CIB - Data Driven CC Marketing

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AGENDA

2)Business Insights

4) Targetted vs Mass Marketing 6) Met Business Goals 8) Technical Journey

1) Problem Statement

3) Customer KPIs

5)Targetted Marketing

7) Why should you hire us?

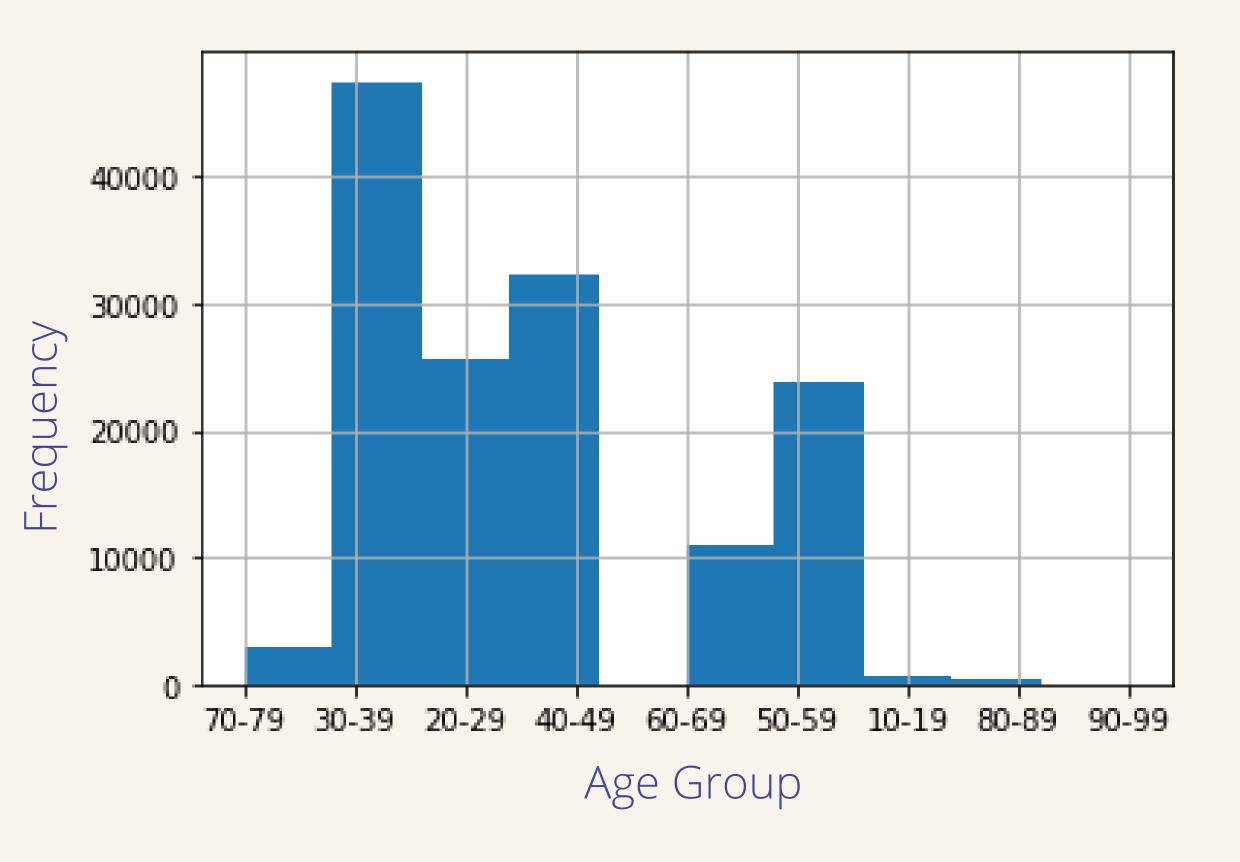
9) References

PROBLEM STATEMENT

- Fintech and credit card industry is growing.
- Competition is high in Credit Card (CC) industry
- Industry studies suggest 57% of its customer base is lost in case of not exceeding customers expectations due to better competitors service. (AcuityKP, n.d.)
- CIB wants to satisfy its customer base and even attract new customers

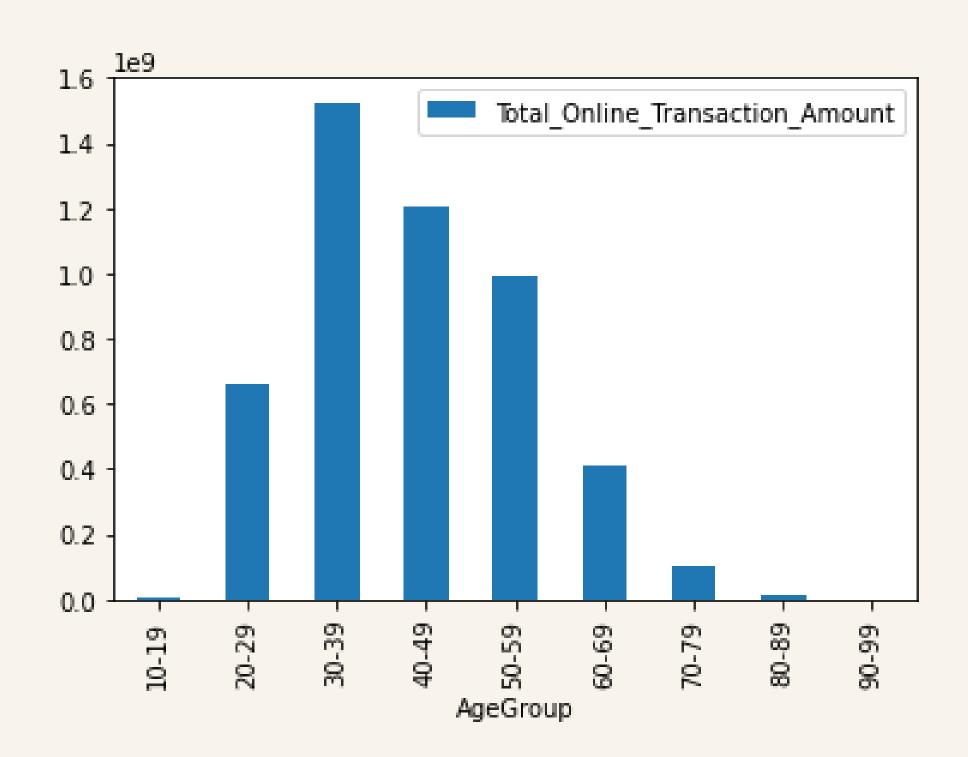
BUSINESS INSIGHTS

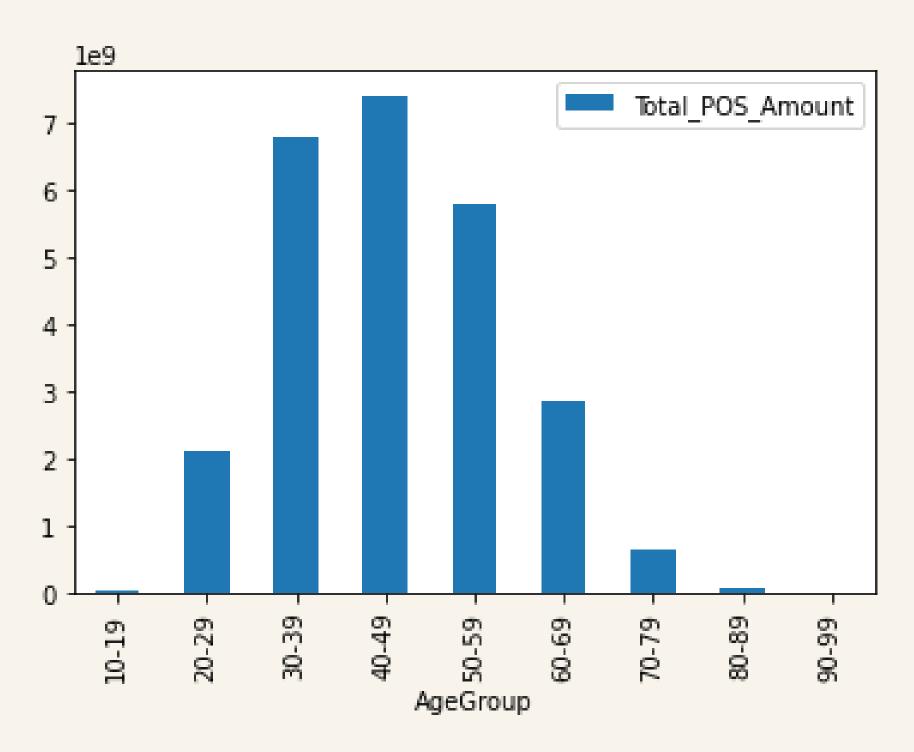
AGE ANALYSIS



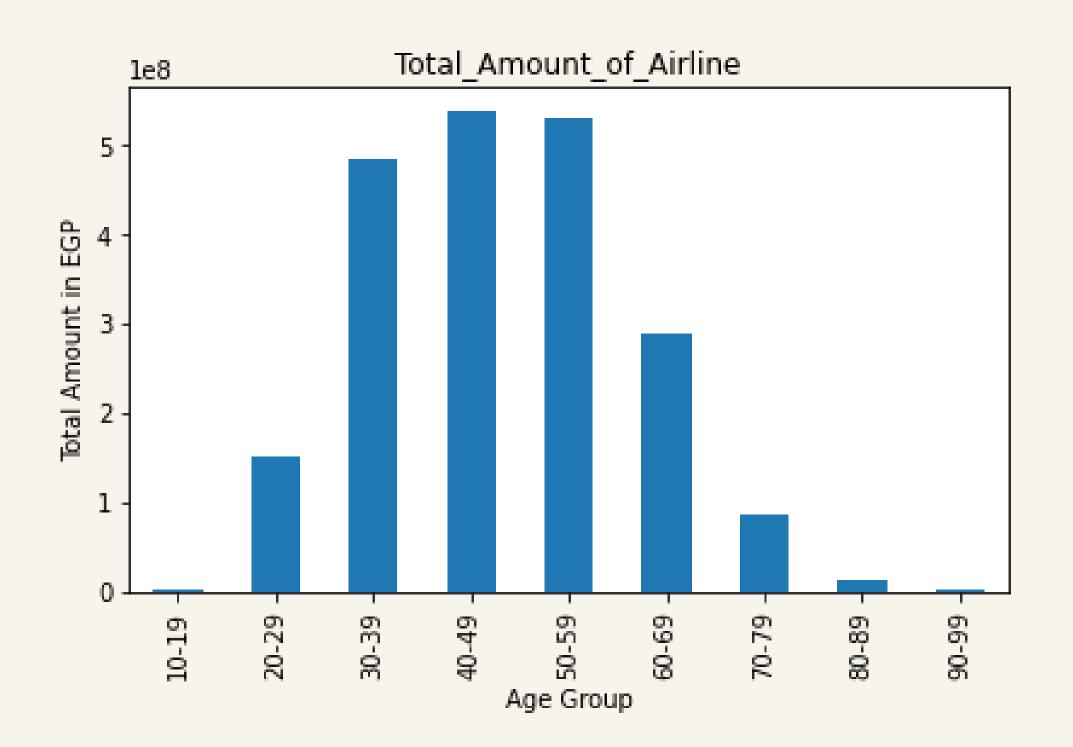
Age	Population%
30-39	33.0 %
40-49	22.4 %
20-29	17.8 %
50-59	16.5 %
60-69	7.7 %
70-79	2.0 %
10-19	0.5 %
80-89	0.2 %

AGE VS POS/ONLINE

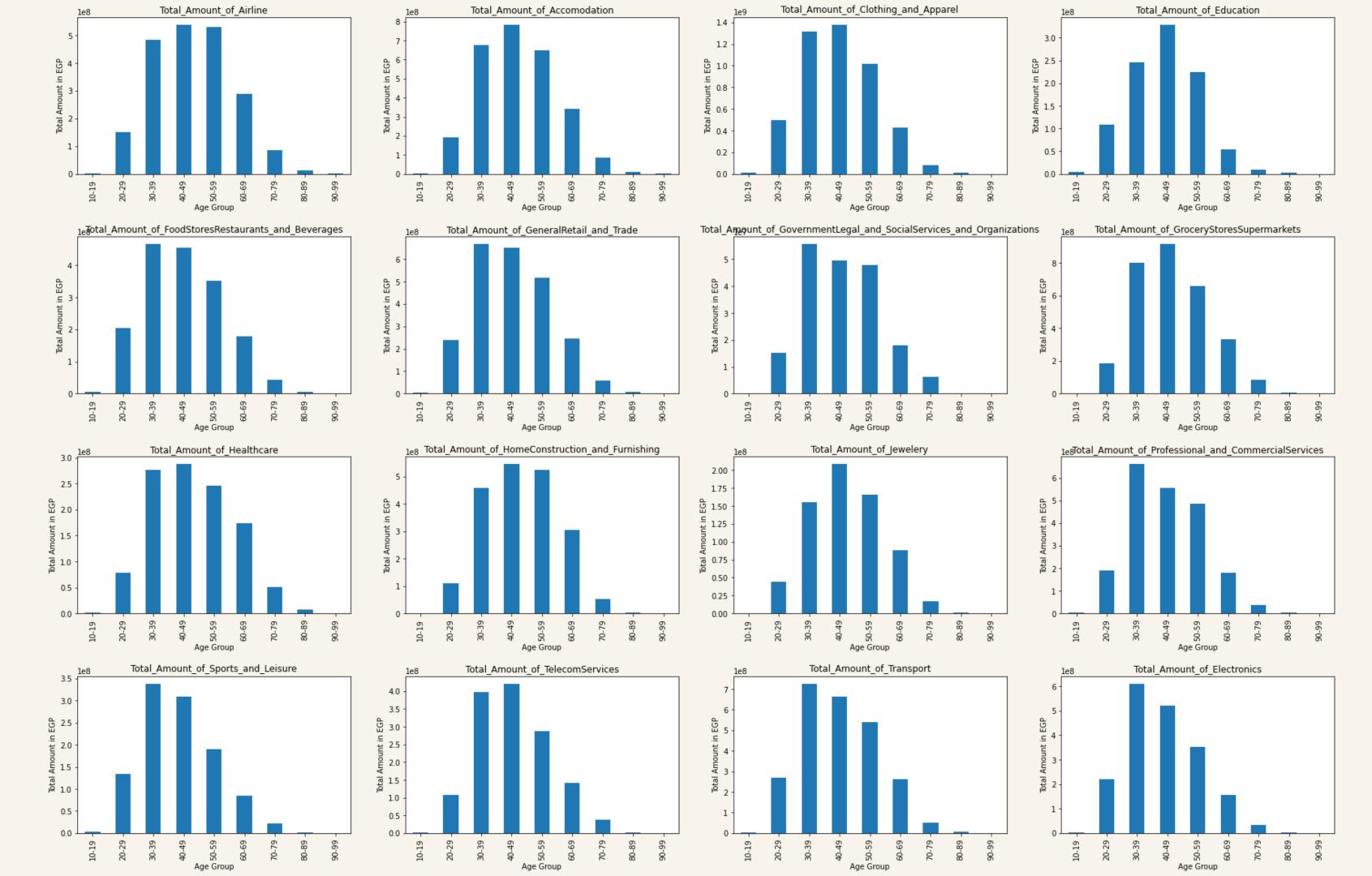




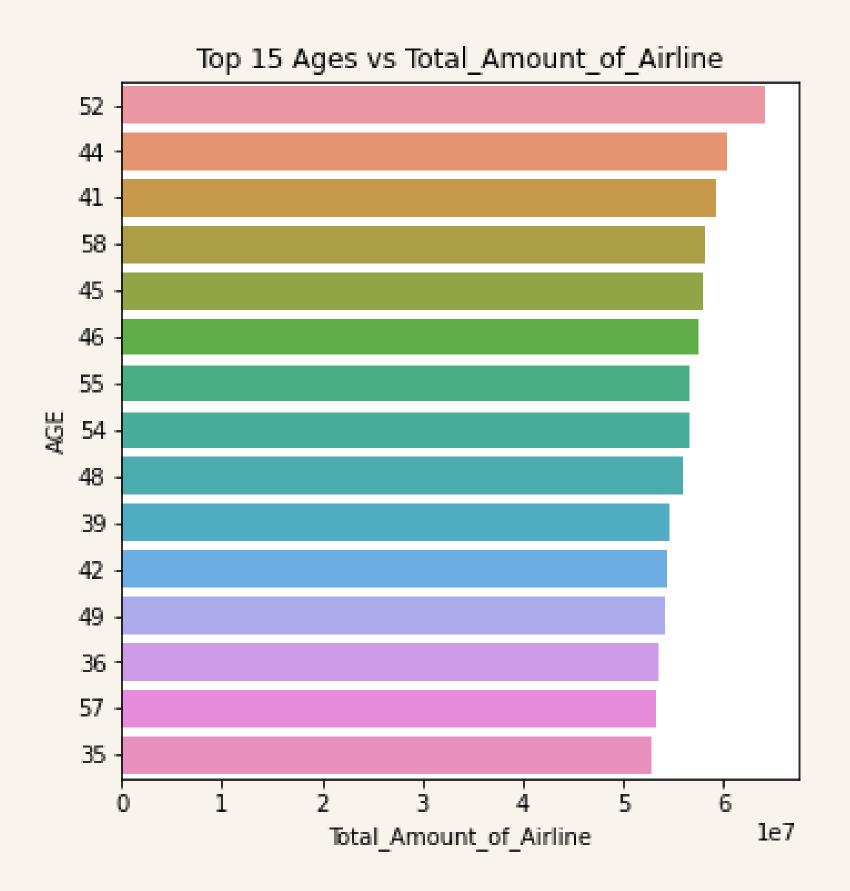
AGE VS MG



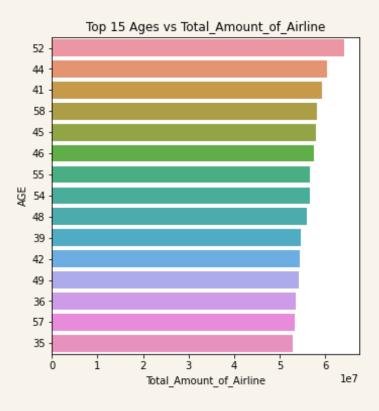
The total amount spent by every age group in all categories

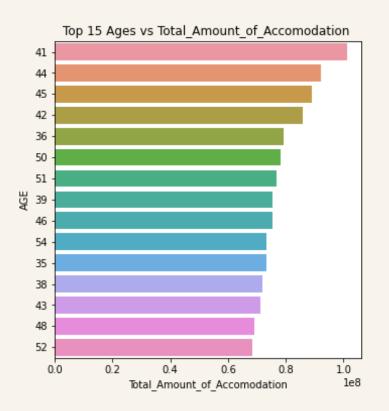


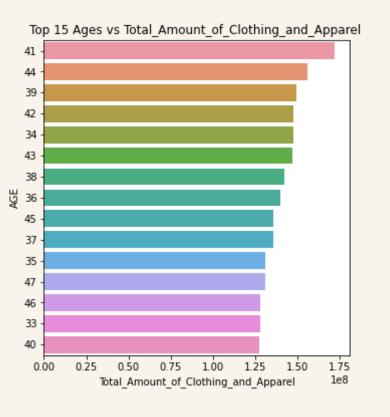
AGE VS MG

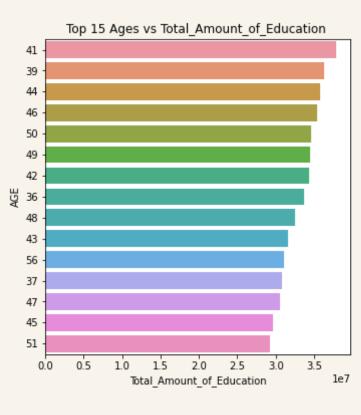


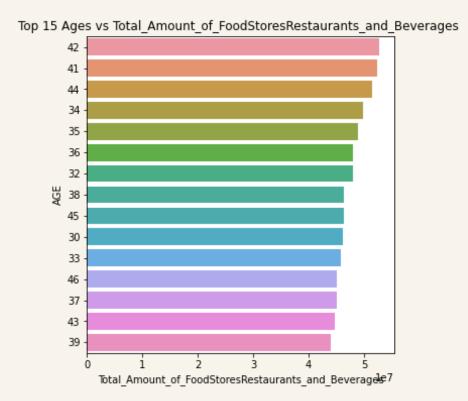
Top 15 Ages that spend in every category

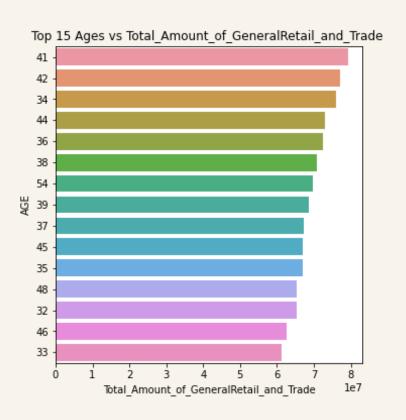


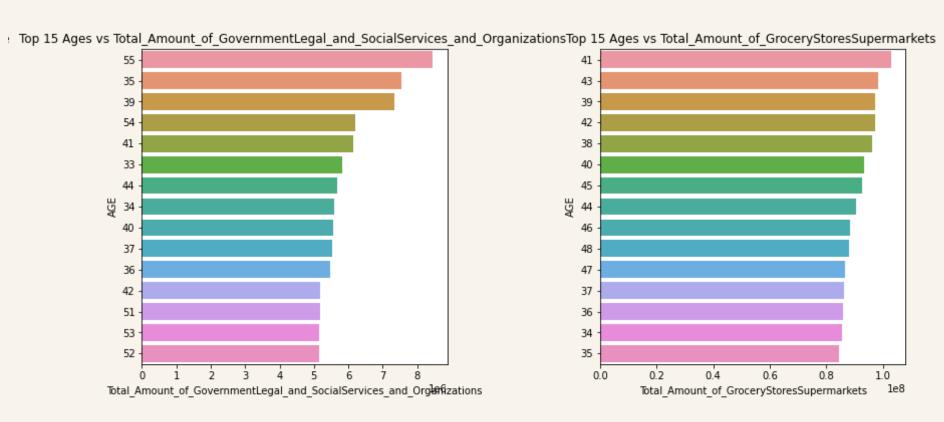






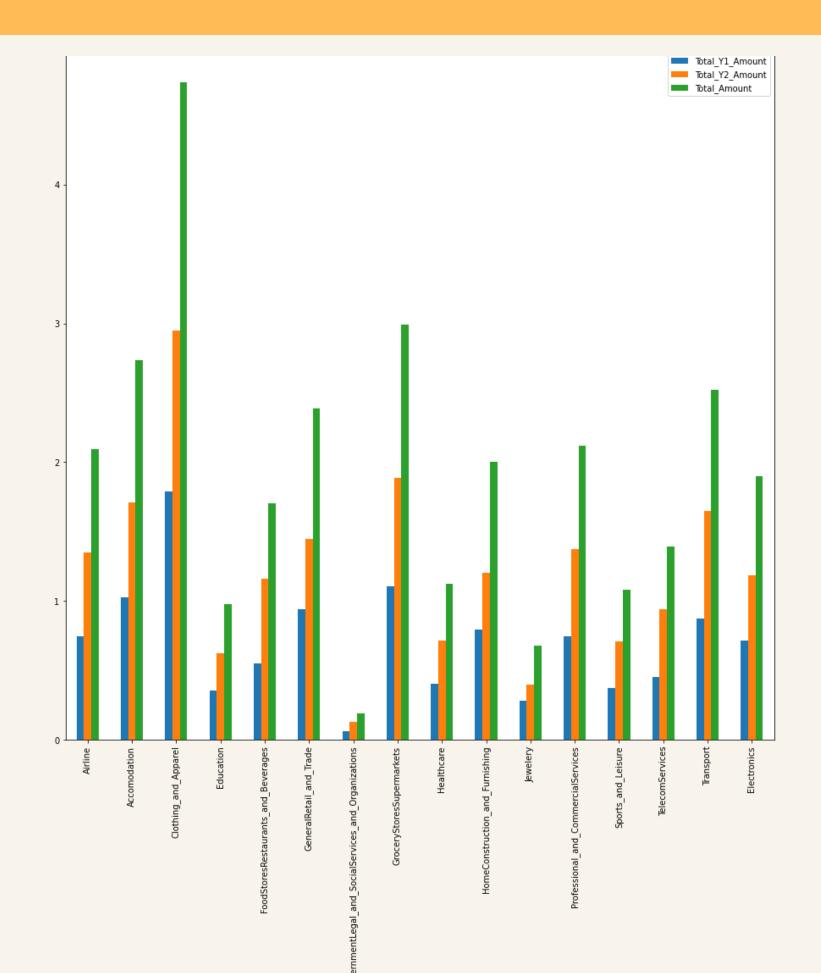






The Age range between 30 and 59 years old are the ones that spend the most in **all** categories.

MERCHANT GROUP ANALYSIS





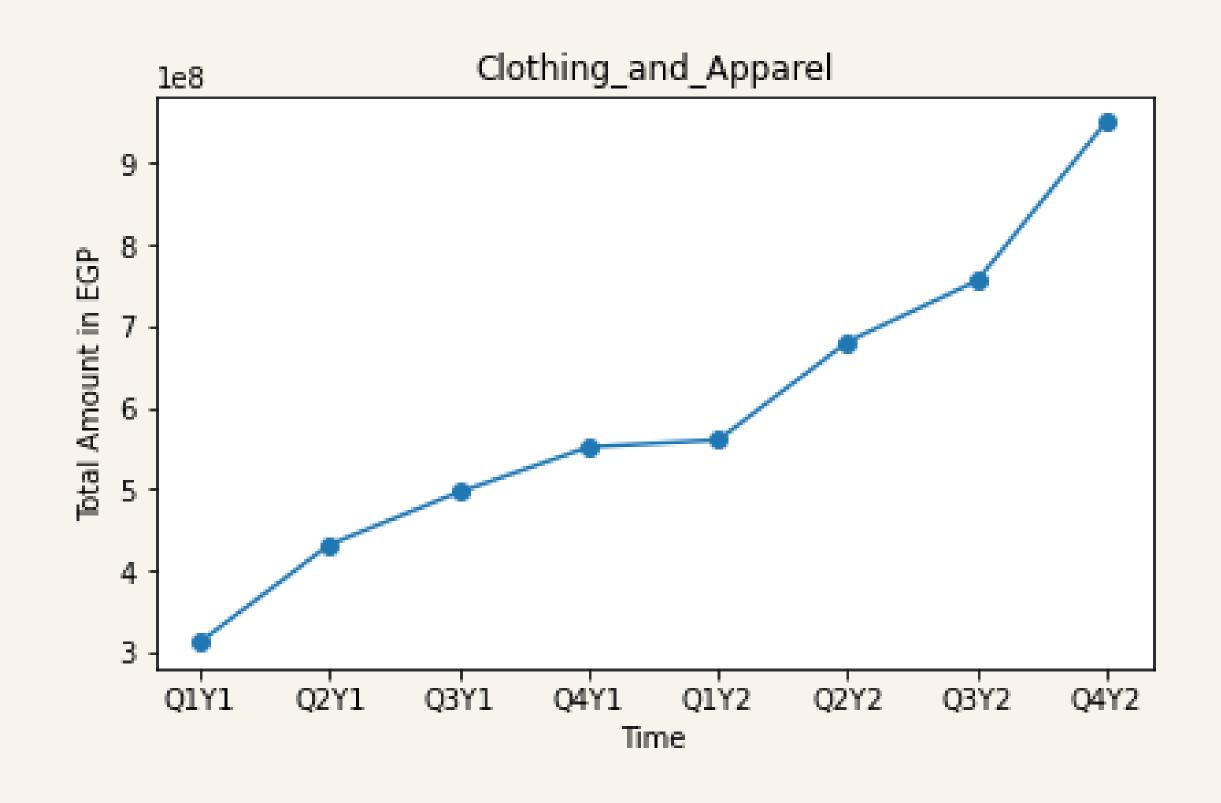
Highest Total Spend

- **GROCERY AND SUPERMARKETS**
- 3 ACCOMODATION
- TRANSPORT

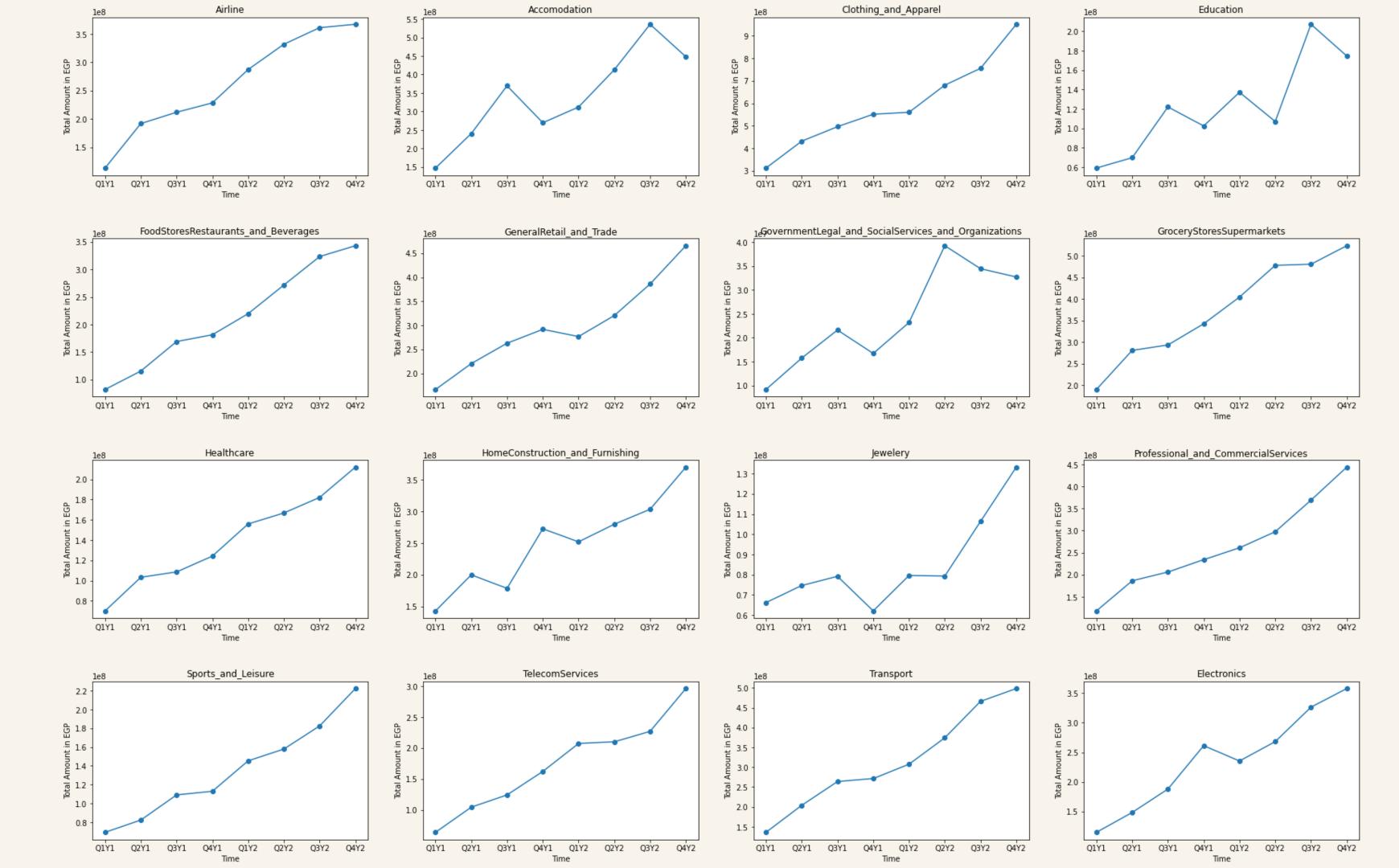
Lowest Total Spend

- GOVERNMENTAL SERVICES
- 2 JEWLERY
- **3** EDUCATION

MG TIME SERIES ANALYSIS



Total Amount
Spent in every
category over time

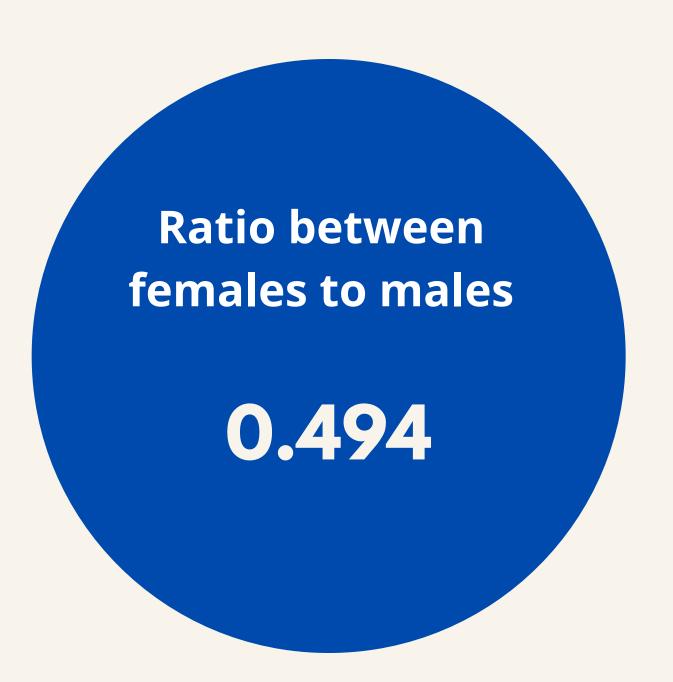


Customers tend to spend more using credit cards in almost all merchant groups by the time especially clothing, grocery stores, food & restaurants, commercial services, transport, sports.

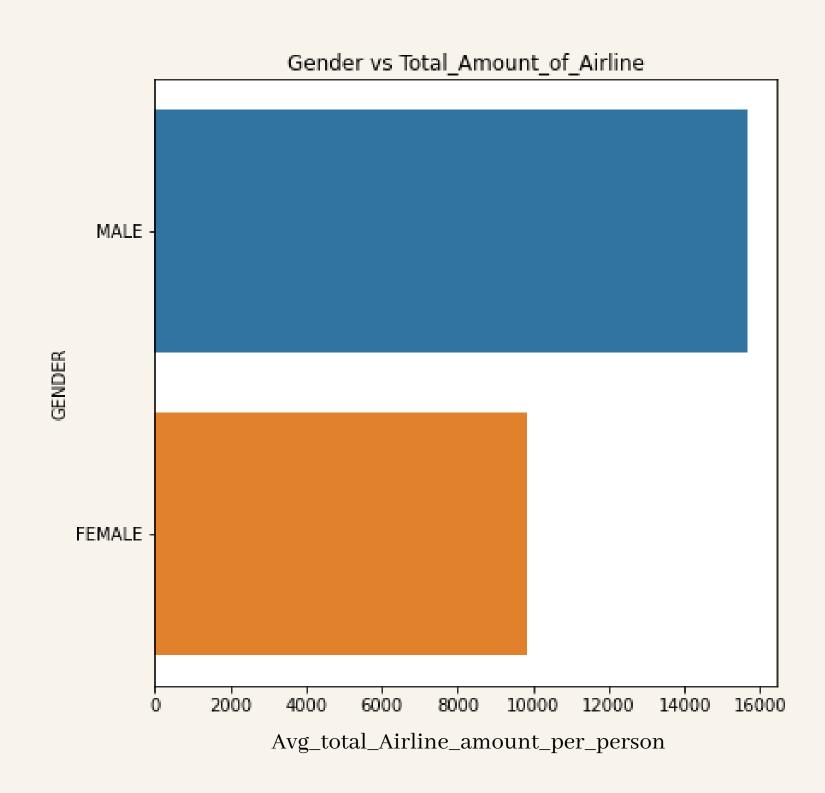
- Education sector is fluctuating
- Decrease in Accomodation total spent amount at the 4th quarter of both years
- A high increse in jewlery group in the second quarter of the second year

