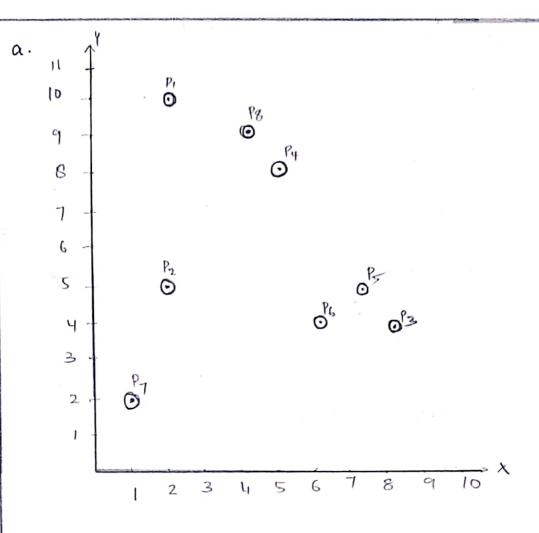
1



 $C_1: \int P_7, P_2 \mathcal{J} \quad C_2: \int P_1, P_8, P_4 \mathcal{J} \quad C_3: \int P_5, P_6, P_3 \mathcal{J}$ $C. \quad Centres: \quad \begin{pmatrix} C_1 & C_2 & C_3 \\ 2,5 & (5,8) & (4,9) \end{pmatrix}$ 3 clusters

		Distance			
	Point	C,	C_2	C3	Belongs to
Pı	(2,10)	5	V13	15	Cg
P2	(2,5)	. 0	-	-	C,
Pg	(8,4)	V37	5	141	C2
Pц	(5,8)	_	6	•	C ₂
P ₅	(7,5)	5	V13	5	C2
PL	(614)	V17	V17	V29	Cl
P-1	(1,2)	VID	(52	V58	C1
Po	(4,9)	_	_	D	C3

C1: { P2, Pc, P73 C2: 1 13, P4, P53 C3: \$ P1, P8 3

d) Centre og cluster after 1st uteration:

Cluster 1: (P2, P6, P7)

Cluster 2: (P3, P4, P5)

Cluster 3: (P1, P8)

Center 1: $\frac{2+6+1}{3}$, $\frac{5+4+2}{3} = 3,3.66$

Center 2: $\frac{8+5+7}{3}$, $\frac{4+8+5}{3}$ = 6.66, 5.66

Cuntur 3: $\frac{2+4}{2}$, $\frac{10+9}{2}$ = 3, \$\frac{3}{5}\$.5

C1 center: {3,3.66}

C2 Center: \$6.66, 5.66}

c3 center: \$3,9.53

e Center of ceuster afect 2nd iteration:

P1- C3, P2-C1, P3-C2, P4-C3, P5-C2, P6-C2 P4-C6, P8-C3

Christia 1: { P2. J Christia 2: { P3, P4, P5 } christia 3: {

clustre 1: §P2, P73

Clusta 2: f P3, P5, P63

Cluster 3: & P1, P4, P83

$$C_1: \frac{3}{2}, \frac{7}{2} = (1.5, 3.5)$$

$$C_2: 8 + 7 + 6, \frac{4 + 5 + 4}{3} = (7, \frac{13}{3})$$

$$C_3: \frac{2+5+4}{3}, \frac{10+8+9}{3} = (\frac{11}{3}, 9)$$

$$P_1 - C3$$
, $P_2 - C1$, $P_3 - C2$, $P_4 - C3$, $P_5 - C2$, $P_6 - C2$, $P_7 - C1$, $P_8 - C3$

Center after 3rd iteration: Same as 2nd iteration centres.

The results are similar to what was quessed in section b.

R. 3 iterations

Resulting Centres: (1.5,3.5), $(7,\frac{13}{3})$, $(\frac{11}{3},9)$

Resulting Clusters:

Single link clustering

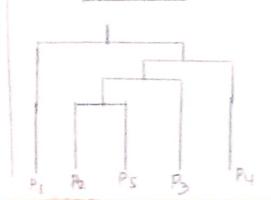
P2, P5 share Right similarity Merge P2 P5

	Pi	P ₃	Py	P2 P5
P,	1.0	0.41	0.55	0.35
P3	0.41	1.0	0.44	0.85
P4	0.55	D-44	1.0	0.76
P2 P5	0.35	0.85	0.76	1.0

P3 P2 P5 share arguest Similarity Merge P3. P2P5

	P_1	24	3/2"
Pı	1.0	0.55	0-41
P4	0.55	1.0	0.76
3 P2 P5	0.41	6.76	1.0

Merge Py, P3P2P5 (0.76)
P1 P4P3P3P5
P1 1.0 0.55
P4P3P4P5 0.55 1.0



Compell Link Clustering

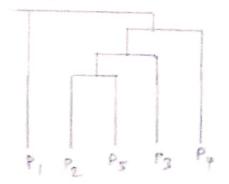
P2 P5 Share high similarity Merge P2P5

	PI	P3	Py	P2 PS
Pi	1.0	0.41	0.55	0.10
93	0.41	1.0	0-44	0.64
Py	0.55	0.44	1.0	0.47
P2 P5	0.10	0.64	0.47	1.0

P3, P2P5 share high sim. Merge P3, P2P5

	Pr	Py	P3 P2	
P_1	1.0	0.55	0.10	
Py	0.55	1.0	0.44	
P3 P2 P5	0.10	0.44	1.0	

Py P3 P2 P5 Share bright Sim. Merge P4, P3 P2 P5 P1 P4 P3 P2 P5 P1 1.0 0.10 P4 P3 P2 P5



Both the methods murge the points in a semilar way as we see that the above two clindograms are same.

Epsilon = 2 Men-Samples = 2

a. There are three clusters discovered.

Chusta 1: (2,10) (2,5) (1,2)

Ceuster 2: (8,4) (7,5) (6,4)

Cluster 3: (5,8) (4,9)

b. Epsilon = Vio

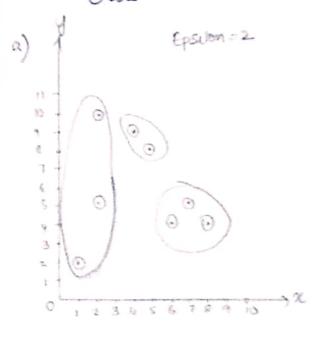
3.

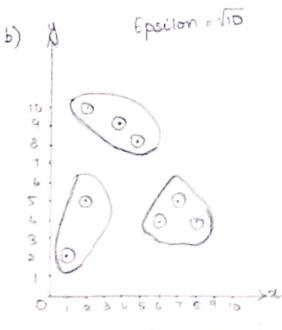
there are three clusters

Clusta 1: (2,10), (5,8), (4,9)

Cluster 2: (2,5), (1,2)

Cluster 3: (8,4), (7,5), (6,4)





i) Cassandra Delongs to "Database" Category of tech Stack

ii) Primary reason: "Distributed"

Rows are organised into tably with a required primary key.

iv) Cassandra automatically Clistributes data across Multiple machines as Cluster size changes Big Table

Big Table Classified as "NOSOL Database as a Service" Primary Reason: "High performance"

A fast, fully managed, Scalable NOSQL database service cideal for web, Mobile, IOT Apps requiring TB to PB by data.

Bigtable has been widely used in Google Analytics & Gmail.

62. Apache cassandra is a distributed DBMS that is built it a handle large amounts of data across multiple data centres and the Cloud.

key features:

- > Highly Scalable
- -> offers high availability
-) Has no single point of failure.

It is a Nosal DB meaning DB stores & retrieves data without requiring data to be stored in Tabular format

08 Tunable Consistency in Cassandra:

Apacine cassandra is a "AP" system which means ut prefers data availability over consistency. To ensure data availability, the data updates should be propagated across networks to remote hosts. If the hosts are down, it may take time ito update the data & lisers may read stale / not up to date information. To avoid flies Tunable Consistency is rused.

When performing a read/write operation a database Client can specify a consistency level. The Consistency level refers to the no. of replicas that ned to respond for a read or vorite operation No be considered complete.

For a less important data, if is set to ONE, For accuracy driven data, Two, THREE OR QUORUM.

Oyl Memfable:

When a Write occurs, Cassandra stores the date un a memory stracture called Memtable. The Memtable is a write-back Cache of data partitions that cassandra looks up by key. The Mentable stores writes in sorted order until reaching a configurable limit, & then it is flushed. Ot. Setably are the immutable data files Abat

Calsandra uses for pursisting data on disk.

When Stables are flushed to disk from Memtables

or are streamed from other nocles, Cassandra

triggers Compactions which combine multiple

Stables into one.

SS- Storting String Table is a felle of key/value String pairs, sorted by keys. SS-table-Immutable, Relational table-Can't be

updated, edited & so on.

Q6. CAP theorem (Brewer's theorem)

CAP theorem status that a distributed system Com only guarantee two out of these three Characteristics: Consistency, Availability and Partition tolerance.

Consistency CP Partition
Tolurance

CA AP

Availability

Ø7

A tablet server stores and serves tablets to Clients. For a given tablet, one tablet server acts as a leader and the others serve follower repeicas of that tablet.

One tablet server can serve multiple tablets, and one tablet can be served by multiple tablet servers

A tablet is a contiguous segment of a table, similar to a partition un other data storage engines or relational clatabases.

A given tablet is repeicated on multiple tablet servers and at a given pt in time, one of these servers and at a given pt in time, one of these replicas is considered the leader tablet.

Any replicas can service reads. Writes require consenses among set of tablet servers serving the tablet.