, 0, 1, 0, 0, 0, 0,
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```

```
In [27]: '''as the prediction above show which record belong to which cluster is mentioned in ab
```

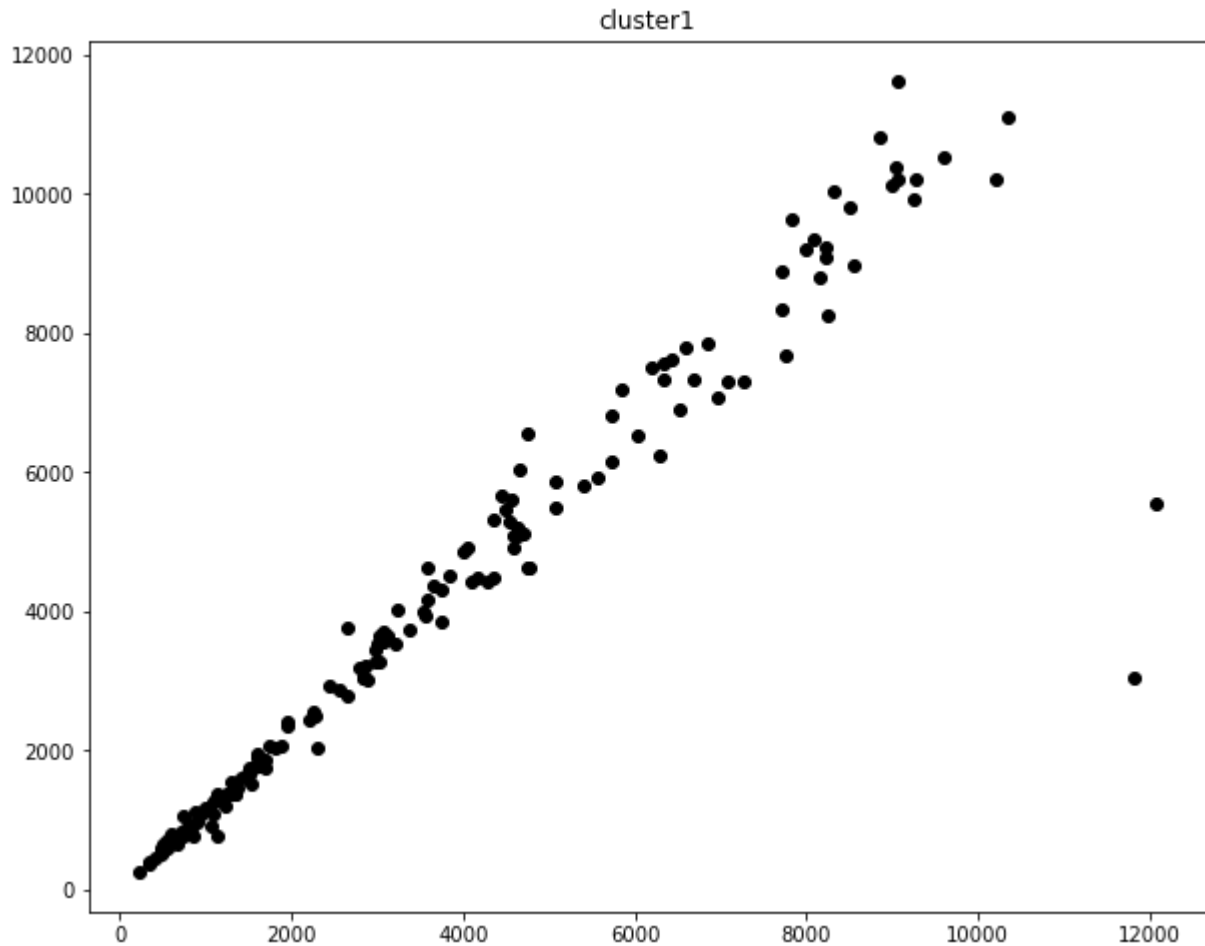
```
