

Introduction

Targeting high value customers based on customer demographics and attributes.

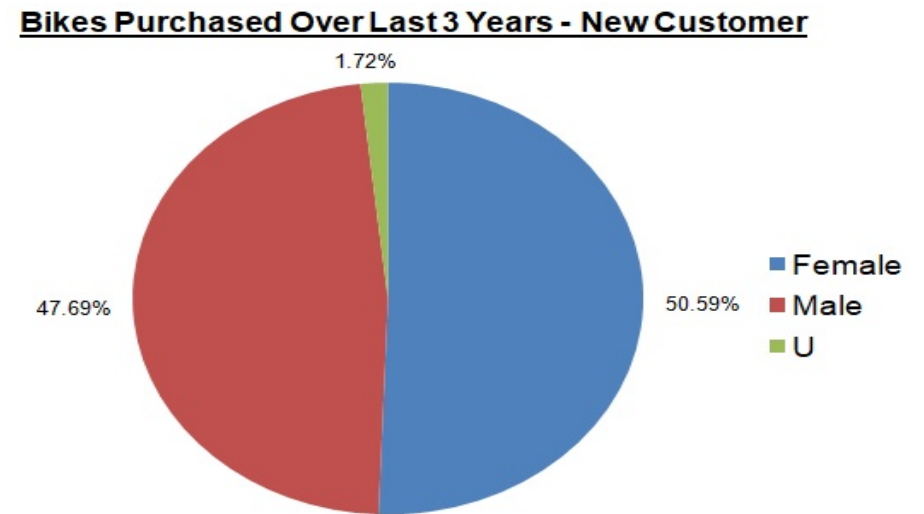
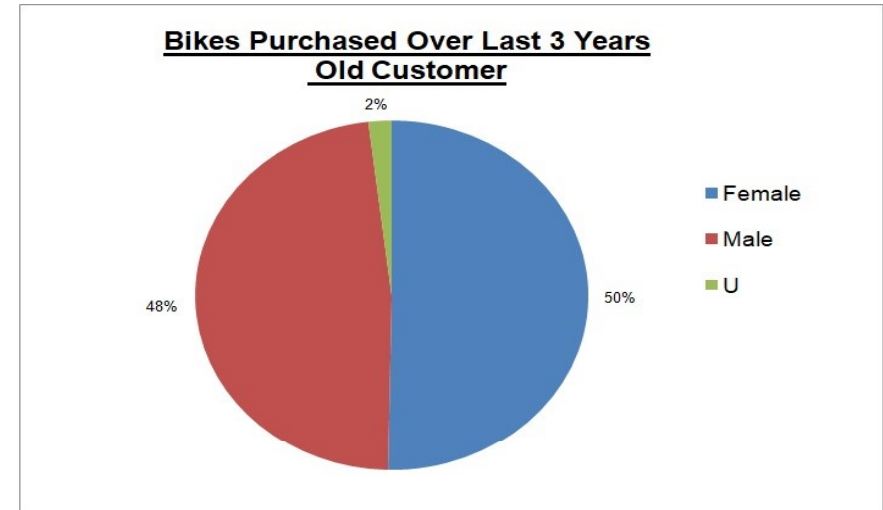
- Sprocket Central Pty Ltd is a company that manufactures high-quality bikes and accessible cycling accessories to riders.
- Their marketing team is looking to boost business by analyzing their existing customer dataset to determine customer trends and behavior.
- Using the existing 3 datasets (Customer demographic, customer address and transactions), analyze and recommend which of these 1000 new customers should be targeted to drive the most value for the organization.



Data Exploration

BIKE RELATED PURCHASES OVER LAST 3 YEARS BASED ON GENDER

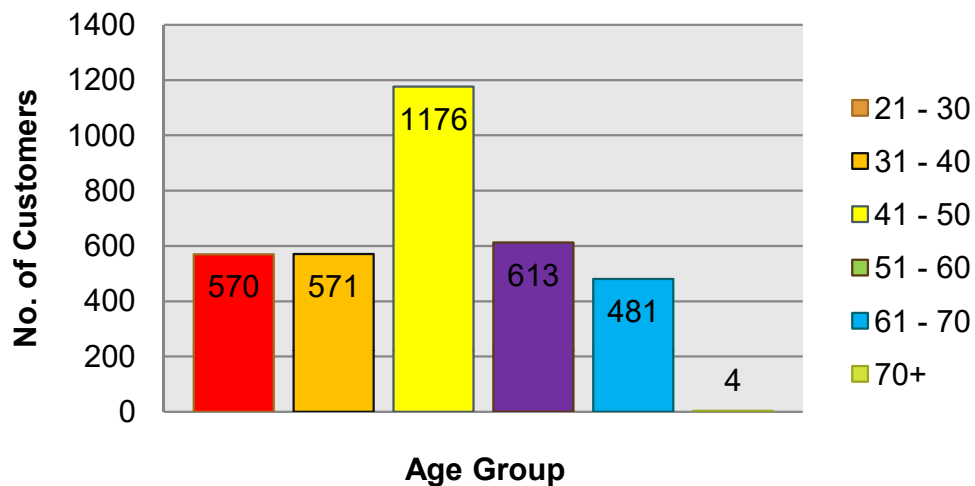
- Over the past 3 years, the percentage of bike-related purchases made by females has been approximately 2% higher than the percentage made by males.
- Additionally, purchases made by individuals of unknown gender accounted for 2% of the total.



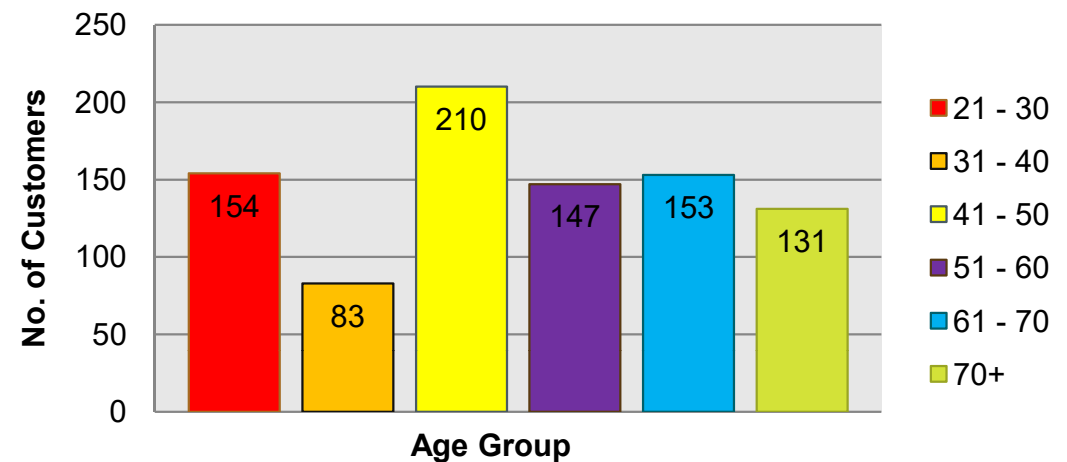
Data Exploration

CUSTOMER AGE DISTRIBUTION

Old Customer Age Distribution



New Customer Age Distribution

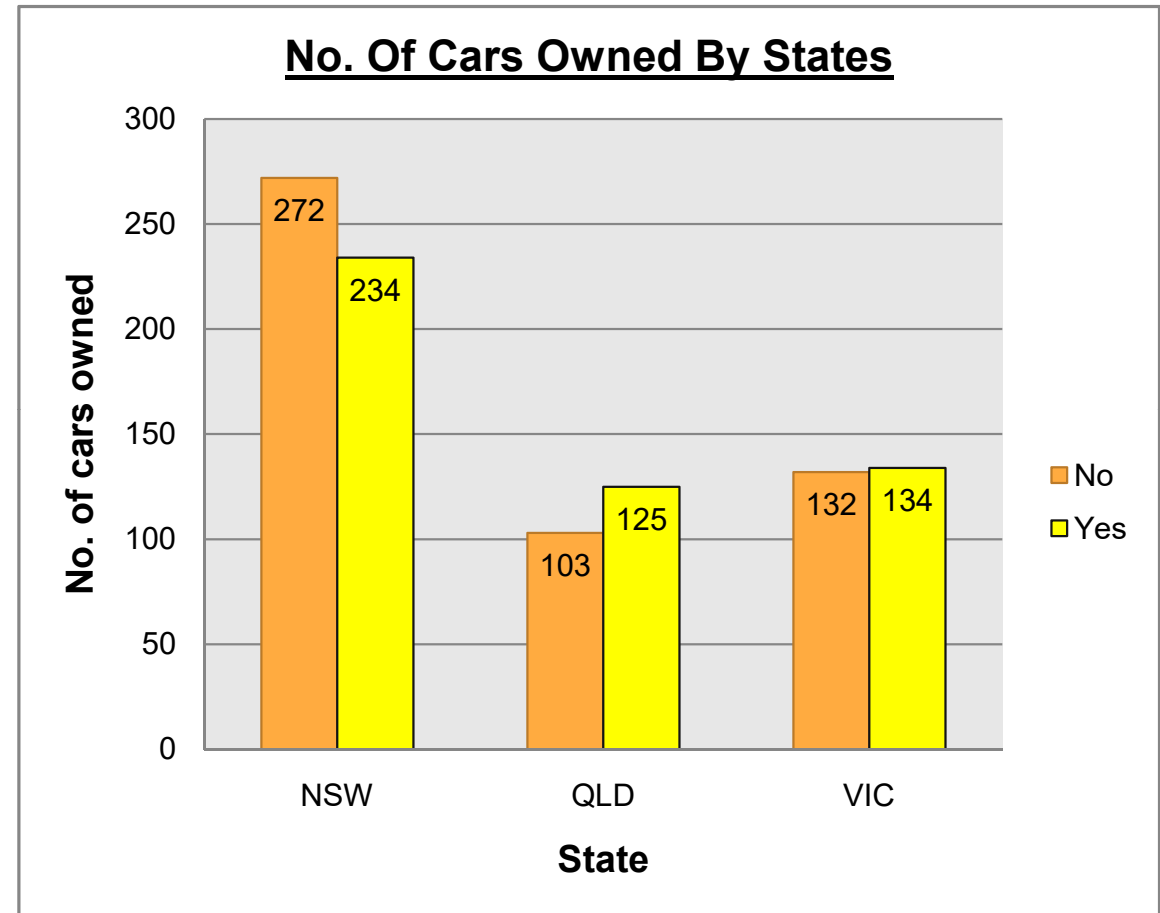


- Most customers are in 41-50 age groups in both old and new customer age.
- Lowest age groups are 70+ in old customer age and 31-40 in new customer age.
- There is a sudden drop of customers in 31-40 age group in new customer age and steep increase in 70+ age group.

Data Exploration

NO. OF CARS OWNED BY STATES

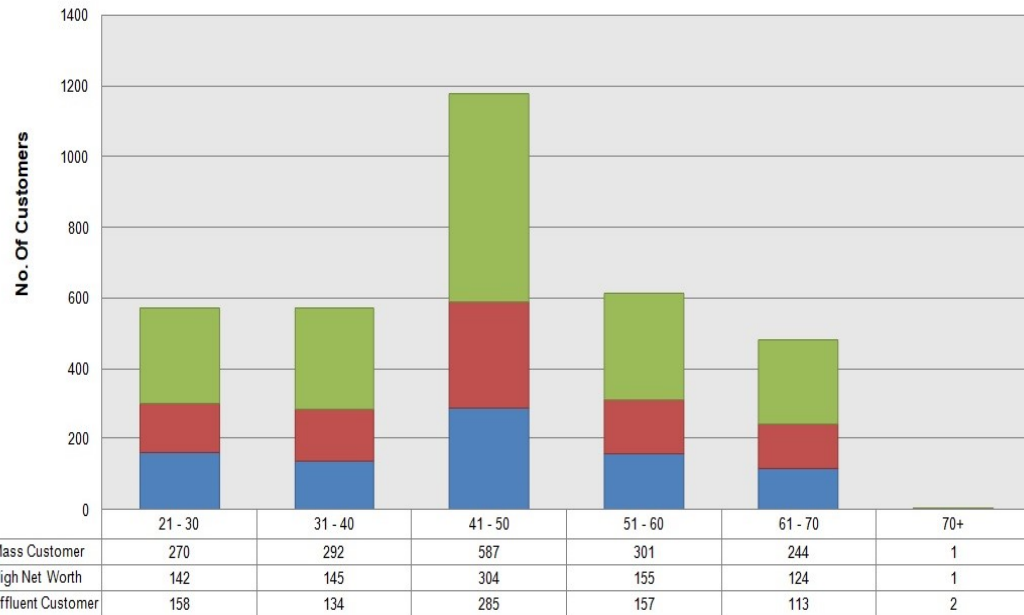
- NSW has the highest number of people who own a car, while QLD has the lowest.
- Both states have a relatively similar number of people who do not own a car,
- Indicating potential for reaching out to this customer segment.
- On the other hand, in VIC, split quite evenly among the population.



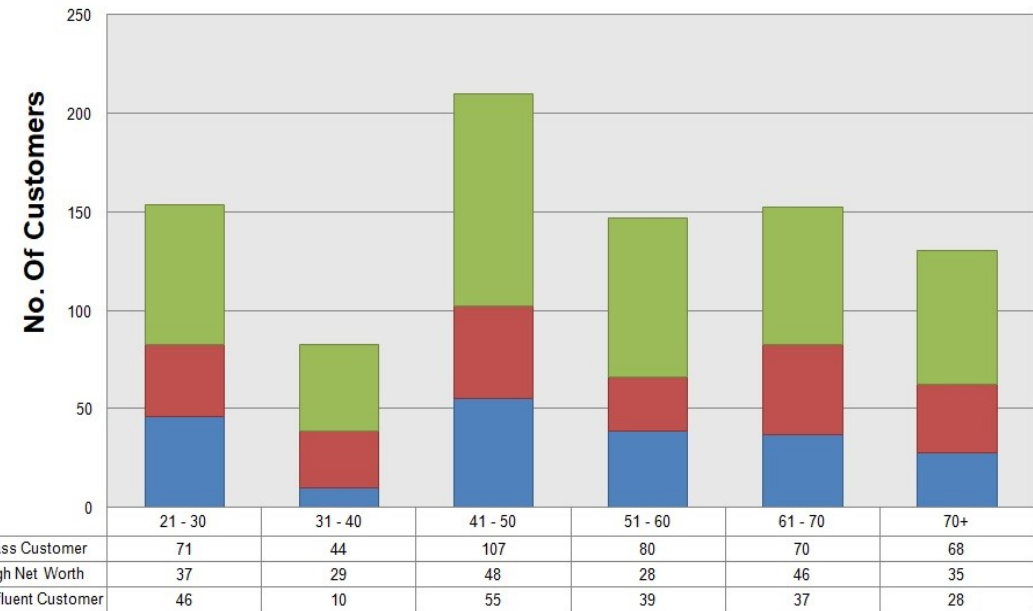
Data Exploration

WEALTH SEGMENT DISTRIBUTION

WEALTH SEGMENTATION - OLD CUSTOMER



WEALTH SEGMENTATION - NEW CUSTOMER

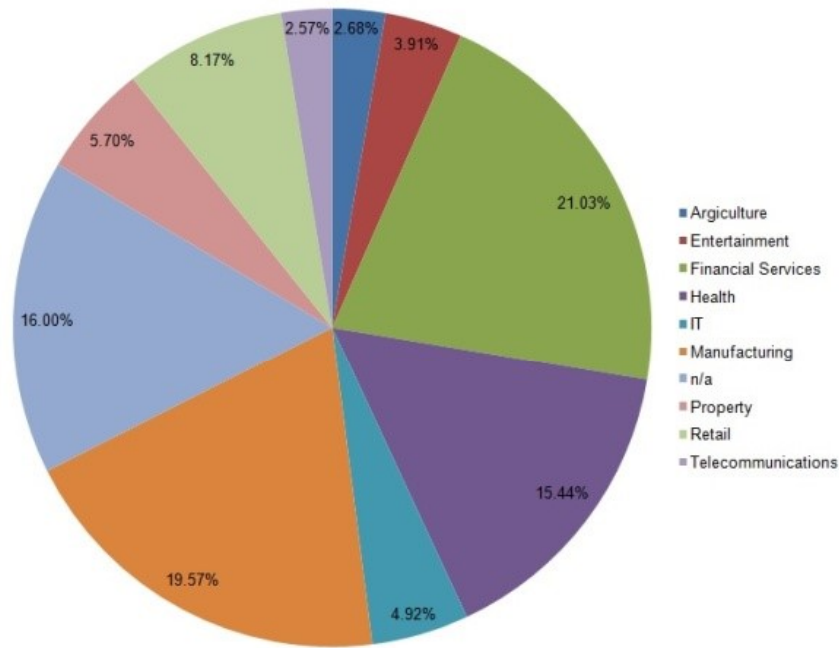


- According to the data, the largest number of customers across all age groups were from the 'mass customer' segment, followed by the 'high net worth'.
- However, in some age groups, the 'affluent customer' segment outperformed the 'high net worth' segment.
- The data also shows that in the old customer segmentation approach, there was a significant drop in the buying trend after the 41-50 age group,
- Whereas in the new customer segmentation approach, this steep decrease is not as prominent.

