In [17]:

```
#Roll No-33238
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
```

In [18]:

```
df = pd.read_csv('/home/pict/Downloads/archive (15)/Mall_Customers.csv')
```

In [19]:

```
plt.rcParams['figure.dpi'] = 300
plt.rcParams['savefig.dpi'] = 300
```

In [20]:

```
plt.rcParams["figure.figsize"] = (8,5)
```

In [21]:

df.head(10)

Out[21]:

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40
5	6	Female	22	17	76
6	7	Female	35	18	6
7	8	Female	23	18	94
8	9	Male	64	19	3
9	10	Female	30	19	72

In [22]:

```
df.isnull().sum()
```

Out[22]:

CustomerID	0
Genre	0
Age	0
Annual Income (k\$)	0
Spending Score (1-100)	0
` ' '	_

dtype: int64

In [23]:

```
for i in df.columns:
    print( i, len( df[df[i] == 0] ) )
```

CustomerID 0 Genre 0 Age 0

Annual Income (k\$) 0

Spending Score (1-100) 0

In [24]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	CustomerID	200 non-null	int64
1	Genre	200 non-null	object
2	Age	200 non-null	int64
3	Annual Income (k\$)	200 non-null	int64
4	Spending Score (1-100)	200 non-null	int64

dtypes: int64(4), object(1)

memory usage: 7.9+ KB

In [25]:

df.describe()

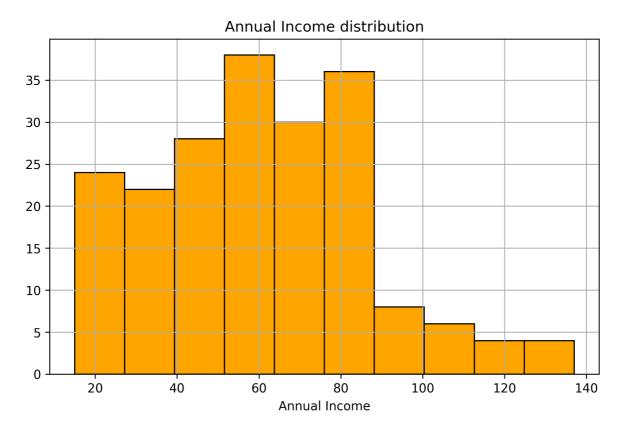
Out[25]:

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
count	200.000000	200.000000	200.000000	200.000000
mean	100.500000	38.850000	60.560000	50.200000
std	57.879185	13.969007	26.264721	25.823522
min	1.000000	18.000000	15.000000	1.000000
25%	50.750000	28.750000	41.500000	34.750000
50%	100.500000	36.000000	61.500000	50.000000
75%	150.250000	49.000000	78.000000	73.000000
max	200.000000	70.000000	137.000000	99.000000

In [26]:

```
plt.plot(figsize=(8,5))
plt.hist(df["Annual Income (k$)"], color='orange', edgecolor='k')
plt.title("Annual Income distribution")
plt.xlabel("Annual Income")
plt.grid(True)
plt.show
```

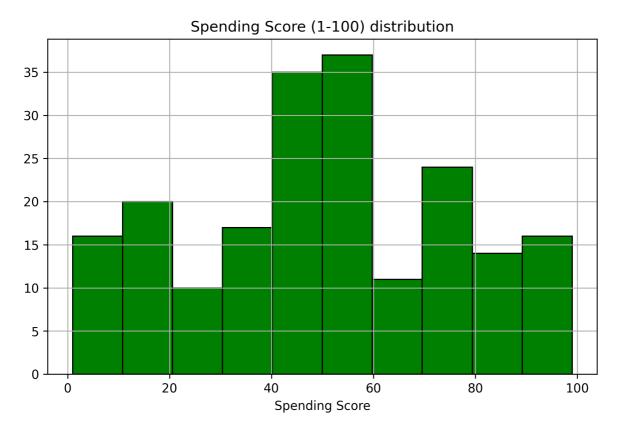
Out[26]:



In [27]:

```
plt.plot(figsize=(8,5))
plt.hist(df["Spending Score (1-100)"], color='green', edgecolor='k')
plt.title("Spending Score (1-100) distribution")
plt.xlabel("Spending Score")
plt.grid(True)
plt.show
```

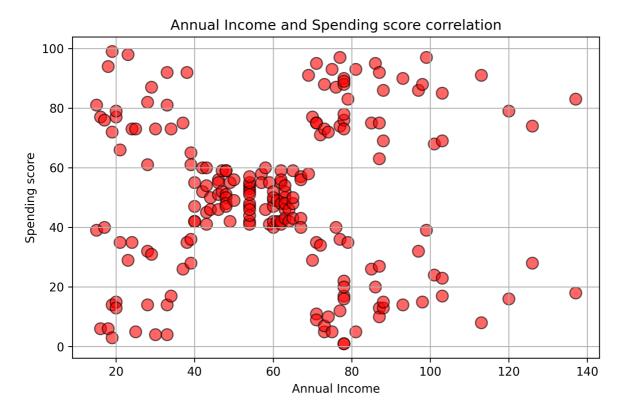
Out[27]:



In [28]:

```
plt.plot(figsize=(8,5))
plt.scatter(df["Annual Income (k$)"], df["Spending Score (1-100)"], color='red', ed
plt.title("Annual Income and Spending score correlation")
plt.xlabel("Annual Income")
plt.ylabel("Spending score")
plt.grid(True)
plt.show
```

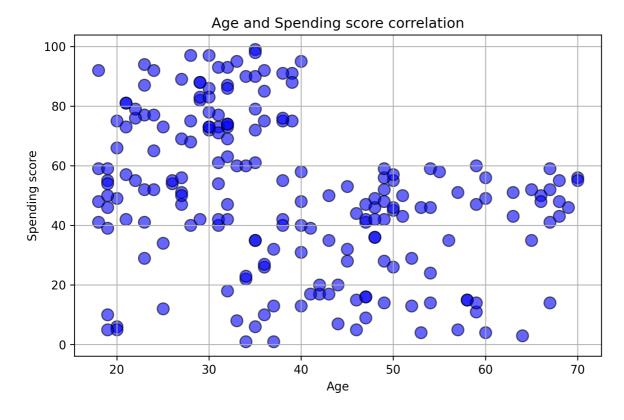
Out[28]:



In [29]:

```
plt.plot(figsize=(8,5))
plt.scatter(df["Age"], df["Spending Score (1-100)"], color='blue', edgecolor='k', a
plt.title("Age and Spending score correlation")
plt.xlabel("Age")
plt.ylabel("Spending score")
plt.grid(True)
plt.show
```

Out[29]:

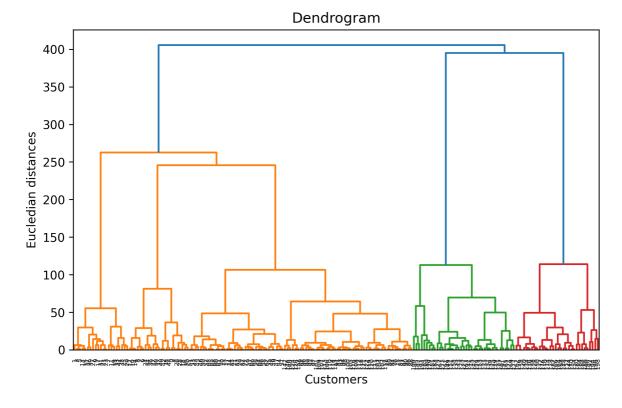


In [30]:

```
X = df.iloc[ : , 3:].values
print(X[:10], "\n\n")
print(X[-10:])
[[15 39]
 [15 81]
 [16 6]
 [16 77]
 [17 40]
 [17 76]
 [18 6]
 [18 94]
 [19 3]
 [19 72]]
[[103
       23]
 [103
       69]
 [113
        8]
 [113
       91]
 [120
       16]
 [120
        79]
 [126
        28]
 [126
       74]
 [137
       18]
 [137
       83]]
```

