* **DBScan Clustering**

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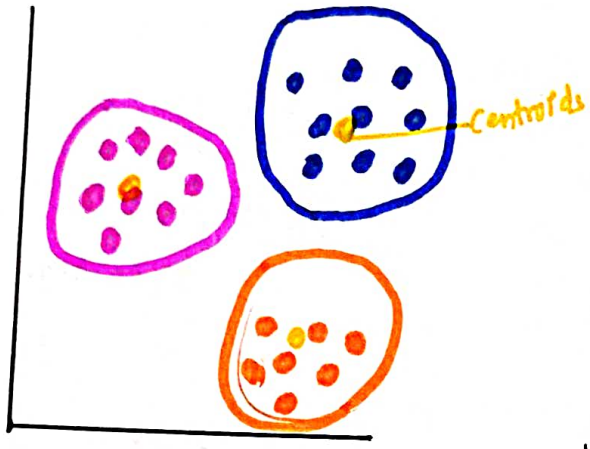
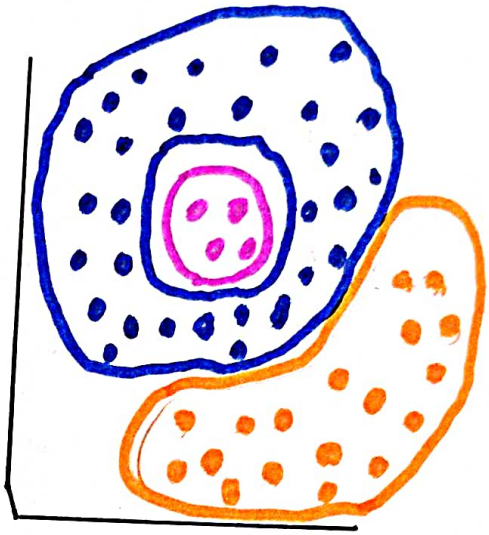
Density Based Spatial Clustering of application with noise.

Density-Based clustering

* Spherical-shape clusters

* Arbitary shape clusters





- K-means assigns all points To a cluster Even if they do not belong in any

- Density based clustering Locates regions of "high density", and seperates outliers.

*** Based on Given density of values it is identifying the clusters

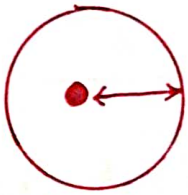
* What is DBSCAN?

* DBSCAN

- it is one of the most common clustering algorithms
- Works based on density of objects.

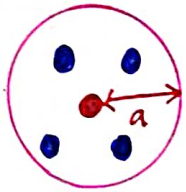
* R (radius of neighborhood)

- Radius (R) that if includes enough number of points within, we call it a dense area.

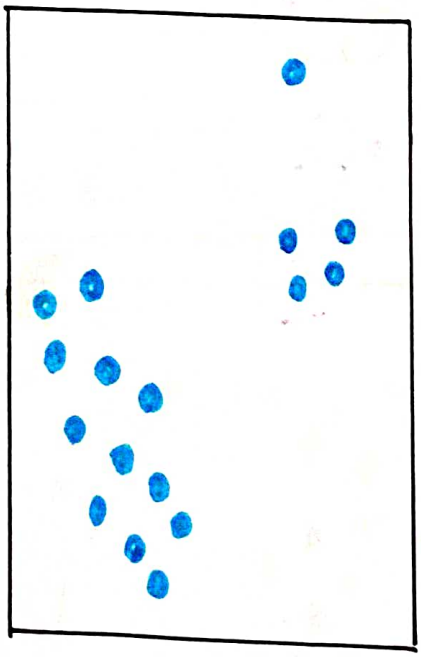


* M (min number of neighbors)

- The minimum number of data points we want in a neighborhood to define a cluster.