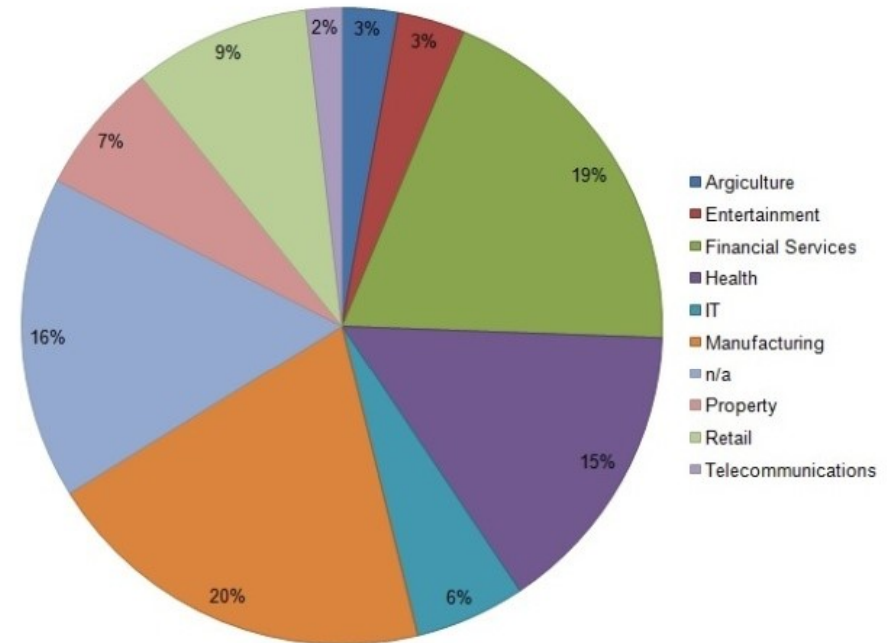
Data Exploration

JOB INDUSTRY DISTRIBUTION

Job Industry Distribution - New Customer



Job Industry Distribution - Old Customer

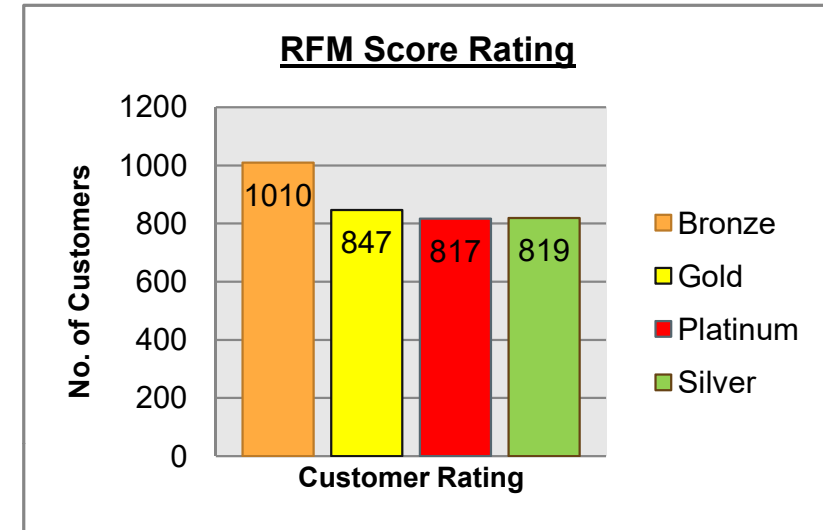


- Based on the data, the top three industry sectors that have contributed to the major profits are financial services, manufacturing, and health, While agriculture, telecommunications, and entertainment are the least contributing sectors.
- The data also shows a difference in the job distribution of old and new customers. In the old customer job distribution, the manufacturing sector was the top industry sector, Whereas in the new customer job distribution, financial services is the leading industry sector.

Model Development

RFM ANALYSIS

- **RFM** analysis is a marketing technique that helps businesses identify and segment their customers based on their purchasing behavior. RFM stands for Recency, Frequency, and Monetary Value, which are the three key metrics used in this analysis.
- **Recency** refers to how recently a customer has made a purchase. Customers who have made a purchase more recently are generally more engaged and likely to make another purchase in the near future.
- **Frequency** refers to how often a customer makes purchases. Customers who make purchases more frequently are more loyal and valuable to a business.
- **Monetary** Value refers to the total amount of money a customer has spent with a business. Customers who have spent more money are generally more valuable and may warrant special attention or rewards.