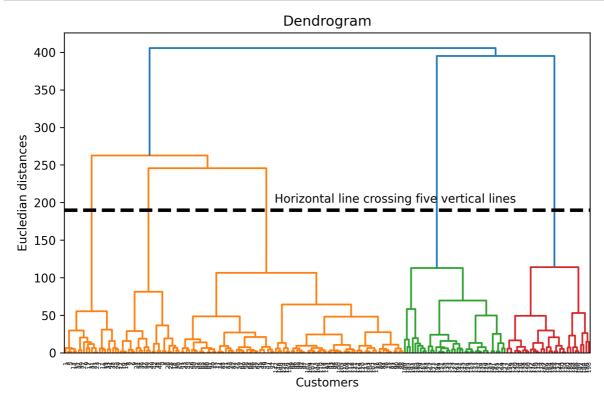
In [48]:

```
import scipy.cluster.hierarchy as sch
sch.dendrogram(sch.linkage(X, method='ward'))
plt.title("Dendrogram")
plt.xlabel("Customers")
plt.ylabel("Eucledian distances")
plt.show()
```



In [58]:

```
sch.dendrogram(sch.linkage(X, method='ward'))
plt.title("Dendrogram")
plt.xlabel("Customers")
plt.ylabel("Eucledian distances")
plt.hlines(y=190,xmin=0 , xmax=2000, lw=3, linestyles="--", color='black')#3
plt.text(x=800, y=200, s="Horizontal line crossing five vertical lines", fontsize=1
plt.show()
```



In [59]:

```
from sklearn.cluster import AgglomerativeClustering
hc = AgglomerativeClustering(n_clusters=5, affinity='euclidean', linkage='ward')
y_hc = hc.fit_predict(X)
```

In [61]:

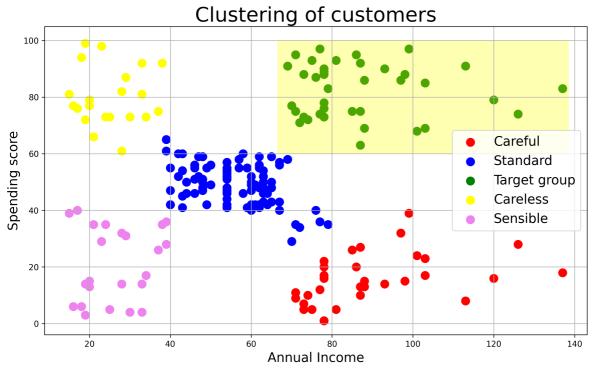
```
print("X: ", X[:10])
print("y_hc: ", y_hc)
X: [[15 39]
     [15 81]
     [16 6]
      [16 77]
      [17 40]
      [17 76]
      [18 6]
      [18 94]
      [19 3]
     [19 72]]
3 4 3 4 3 4
     1 1
    0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 1 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 2 \; 0 \; 
2 0
     2 0 2 0 2 0 2 0 2 0 2 0 2 0 2 0 2]
```