GENDER ANALYSIS

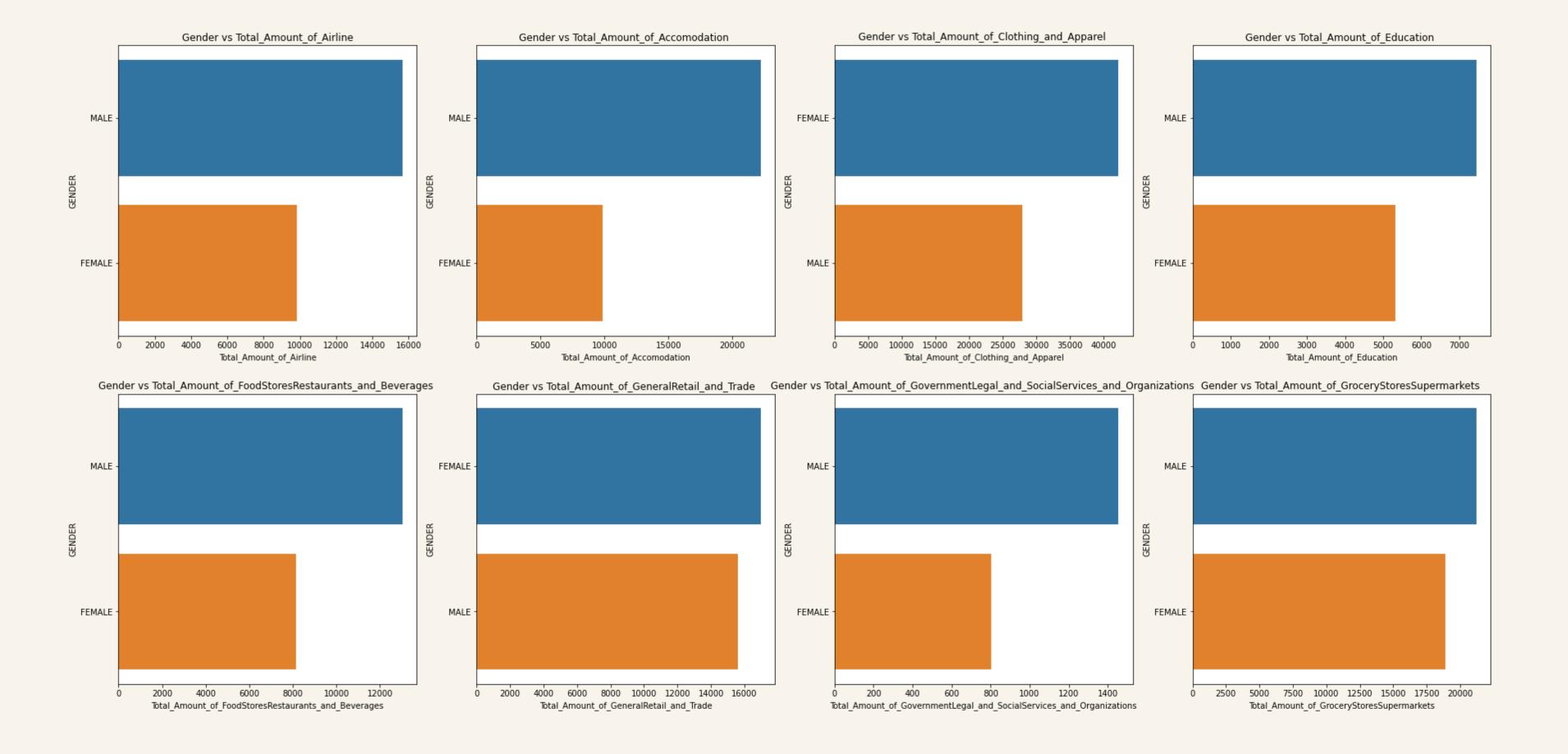
MALE 84956 FEMALE 41970 NA 16991

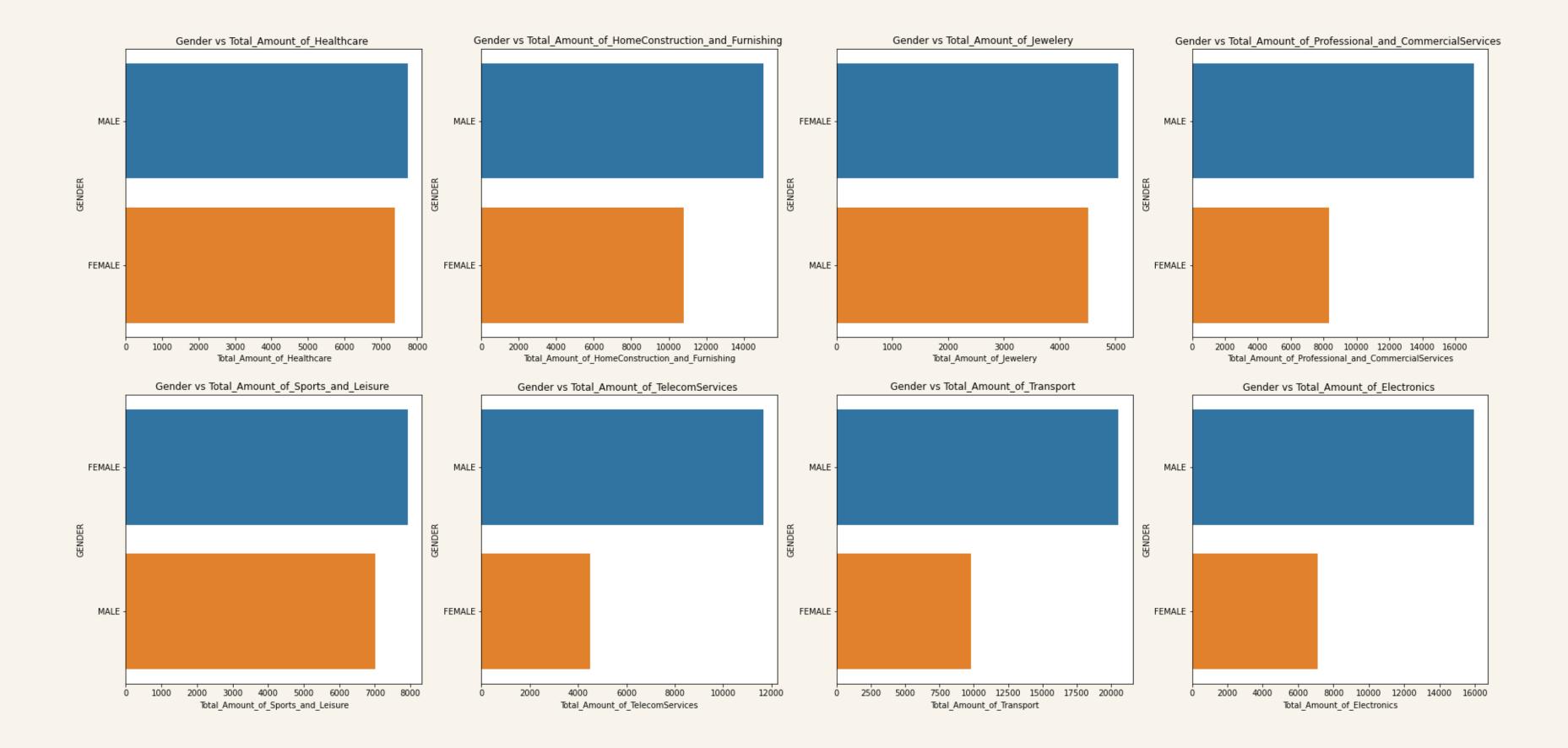


GENDER VS MG



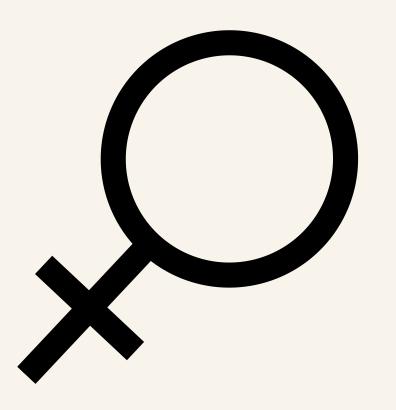
Average Total Amount spent by gender in all categories





males spend much more in Electronics, Telecom Services, Transport, Commercial services, governmental services, Acommodation, and Airline females spend much more in the Clothing and Apparel group than males.





CUSTOMER KPIs

PROFATIBILITY KPI

a feature called transaction value score was created to measure the profitability of the customer. This KPI was calculated by:

(Total Transaction Amount / Total Transaction Count)

Then, it was normalized to have a score from 0 to 1 for each user.

ACTIVITY KPI

a feature called activity score was created to measure the activity of each user according to the following formula:

(Total Retail Transactions count *
(Number of Active months in two
years / 24) + Total Cash
Transcations Count)

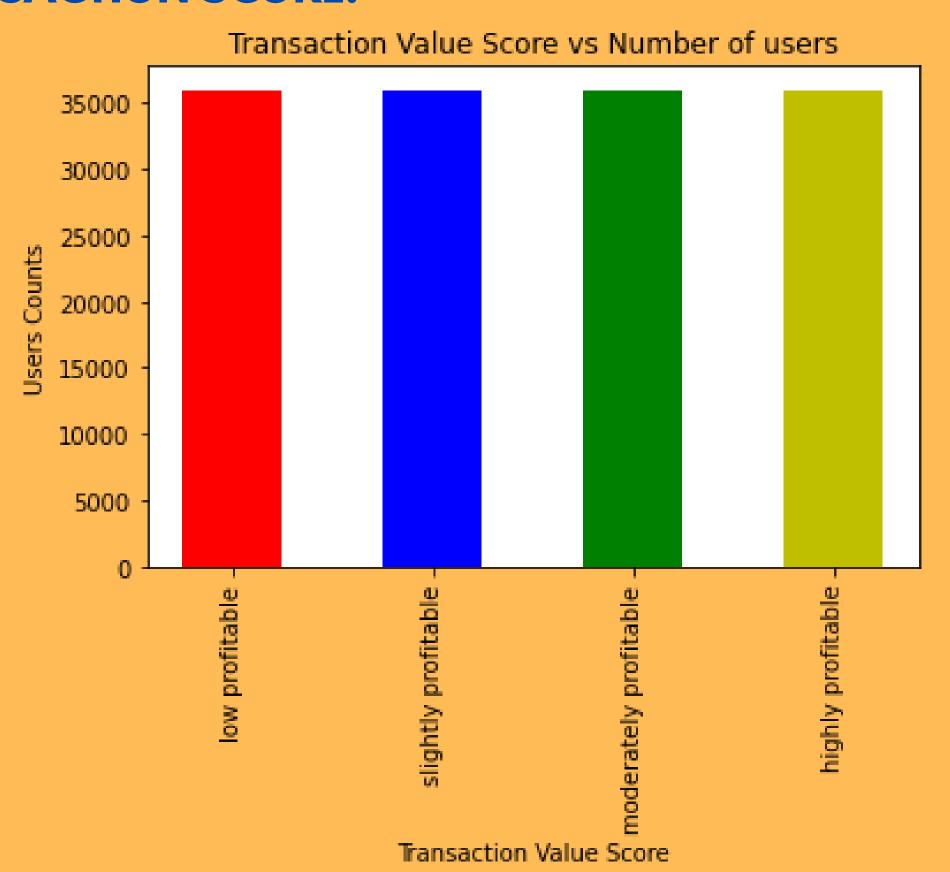
Then, it was normalized to have a score from 0 to 1 for each user.

we decided to segment to customers according to the following criterial with tailored targetted marketing plans to each segment:

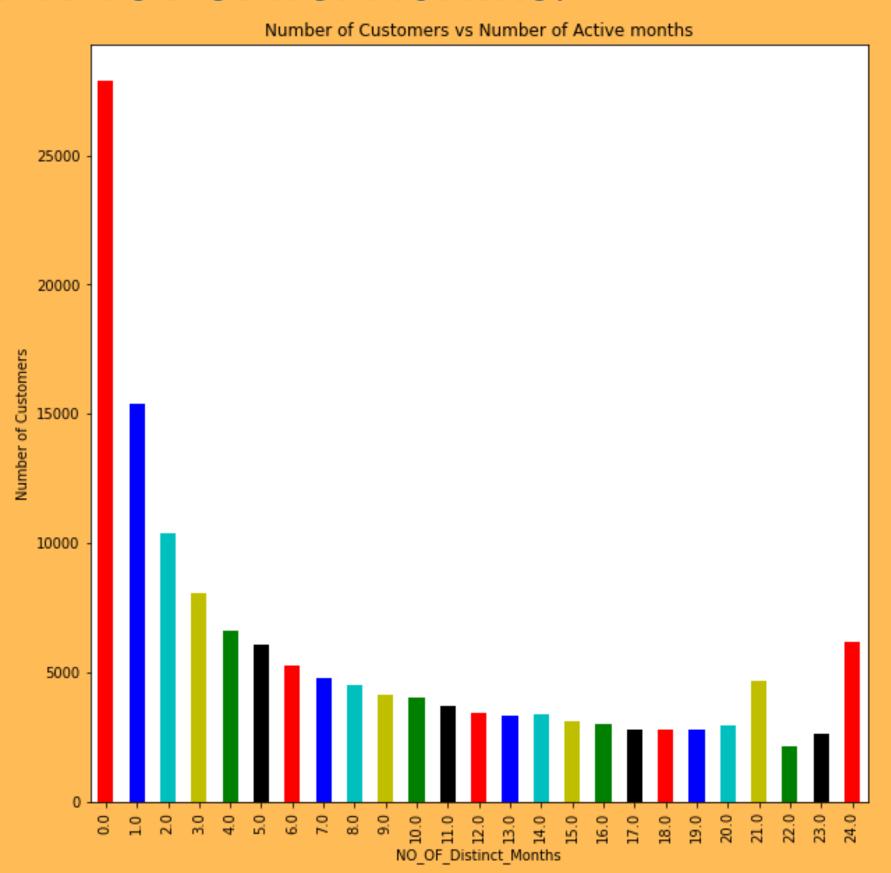
- Profitability
- Activity



BASED ON TRANSACTION SCORE:

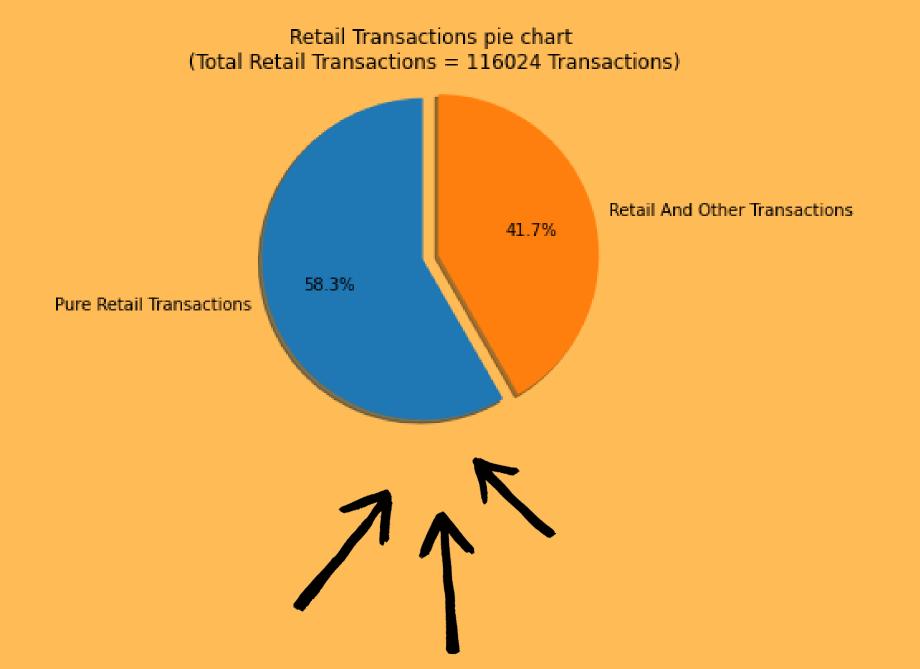


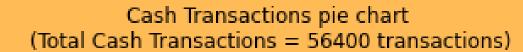
CUSTOMER ACTIVITY VS DISTINCT MONTHS:

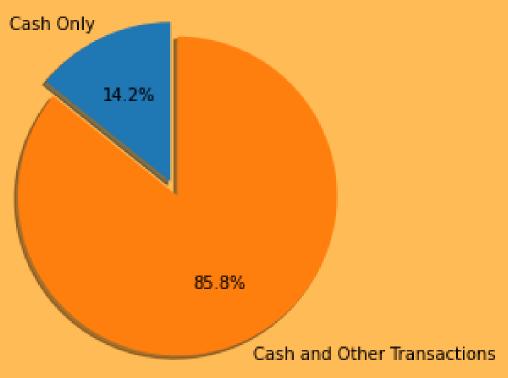


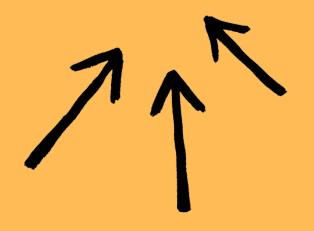
there are many customers with 0 active months since number of active months is only calculated for **retail** transactions

CASH VS RETAIL TRANSACTIONS

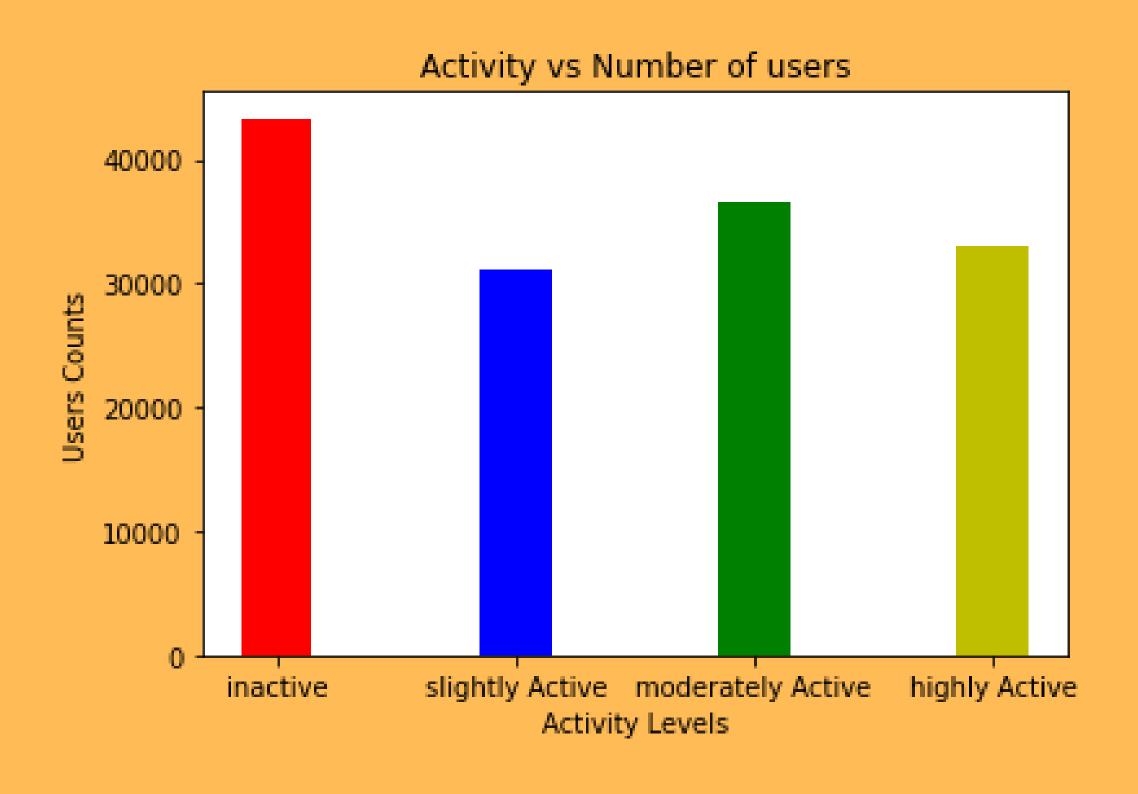








BASED ON ACTIVITY:



TARGETTED VS MASSIVE MARKETING

TARGETTED MARKETING

Pros

- cheaper and more affordable
- offers a great profit margin
- offers a better chance for leads

Cons

- Takes time to see the result
- Requires planning and strategy to determine your segments.



MASSIVE MARKETING

Pros

- It reaches a wide range of audiences
- Maximum product/services exposure

Cons

- Expensive
- Hard to measure its effectiveness
- low profit margin
- requires a lot of manpower

(Staff,2021)

(Staff, 2021)

TARGETTED MARKETING

PROFIT SEGMENTS:

low profit: offer them discounts in order to use it more - investigate their behavior more - partnerships

slightly profitable: partnerships with favorite MG - reduce fees based on number of transactions and count of transactions

moderately profitable: offer rewards - increase out of balance Credit

Highly profitable: offer rewards - reduce Credit interest - enhance their most used services - partnerships

TARGETTED MARKETING

ACTIVITY SEGMENTS:

inactive: increase awareness of credit card benefits - credit card score and usage by social media ads

slightly active: offer customized promotions based on favorite categories

moderately active: offer rewards - Increase awareness of the benefits of Highly active users

Highly active: Bonus credit - Increase credit card tenure - Remove Fees

TARGETTED MARKETING

RECOMMENDED COURSES OF ACTION:

- Create Credit Cards that are tailored for each merchant with lower credit interest and more outbalance credit than normal credit cards. for example: clothing has the most spendings in egypt. hence, clothing credit cards would be marketed well.
- Partnerships with Brands from each MG category. In clothing we can have CIB partner with Zara, H&M, Bershka, ..
- Awareness Campaigns on Credit Cards usage and benefits in collaboration with other CC competitors to reduce competition since according to CBE, in 2021, only 3M egyptians has credit cards.
- Offer Bonus Credits for Customers based on their Activity and profitability.
- Create Marketing Campaigns based on Gender, Age & MG. for example, Jewelery campaigns that targets women whereas Sports Campaigns targeting men. similar trends are found in our Analysis. The marketing channels will be tailored as well.

MET BUSINESS GOALS:

- Increase customer Satisfaction
- Increase Customer engagement using CIB credit cards
- Reduce Marketing campaign costs (Mass Campaigns vs Target Campaigns)
- Reduce Merchants transactional costs

WHY SHOULD YOU HIRE US?

Customer KPIs

Business Improvement Increase Customer Base

Exceed Customer Needs

Thorough Market Analysis

Future work recommendations

EDA AND HANDLING MISSING VALUES:

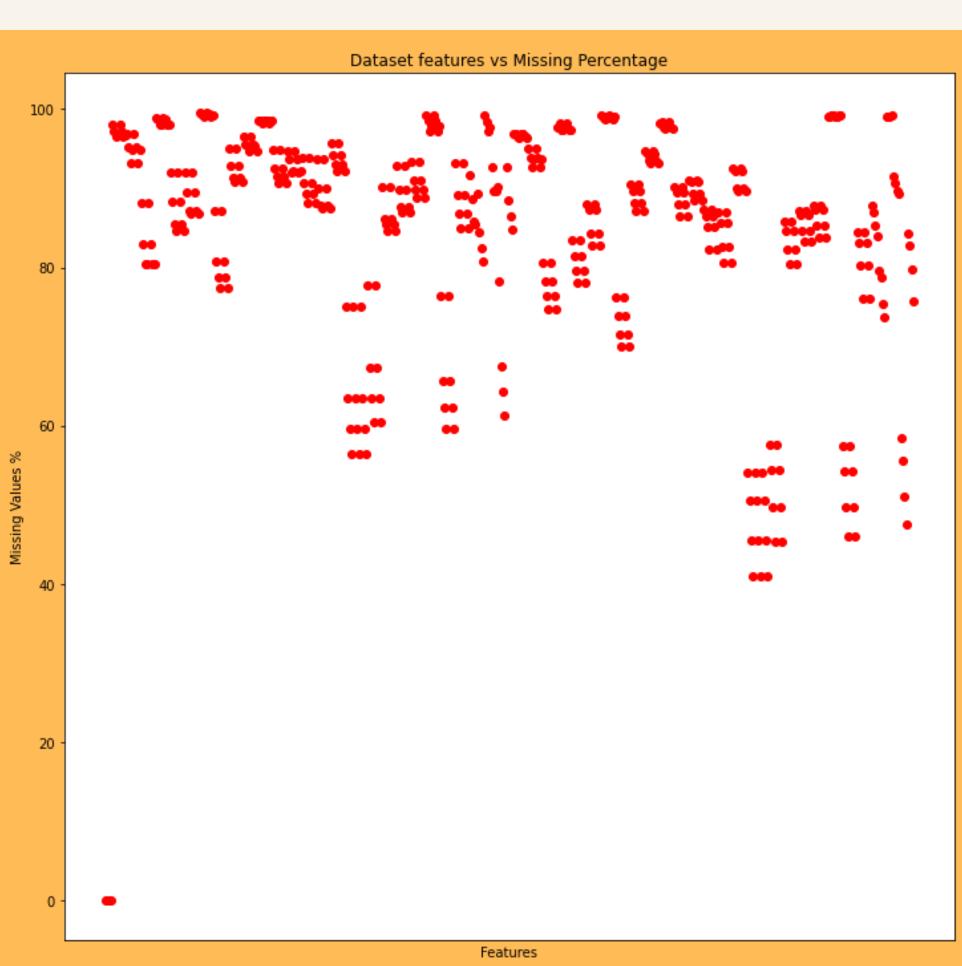
- 95.72% of the numerical columns has more than 50 % missing values
- 84.46% of the numerical columns has more than 75 % missing values
- 42.79% of the numerical columns has more than 90 % missing values

Hence, dropping the columns with most missing values does not seem like a reasonable choice.

After careful consideration:

- Regarding the Categorical values, we had only two categorical values
 [gender, batch_id]. we dropped batch_id and had a category NA for
 the missing gemders.
- Regarding the **Numerical** missing values, we decided to fill them with 0 indication that there was no any amount spent or count of transaction for this record.

The absence of information is information!



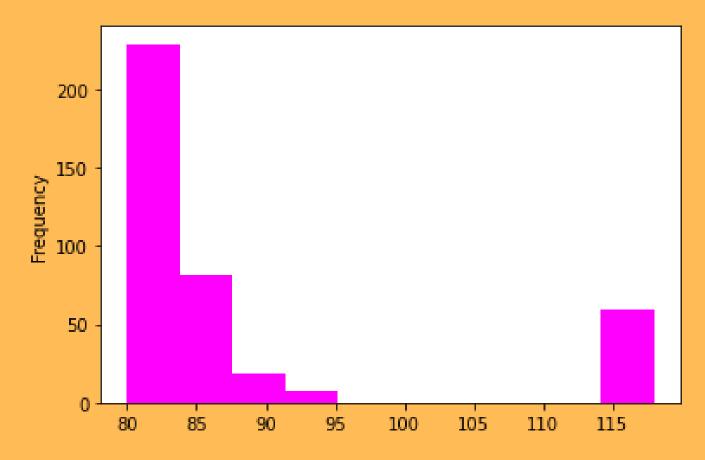
OUTLIERS DETECTION AND HANDLING:

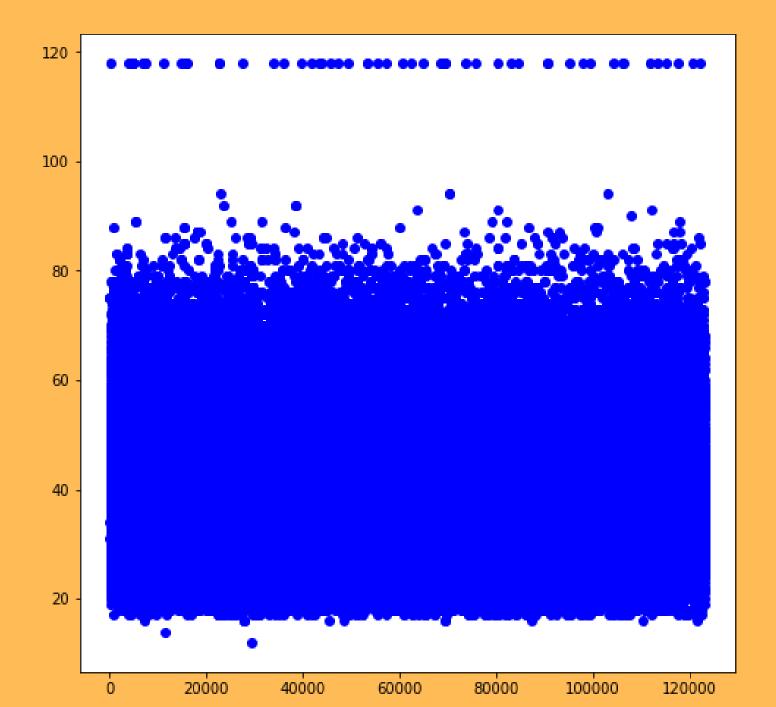
AGE:

we found a fair and reasonable for ages that were determined as outliers by calculating the z-score.

hence, we decided to go for removing the ages that are greater than 100 due to the following reasons:

- All the ages greater than 100 years old are equal to 118 years old which is kind of suspicious.
- All the customers with ages greater than 100 years old have no gender specified except for 2 customers.
- The count of the customers greater than 100 years old is only 59 customers.





FEATURE ENGINEERING

Here are some of the features that we added:

(Total_Transactions_Amount_Y1, Total_Transactions_Amount_Y2, Total_Transactions_Amount): the total amount spent by a single credit card in Y1 and Y2 and the two years sum as well.

(Total_Transactions_Count_Y1, Total_Transactions_Count_Y1, Total_Transactions_Count): the number of transactions done by a single card in Y1 and Y2 and the two years count as well.

(NO_OF_Distinct_Months_Y1 , NO_OF_Distinct_Months_Y1 , NO_OF_Distinct_Months_Y2): the total number of distinct active months for a specific card in a year and their sum as well.

Transaction value score: we defined this measure as being the (Total_Transactions_Amount / Total_Transactions_count) and will be normalized.

Total_merchant_X_score_Y1: for each merchant X of the 16 merchants, we will add a feature to represent the total in year 1 by a single card

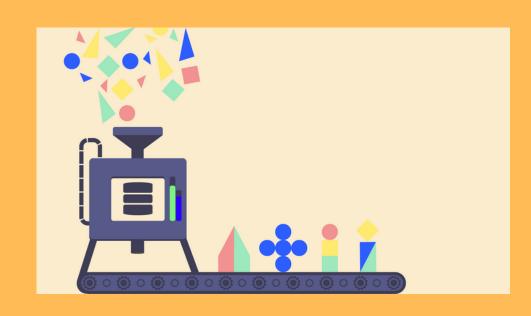
Total_merchant_X_score_Y2: for each merchant X of the 16 merchants, we will add a feature to represent the total in year 2 by a single card

Total_merchant_X_score: for each merchant X of the 16 merchants, we will add a feature to represent the total spent to this merchant over the two years by a single card.

Total_online_transactions: two new features were added for the total amount and count for online transactions.

Total_POS_transaction: two new features were added for the total amount and count for online transactions.

Activity Score : for each card, it is measure as (Total Retail Transactions count * (Number of Active months in two years / 24) + Total Cash Transcations Count) in order to be a measure of activity of the card, then it will be normalized.



DESCRIPTIVE ANALYTICS

we tackled the following Analysis in our work:

- Merchant Group Analysis by Gender and Age.
- Merchant Group Analysis wholistic spendings
- Merchant Group Analysis Trends
- Age group spendings
- Gender Analysis
- IQR analysis
- Activity Analysis
- Profitability Analysis

Future Work recommended Analysis:

• Customer level Analysis (more behavioral trends for example: users with multiple credit cards)



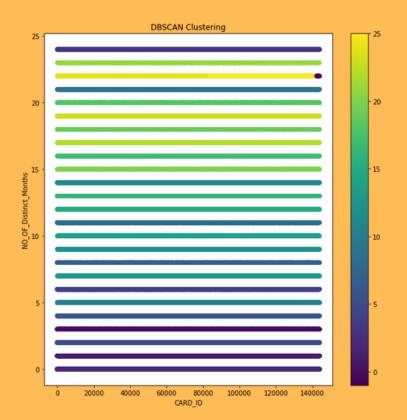
PREDICTIVE ANALYTICS

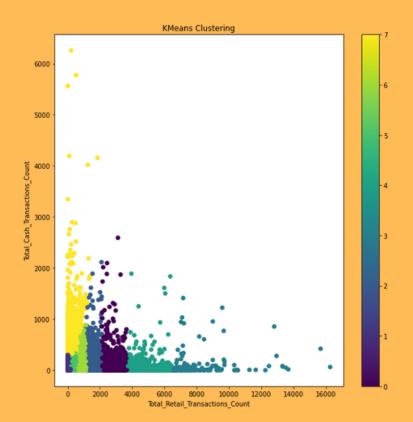
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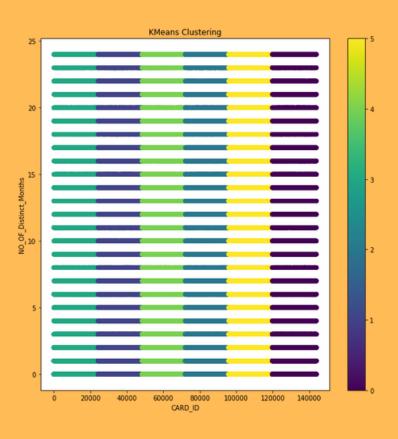
• We tackled clustering using DBScan and Kmeans. However, we did not have enough time to fine tune our clustering models in order to have meaningful clusters that can be used. however, some clustering attempts are shown below:

Recommended future work:

- Fine tune clustering models to have more meaningful clusters that can be used for customer segmentation.
- Develop a classification model that's able to lable a new customer as being (active/inactive), (profitable/not profitable)
- Develop multiple regression models that are able to predict customer's future spendings based on their History.







Script link

https://colab.research.google.com/drive/1NsJSYbLTeS44cpxsYzBJ_N9XZcSMrJ_n?usp=sharing



REFERENCES

- Effective data-driven campaigning for credit cards: Target, attract, retain and engage. AcuityKP. (n.d.). Retrieved October 23, 2021, from https://www.acuitykp.com/blog/effective-data-driven-campaigning-for-credit-cards/
- Staff, E. (2021, August 23). Mass marketing vs target marketing: Which one is better for startups. WordPress ERP. Retrieved October 23, 2021, from https://wperp.com/39946/ultimate-guide-on-mass-marketing-vs-target-marketing/.