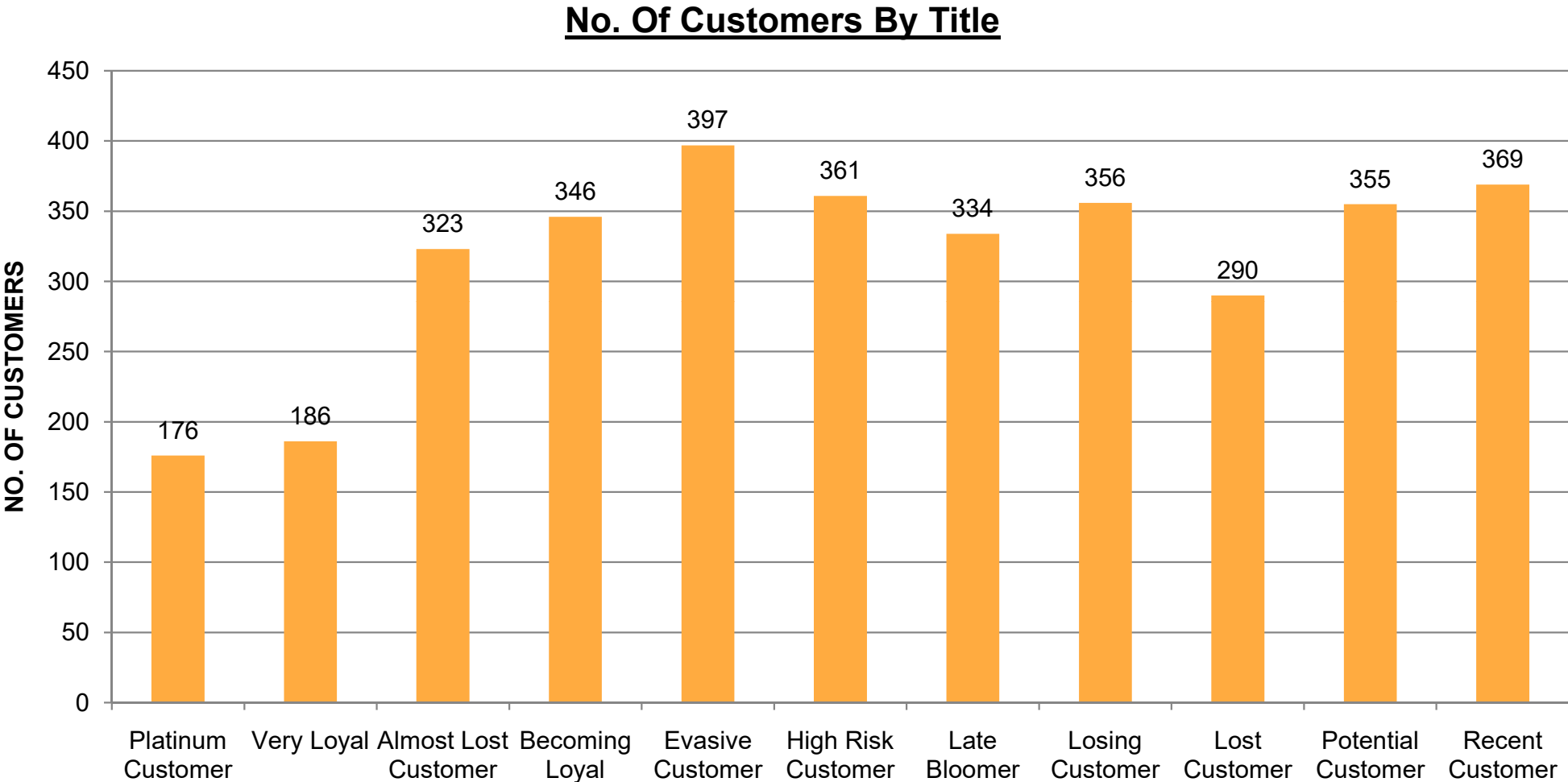


Model Development

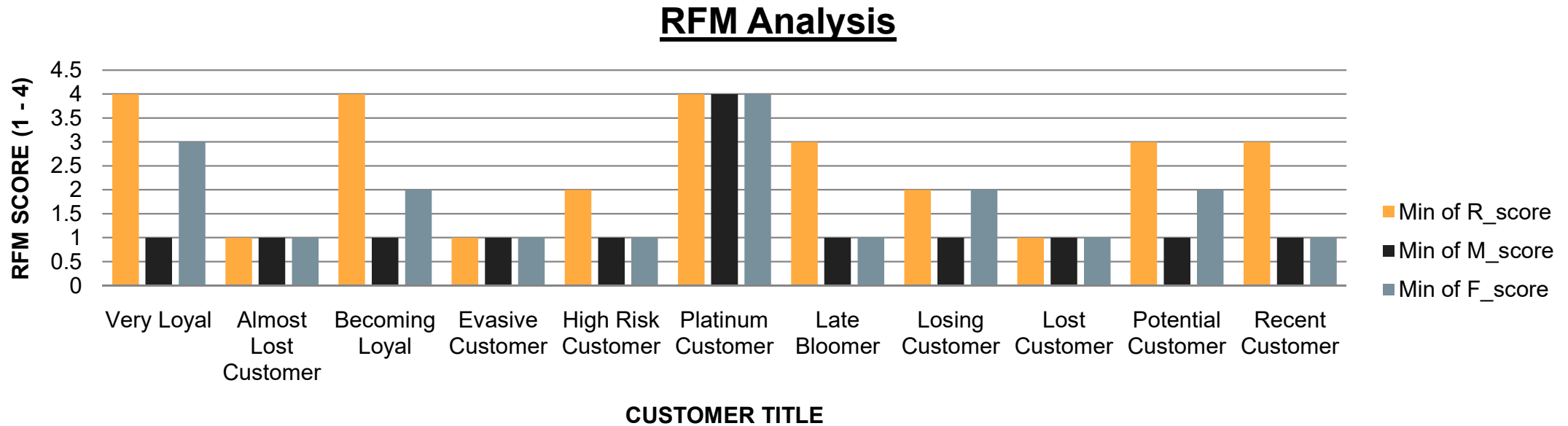
Rank	Customer Title	Description	RFM Score
1	Platinum Customer	Most recent buy, buys often, most spent	444
2	Very Loyal	Most recent, buys often, spends large amount of money	433
3	Becoming Loyal	Relatively recent, bought more than once, spends large amount of money	421
4	Recent Customer	Bought recently, not very often, average money spent	344
5	Potential Customer	Bought recently, never bought before, spent small amount	323
6	Late bloomer	No purchases recently, but RFM value is larger than average	311
7	Losing Customer	Purchases was a while ago, below average RFM value	224
8	High Risk Customer	Purchase was long time ago, frequency is quite high, amount spent is high	212
9	Almost Lost Customer	Very low recency, low frequency, but high amount spent	124
10	Evasive Customer	Very low recency, Very low frequency, small amount spent	112
11	Lost Customer	Very Low RFM	111

- By combining all three metrics, business can segment their customers into different groups based on their overall value to the business.
- For example, a customer who has made a large purchase recently and frequently would be considered a high-value customer and may warrant special marketing efforts to retain their loyalty.
- In contrast, a customer who has not made a purchase in a long time and has spent very little money would be considered a low-value customer and may not warrant as much attention.

Model Development



Model Development



- The RFM model score provides a comprehensive assessment of a customer's level of engagement with a business across all three categories of Recency, Frequency, and Monetary Value, provide a holistic view of customers overall value to the business.
- RFM analysis can be used to inform marketing and customer engagement strategies, such as targeted promotions and personalized communications.

Interpretation

LIST OF 1000 POTENTIAL CUSTOMERS WITH THEIR DEMOGRAPHICS AND ATTRIBUTES

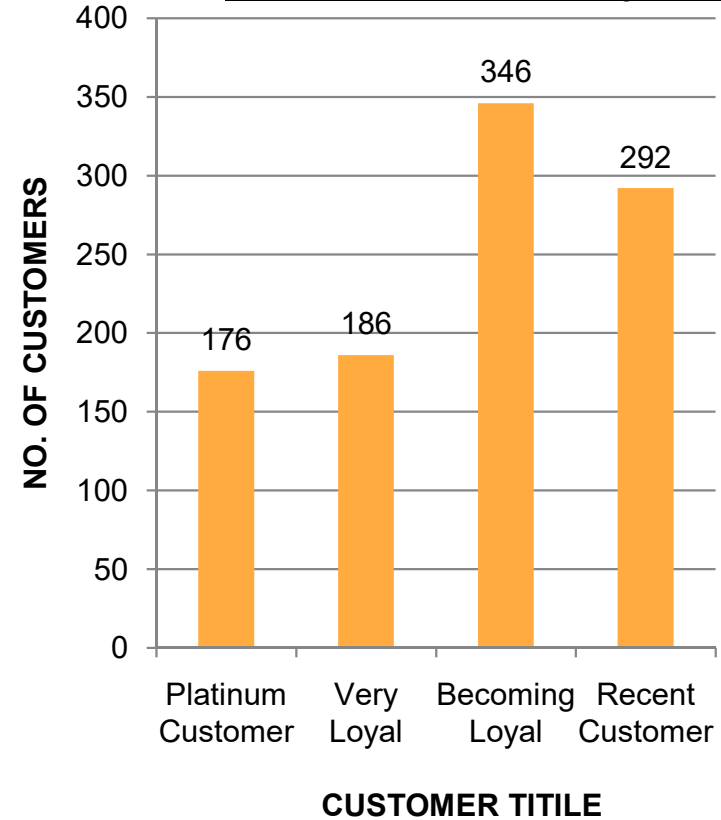
Sample list of top 1000 potential customers:

Customer_id	RFM Value	First name	Last Name	gender	State	Postcode	Customer Title
25	444	Geoff	Assaf	Male	QLD	4413	Platinum Customer
37	444	Laurie	Dwerryhouse	Female	QLD	4726	Platinum Customer
89	444	Benedicto	Hoxey	Male	NSW	2100	Platinum Customer
91	444	Dylan	Meaker	Male	NSW	2166	Platinum Customer
99	444	Reggie	Mulliner	Male	NSW	2160	Platinum Customer
109	444	Cody	Blabey	Male	NSW	2217	Platinum Customer
117	444	Nance	Suttling	Female	QLD	4226	Platinum Customer
144	444	Jory	Barrabeale	U	NSW	2315	Platinum Customer
151	444	Donnie	Brimson	Male	NSW	2759	Platinum Customer
165	444	Aldon	Roelofs	Male	NSW	2032	Platinum Customer
167	444	Nathalie	Tideswell	Female	QLD	4352	Platinum Customer
171	444	Halli	Davidoff	Female	NSW	2099	Platinum Customer
173	444	Ebba	Hanselmann	Female	NSW	2766	Platinum Customer
174	444	Catie	Tosspell	Female	NSW	2148	Platinum Customer
200	444	Hube	Weald	Male	QLD	4005	Platinum Customer
234	444	Mead	Alpes	Female	NSW	2036	Platinum Customer
235	444	Leona	Phateplace	Female	NSW	2018	Platinum Customer
239	444	Wells	Pressman	Male	QLD	4514	Platinum Customer
298	444	Maurise	Lenglet	Male	VIC	3029	Platinum Customer
320	444	Aldous	Cubin	Male	QLD	4701	Platinum Customer
322	444	De	Denington	Female	QLD	4350	Platinum Customer
344	444	Carlene	Eyckelbeck	Female	NSW	2147	Platinum Customer
346	444	Timmie	Eisikowitz	Male	NSW	2155	Platinum Customer
357	444	Betta	Shrimpton	Female	VIC	3199	Platinum Customer
377	444	Levy	Bonavia	Male	NSW	2265	Platinum Customer
381	444	Leesa	Castleman	Female	NSW	2062	Platinum Customer
404	444	Sinclair	Sustins	Male	NSW	2536	Platinum Customer
515	444	Winna	Verny	Female	NSW	2026	Platinum Customer
534	444	Madel	Palfrey	Female	NSW	2218	Platinum Customer

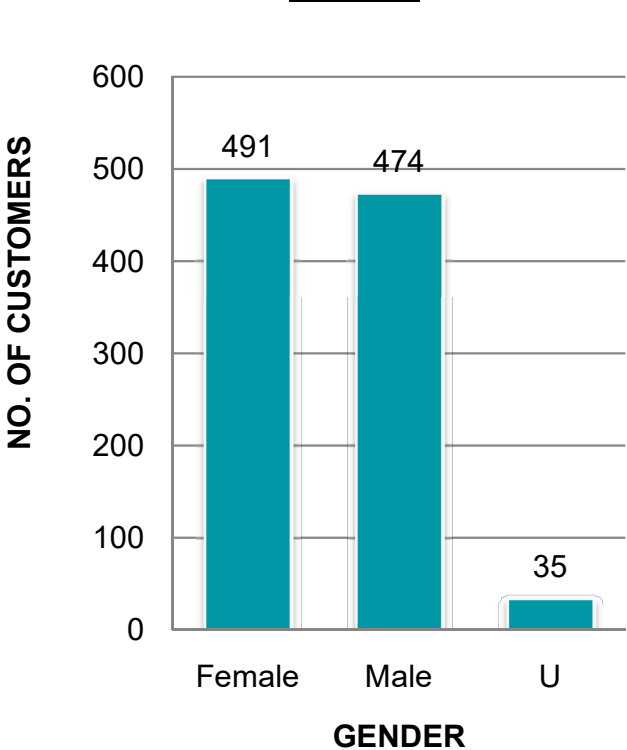
Interpretation

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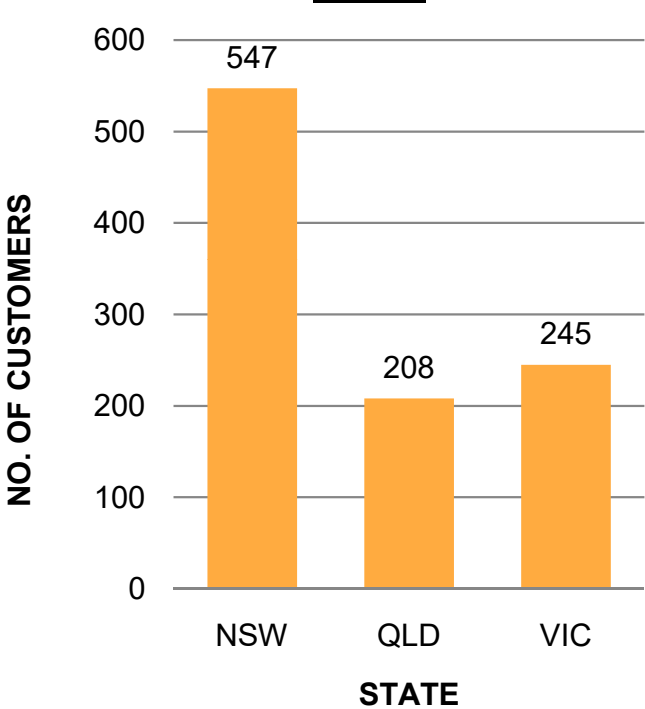
Customer Distribution By Title

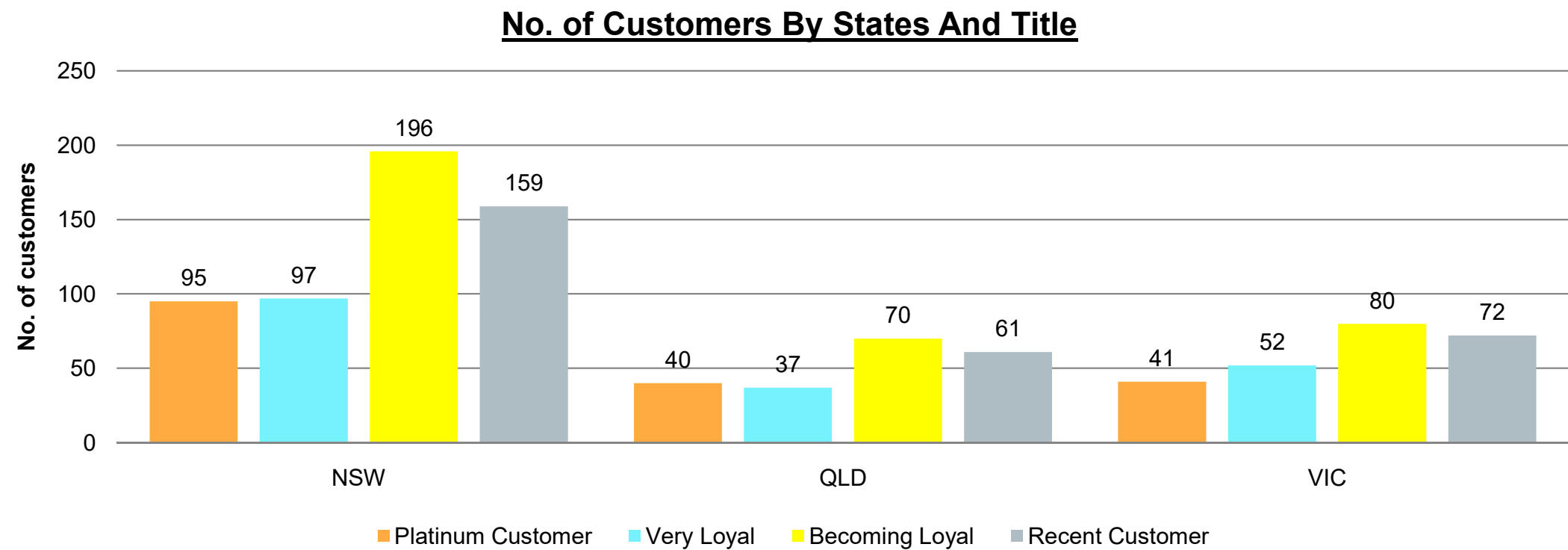


Customer Distribution By Gender



Customer Distribution By States





• Identified top 1000 high-value customers through RFM analysis can help businesses focus their resources on retaining their most valuable customers and drive long-term growth.