plt.figure(figsize = (10,8))
plt.title("K Mean Clustering")
plt.scatter(X[y_kmeans==0,0], X[y_kmeans==0,1], s=100, c="black")
plt.scatter(X[y_kmeans==1,0], X[y_kmeans==1,1], s=100, c="purple")
plt.scatter(X[y_kmeans==2,0], X[y_kmeans==2,1], s=100, c="orange")
plt.scatter(X[y_kmeans==3,0], X[y_kmeans==3,1], s=100, c="red")
plt.scatter(X[y_kmeans==4,0], X[y_kmeans==4,1], s=100, c="blue")
plt.show()
```

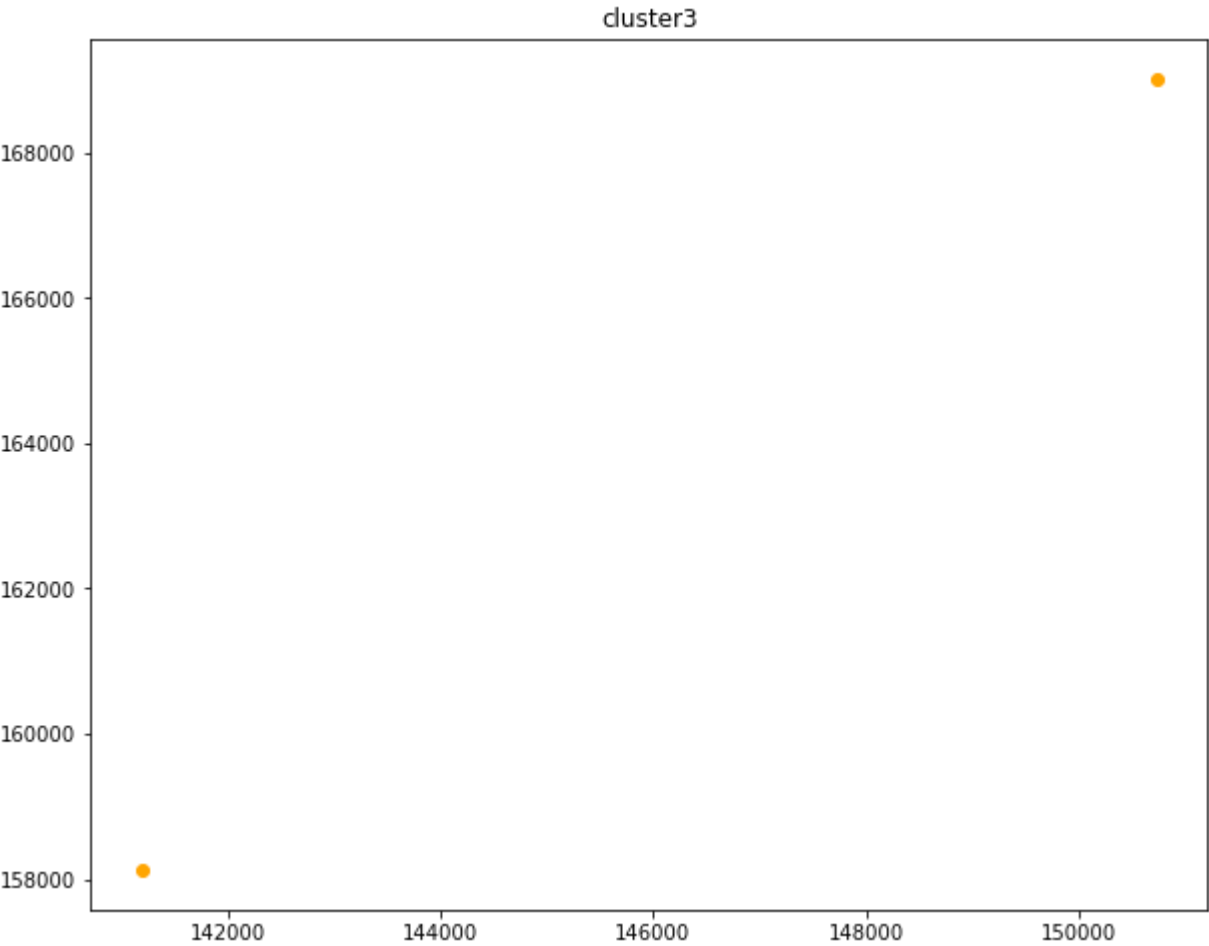
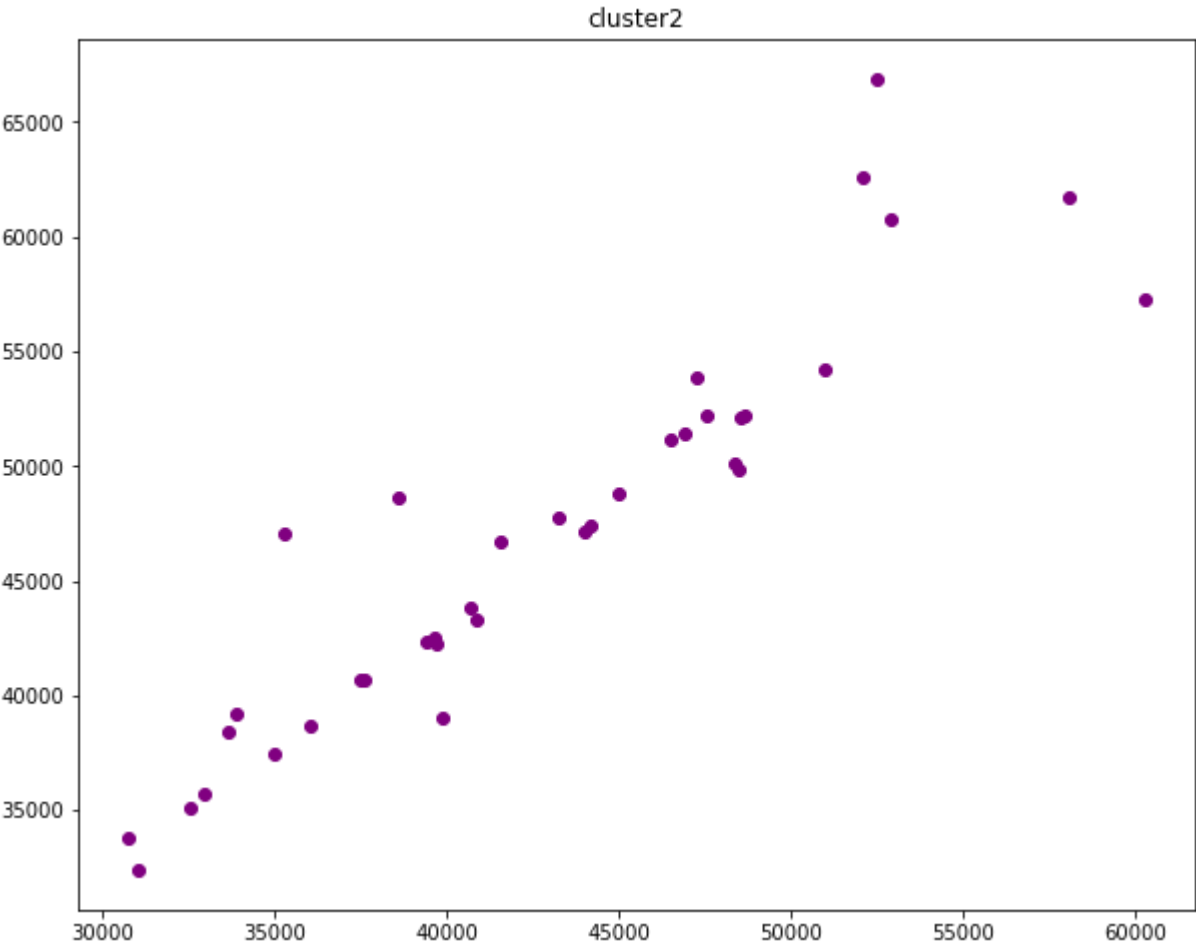


