

MARKET SEGMENTATION

Using Machine Learning

10 Steps Study & McDonalds Dataset Analysis

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Summary

Chapter 1: Introduction to Market Segmentation Analysis

Chapter 1 serves as an introduction to the concept of market segmentation. It defines segmentation and its importance in marketing. The chapter emphasizes the need for data-driven decision-making and highlights how segmentation can lead to more effective marketing strategies. It sets the stage for understanding the entire segmentation process.

Chapter 2: Basic Concepts in Market Segmentation

Chapter 2 dives into the fundamental concepts of market segmentation. It explains the criteria for effective segmentation, such as homogeneity within segments and heterogeneity between segments. The chapter introduces the concept of segment attractiveness and highlights the role of segmentation variables. It emphasizes the importance of data in segmentation and lays the groundwork for the subsequent steps in the segmentation process.

Chapter 3: Segmenting Consumer Markets

This chapter focuses on segmenting consumer markets. It discusses the process of deciding whether or not to segment and the implications of committing to a segmentation strategy. The chapter identifies implementation barriers, including the role of senior management and organizational culture. It stresses the importance of making the decision to investigate segmentation potential at the highest executive level. This chapter lays the foundation for understanding the complexities and challenges of consumer market segmentation.

Chapter 4: Data Collection Techniques and Issues

This chapter emphasizes the importance of data collection in market segmentation. It discusses various techniques, including surveys, internal data sources, and experimental studies. Challenges like response bias and sample size are addressed, stressing the need for high-quality, unbiased data.

Chapter 5: Segmentation Algorithm Selection

Chapter 5 explores segmentation algorithms, including hierarchical, non-hierarchical, and model-based methods. It advises on algorithm selection based on data characteristics and objectives, highlighting the significance of understanding the strengths and limitations of each method.

Chapter 6: Profiling and Targeting

This chapter focuses on developing marketing strategies for segments. It covers the creation of profiles for each segment, emphasizing the importance of tailoring product offerings and marketing messages to meet the specific needs of target audiences.

Chapter 7: Segmentation Evaluation

Chapter 7 discusses methods for evaluating segmentation solutions. It introduces statistical measures like Homogeneity and Separation Indices and practical measures such as managerial relevance. It emphasizes that segmentation should improve decision-making and offer actionable insights.

Chapter 8: Market Segmentation in Action

Real-world examples and case studies are featured in this chapter, showcasing successful market segmentation strategies in various industries. It highlights how segmentation can lead to enhanced marketing effectiveness and increased profitability.

Chapter 9: Advanced Topics in Market Segmentation

Chapter 9 explores advanced topics in segmentation, including dynamic segmentation for evolving markets, digital-age segmentation techniques, and international segmentation. It addresses the challenges and opportunities of these advanced approaches.

Chapter 10: Conclusions and Future Directions

The final chapter summarizes key takeaways, stressing that market segmentation should be seen as an ongoing process. It discusses future trends in segmentation, such as the integration of big data and artificial

intelligence, and encourages a proactive approach to adapting to changing markets.

SOMETHING ABOUT PCA:

Why PCA?

In PCA (Principal Component Analysis), the primary goal is to maximize variance because it helps retain as much information as possible when reducing the dimensionality of a dataset. Maximizing variance serves several important purposes:

- 1. **Information Retention:** Variance represents the spread or the amount of information in the data. By maximizing variance, PCA ensures that the first principal component (PC1) captures the most significant source of variability in the data. This means that PC1 retains as much information as possible from the original dataset.*
- 2. **Dimension Reduction:** When you reduce the dimensionality of a dataset by selecting a subset of principal components, retaining high variance components ensures that you preserve the most critical patterns and relationships among the original features while eliminating noise and less important information.*
- 3. **Feature Selection:** PCA indirectly performs feature selection by giving higher importance to the features that contribute more to the variance. Features with low variance are less informative and tend to have less impact on the principal components.*

4. **Visualization:** High variance components often represent the dominant patterns in the data. By maximizing variance, you increase the chances of visualizing and understanding the most significant aspects of the data when you plot it in lower-dimensional space.
5. **Reducing Redundancy:** Maximizing variance helps PCA create orthogonal (uncorrelated) principal components. These components capture different aspects of the data, reducing redundancy. This orthogonality simplifies the interpretation of the components and makes them more useful for subsequent analyses.

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