

```
In [1]: import pandas as pd
import numpy as np
```

```
In [2]: data=pd.read_csv(r"C:\Users\prasa\OneDrive\Desktop\Wholesale customers data.csv")
data
```

```
Out[2]:
```

| | Channel | Region | Fresh | Milk | Grocery | Frozen | Detergents_Paper | Delicassen |
|-----|---------|--------|-------|-------|---------|--------|------------------|------------|
| 0 | 2 | 3 | 12669 | 9656 | 7561 | 214 | 2674 | 1338 |
| 1 | 2 | 3 | 7057 | 9810 | 9568 | 1762 | 3293 | 1776 |
| 2 | 2 | 3 | 6353 | 8808 | 7684 | 2405 | 3516 | 7844 |
| 3 | 1 | 3 | 13265 | 1196 | 4221 | 6404 | 507 | 1788 |
| 4 | 2 | 3 | 22615 | 5410 | 7198 | 3915 | 1777 | 5185 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 435 | 1 | 3 | 29703 | 12051 | 16027 | 13135 | 182 | 2204 |
| 436 | 1 | 3 | 39228 | 1431 | 764 | 4510 | 93 | 2346 |
| 437 | 2 | 3 | 14531 | 15488 | 30243 | 437 | 14841 | 1867 |
| 438 | 1 | 3 | 10290 | 1981 | 2232 | 1038 | 168 | 2125 |
| 439 | 1 | 3 | 2787 | 1698 | 2510 | 65 | 477 | 52 |

440 rows × 8 columns

```
In [3]: data.head()
```

```
Out[3]:
```

| | Channel | Region | Fresh | Milk | Grocery | Frozen | Detergents_Paper | Delicassen |
|---|---------|--------|-------|------|---------|--------|------------------|------------|
| 0 | 2 | 3 | 12669 | 9656 | 7561 | 214 | 2674 | 1338 |
| 1 | 2 | 3 | 7057 | 9810 | 9568 | 1762 | 3293 | 1776 |
| 2 | 2 | 3 | 6353 | 8808 | 7684 | 2405 | 3516 | 7844 |
| 3 | 1 | 3 | 13265 | 1196 | 4221 | 6404 | 507 | 1788 |
| 4 | 2 | 3 | 22615 | 5410 | 7198 | 3915 | 1777 | 5185 |

```
In [4]: data.tail()
```

```
Out[4]:
```

| | Channel | Region | Fresh | Milk | Grocery | Frozen | Detergents_Paper | Delicassen |
|-----|---------|--------|-------|-------|---------|--------|------------------|------------|
| 435 | 1 | 3 | 29703 | 12051 | 16027 | 13135 | 182 | 2204 |
| 436 | 1 | 3 | 39228 | 1431 | 764 | 4510 | 93 | 2346 |
| 437 | 2 | 3 | 14531 | 15488 | 30243 | 437 | 14841 | 1867 |
| 438 | 1 | 3 | 10290 | 1981 | 2232 | 1038 | 168 | 2125 |
| 439 | 1 | 3 | 2787 | 1698 | 2510 | 65 | 477 | 52 |

```
In [5]: # Normalize the data i.e scales each variable to range 0 to 1
# why we need to scale?
# if data is not scaled, then the model is biased towards high magnitudes values which
```

```
In [6]: #from sklearn.preprocessing import MinMaxScaler
#scaler=MinMaxScaler()
#new_data=scaler.fit_transform(data)
#new_data
```

```
In [7]: from sklearn.preprocessing import normalize
new_data=normalize(data)
new_data
```

```
Out[7]: array([[1.11821406e-04, 1.67732109e-04, 7.08332695e-01, ...,
1.19648904e-02, 1.49505220e-01, 7.48085205e-02],
[1.25321880e-04, 1.87982820e-04, 4.42198253e-01, ...,
1.10408576e-01, 2.06342475e-01, 1.11285829e-01],
[1.24839188e-04, 1.87258782e-04, 3.96551681e-01, ...,
1.50119124e-01, 2.19467293e-01, 4.89619296e-01],
...,
[5.01633106e-05, 7.52449659e-05, 3.64461533e-01, ...,
1.09606834e-02, 3.72236846e-01, 4.68274505e-02],
[9.11309417e-05, 2.73392825e-04, 9.37737390e-01, ...,
9.45939175e-02, 1.53099982e-02, 1.93653251e-01],
[2.41225630e-04, 7.23676891e-04, 6.72295832e-01, ...,
1.56796660e-02, 1.15064626e-01, 1.25437328e-02]])
```

```
In [8]: new_data=pd.DataFrame(new_data,columns=data.columns)
new_data
```

