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TheAnalyticsTeam

Sprocket Central Pty Ltd

Data analytics approach

Created by Tram Nguyen

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Agenda

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

Introduction

Module task 1

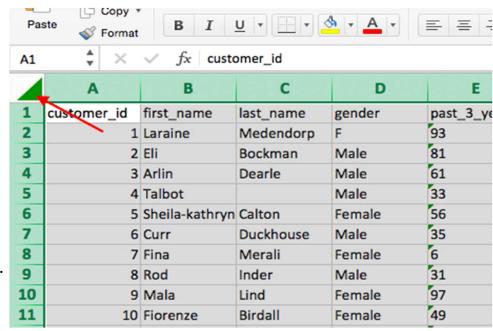
Module task 2

Module Task 3

Task1: identify all data quality issues.

- Data incompleteness
- Data inconsistency
- Data Inaccuracy
- Data not coming from the same source

Recommendations are attached in email to clients.

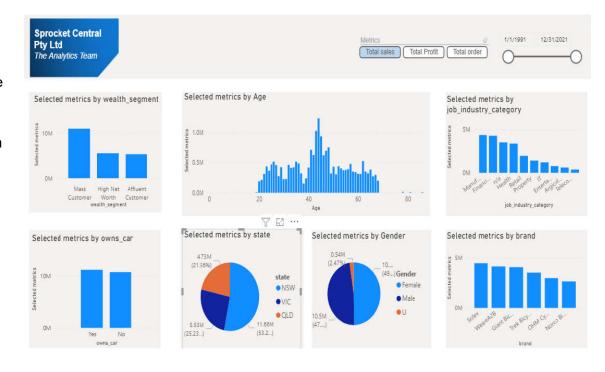


Data Exploration

Module task 2 : Exploration into customer demographics, geographics, psychology, and behaviour. <u>Links</u>

Demographics metrics:

- Age: 20-70 focus group: 40-50 (highest frequency)
- Wealth segment: Mass customer occupies the largest proportion of sales and profit of 10 million and 5.5 million.
- Gender: almost the same distribution between male and female customers.
- Job industry: Customer working in manufacturing, financial service and health sector generate the highest sales and profit (75%),
- Car ownership: 50% of customer owning car. **Geographics** metrics:
- Most of sales derived from New South Wales (53%), Victoria(25%), Queensland(21%)



Data Exploration

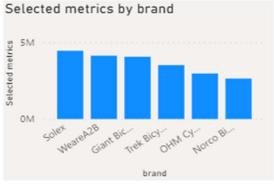
Module task 2: Exploration into customer demographics, geographics, psychology, and behaviour.

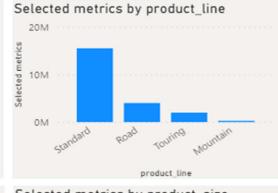
Psychology metrics:

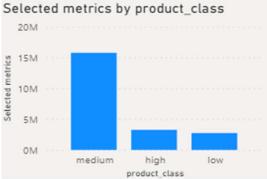
- Customers prefer buying products belonging to normal brands Solex, WeareA2B, Giant Bicycle.
 Other special purposed brands are demanded by a smaller group of customers.
- Standard product line, medium product class and medium product size are the main product attributes when it comes to sales and profit.

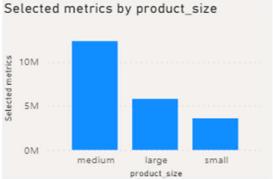
Behaviour metrics:

 There is a balance distribution between online and offline channels.









Model Development

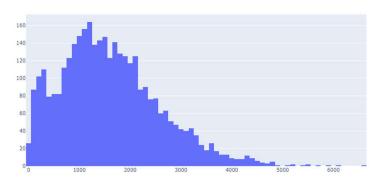
Module Task 3: Business model using Recency, Frequency, Monetary Value (RFM) Links

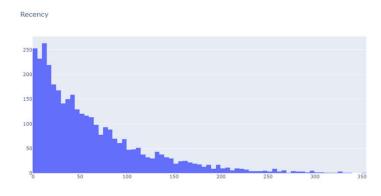
To calculate **recency**, we need to find out most recent purchase date of each customer and see how many days they are inactive for.

To create **frequency** clusters, we need to find total number orders for each customer.

Finally, we cluster them based on revenue (**monetary**) for each customer.





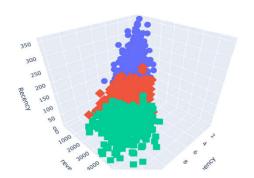


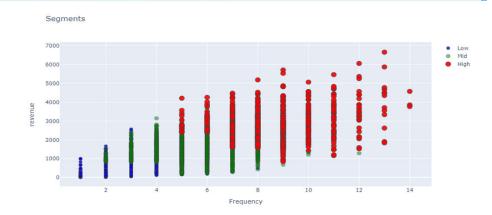


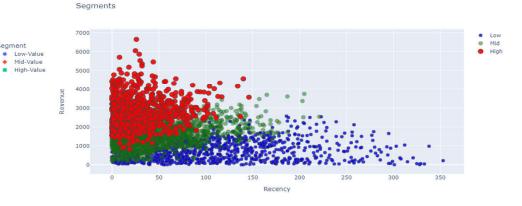
Model Development

Module Task 3: Overall Segmentation **Links**

- **Low Value**: Customers who are less active than others, not very frequent buyer/visitor and generates very low zero maybe negative revenue.
- •Mid Value: In the middle of everything. Often using our platform (but not as much as our High Values), fairly frequent and generates moderate revenue.
- •**High Value**: The group we don't want to lose. High Revenue, Frequency and low Inactivity.







Interpretation

Module Task 3: Overall Segmentation **Links**

Cluster 1: low value customers having overall score from 0-2.

- Made last purchase from 100 days on earlier (more than 50 days inactive).
- Total revenue per user less than \$800.
- Total purchase of less than 4.
- Recommendation: Increase frequency.

Cluster 2: mid value customers having overall score from 3-6.

- Made last purchase within 40 and 70 days ago.
- Total revenue per user between \$800 and \$2000.
- Total purchase between 4 and 6.5.
- Recommendation: Improve retention + Increase Recency.

Cluster 3: high value customers having overall score from 7-10.

- Made last purchase less than 40 days ago.
- Total revenue per user more than \$2200.
- Total purchase more than 7.
- Recommendation: Improve retention.

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