In [39]:

```
plt.figure(figsize=(12,7))
plt.scatter(X[y_hc == 0, 0], X[y_hc == 0, 1], s=100, c='red', label='Careful')
plt.scatter(X[y_hc == 1, 0], X[y_hc == 1, 1], s=100, c='blue', label='Standard')
plt.scatter(X[y_hc == 2, 0], X[y_hc == 2, 1], s=100, c='green', label='Target group
plt.scatter(X[y_hc == 3, 0], X[y_hc == 3, 1], s=100, c='yellow', label='Careless')
plt.scatter(X[y_hc == 4, 0], X[y_hc == 4, 1], s=100, c='violet', label='Sensible')

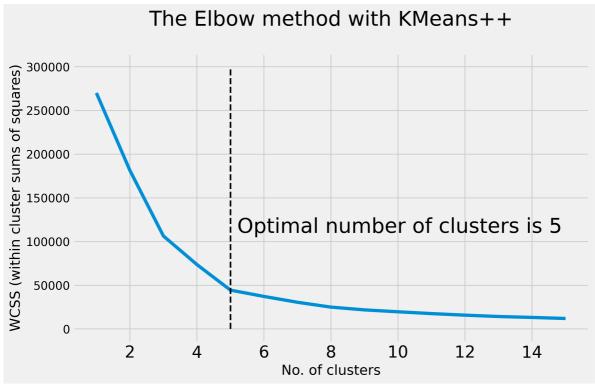
plt.title("Clustering of customers", fontsize=25)
plt.xlabel("Annual Income", fontsize=16)
plt.ylabel("Spending score", fontsize=16)
plt.legend(fontsize=16)
plt.legend(fontsize=16)
plt.grid(True)

plt.axhspan(ymin=60, ymax=100, xmin=0.43, xmax=0.965, alpha=0.3, color="yellow")
plt.show()
```



In [36]:

```
from sklearn.cluster import KMeans
wcss = []
for i in range(1,16):
    kmeans = KMeans(n clusters=i, init='k-means++')
    kmeans.fit(X)
    wcss.append(kmeans.inertia )
with plt.style.context(('fivethirtyeight')): #To set background style of plot
    plt.figure(figsize=(10,6))
    plt.plot(range(1,16), wcss)
    plt.title("The Elbow method with KMeans++\n", fontsize=25)
    plt.xlabel("No. of clusters")
    plt.ylabel("WCSS (within cluster sums of squares)")
    plt.xticks(fontsize=20)
    plt.vlines(x=5, ymin=0, ymax=300000, linestyle="--", color="black", lw=2)
    plt.text(x=5.2, y=110000, s="Optimal number of clusters is 5", fontsize=25)
    plt.show()
```



In [37]:

```
km = KMeans(n_clusters=5, max_iter=100)
km.fit(X)
y_km = km.fit_predict(X)
```

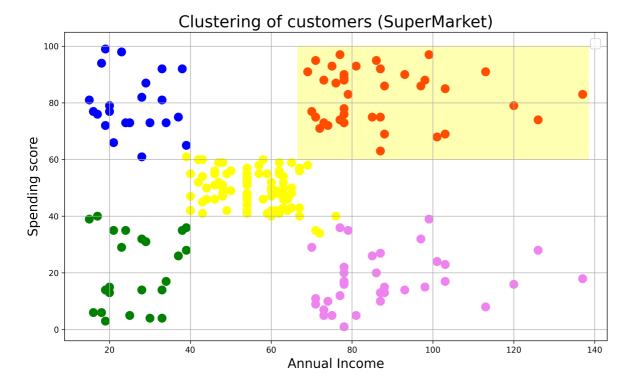
In [38]:

```
plt.figure(figsize=(12,7))
plt.scatter(X[y_km == 0, 0], X[y_km == 0, 1], s=100, c='blue')
plt.scatter(X[y_km == 1, 0], X[y_km == 1, 1], s=100, c='red')
plt.scatter(X[y_km == 2, 0], X[y_km == 2, 1], s=100, c='violet')
plt.scatter(X[y_km == 3, 0], X[y_km == 3, 1], s=100, c='green')
plt.scatter(X[y_km == 4, 0], X[y_km == 4, 1], s=100, c='yellow')

plt.title("Clustering of customers (SuperMarket)", fontsize=20)
plt.xlabel("Annual Income", fontsize=16)
plt.ylabel("Spending score", fontsize=16)
plt.legend(fontsize=16)
plt.grid(True)

plt.axhspan(ymin=60, ymax=100, xmin=0.43, xmax=0.965, alpha=0.3, color="yellow")
plt.show()
```

No artists with labels found to put in legend. Note that artists whos e label start with an underscore are ignored when legend() is called w ith no argument.



In []: