

# **WHAT IS K-MEANS CLUSTERING?**

- **Definition:** K-Means is a clustering algorithm that groups data into  $k$  clusters based on similarity.
- 
- **Purpose:** To partition the data such that points within each cluster are similar to each other and dissimilar to points in other clusters.
- 
- 

## **~Applications**

- \*Market segmentation**
- \*Image compression**
- \*Anomaly detection**

# GAUSSIAN MIXTURE MODELS (GMM).

## OVERVIEW:

- MODELS DATA AS A MIXTURE OF GAUSSIAN DISTRIBUTIONS.
- EACH CLUSTER IS REPRESENTED BY A GAUSSIAN COMPONENT.
- USES EXPECTATION-MAXIMIZATION (EM) ALGORITHM FOR PARAMETER ESTIMATION.

STRENGTHS: FLEXIBLE, CAN HANDLE OVERLAPPING CLUSTERS,  
PROBABILISTIC APPROACH

WEAKNESSES: CAN BE COMPUTATIONALLY EXPENSIVE, SENSITIVE  
TO INITIALIZATION

# DBSCAN (DENSITY-BASED SPATIAL CLUSTERING OF APPLICATIONS WITH NOISE).

## OVERVIEW:

- GROUPS TOGETHER POINTS THAT ARE CLOSELY PACKED TOGETHER.
- IDENTIFIES CLUSTERS BASED ON DENSITY THRESHOLDS CAN HANDLE CLUSTERS OF ARBITRARY SHAPE.

STRENGTHS: ROBUST TO NOISE, DISCOVERS CLUSTERS OF VARYING SHAPES AND SIZES

WEAKNESSES: SENSITIVE TO PARAMETER SELECTION (EPSILON AND MINPTS), CAN BE INEFFICIENT FOR LARGE DATASETS

# COMPARISON: K-MEANS VS. GMM

<b>Feature</b>		<b>K-Means</b>		<b>Gaussian Mixture Models</b>	
Model Assumption		Spherical clusters		Arbitrary-shaped clusters	
Clustering Approach		Partitional		Probabilistic	
Parameter Estimation		Simple		EM algorithm	
Sensitivity to Initialization		Yes		Yes	

# COMPARISON: K-MEANS VS. DBSCAN

Feature	K-Means	DBSCAN
Cluster Shape	Spherical	Arbitrary
Noise Handling	Less robust	Robust
Parameter Sensitivity	Sensitive to K	Sensitive to epsilon and minPts
Outlier Detection	Implicit	Explicit

## **WHEN TO USE WHICH ALGORITHM:**

**K-MEANS: SIMPLE, EFFICIENT, WORKS WELL FOR  
WELL-SEPARATED CLUSTERS**

**GMM: FLEXIBLE, SUITABLE FOR OVERLAPPING  
CLUSTERS, PROBABILISTIC APPROACH**

**DBSCAN: ROBUST TO NOISE, HANDLES CLUSTERS  
OF ARBITRARY SHAPES, GOOD FOR OUTLIER  
DETECTION**