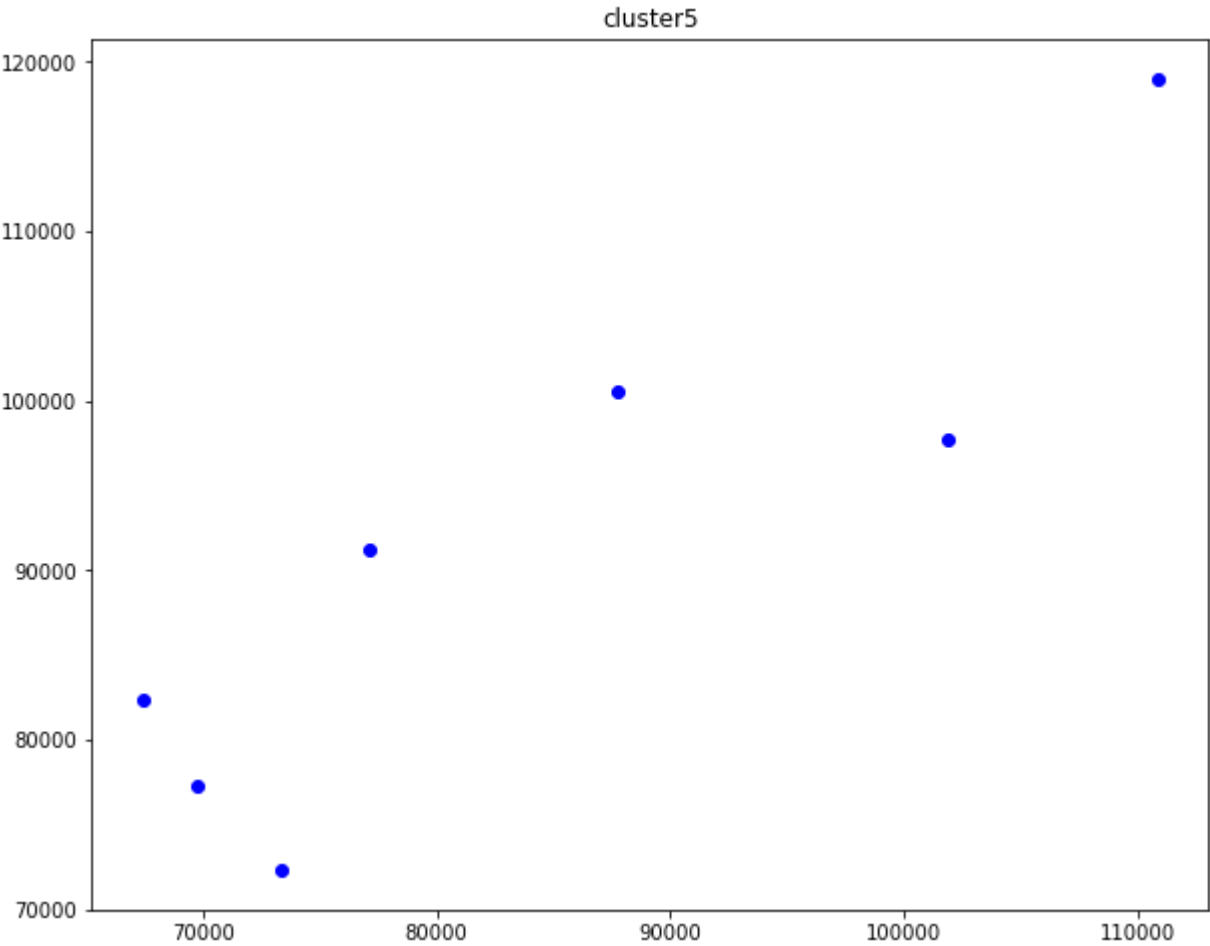
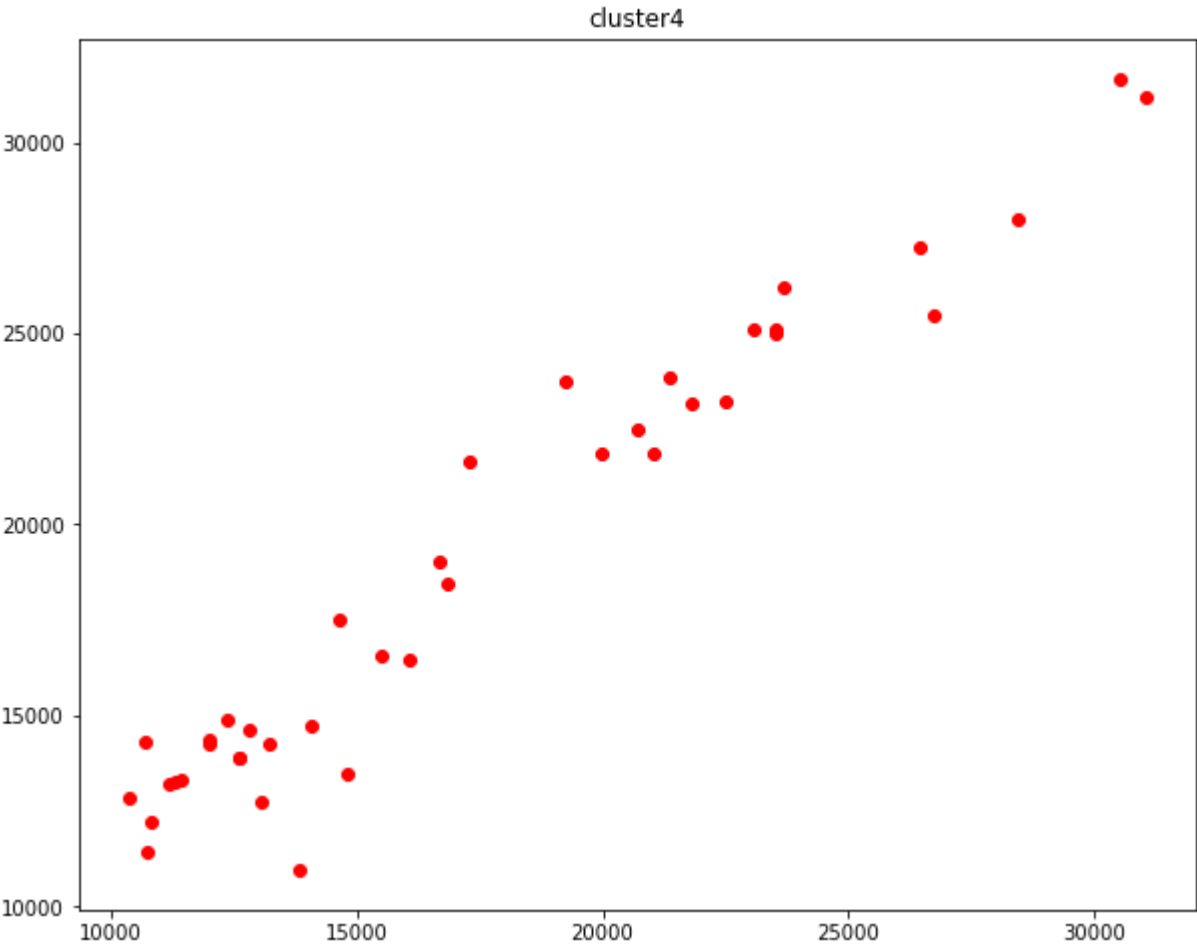
```
In [26]: # Interpretation of the results.
'''black show most elements of dataset,
red show after black in rich quantity
```



```
purple placed after red
then blue and orange'''
clor = ["black", "purple", "orange", "red", "blue"]
for i in range(5):
    plt.figure(figsize=(10,8))
    plt.title(f"cluster{i+1}")
    plt.scatter(X[y_kmeans==i,0], X[y_kmeans==i,1], color=clor[i])
    plt.show()
```







In []: