

## Coffee Shop Sales: Model Results and Analysis

This document outlines the results of two predictive models built to analyze the coffee shop sales data:

1. **A Decision Tree Classifier** to predict sales by time of day.
2. **A KMeans Clustering Model** to segment customers for targeted marketing.

### Model 1: Sales Prediction (Decision Tree)

#### 1. Classification Report

- **Accuracy: 27%** The model's overall accuracy is approximately 27%. While this figure appears low, it is important to consider the following factors:
  - **Task Difficulty:** Predicting one of eight distinct coffee types is an inherently complex classification task.
  - **Insufficient Features:** The model only used time-based features (hour, day of week, month). Customer preference is highly personal and not solely dictated by the time of day.
  - **Data Imbalance:** As shown by the f1-scores, the model is slightly better at predicting common items like 'Americano' (0.33) and 'Americano with Milk' (0.34), but performs very poorly (0.00) on rare items like 'Espresso'.

#### 2. Example Prediction

- **Prediction:** Example Prediction for 10 AM on a Friday in March: Latte
- **Interpretation:** The model predicts that a customer at 10 AM on a Friday in March is most likely to buy a **Latte**. This aligns with our previous heatmap analysis, which identified 'Latte' and 'Americano with Milk' as popular choices during this peak morning hour.

#### 3. How to Use This Model

This model should be viewed as an **automated supplement to the sales heatmaps**. It can be used to provide a quick "best guess" for what to prepare for an upcoming hour. Its low accuracy means it should **not be exclusively relied upon**, but it can serve as a useful "guide" for production planning and inventory management.

### Model 2: Customer Clustering (KMeans)

To segment customers, we used an RFM (Recency, Frequency, Monetary) model based on the following features:

- **Recency:** days\_since\_last\_visit
- **Frequency:** total\_visits
- **Monetary:** total\_spent

#### 1. The "Elbow Plot" (kmeans\_elbow\_plot.png)

This plot shows the model's "Inertia" (a measure of cluster density). We look for the "elbow," the point where adding more clusters yields diminishing returns. Based on the plot, a **k-value of 4** was selected to provide a granular and meaningful segmentation of the customer base.

## 2. Cluster Analysis (Group Means)

This table shows the most critical output: the average characteristics of our four customer segments. These values represent the "profile" of each group.

| Cluster | days_since_last_visit | total_visits | total_spent |
|---------|-----------------------|--------------|-------------|
| 0       | 272.09                | 1.58         | 50.79       |
| 1       | 99.88                 | 1.88         | 60.24       |
| 2       | 71.50                 | 74.75        | 2259.01     |
| 3       | 82.66                 | 15.77        | 497.53      |

## 3. Interpretation of Customer Segments

### Cluster 0: "Churned / One-Time Customers"

- **Profile:** This group has not visited in an average of **272 days** (high recency), has a very low visit frequency (avg. 1.58 visits), and the lowest monetary value.
- **Marketing Strategy:** This group is almost entirely lost. A deep-discount "we miss you" campaign (e.g., 50% off) could be attempted, but marketing budget should be limited here.

### Cluster 1: "Recent New Customers / At-Risk"

- **Profile:** This group is composed of new or low-frequency customers who visited more recently than Cluster 0 (avg. 100 days ago) but have not formed a habit (avg. 1.88 visits).
- **Marketing Strategy:** This is the "**At-Risk**" group. The goal is to convert them into regulars. Target them with a "20% off your second visit" or "bring a friend" coupon to encourage a return visit and move them into Cluster 3.

### Cluster 2: "VIPs / Super-Loyal"

- **Profile:** This is a small but elite group, defined by extremely high frequency (avg. **74.75 visits**) and extremely high spending (avg. **\$2259.01**). They visit relatively often (avg. 71.5 days).
- **Marketing Strategy:** These are your super-fans. **Do not give them discounts;** they don't need them. Give them **status:** free size upgrades, first access to new products, or holiday cards. Their loyalty is priceless and should be rewarded with recognition, not discounts.

### Cluster 3: "Loyal Regulars"

- **Profile:** This is your core, high-value customer base. They are frequent (avg. **15.77 visits**), spend significantly (avg. **\$497.53**), and visit regularly.

- **Marketing Strategy:** This is the "morning coffee" crowd. A "Buy 5, Get 1 Free" loyalty card or a