

# Sprocket Central Pty Ltd

Data Analytics Approach

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## Agenda

- 1. Introduction
- 2. Data Exploration
- 3. Model Development
- 4. Interpretation

## Introduction

## **Data Quality Assessment**

	Customer Demographic	Customer Addresses	Transactions
Accuracy	DOB: Inaccuracy Default: Inaccuracy		
Completeness	Customer IDs: Not in Sync Last Name: Null DOB: Null Job Title: Null Job Industry Category: Null Tenure: Null	Customer IDs: Not in Sync	Customer IDs: Not in Sync Online Order: Null Brand: Null Product Line: Null Product Class: Null Product Size: Null Standard Cost: Null Product First Sold Date: Null
Consistency	Gender: Inconsistency Default: Inconsistency	State: Inconsistency	
Currency	Deceased Indicator: Filter Out		
Relevancy	Default: Exclude Field	Country: Filter Out	Order Status: Exclude Cancelled
Validity	Default: Format		Product First Sold Date: Format

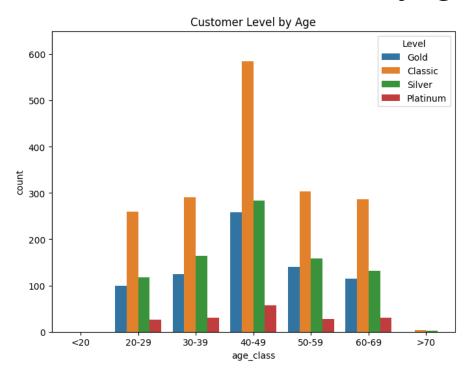
#### Introduction

## **Data Cleaning**

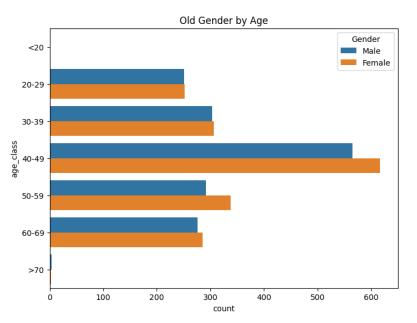
- Data containing missing values were removed from the records.
- Conflicting records were also dropped by considering the join keys between tables.
- Age, Last Purchases (Days Ago) and Profit features were added.
- Records related to the deceased indicator were excluded.

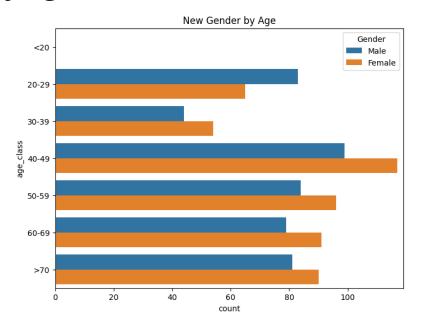
Distinct Customer IDs (Before)	4,000
Distinct Customer IDs (After)	3,491

## Customer Level Old Customer by Age

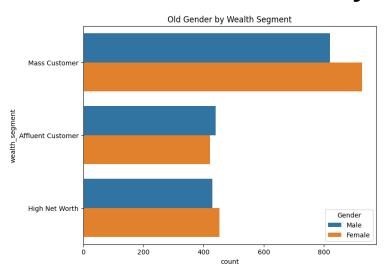


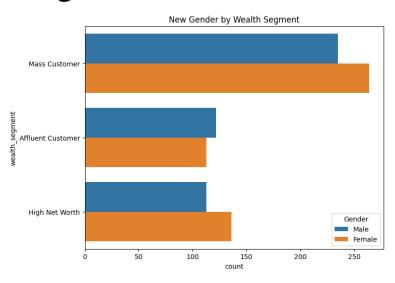
## Gender by Age



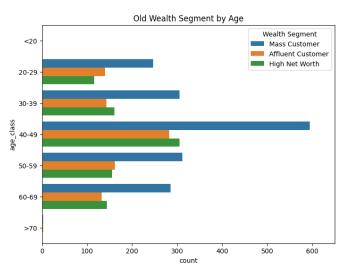


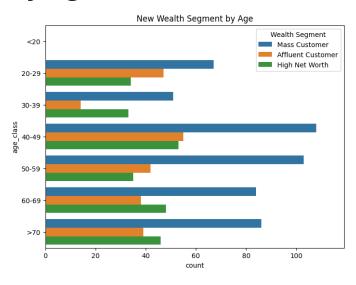
## Gender by Wealth Segment



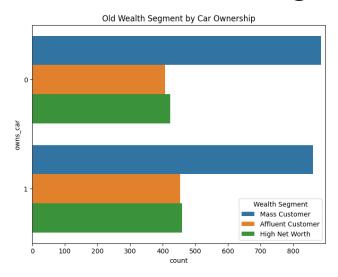


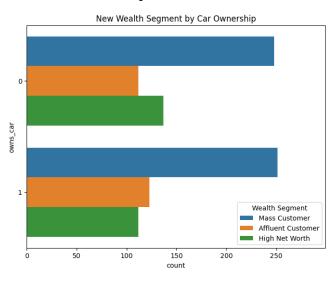
## Wealth Segment by Age



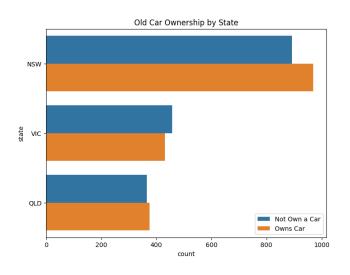


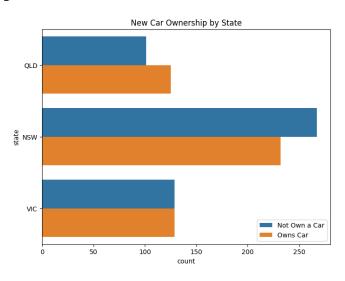
## Wealth Segment by Car Ownership



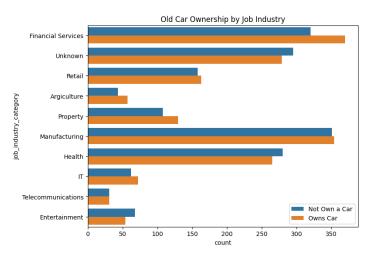


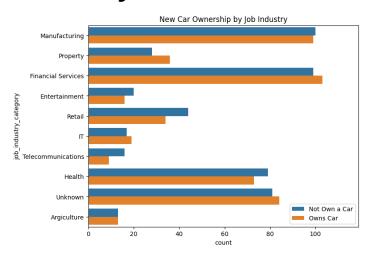
## Car Ownership by State



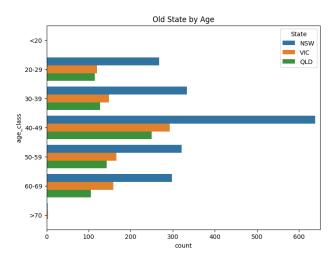


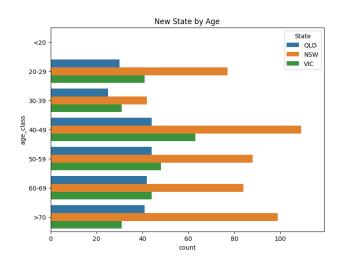
## Car Ownership by Job Industry





## State by Age





## **RFM Analysis**

#### Recency

➤ Based on the most recent transaction date of a customer, where they were divided into 4 quartiles and given a *R\_Score*.

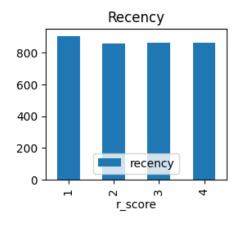
#### Frequency

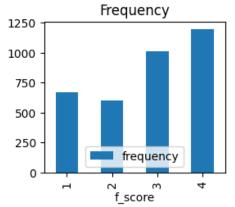
➤ Based on the number of transactions made by a customer, where they were divided into 4 quartiles and given a F\_Score.

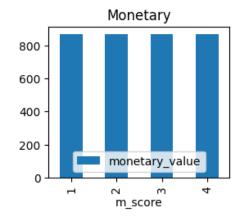
#### Monetary Value

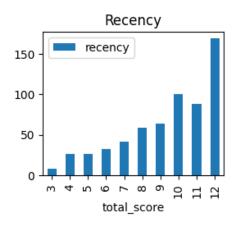
▶ Based on the average profit generated by each customer, where they were divided into 4 quartiles and given a *M\_Score*.

## **RFM Analysis**

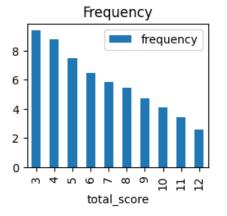


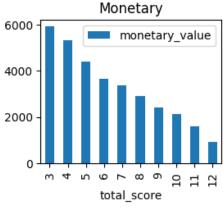




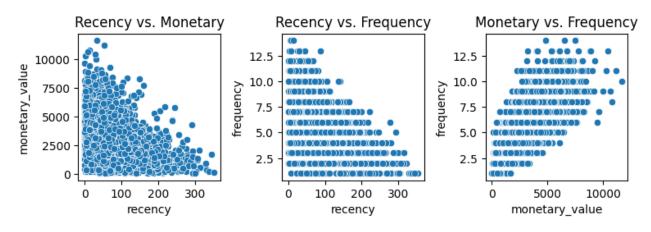


## **RFM Analysis**





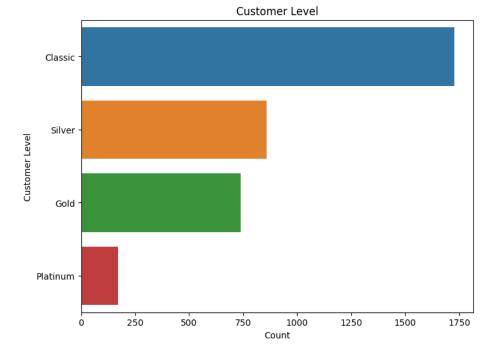
## **RFM Analysis**



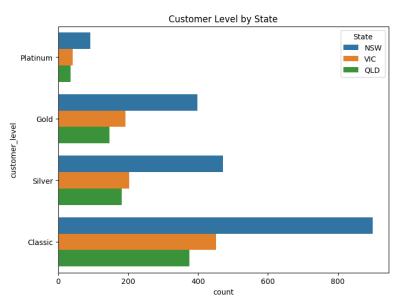
## **RFM Analysis**

4 customer tiers were identified based on the RFM Segmentation:

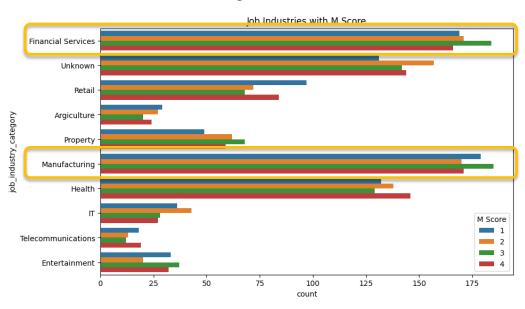
- 1. Platinum Class: Who have made recent purchases, are frequent, and contribute the most profits.
- 2. Gold Class
- 3. Silver Class
- 4. Classic Class: Who have not made recent purchases, are not frequent, and do not significantly contribute to profits.



## **Customer Tiers by State**



## **Profitable Job Industries**



## **Customer Segments**

Segment	RFM Score
Champions	3
Loyalists	4
Potential Loyalists	5
New Customers	6
Promising	7
Need Attention	8
About to Sleep	9
High Risk	10
Hibernating	11
Lost Customers	12

## **Targeting Methodology**

- Customers with high RFM scores can be selected and focused on.
- These customers have made purchases recently, frequently, and generate significant profits.

## Thank You