

# Spending Analysis Project Report

## Objective

The primary objective of this project was to analyze quarterly relationships between economic indicators - Consumer Price Index (CPI) and Spending Momentum Index (SMI) - and e-commerce spending. By leveraging unsupervised learning methods like K-Means and DBSCAN, we sought to uncover meaningful clusters, identify outliers, and provide actionable insights for businesses and stakeholders.

## Summary of Key Findings

### 1. K-Means Clustering:

- Optimal number of clusters: 5
- Silhouette Score: 0.63, indicating good cluster separation.
- Captured distinct groupings of quarters based on CPI, SMI, and spending but forced all data points into clusters, including outliers.

### 2. DBSCAN Clustering:

- Optimal Parameters:  $\text{eps}=1.5$ ,  $\text{min\_samples}=5$
- Silhouette Score: 0.63, slightly outperforming K-Means.
- Strengths:
  - Identified meaningful clusters while preserving outliers as a separate category.
  - Highlighted anomalies in quarters (Cluster -1) with significantly low CPI, SMI, and e-commerce spending.

### Cluster Insights:

- Cluster 0 (Majority Behavior): Represents typical quarters with moderate CPI, SMI, and spending.
- Cluster -1 (Outliers): Captures quarters characterized by low CPI, SMI, and e-commerce spending,

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such as Q2 and Q3 2020 during the COVID-19 pandemic.

### Recommendations

#### 1. Analyze Outliers:

- Investigate further economic events contributing to anomalies in Q2 and Q3 2020.
- Consider external factors like policy changes, stimulus efforts, and consumer confidence.

#### 2. Data Enhancement:

- Include additional features like employment rates, wage growth, and consumer confidence index.

#### 3. Modeling Improvements:

- Apply time-series models to explore lagged relationships between CPI, SMI, and spending.
- Experiment with advanced clustering techniques, such as Gaussian Mixture Models (GMM) and Hierarchical Clustering.

#### 4. Business Actions:

- Use trends in CPI and SMI to predict and adapt to future spending behavior.
- Tailor marketing strategies during low-SMI periods to focus on essential goods or promotions.

### Conclusion

This project successfully demonstrated the value of unsupervised learning techniques in analyzing economic indicators and consumer behavior. The combination of K-Means and DBSCAN clustering provided meaningful insights into quarterly trends and outlier periods. These findings offer actionable recommendations for businesses and policymakers to adapt to changing economic conditions.

## **Spending Analysis Project Report**

By incorporating additional data and modeling techniques in future analyses, this approach can be further refined to drive deeper insights and more robust strategies for stakeholders.