	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	rea s
0	male	group A	high school	standard	completed	67	
1	female	group D	some high school	free/reduced	none	40	
2	male	group E	some college	free/reduced	none	59	
3	male	group B	high school	standard	none	77	

#printing information of dataset
exams\_data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	gender	1000 non-null	object
1	race/ethnicity	1000 non-null	object
2	parental level of education	1000 non-null	object
3	lunch	1000 non-null	object
4	test preparation course	1000 non-null	object
5	math score	1000 non-null	int64
6	reading score	1000 non-null	int64
7	writing score	1000 non-null	int64

dtypes: int64(3), object(5)
memory usage: 62.6+ KB

```
11/17/22, 9:11 PM
                                                ML_Pract4A.ipynb - Colaboratory
   #checking whether the dataset contains null value or not
   exams_data.isnull().sum()
        gender
                                         0
        race/ethnicity
                                         0
        parental level of education
                                         0
        lunch
        test preparation course
                                         0
        math score
        reading score
                                         0
        writing score
        dtype: int64
   #choosing reading and writing scores columns from dataset
   X = exams_data.iloc[:,[6,7]].values
   print(X)
         [[67 63]
          [59 55]
          [60 50]
          . . .
          [35 41]
          [74 82]
          [60 62]]
   #Within Clusters : Sum of Squares
   #For diff number of clusters like 2,3, .... 10.
   #Creating an empty list
   wcss = []
   for i in range(1,11):
     kmeans = KMeans(n clusters=i, init='k-means++', random state = 42)
     kmeans.fit(X)
     wcss.append(kmeans.inertia_)
   #plotting elbow graph
   sns.set()
   plt.plot(range(1,11), wcss)
```

plt.title('The Elbow Point Graph') plt.xlabel('Number of Clusters')

plt.ylabel('WCSS')

plt.show()

## The Elbow Point Graph

```
400000
300000
8
200000
```

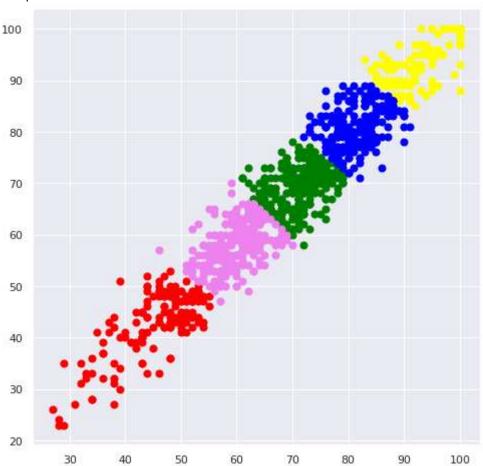
#Training the KMeans cluster model
kmeans = KMeans(n\_clusters=5, init='k-means++', random\_state = 0)

#Return a lable for each datapoint based on their cluster
Y = kmeans.fit\_predict(X)
print(Y)

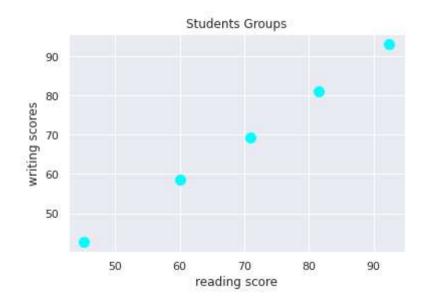
[0 4 4 0 0 3 4 3 4 1 3 2 3 0 3 4 1 4 1 1 3 4 0 4 1 1 4 4 2 4 3 0 1 0 2 0 3 4 0 3 0 0 3 0 2 1 4 3 0 4 0 4 3 1 4 4 3 3 4 1 0 4 4 3 2 0 4 1 4 4 0 1 0 3 0 0 4 3 0 1 4 1 4 0 1 0 0 2 4 0 1 0 3 2 0 0 2 1 1 4 4 3 3 0 4 3 0 1 0 1 3 1 3 1 0 3 4 4 4 2 0 2 4 1 3 4 0 1 3 2 3 0 4 4 4 0 3 0 1 0 4 3 3 4 0 4 4 1 2 1 3 4 3 0 4 1 4 3 2 3 2 4 2 0 0 2 4 0 2 3 4 0 0 4 0 4 4 4 2 3 4 1 2 3 3 2 0 4 4 4 4 1 4 2 0 1 1 4 0 4 4 1 0 0 3 0 3 3 3 4 0 3 0 0 3 4 0 3 4 4 3 0 1 3 2 0 1 0 3 3 3 0 1 0 3 2 0 2 4 0 3 3 4 4 1 0 1 0 4 3 0 0 4 0 4 4 3 0 0 2 0 1 4 4 3 0 0 3 4 0 0 0 0 2 2 3 0 3 1 0 3 2 0 1 1 3 3 1 0 4 0 3 0 0 4 3 1 4 4 1 1 1 4 0 4 4 4 0 3 3 0 4 1 3 0 2 0 3 2 3 4 3 4 3 3 4 3 4 4 4 0 3 1 3 1 0 4 1 0 3 4 1 4 0 3 4 3 0 4 2 4 1 3 0 0 4 0 4 0 1 3 1 4 3 0 0 4 0 4 0 0 0 0 4 3 0 2 1 2 2 1 2 3 3 3 3 4 4 3 3 4 4 1 4 4 3 2 3 3 0 0 0 0 4 3 4 2 0 1 0 0 0 4 4 4 0 4 0 4 1 0 0 1 0 1 3 0 0 1 3 3 0 0 2 4 3 0 1 2 0 4 2 0 1 2 0 0 0 4 0 4 4 3 1 0 4 2 3 4 4 4 1 1 0 4 4 3 0 0 4 2 3 0 1 0 0 3 4 0 1 3 3 3 3 2 2 4 4 3 0 0 0 1 3 4 4 1 0 2 3 4 2 4 3 1 2 0 4 4 0 0 3 0 3 3 4 1 1 3 2 3 3 0 3 1 0 0 2 4 0 2 0 4 0 3 2 4 0 0 0 0 3 3 4 3 0 4 4 1 4 2 0 2 4 0 1 1 0 3 1 4 4 0 4 2 3 4 3 1 0 4 2 1 3 4 2 4 4 0 1 3 0 0 3 3 4 0 2 2 0 3 2 0 2 4 4 2 4 1 2 4 0 1 0 2 1 3 0 1 0 4 3 0 3 0 1 2 0 4 0 4 1 3 3 3 4 2 3 3 4 0 4 4 4 2 2 4 0 3 0 3 0 0 4 1 4 3 0 2 1 0 0 2 1 4 3 0 0 0 1 2 0 2 3 3 0 3 4 1 1 1 3 1 0 4 0 4 0 0 4 1 1 3 0 0 4 0 0 0 1 2 3 3 4 2 4 4 0 3 0 1 1 0 0 0 4 2 3 3 0 4 3 1 4 0 4 2 1 4 1 1 2 4 4 4 1 3 0 4 1 3 0 2 0 4 2 2 3 0 0 4 0 3 2 3 0 1 2 2 0 0 2 4 4 2 3 2 4 4 3 0 3 4 2 0 1 0 3 1 4 0 3 3 3 0 1 1 0 1 1 2 3 2 4 3 4 0 3 4 4 3 0 0 0 3 1 2 4 4 4 3 4 2 1 0 1 0 0 3 4 1 4 4 4 2 4 2 0 3 0 3 4 3 0 0 2 2 3 3 1 1 1 1 3 3 3 0 0 0 0 3 1 0 2 3 0 0 4 0 2 0 3 3 4 0 2 2 4 2 2 3 2 3 3 0 3 0 4 3 4 4 4 1 3 4 0 4 4 0 4 1 0 3 4 3 3 1 0 0 4 3 4 0 1 1 3 3 4 0 3 1 1 0 0 4 0 3 3 0 1 0 4 4 4 0 4 0 0 0 0 3 4 0 4 4 0 4 0 0 4 3 3 1 3 0 4 0 1 1 1 4 3 3 3 0 0 4 0 4 1 4 2 0 4 4 4 3 0 4 4 3 4 3 1 0 3 2 0 2 1 4 2 1 0 0 4 0 1 4 1 0 1 1 0 1 0 0 3 3 3 4 4 2 4 4 3 0 2 1 3 4]

```
#plotting graphs for all clusters
plt.figure(figsize=(8,8))
plt.scatter(X[Y==0,0], X[Y==0,1], s=50, c='green', label='Cluster 1')
plt.scatter(X[Y==1,0], X[Y==1,1], s=50, c='red', label='Cluster 2')
plt.scatter(X[Y==2,0], X[Y==2,1], s=50, c='yellow', label='Cluster 3')
plt.scatter(X[Y==3,0], X[Y==3,1], s=50, c='blue', label='Cluster 4')
plt.scatter(X[Y==4,0], X[Y==4,1], s=50, c='violet', label='Cluster 5')
```

<matplotlib.collections.PathCollection at 0x7fdd69cd1090>



```
#plotting centriods
plt.scatter(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], s=100, c='cyan', labe
plt.title('Students Groups')
plt.xlabel('reading score')
plt.ylabel('writing scores')
plt.show()
```



```
#plotting clusters along with centroids
plt.figure(figsize=(8,8))
plt.scatter(X[Y==0,0], X[Y==0,1], s=50, c='green', label='Cluster 1')
plt.scatter(X[Y==1,0], X[Y==1,1], s=50, c='red', label='Cluster 2')
plt.scatter(X[Y==2,0], X[Y==2,1], s=50, c='yellow', label='Cluster 3')
plt.scatter(X[Y==3,0], X[Y==3,1], s=50, c='blue', label='Cluster 4')
plt.scatter(X[Y==4,0], X[Y==4,1], s=50, c='violet', label='Cluster 5')

plt.scatter(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], s=100, c='cyan', labe
plt.title('Students Groups')
plt.xlabel('reading scores')
plt.ylabel('writing scores')
plt.show()
```

