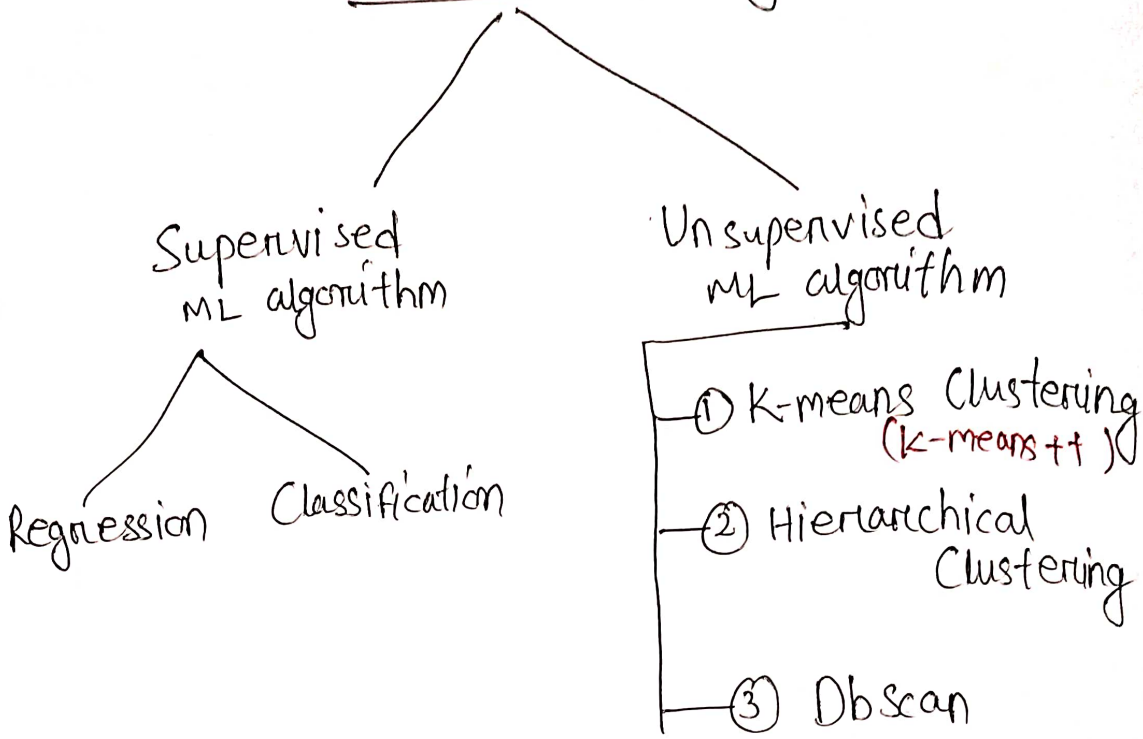
Clustering

Machine Learning



In case of Supervised

Height (I/P)	weight (I/P)	BMI (O/P)
180	70	21
160	80	23
150	75	23
140	90	24
130	100	25
170	60	19

In case of Unsupervised

Height	weight	BMI	
180	70	normal	①
160	45	underweight	②
150	70	normal	①
140	90	normal	①
130	100	overweight	③
170	60	normal.	①

→ we can make cluster by using BMI values.

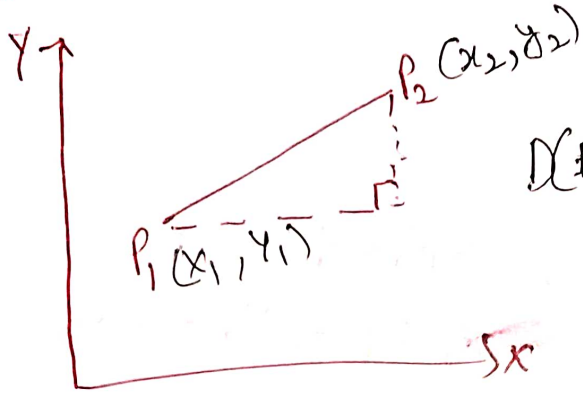
-) ③ groups

① K-means

→ Unsupervised algorithm works by calculating similarity score.

-) ① Euclidean distance. ③ cosine similarity.
② manhatten distance.

① Euclidean distance



$$D(P_1, P_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

② Manhatten distance

$$D(P_1 - P_2) = |x_1 - x_2| + |y_1 - y_2|$$

③ cosine similarity

$$\cos(\theta) = \frac{A \cdot B}{|A| |B|}$$

	Height	weight	
	185	72	C_1
	170	56	C_2
3	168	60	
4	179	68	
5	182	72	
6	188	77	
7	180	71	
8	183	84	
9	180	88	
10	180	67	
11	167	76	

→ Clustering using K-means.

Step ① Initialization of centroid

In case of k-means centroid initialization is random. Let

Let K = 2 (K = no. of centroid)

Lets assume 1st and 2nd row as C_1 and C_2

Step ② Euclidean distance.

So to make cluster we have to find euclidean distance to other point from C_1 and C_2 (two cluster)

$$D(C_1, 3) = \sqrt{(185-168)^2 + (72-60)^2}$$

$$= \sqrt{(17)^2 + (12)^2} = 20.8$$

$$D(C_2, 3) = \sqrt{(168-170)^2 + (60-56)^2}$$

$$= \sqrt{2^2 + 4^2} = 4.4$$

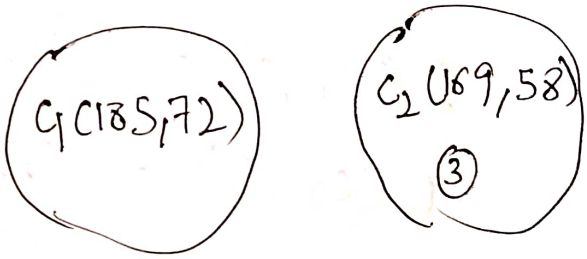
so $D(C_2, 3) < D(C_1, 3)$

As (3) belongs to (C2) (cluster-2) C_2 need to be updated.

step-3

$$\text{new } C_2 = \left(\frac{170+168}{2}, \frac{56+60}{2}\right)$$

$$= (169, 58)$$



Again step (2)

$$D(C_1, 4) = \sqrt{(185-179)^2 + (72-68)^2}$$

$$= \sqrt{6^2 + 4^2} = 7.07$$

$$D(C_2, 4) = \sqrt{(169-179)^2 + (58-68)^2}$$

$$= \sqrt{10^2 + 10^2} = 14.14$$

As $D(C_1, 4) < D(C_2, 4)$

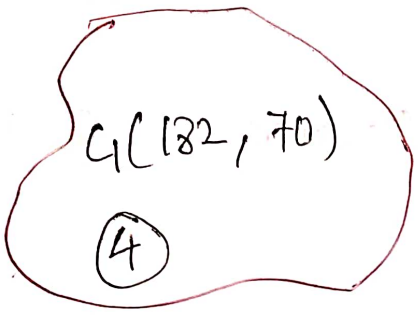
so (4) belongs to C_1 (centroid-1)

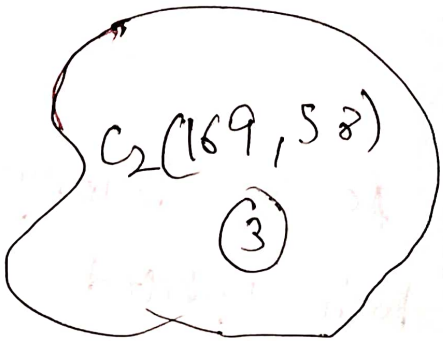
Step-3

As 4 belongs to c_1, c_1 centroid needs to update

new $c_1 = \frac{185+179}{2}, \frac{72+68}{2}$

$= (182, 70)$





Again step (2)

$D(C_1, 5) = \sqrt{(182-182)^2 + (72-70)^2} = 2$

$D(C_2, 5) = \sqrt{(182-169)^2 + (72-58)^2}$

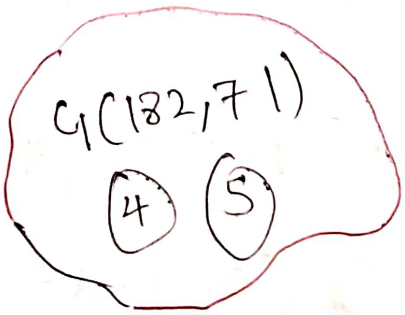
$= \sqrt{(13)^2 + (14)^2} = 19.1$

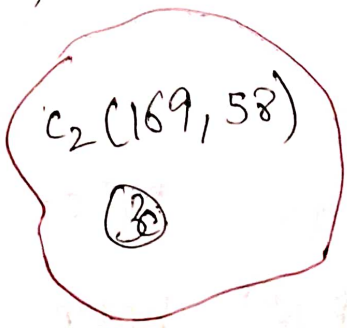
As $D(C_1, 5) < D(C_2, 5)$

↓ step (3) Repeted.

new $c_1 = \left(\frac{182+182}{2}, \frac{70+72}{2}\right)$

$= (182, 71)$





Steps

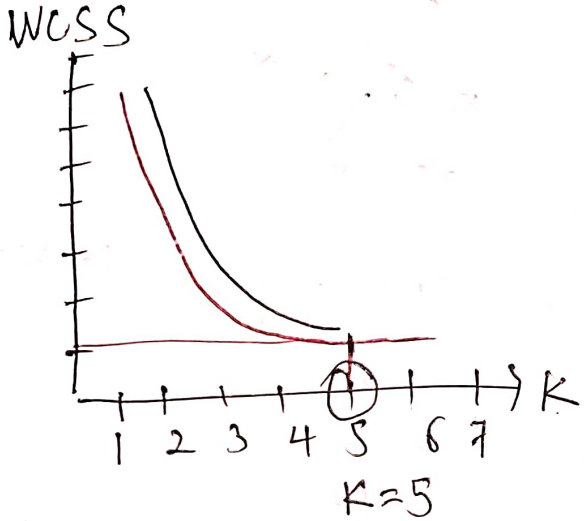
(1) centroid

(2) Distance (compare minimum distance)

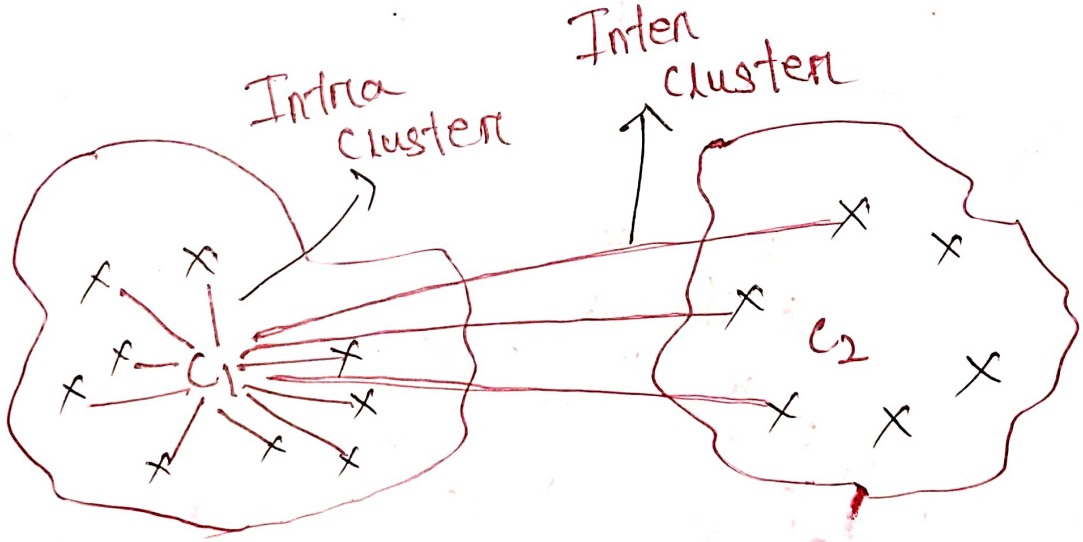
(3) include point in cluster & update centroid.

Q. How to decide k value ?

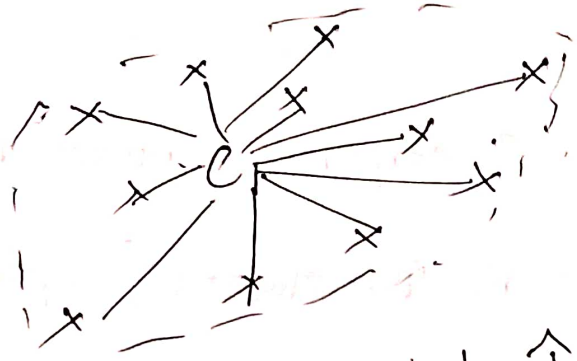
Ans. ELBOW method.



WCSS → within-cluster sum of square.

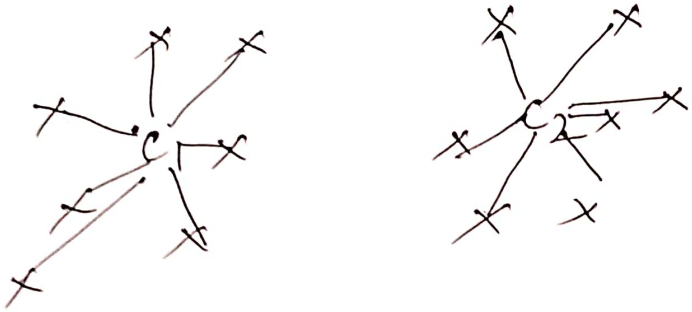


let K = 1



$WCSS_1$ = high ↑

K = 2

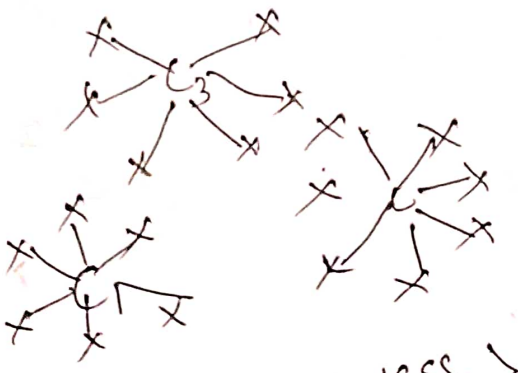


$WCSS_2$ = Low (As compare to K=1)

$$WCSS = \sum_{i=1}^{n} d(C_1, x_i)^2$$

$WCSS_1 > WCSS_2$

K = 3



$WCSS_1 > WCSS_2 > WCSS_3$

wcss is minimum at specific value of K
assume (K=5)

* How to validate cluster value (K=5)

① Dunn index.

② Silhouette score

① Dunn index $= \frac{\max dist(x_i, x_j)}{\max dist(y_i, y_j)}$

② Silhouette score $= \frac{b_i - a_i}{\max(b_i - a_i)} \rightarrow [-1 \text{ to } 1]$

In silhouette score

a_i = intra cluster distance

b_i = inter cluster distance.

If $a_i > b_i$

silhouett score -ve (worst value)

→ In case of K-means ++, random initialization of K-means problem is solved.

→ Kmeans is a centroid based algorithm.