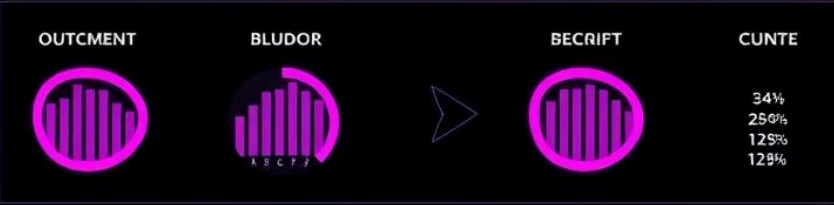
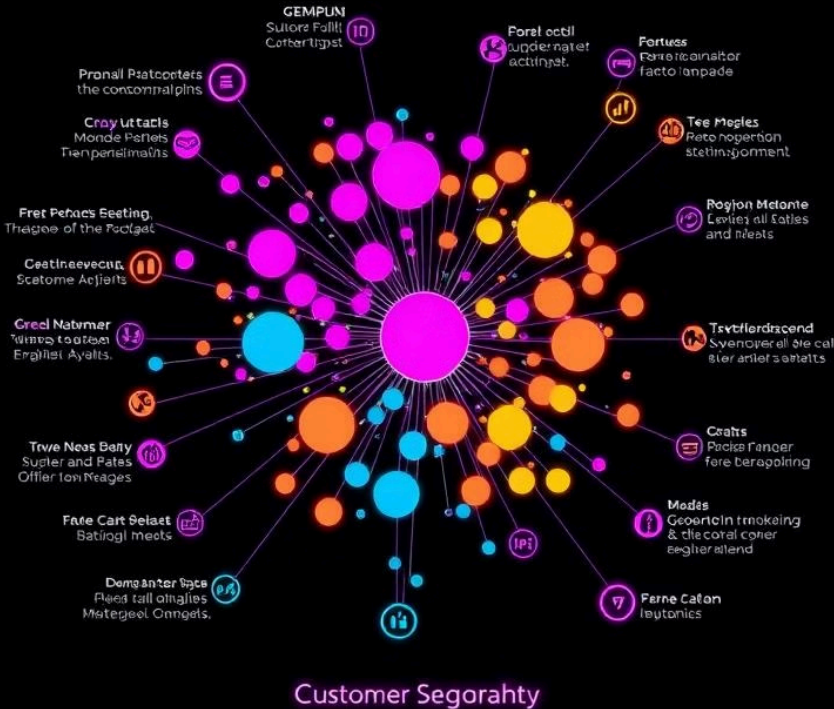


Customer Segmentation Analysis

This presentation will guide you through a customer segmentation analysis utilizing clustering techniques. We will explore the process of identifying distinct customer groups based on their profile and transaction data, and then discuss the evaluation of these clusters using relevant metrics. We will finish by visualizing the clusters for insightful interpretation.

CUSTOMER SEGINTATION USCOMES SEESDNATION

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Data Preparation and Feature Engineering

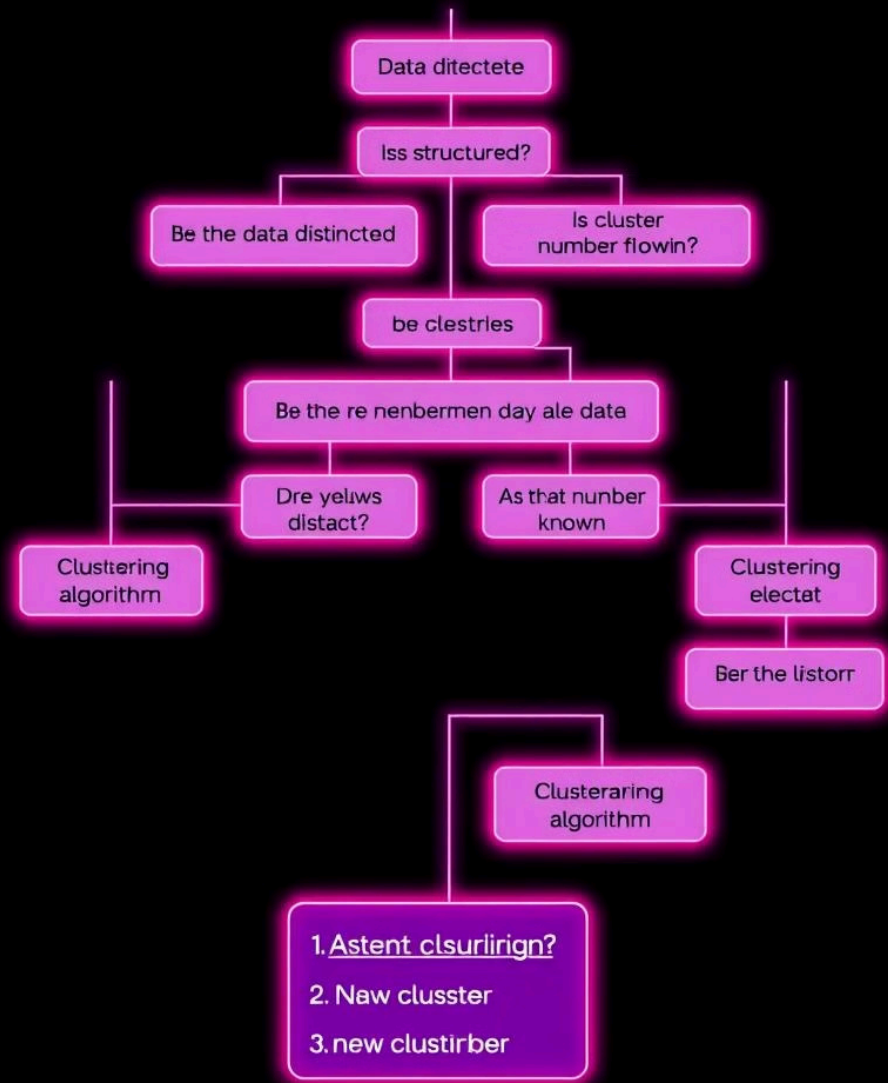
Data Acquisition

We begin by acquiring the customer and transaction datasets, ensuring they are complete and clean. Missing values are addressed, and data types are verified for consistency.

Feature Engineering

Features are extracted from the raw data to create a comprehensive representation of each customer. This includes demographics, transaction history, purchase behavior, and other relevant variables.

Clustering Algorithm Selection



1

K-Means Clustering

K-Means is a popular choice due to its simplicity and scalability. It partitions data into K clusters based on distance from cluster centroids.

2

Hierarchical Clustering

Hierarchical clustering builds a hierarchy of clusters based on similarity. It offers flexibility in exploring cluster structures at different levels.

3

DBSCAN (Density-Based Spatial Clustering of Applications with Noise)

DBSCAN is particularly effective for identifying clusters of varying densities and shapes. It groups data points based on their density and proximity.

Clustering Implementation and Evaluation

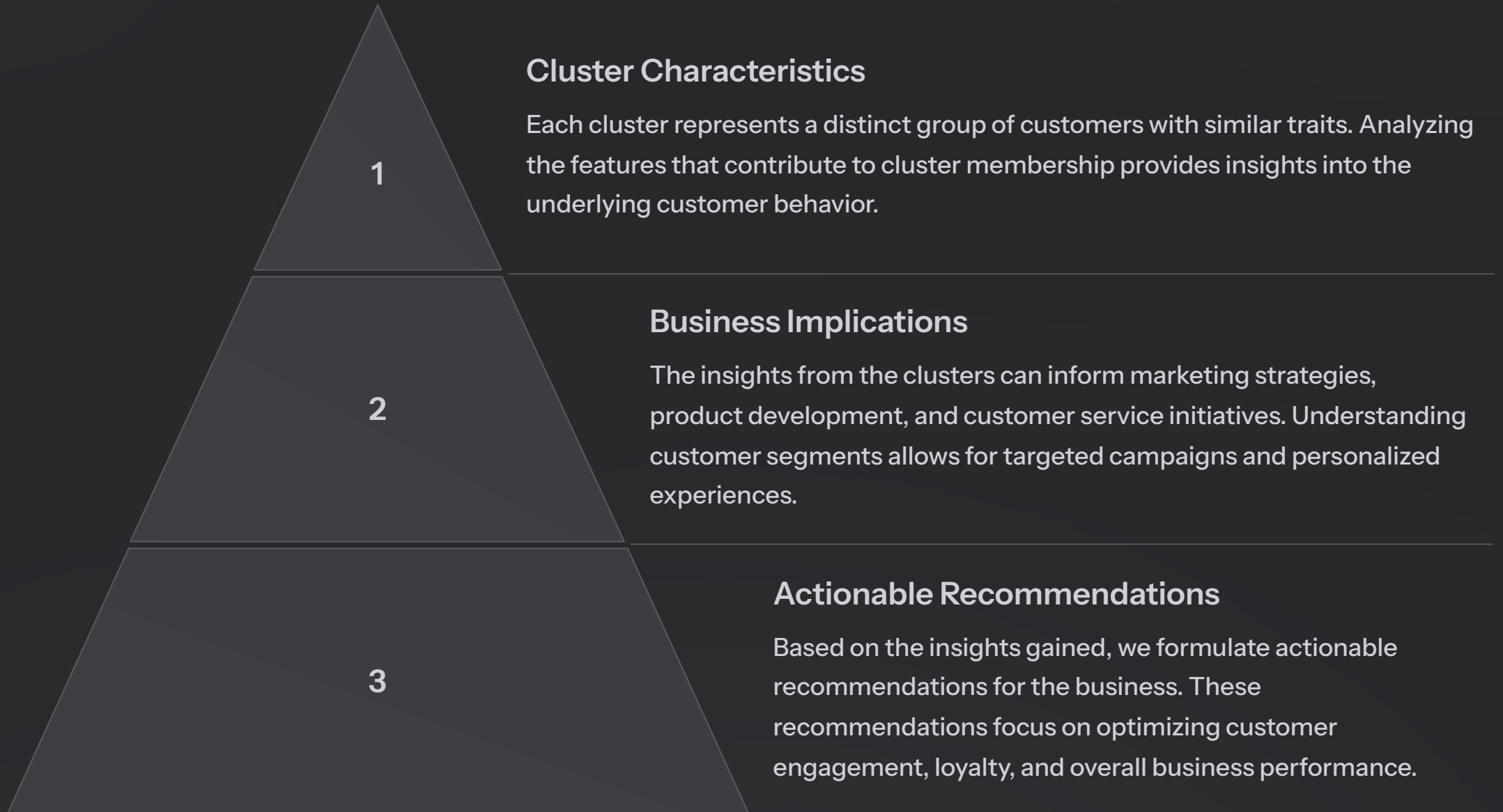
Clustering Execution

The chosen algorithm is implemented with appropriate parameter settings. We select an optimal number of clusters based on the nature of the data and the desired segmentation granularity.

Clustering Metrics

The quality of the clusters is evaluated using metrics like the Davies-Bouldin Index, Silhouette Score, and Calinski-Harabasz Index. These metrics assess the separation and compactness of the clusters.

Interpreting the Clusters



Visualizing the Clusters



Scatter Plots

Scatter plots visualize the distribution of customers in a two-dimensional space, highlighting cluster separation and identifying outliers.



Pie Charts

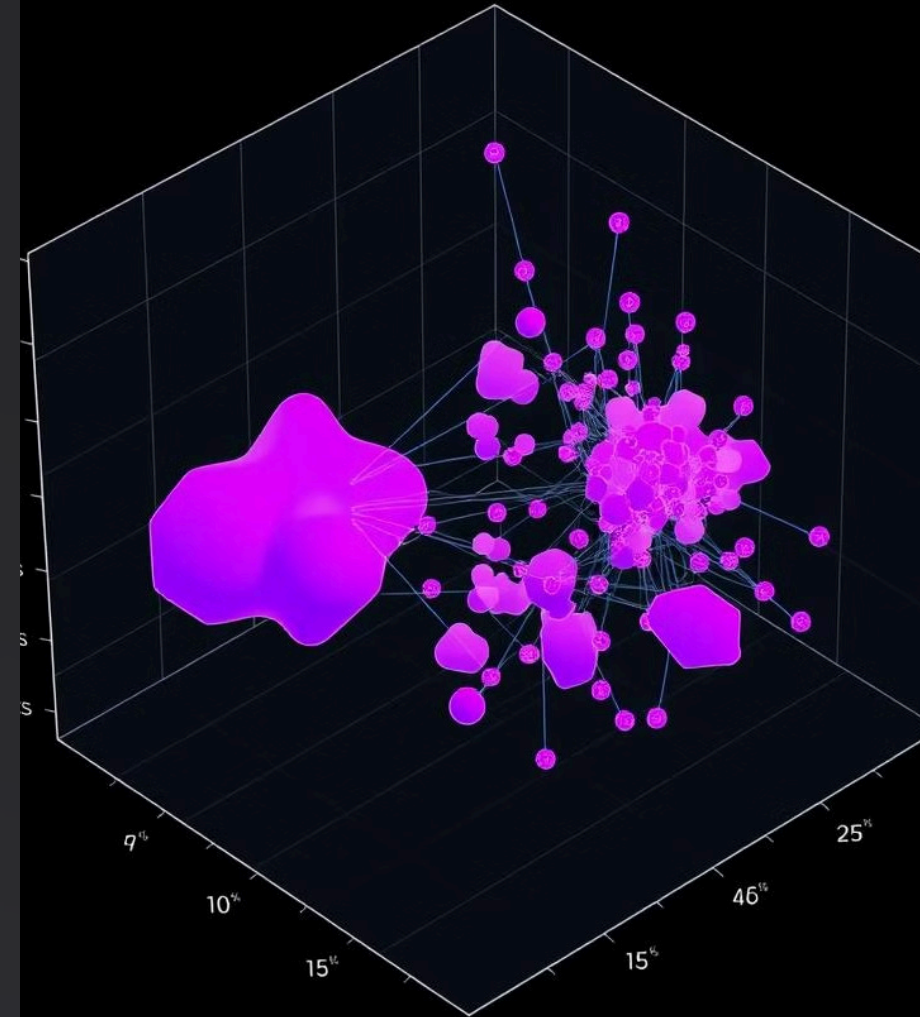
Pie charts provide a visual representation of the percentage of customers belonging to each cluster, offering a quick overview of the segment distribution.



Bar Charts

Bar charts represent the distribution of customers across different clusters, illustrating the relative sizes and proportions of each segment.

Cusmer Clusters



Report and Deliverables

1

Report Structure

The final report includes an introduction, a detailed description of the methodology, the clustering results, cluster interpretations, and actionable recommendations.

2

Clustering Metrics

The report presents key clustering metrics, including the Davies-Bouldin Index, Silhouette Score, and Calinski-Harabasz Index, to quantify cluster quality.

3

Visualizations

Supporting visualizations such as scatter plots, bar charts, and pie charts are included to visually represent the clusters and their characteristics.



Conclusion and Next Steps

Through careful data preparation, clustering algorithm selection, and comprehensive evaluation, we have successfully identified distinct customer segments. These segments offer valuable insights into customer behavior and preferences. The findings provide a solid foundation for developing targeted marketing campaigns, improving customer service, and optimizing business strategies. In the future, we can explore advanced clustering techniques, incorporate additional data sources, and continuously monitor cluster dynamics to ensure the insights remain relevant and actionable.

Thankyou