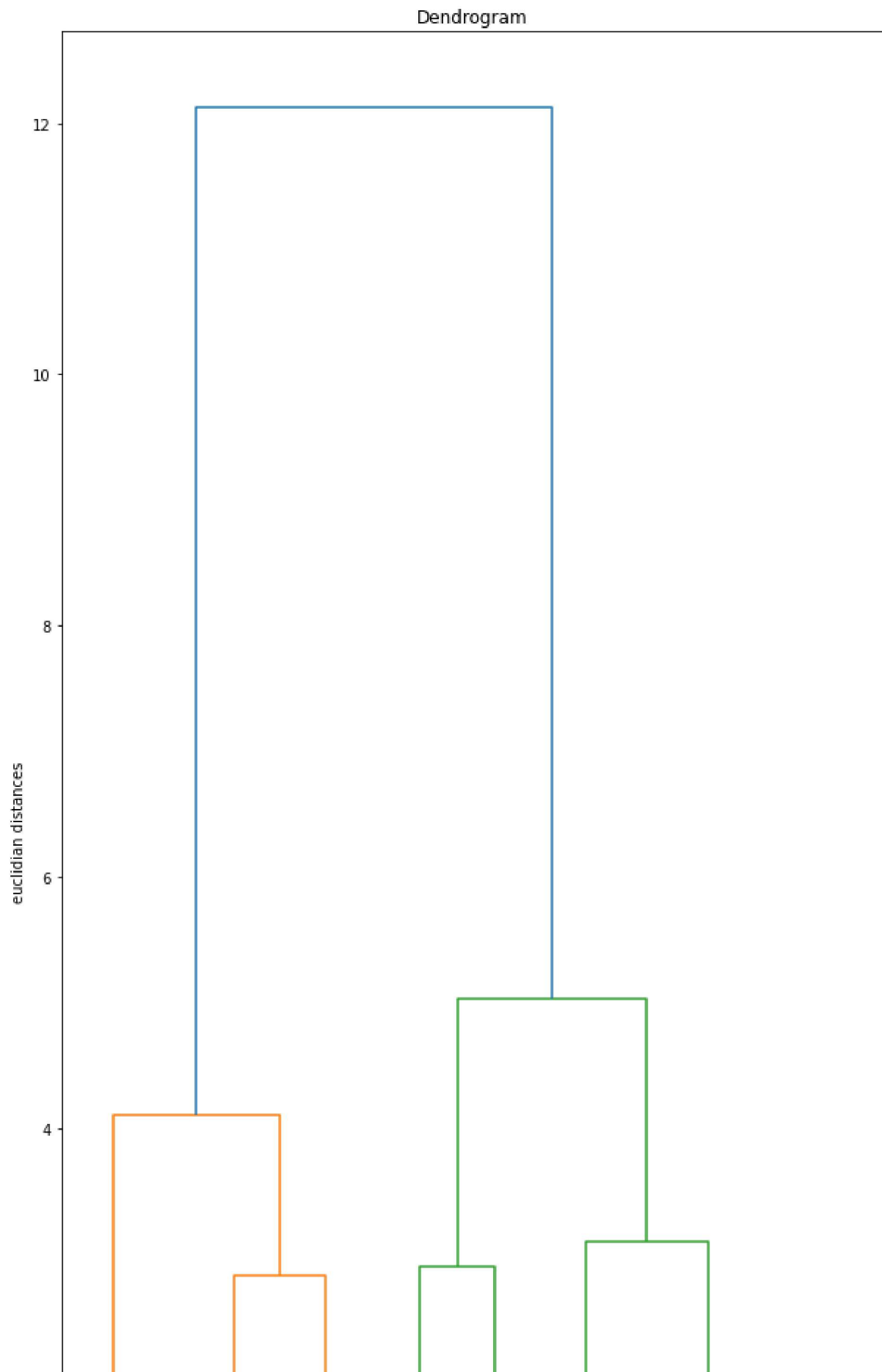
```
Out[8]:
```

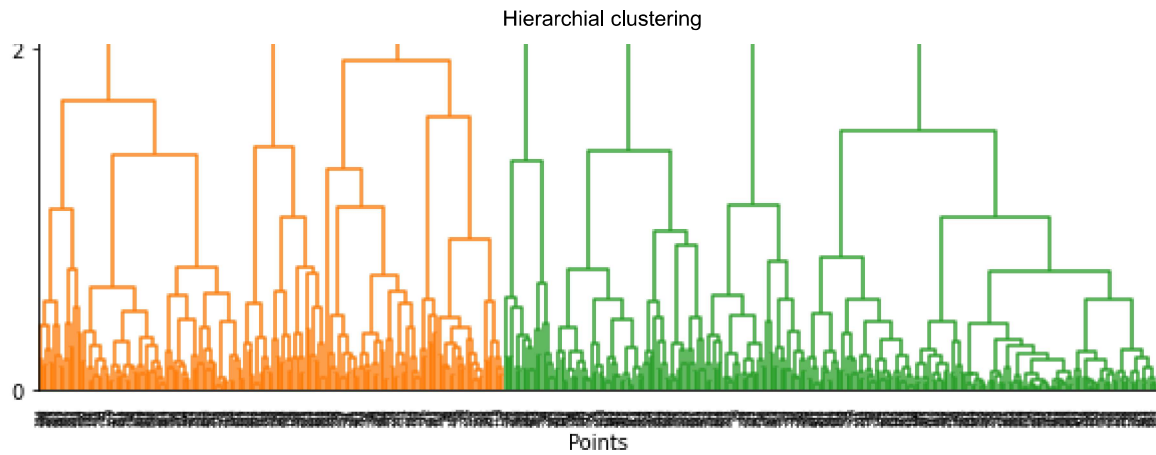
| | Channel | Region | Fresh | Milk | Grocery | Frozen | Detergents_Paper | Delicassen |
|-----|----------|----------|----------|----------|----------|----------|------------------|------------|
| 0 | 0.000112 | 0.000168 | 0.708333 | 0.539874 | 0.422741 | 0.011965 | 0.149505 | 0.074809 |
| 1 | 0.000125 | 0.000188 | 0.442198 | 0.614704 | 0.599540 | 0.110409 | 0.206342 | 0.111286 |
| 2 | 0.000125 | 0.000187 | 0.396552 | 0.549792 | 0.479632 | 0.150119 | 0.219467 | 0.489619 |
| 3 | 0.000065 | 0.000194 | 0.856837 | 0.077254 | 0.272650 | 0.413659 | 0.032749 | 0.115494 |
| 4 | 0.000079 | 0.000119 | 0.895416 | 0.214203 | 0.284997 | 0.155010 | 0.070358 | 0.205294 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 435 | 0.000026 | 0.000078 | 0.776890 | 0.315197 | 0.419191 | 0.343549 | 0.004760 | 0.057646 |
| 436 | 0.000025 | 0.000076 | 0.990872 | 0.036146 | 0.019298 | 0.113919 | 0.002349 | 0.059258 |
| 437 | 0.000050 | 0.000075 | 0.364462 | 0.388465 | 0.758545 | 0.010961 | 0.372237 | 0.046827 |
| 438 | 0.000091 | 0.000273 | 0.937737 | 0.180530 | 0.203404 | 0.094594 | 0.015310 | 0.193653 |
| 439 | 0.000241 | 0.000724 | 0.672296 | 0.409601 | 0.605476 | 0.015680 | 0.115065 | 0.012544 |

440 rows × 8 columns

```
In [9]: #determine no.of clusters by using the Dendrogram
import scipy.cluster.hierarchy as shc
import matplotlib.pyplot as plt
plt.figure(figsize=(10,20))
plt.title("Dendrogram")
plt.xlabel("Points")
```

