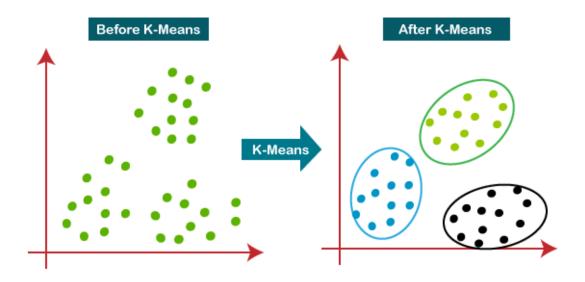
# Data Science Clustering Techniques



# **K-MEANS CLUSTERING**

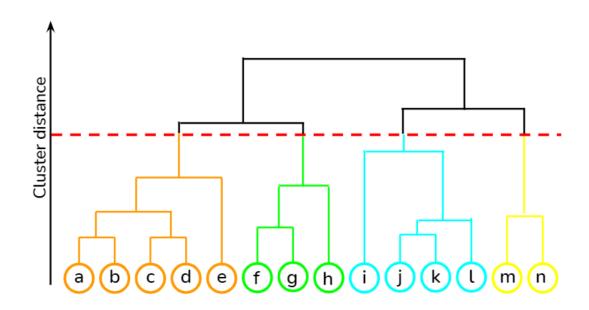


A centroid-based method that partitions data points into a pre-defined number of clusters (k). Simple, efficient, but sensitive to outliers and initialization.





# HIERARCHICAL CLUSTERING

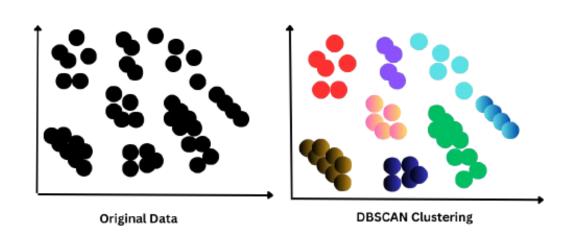


Builds a hierarchy of clusters, either merging (agglomerative) or splitting (divisive). Good for exploratory analysis and identifying natural groupings of varying sizes.





#### **DBSCAN**

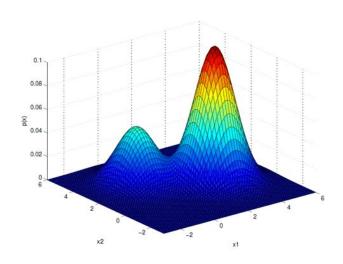


Identifies clusters based on density of data points. Robust to outliers and noise, but may struggle with clusters of varying densities.





# **GAUSSIAN MIXTURE MODELS**

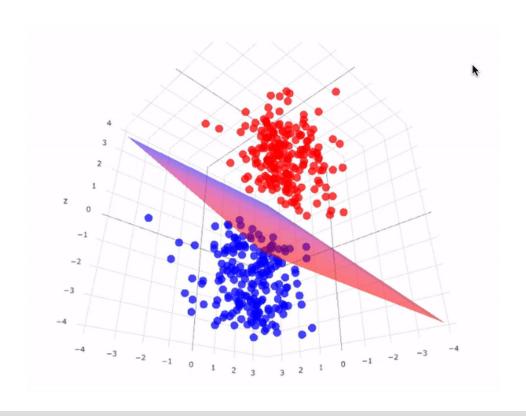


Assumes data points are generated from a mixture of Gaussian distributions, where each distribution represents a cluster. Flexible for complex data shapes, but requires parameter tuning.





# **SVM CLUSTERING**

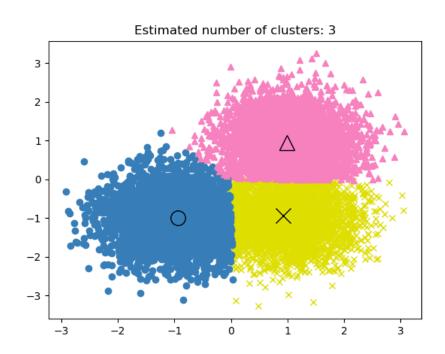


Uses support vectors to define cluster boundaries. Effective for high-dimensional data and non-linear clusters, but computationally expensive.





# **MEANSHIFT CLUSTERING**

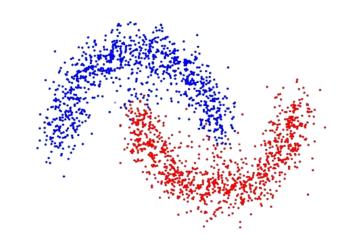


Iteratively shifts data points towards regions of higher density, ultimately converging to cluster centers. Useful for complex data shapes and identifying overlapping clusters.





# **SPECTRAL CLUSTERING**



Projects data into a lower-dimensional space using dimensionality reduction techniques and then performs clustering in that space. Effective for high-dimensional data and non-linear clusters.



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