# Market Segmentation Analysis

## **Step 1: Deciding (not) to Segment**

Based on the "Market Segmentation Analysis" PDF, here is the detailed evaluation for deciding whether to proceed with segmentation for McDonald's:

#### 1. Assess the Need for Segmentation

Market Diversity: The McDonald's customer base is diverse, comprising various age groups, income levels, and preferences. This diversity indicates the need for tailored marketing strategies to cater to different segments effectively.

Distinct Preferences: Different customer groups have distinct preferences, such as menu items, dining frequency, and spending habits. For example, families might prefer value meals, while young adults might opt for quick snacks.

#### 2. Evaluate Benefits and Costs

Improved Targeting: Segmentation allows McDonald's to create targeted marketing campaigns that resonate with specific customer groups, leading to higher engagement and conversion rates.

Increased Customer Satisfaction: By addressing the unique needs and preferences of each segment, McDonald's can enhance customer satisfaction and loyalty.

Higher Profitability: Tailored marketing efforts can lead to increased sales and higher profitability as customers are more likely to respond positively to personalized offers.

Data Collection and Analysis: Collecting and analysing data for segmentation requires investment in tools and resources, including data analysts and marketing experts.

Implementation Efforts: Developing and executing segment-specific marketing strategies involves additional costs in terms of time and resources.

Monitoring and Evaluation: Continuous monitoring and evaluation of the segmentation strategy are necessary to ensure its effectiveness, which adds to the overall cost.

#### 3. Identify Implementation Barriers

Possible Barriers: Lack of Management Support: If the management is not fully committed to the segmentation strategy, it may hinder the implementation process.

Insufficient Resources: Limited budget and resources can restrict the ability to collect data, analyse segments, and implement tailored marketing campaigns.

Organizational Resistance: Resistance to change within the organization can pose a significant barrier to implementing a new segmentation strategy.

#### 4. Make the Decision

Decision: Based on the assessment of market diversity, distinct customer preferences, potential benefits, costs, and implementation barriers, it is strategically beneficial for McDonald's to proceed with market segmentation. The potential benefits of improved targeting, increased customer satisfaction, and higher profitability outweigh the costs and barriers. By effectively addressing the implementation barriers, McDonald's can leverage segmentation to enhance its marketing efforts and achieve better business outcomes.

#### **Summary:**

- Market Diversity: High
- Distinct Preferences: Yes
- Potential Benefits: Improved targeting, increased customer satisfaction, higher profitability
- Costs: Data collection and analysis, implementation efforts, monitoring and evaluation
- Implementation Barriers: Lack of management support, insufficient resources, organizational resistance
- Decision: Proceed with market segmentation

# Step 2: Specifying the Ideal Target Segment:

To specify the ideal target segment for McDonald's, we need to define criteria for evaluating and selecting segments. Here are the criteria based on the instructions from the "Market Segmentation Analysis" PDF

#### 1. Define Segment Evaluation Criteria

- Size: Number of customers in the segment.
- Growth Potential: Expected growth rate.
- Profitability: Potential revenue and profit.
- Accessibility: Ease of reaching the segment.
- Compatibility: Alignment with McDonald's brand and capabilities.

#### 2. Establish Knock-Out Criteria

- Minimum Size: Must have a sufficient number of customers.
- Minimum Revenue Potential: Must have a sufficient revenue potential.

#### 3. Identify Attractiveness Criteria

- High Spending: Higher average spending per visit.
- Frequent Visits: High visit frequency.
- Strong Loyalty: Strong brand loyalty and low churn rates.
- Low Competition: Less competition from other brands.

#### 4. Implement a Structured Evaluation Process

- Scoring System: Rank segments based on evaluation criteria.
- Weighting: Assign importance to each criterion.
- Aggregate Scores: Calculate scores to identify promising segments

#### Summary on specifying the ideal Target Market:

- Define Segment Evaluation Criteria: Size, growth potential, profitability, accessibility, and compatibility.
- Establish Knock-Out Criteria: Minimum size and revenue potential.
- Identify Attractiveness Criteria: High spending, frequent visits, strong loyalty, and low competition.
- Implement a Structured Evaluation Process: Develop a scoring system, assign weights, and calculate aggregate scores for segments.

### **Step 3: Collecting Data:**

For this project, we need to gather data from relevant sources. The primary data source for the McDonald's case study is the provided dataset. This dataset includes various customer information that can be used for segmentation analysis.

#### Data-Set: McDonald's dataset

We need to select relevant variables from the dataset that will help in segmenting the market effectively. Key variables include:

- **♣** Demographic information like age and gender.
- ♣ Behavioural data such as purchase history and frequency of visits.
- Psychographic data including preferences and interests.

#### **Collect Data:**

We load the dataset and prepare it for analysis. Here's how to do it in Python:

```
In [1]: import pandas as pd
          # Load the McDonald's dataset
         url = 'https://homepage.boku.ac.at/leisch/MSA/datasets/mcdonalds.csv'
          mcdonalds = pd.read_csv(url)
          # Display basic information about the dataset
          print(mcdonalds.info())
          print(mcdonalds.describe())
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 1453 entries, 0 to 1452
         Data columns (total 15 columns):
                                Non-Null Count Dtype
          # Column
              yummy 1453 non-null object
convenient 1453 non-null object
spicy 1453 non-null object
          0 yummy
           1
              spicy
fattening
                                1453 non-null object
                                1453 non-null
1453 non-null
               greasy
                                                     object
               fast
          6 cheap 1453 non-null
7 tasty 1453 non-null
8 expensive 1453 non-null
9 healthy 1453 non-null
10 disgusting 1453 non-null
11 Like 1453 non-null
                                                     object
                                                      object
                                                      object
                                                      object
                                                      object
                                                      object
          11 Like 1453 non-null
12 Age 1453 non-null
                                                      object
                                                      int64
          13 VisitFrequency 1453 non-null object
           14 Gender
                                  1453 non-null object
          dtypes: int64(1), object(14)
          memory usage: 170.4+ KB
```

To ensure the data is accurate and usable, we need to clean it. This involves handling missing values, removing duplicates, and normalizing numerical data. Then, we can any missing values by filling them with 0. We remove any duplicate rows to ensure each entry is unique. For numerical columns like 'Age' and 'Income', we use standard scaling to normalize the data, making it easier to compare across segments.

```
In [2]: # Handle missing values by filling with 0
        mcdonalds.fillna(0, inplace=True)
        # Remove duplicate rows if any
        mcdonalds.drop_duplicates(inplace=True)
        # Normalizing numerical data, if applicable
        from sklearn.preprocessing import StandardScaler
        # Assuming 'Age' and 'Income' are numerical columns to be normalized
        scaler = StandardScaler()
        if 'Age' in mcdonalds.columns and 'Income' in mcdonalds.columns:
            mcdonalds[['Age', 'Income']] = scaler.fit_transform(mcdonalds[['Age', 'Income']])
        # Save the cleaned data to a new file
        mcdonalds.to_csv('cleaned_mcdonalds_data.csv', index=False)
        # Display the first few rows of the cleaned data
        print(mcdonalds.head())
          yummy convenient spicy fattening greasy fast cheap tasty expensive healthy \
                                                             No
                 Yes No Yes No Yes Yes
Yes No Yes Yes Yes Yes
                                                                                 No
            No
                                                                        Yes
                                                              Yes
                                              Yes Yes
        1
            Yes
                                                                        Yes
                                                                                 No
                                    Yes Yes Yes
                     Yes Yes
                                                        No Yes
                                                                       Yes
                                     Yes Yes Yes Yes Yes
                                                                                Yes
                     Yes No
Yes No
                                                                       No
        3
           Yes
                                                                                 No
                                             Yes Yes Yes
        4
                                                                        No
                                                                                Yes
          disgusting Like Age VisitFrequency Gender
        0 No -3 61 Every three months Female
1 No +2 51 Every three months Female
2 No +1 62 Every three months Female
```

#### Summary of Step 3

- **♣** We load the McDonald's dataset and examine its structure.
- ♣ We clean the dataset by handling missing values, removing duplicates, and normalizing numerical data.
- ♣ The cleaned data is then saved for further analysis.

This process ensures that the data is prepared and ready for segmentation analysis, allowing us to move on to the next steps of identifying and profiling customer segments.

#### GitHub Link:

https://github.com/priya99karn/Summary-Report---Market-Segmentation-Analysis.git