Ex:-



Each point is either:

- core Point
- border point
- Outlier point

$\therefore R = 2 unit, M = 6$

* DBSCan algorithm :: Core point

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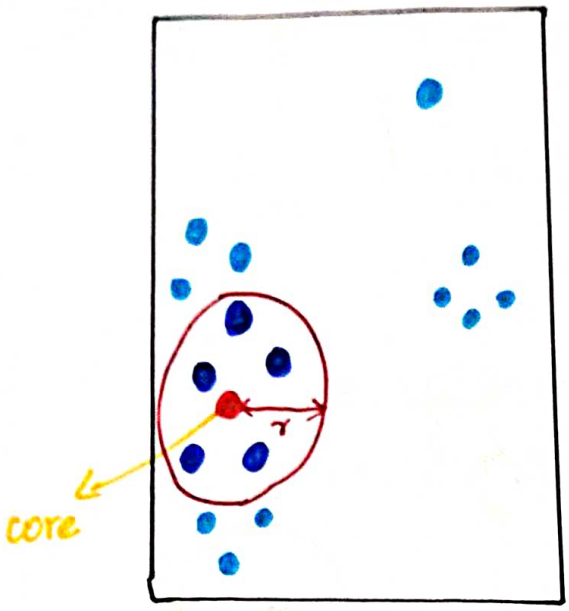
which is satisfy minimum no. of neighbours.



R = 2 unit, M = 6

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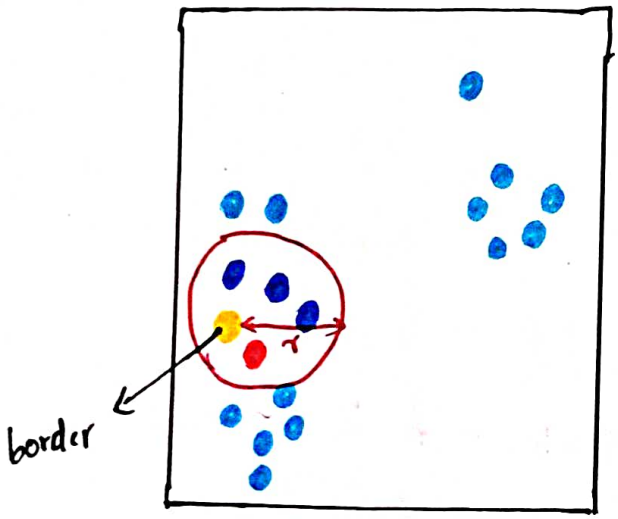
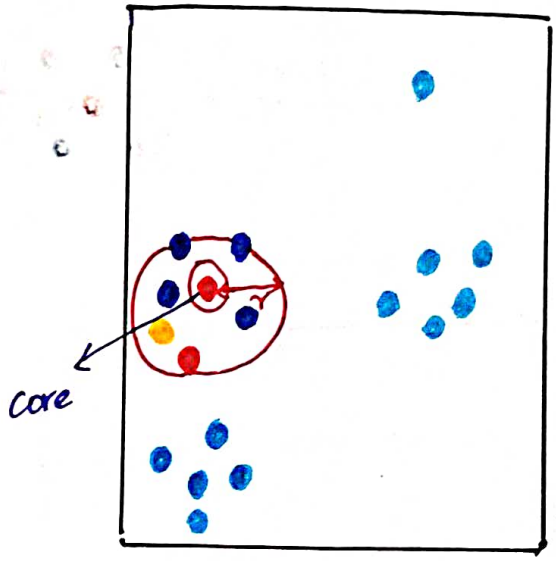
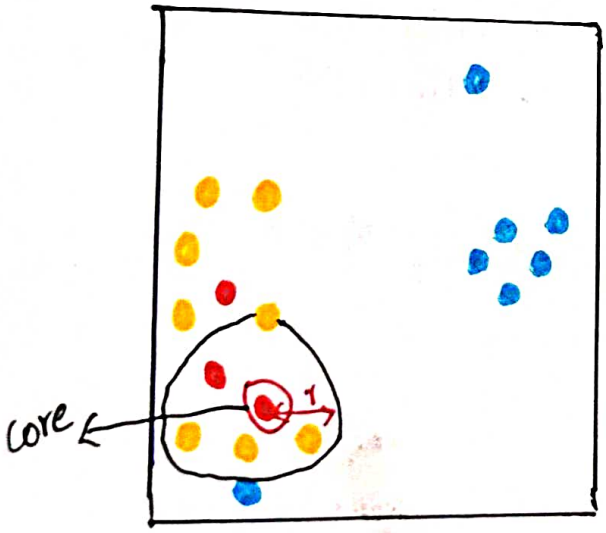
minimum.

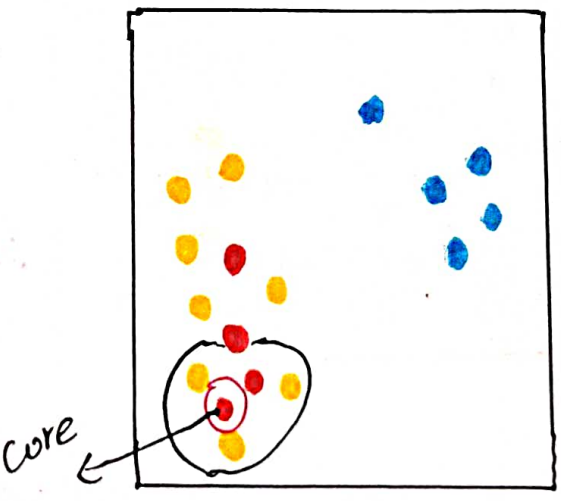
* DBSCan algorithm : border point

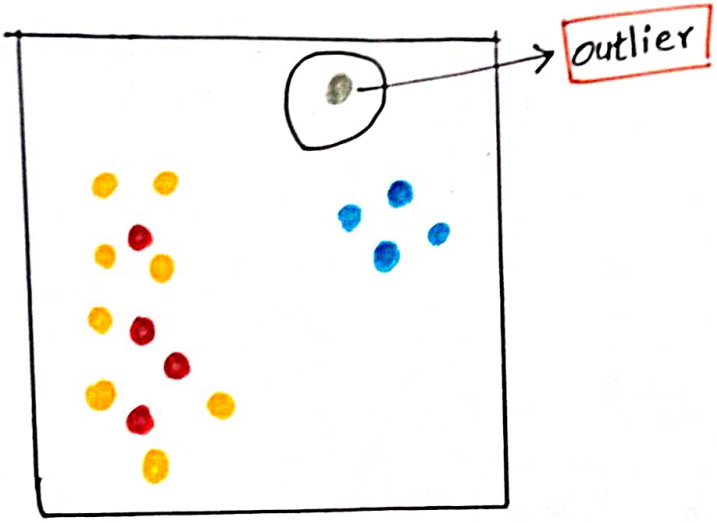
R = 2 unit , M = 6

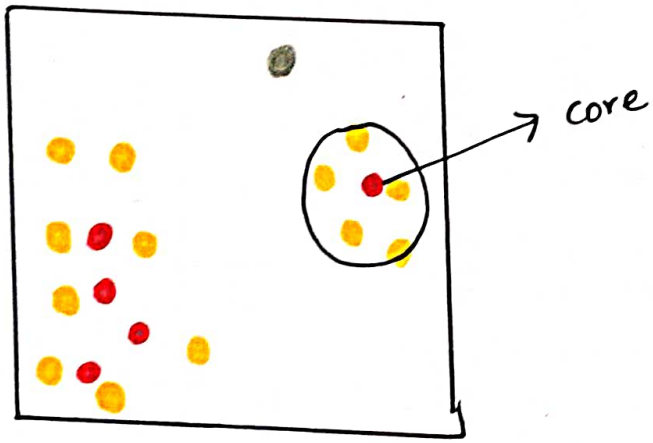


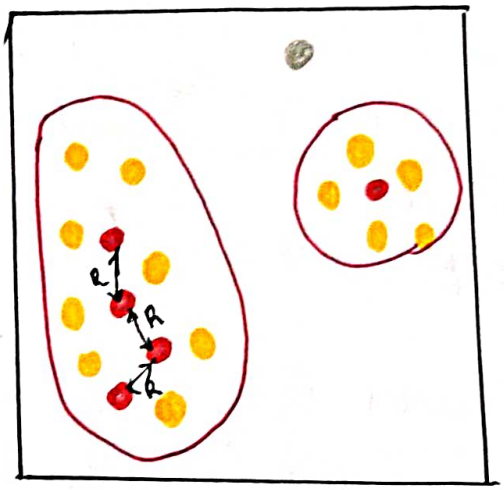












R = 2unit , M = 6

CODE : Same data & Same steps till modelling

MODEL: DBSCAN

```
from sklearn.cluster import DBSCAN
# dbs = DBSCAN (eps = 5 , min_samples = 5)
                ↳ Radius.
```

Predict

```
# y_dbs = dbs.fit_Predict (x)
```

y-dbs

Out : array([-1, 0, -1, 0, -1, 0, -1, -1, 0, 0, -1
- - - - - - - - - - - - -
- - - - - - - - - - - -])].

np. unique (y-dps)

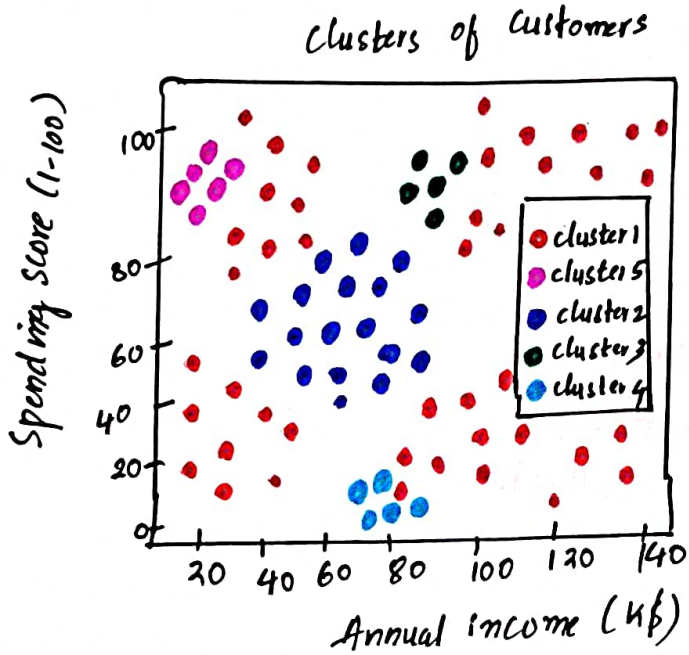
Out: array ([-1, 0, 1, 2, 3, 4]).

Visualising the clusters

```
# cluster 1
# plt.scatter (x[y-dbs == -1, 0], x[y-dbs == -1, 1],
      s=100, c = "red", label= "cluster 1")

# cluster 5
# plt.scatter (x[y-dbs == 0, 0], x[y-dbs == 0, 1],
      s = 100, c = "magenta", label= "cluster 5")

# cluster 2
# plt.scatter (x [y-dbs == 1, 0], x[y-dbs == 1, 1],
      s= 100, c = "blue", label= "cluster 2")

# cluster 3
# plt.scatter (x [y-dbs == 2, 0], x[y-dbs == 2, 1],
      s=100, c = "green", label = "cluster 3")

# cluster 4
# plt.scatter (x[y-dbs == 3, 0], x[y-dbs == 3, 1],
      s=100, c = "cyan", label = "cluster 4")
```

```
#plt.title("clusters of customers")
# plt.xlabel("Annual income (k$)")
# plt.ylabel("spending score (1-100)")
#plt.legend()
# plt.show()
```

Out:



04/05/22
5:00pm.