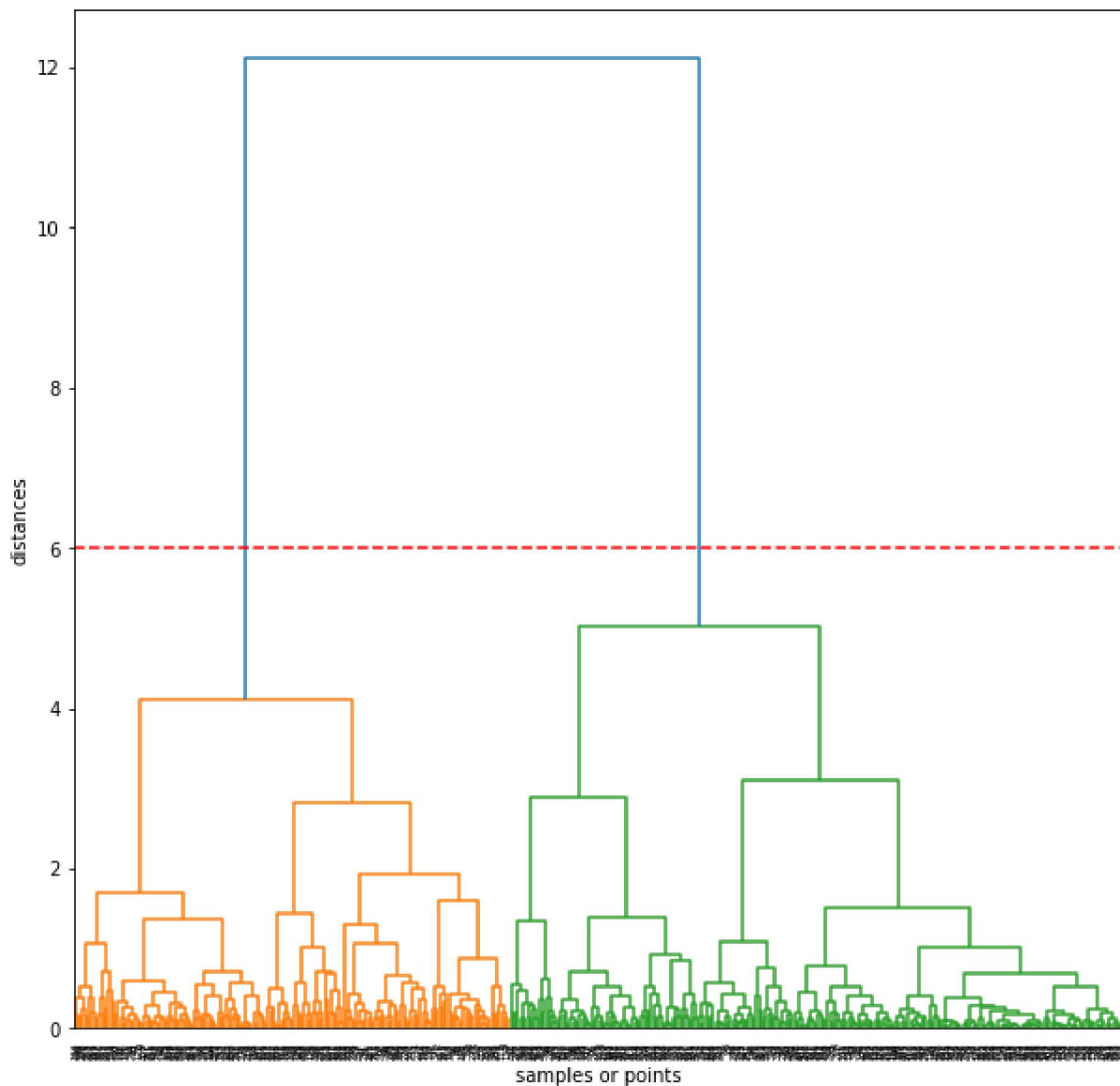
```
plt.ylabel("euclidian distances")  
dend=shc.dendrogram(shc.linkage(new_data,method="ward"))
```





```
In [10]: # how do u decide the threshold to cut the dendrogram to know the no.of clusters.
plt.figure(figsize=(10,10))
plt.ylabel("distances")
plt.xlabel("samples or points")
dend=shc.dendrogram(shc.linkage(new_data,method="ward"))
plt.axhline(y=6,color="red",linestyle="--") # draw a horizontal line
```

```
Out[10]: <matplotlib.lines.Line2D at 0x2af5cf43040>
```



since the horizontal line crosses the vertical axis at 2 different points. the no.of clusters is 2

```
In [11]: from sklearn.cluster import AgglomerativeClustering
cluster=AgglomerativeClustering(n_clusters=2,affinity="euclidean",memory=None,linkage="
cluster
```

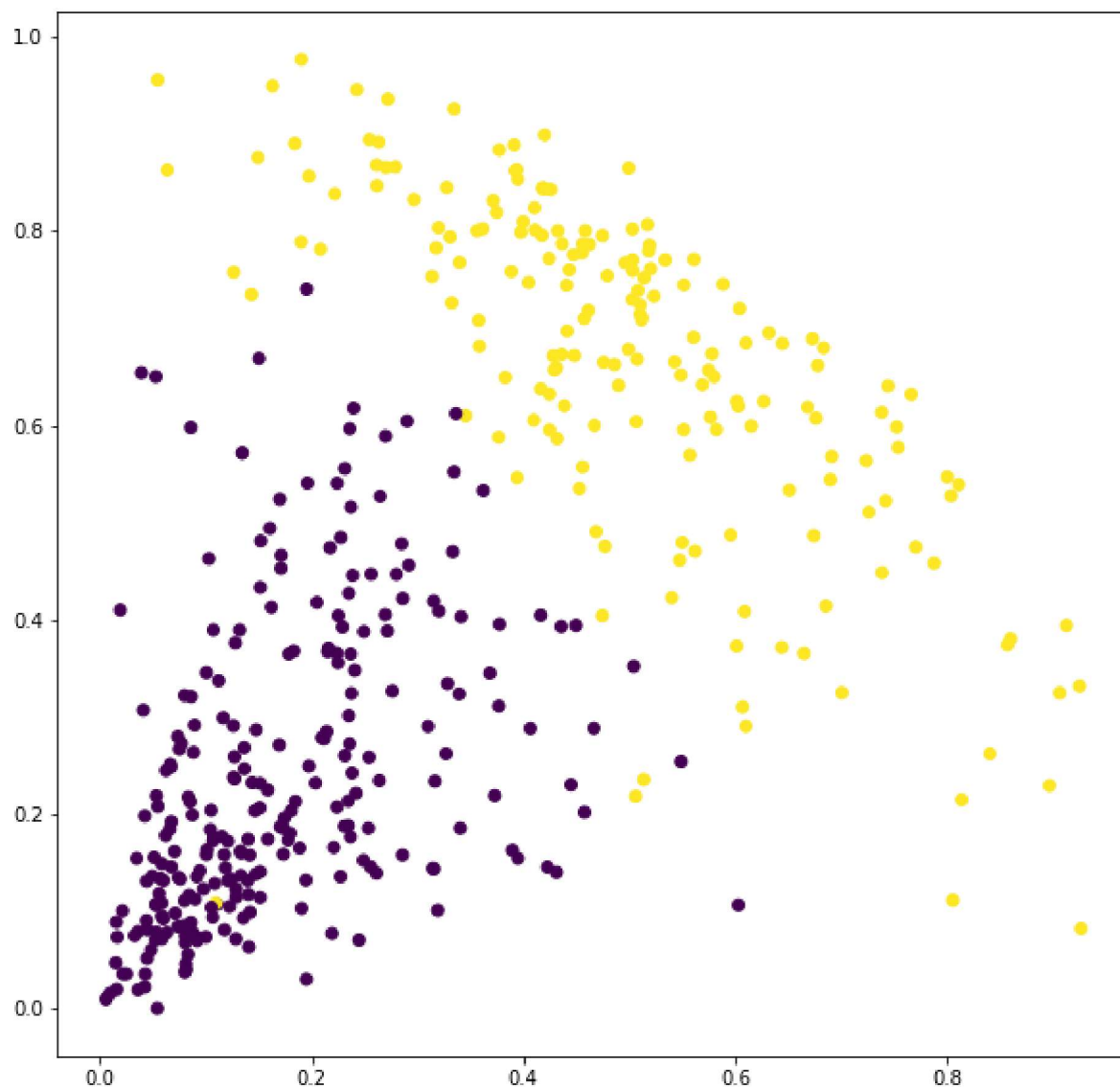
```
Out[11]: AgglomerativeClustering()
```

```
In [12]: cluster.fit_predict(new_data)
```

```
Out[12]: array([1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0,
                0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1,
                1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 1,
                1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0,
                0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1,
                0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0,
                0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1,
                0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1,
                0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1,
                0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0,
                0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0,
                0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1,
                1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
                0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0,
                1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1,
                1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1],
                dtype=int64)
```

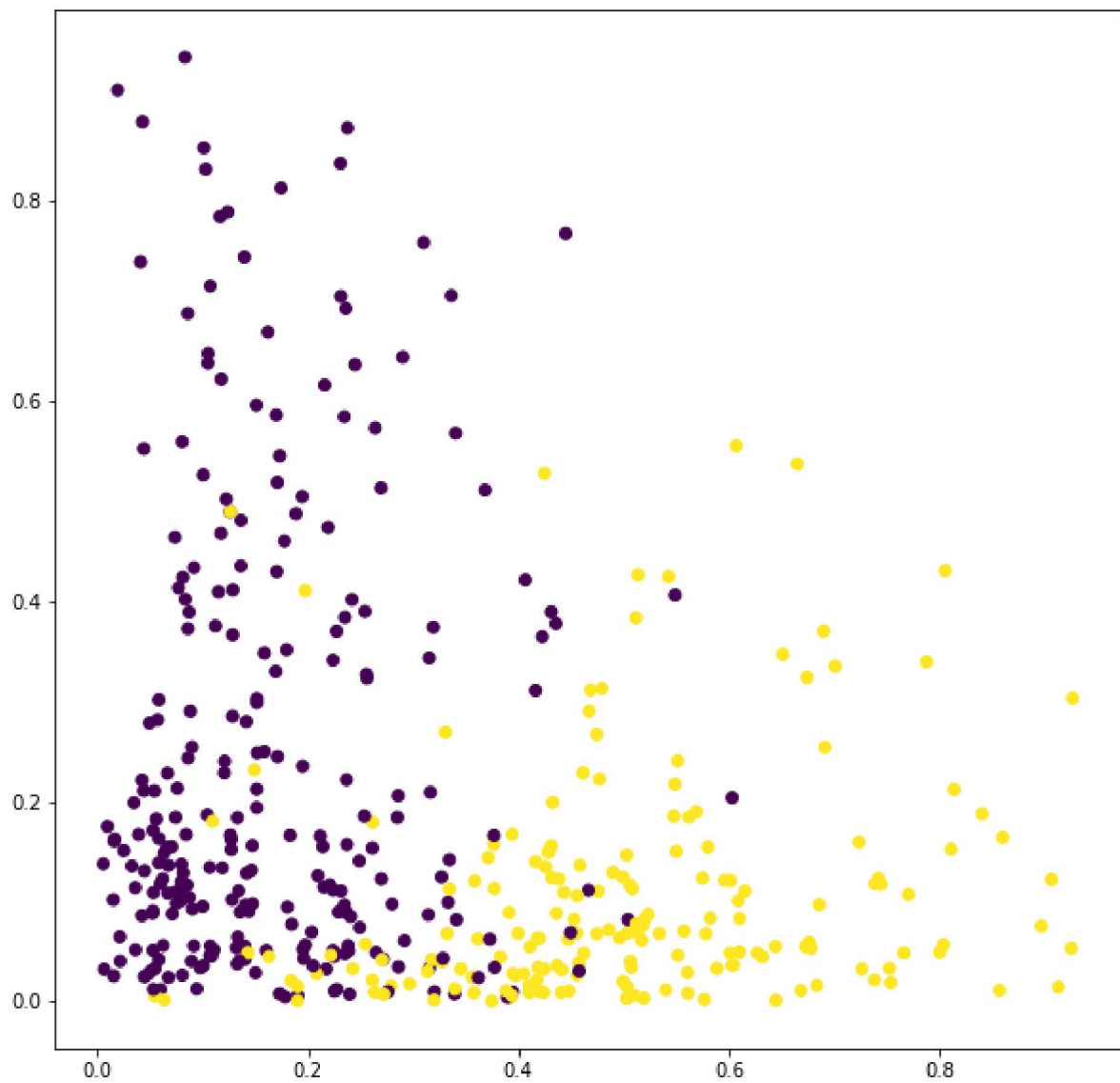
```
In [13]: plt.figure(figsize=(10,10))
plt.scatter(new_data["Milk"],new_data["Grocery"],c=cluster.labels_)
```

```
Out[13]: <matplotlib.collections.PathCollection at 0x2af5ef46130>
```



```
In [14]: plt.figure(figsize=(10,10))  
plt.scatter(new_data["Milk"],new_data["Frozen"],c=cluster.labels_)
```

```
Out[14]: <matplotlib.collections.PathCollection at 0x2af5d065f40>
```



In []: