# Data Analysis Project

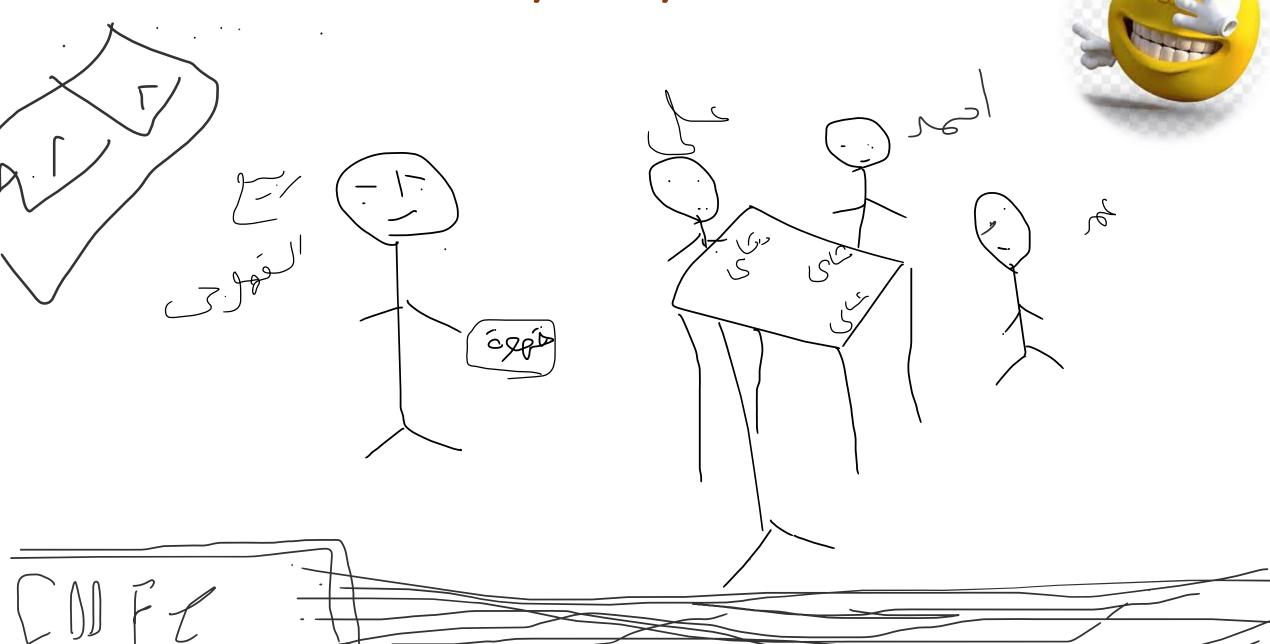
Final project of Engineer Mustafa Othman's course Data Analysis using Python





### **CAFE REWARDS**

### Funny story







The task was to perform as a Senior Marketing Analyst at Maven Cafe, who just run a test by sending different combinations of promotional offers to existing rewards members. Now that the 30-day period for the test has concluded, the task is

- to identify key customer segments
- and develop a data-driven strategy for future promotional messaging & targeting.

### The Objective

My mission as Sr. Marketing Analyst was two-fold:

- Identify Key Customer Segments: Determine which customer groups were most (and least) engaged with the different promotional offers.
- Develop a Data-Driven Strategy: Use insights gathered from the test to recommend a strategic approach for future promotional messaging and offer targeting, ensuring increased customer engagement.

### About the Data Set



#### The data consists of a table with 17'000 members

The analysis is based on data simulating customer behavior over a 30-day period, capturing transactions and interactions with different promotional offers. The dataset is split into three key tables:

- Offers table: Information on the different promotional offers, including type, difficulty, and rewards.
- Customers table: Age, gender, income, and membership details of the cafe's rewards members.
- Events table: Capturing how customers engaged with offers—whether they received, viewed, or completed an offer, and whether they made a transaction related to it.
- For a transaction to be tied to an offer, it must occur at the same time the offer was "completed" by the customer. This nuanced behavior offers an important insight into how customers interacted with the promotions.

# df1

## customers

customer\_id gender age

income

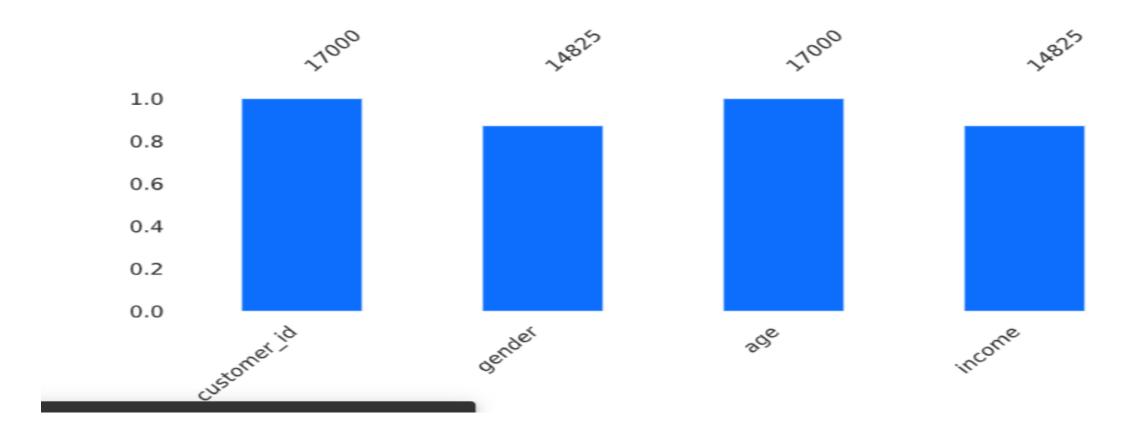
| became_member_on |                                  |     | _   |          |
|------------------|----------------------------------|-----|-----|----------|
| 2017-02-12       | 68be06ca386d4c31939f3a4f0e3dd783 | NaN | 118 | NaN      |
| 2017-07-15       | 0610b486422d4921ae7d2bf64640c50b | F   | 55  | 112000.0 |
| 2018-07-12       | 38fe809add3b4fcf9315a9694bb96ff5 | NaN | 118 | NaN      |
| 2017-05-09       | 78afa995795e4d85b5d9ceeca43f5fef | F   | 75  | 100000.0 |
| 2017-08-04       | a03223e636434f42ac4c3df47e8bac43 | NaN | 118 | NaN      |

 The first thing I knew from the data consists of 5 columns and 17,000 rows

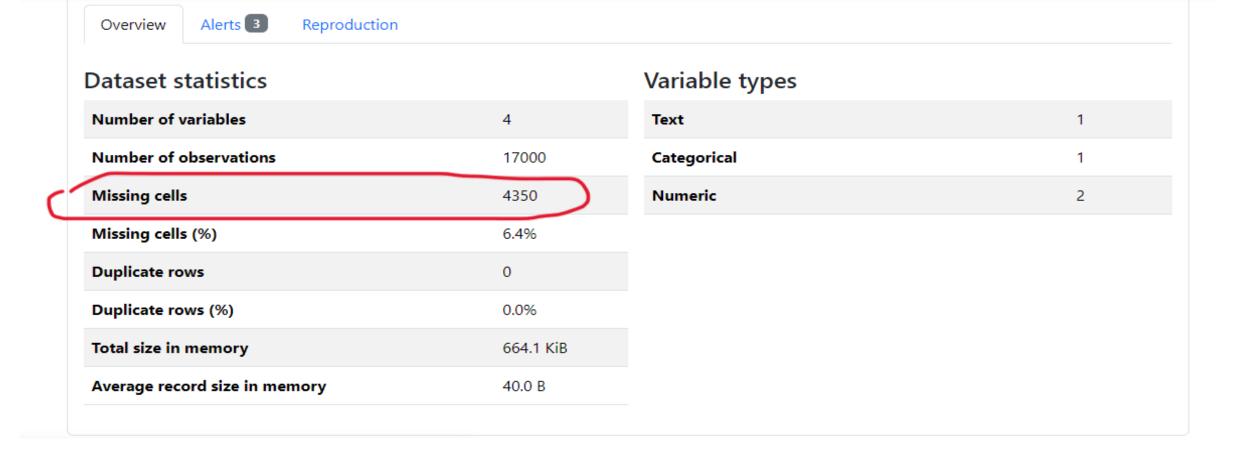
 After I called the data, the second thing I noticed was that I had a problem with the data and it was clear to everyone that in the income column there were empty cells.

 Thirdly, I made a quick report to explain everything about the data through a library.

YDATA\_PROFILING

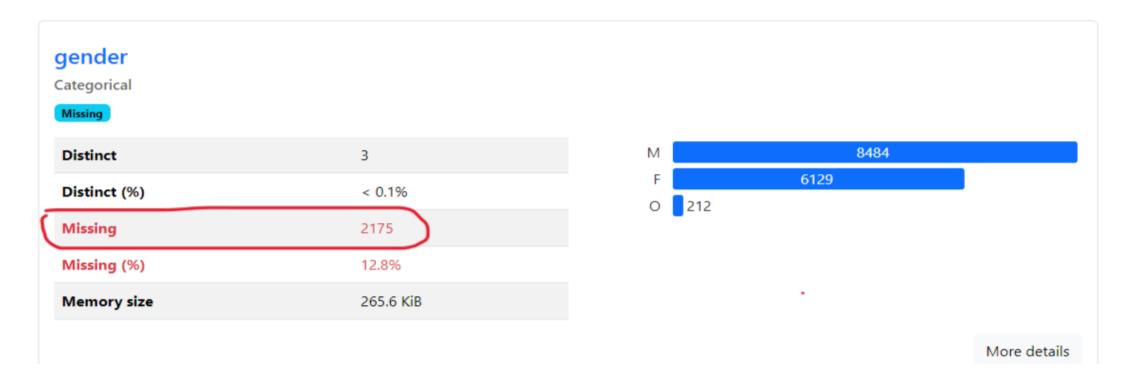


I knew that my data has a problem in the number of rows is not equal in all columns and in the difference from 14225 to 17000



I also knew that I have 4350 cells missing by 6.4%

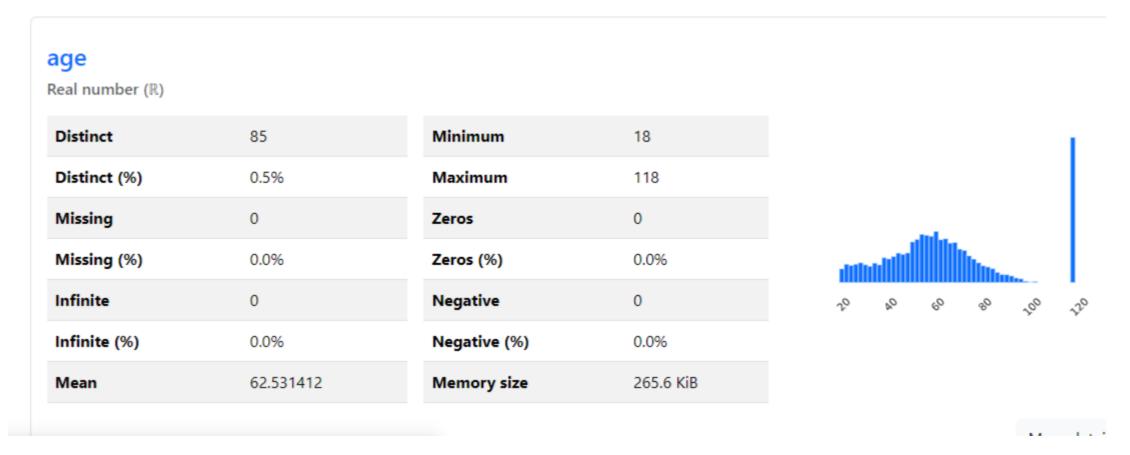
#### In the gender column:



I noticed that I have 2175 missing in the gender column

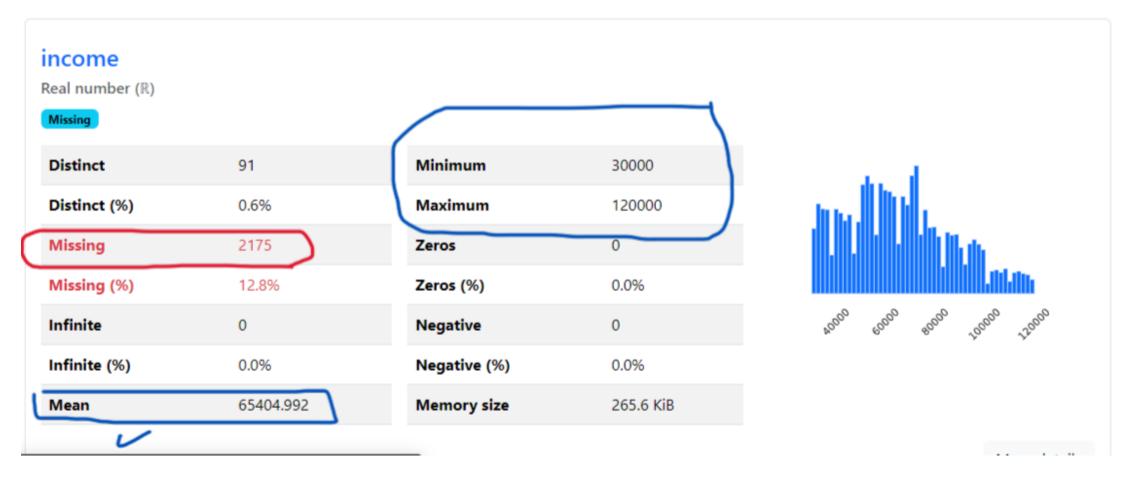
• and that there are more male customers than female customers

#### In the age column:

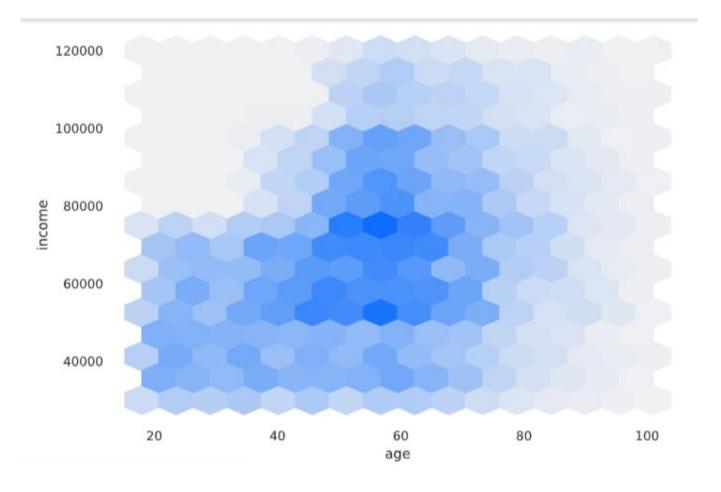


- I noticed that the ages of our customers range from 18 to 118
- I also learned that the average age of our customers is 62

#### In the income column:



- Here again there were missing 2175 cells
- and that the income of customers starts from 30,000 to 120,000 \$ with an average of 65,404 \$



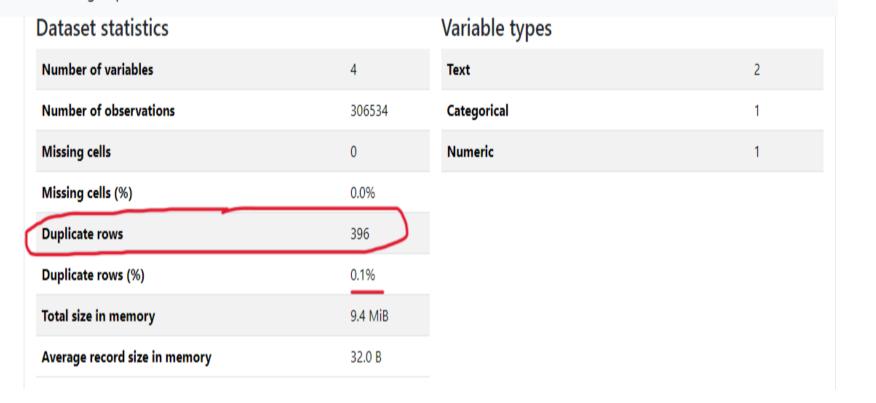
This was a chart showing the relationship between income and age.

• We notice that the number of customers is concentrated at the age of 60 and the income is about 70,000.

# df2

### events

| 1  | customer_id                      | event 🔻        | value  | ▼ tir | me | J   |
|----|----------------------------------|----------------|--|-------|----|-----|
| 2  | 78afa995795e4d85b5d9ceeca43f5fef | offer received | {'offer id': '9b98b8c7a33c4b65b9aebfe6a799e6d9'} |       |    |     |
| 3  | a03223e636434f42ac4c3df47e8bac43 | offer received | {'offer id': '0b1e1539f2cc45b7b9fa7c272da2e1d7'} |       |    |     |
| 4  | e2127556f4f64592b11af22de27a7932 | offer received | {'offer id': '2906b810c7d4411798c6938adc9daaa5'} |       |    | -   |
| 5  | 8ec6ce2a7e7949b1bf142def7d0e0586 | offer received | {'offer id': 'fafdcd668e3743c1bb461111dcafc2a4'} |       |    | -   |
| 6  | 68617ca6246f4fbc85e91a2a49552598 | offer received | {'offer id': '4d5c57ea9a6940dd891ad53e9dbe8da0'} |       |    | -   |
| 7  | 389bc3fa690240e798340f5a15918d5c | offer received | {'offer id': 'f19421c1d4aa40978ebb69ca19b0e20d'} |       |    | -   |
| 8  | c4863c7985cf408faee930f111475da3 | offer received | {'offer id': '2298d6c36e964ae4a3e7e9706d1fb8c2'} |       |    | - ( |
| 9  | 2eeac8d8feae4a8cad5a6af0499a211d | offer received | {'offer id': '3f207df678b143eea3cee63160fa8bed'} |       |    | - ( |
| 10 | aa4862eba776480b8bb9c68455b8c2e1 | offer received | {'offer id': '0b1e1539f2cc45b7b9fa7c272da2e1d7'} |       |    | - ( |
| 11 | 31dda685af34476cad5bc968bdb01c53 | offer received | {'offer id': '0b1e1539f2cc45b7b9fa7c272da2e1d7'} |       |    | -   |
| 12 | 744d603ef08c4f33af5a61c8c7628d1c | offer received | {'offer id': '0b1e1539f2cc45b7b9fa7c272da2e1d7'} |       |    |     |
| 10 | 24022455015545016762006555200070 | offerroseived  | ('offerid', '0b1e1520f2ee45b7b0fe7e272de2e1d7')  |       |    |     |

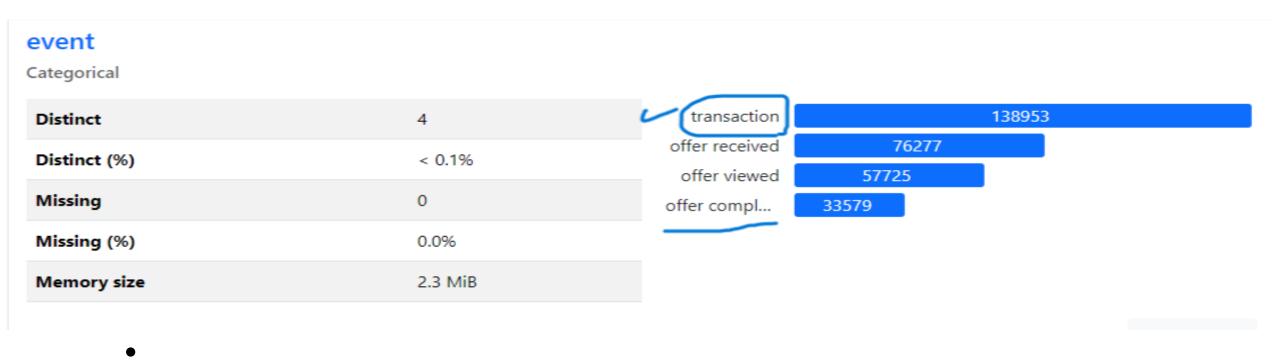


I made a quick report to explain everything about the data through a library ... YDATA\_PROFILING

I discovered that the data consists of 306534 rows and 4 columns

• I noticed that I have 396 duplicate rows, which is 0.1%.

#### In the event column:



- I noticed that I have transaction 1389953 completed
- and 33579 offer completed only.

# df3

# offers

#### 10\*6

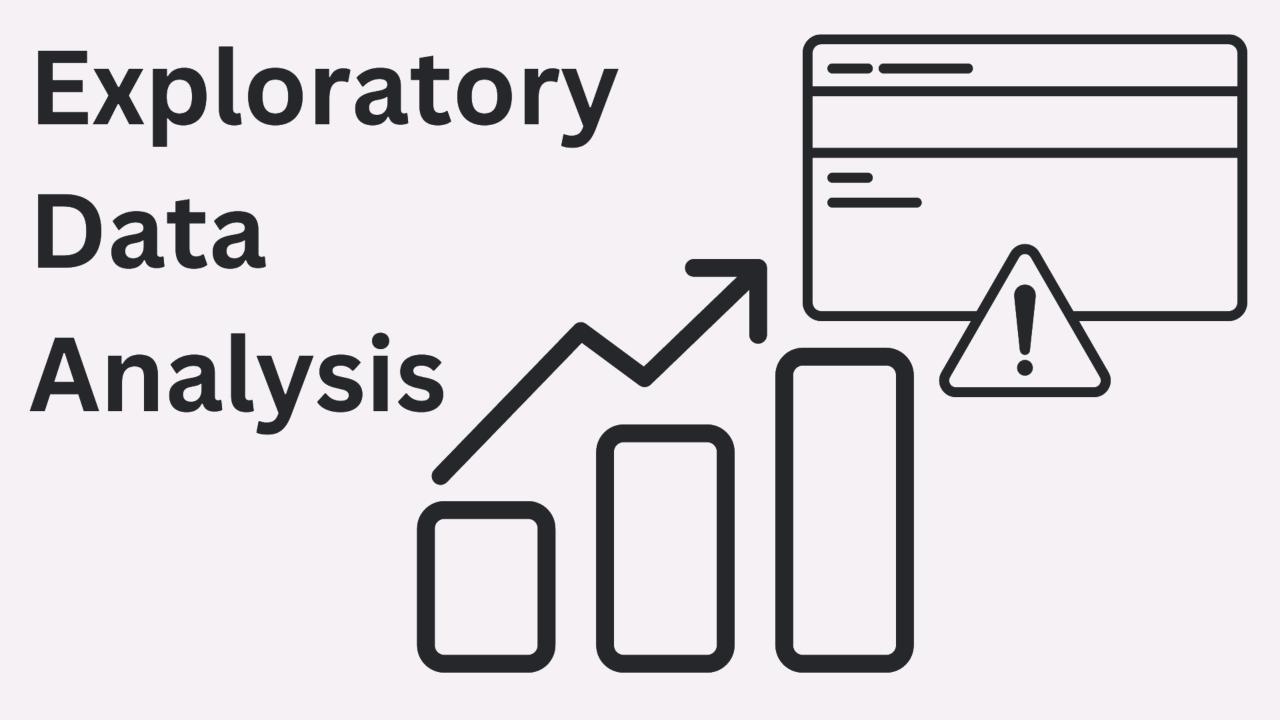
| 1  | offer_id 🔻                       | offer_type 💌  | difficulty 💌 | reward 💌 | duration 💌 | channels                             |
|----|----------------------------------|---------------|--------------|----------|------------|--------------------------------------|
| 2  | ae264e3637204a6fb9bb56bc8210ddfd | bogo          | 10           | 10       | 7          | ['email', 'mobile', 'social']        |
| 3  | 4d5c57ea9a6940dd891ad53e9dbe8da0 | bogo          | 10           | 10       | 5          | ['web', 'email', 'mobile', 'social'] |
| 4  | 3f207df678b143eea3cee63160fa8bed | informational | 0            | 0        | 4          | ['web', 'email', 'mobile']           |
| 5  | 9b98b8c7a33c4b65b9aebfe6a799e6d9 | bogo          | 5            | 5        | 7          | ['web', 'email', 'mobile']           |
| 6  | 0b1e1539f2cc45b7b9fa7c272da2e1d7 | discount      | 20           | 5        | 10         | ['web', 'email']                     |
| 7  | 2298d6c36e964ae4a3e7e9706d1fb8c2 | discount      | 7            | 3        | 7          | ['web', 'email', 'mobile', 'social'] |
| 8  | fafdcd668e3743c1bb461111dcafc2a4 | discount      | 10           | 2        | 10         | ['web', 'email', 'mobile', 'social'] |
| 9  | 5a8bc65990b245e5a138643cd4eb9837 | informational | 0            | 0        | 3          | ['email', 'mobile', 'social']        |
| 10 | f19421c1d4aa40978ebb69ca19b0e20d | bogo          | 5            | 5        | 5          | ['web', 'email', 'mobile', 'social'] |
| 11 | 2906b810c7d4411798c6938adc9daaa5 | discount      | 10           | 2        | 7          | ['web', 'email', 'mobile']           |

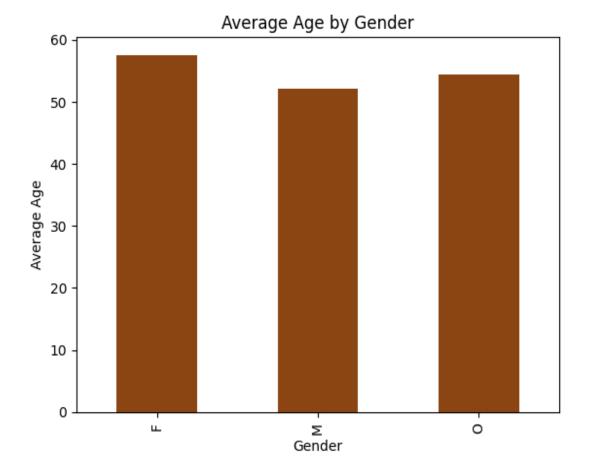
### Daata Cleaning:

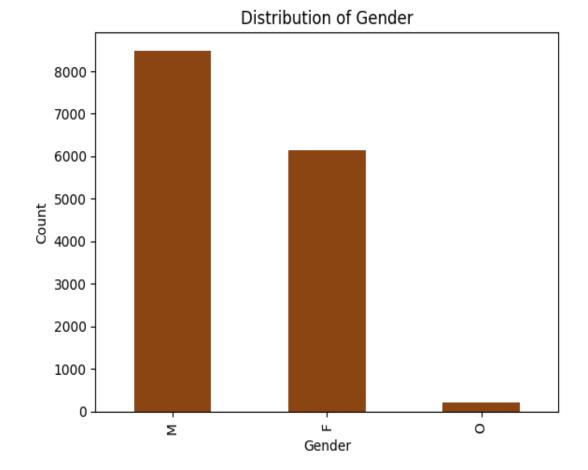


After solving the missing problem in df1 and duplicating in f2,

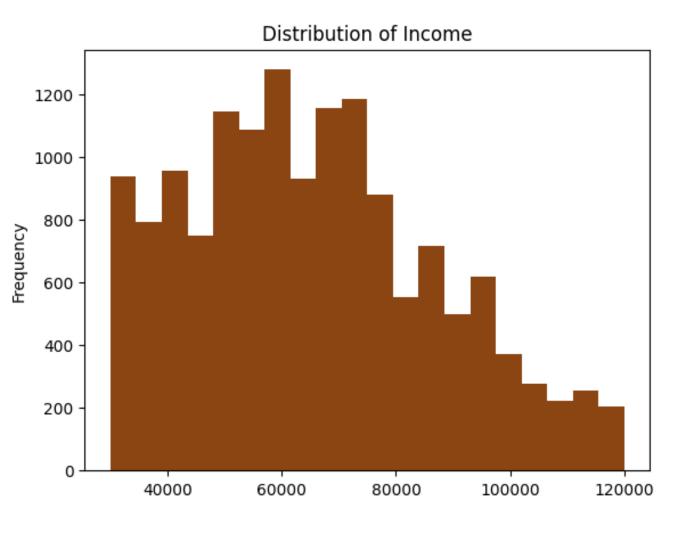
- df1 now consists of **5\*14825**
- and df2 consists of **4\*306137**



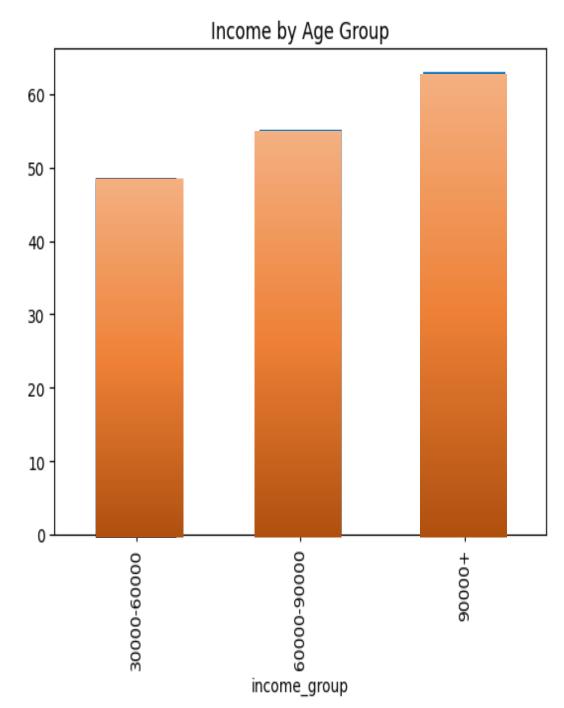


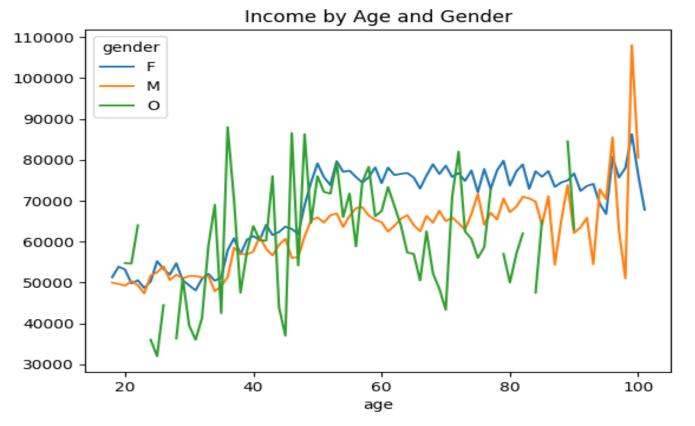


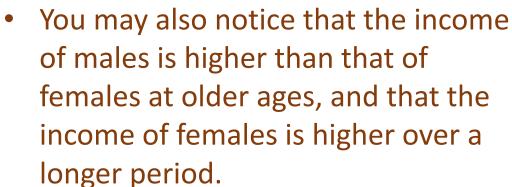
 It shows the average age of men and women and the slight difference for women, with the average age of women being 57.5 and the average age of men being 52.1.  It shows that the number of women is more than that men, as their number is 8484 and the number of men is 6129.

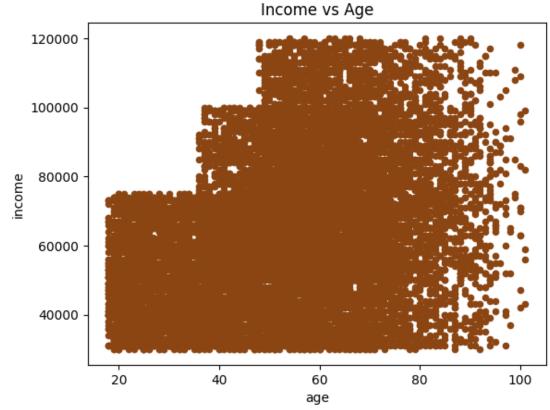


It shows that the average income is
 65404

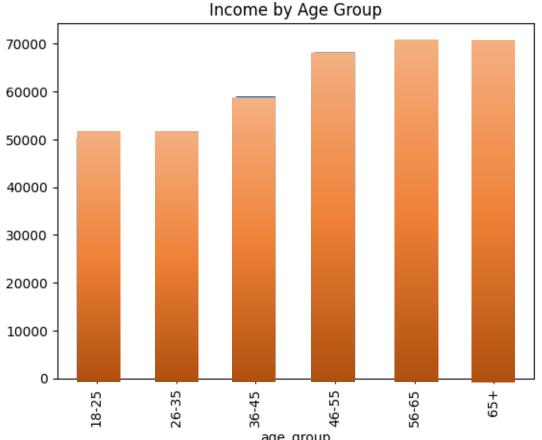


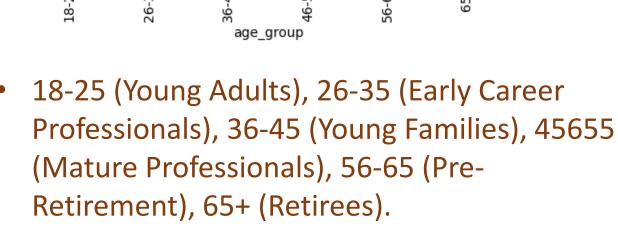


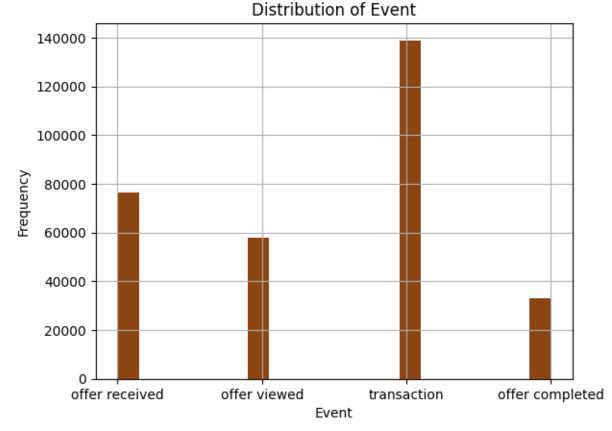




 We notice that the largest number of customers is around 50-70 years old.





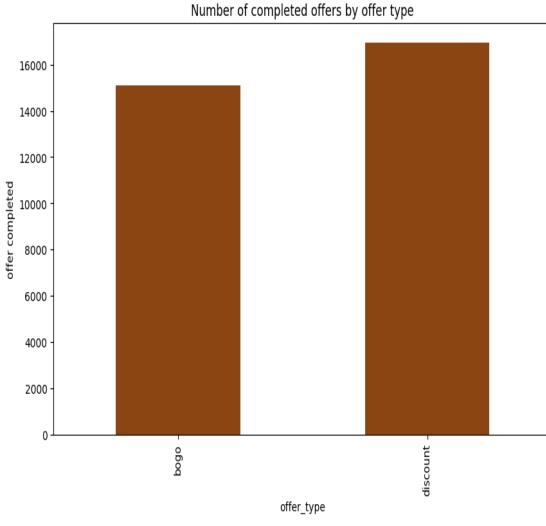


| • | transaction     | 138953 |
|---|-----------------|--------|
| • | offer received  | 76277  |
| • | offer viewed    | 57725  |
| • | offer completed | 33182  |

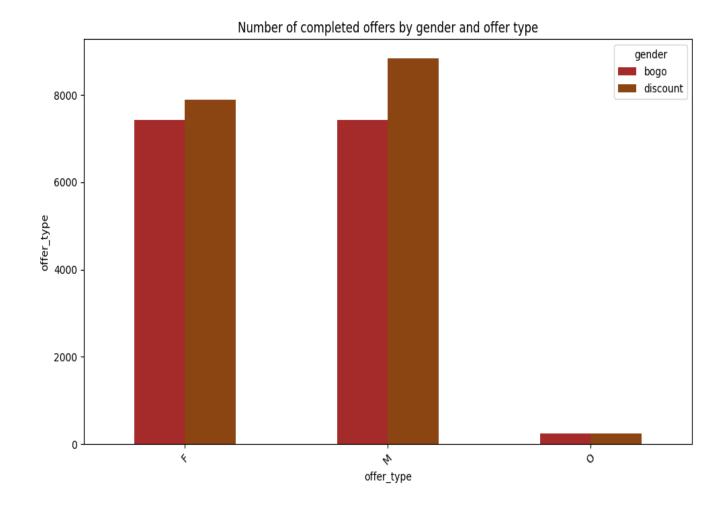
### **Analytical Questions**



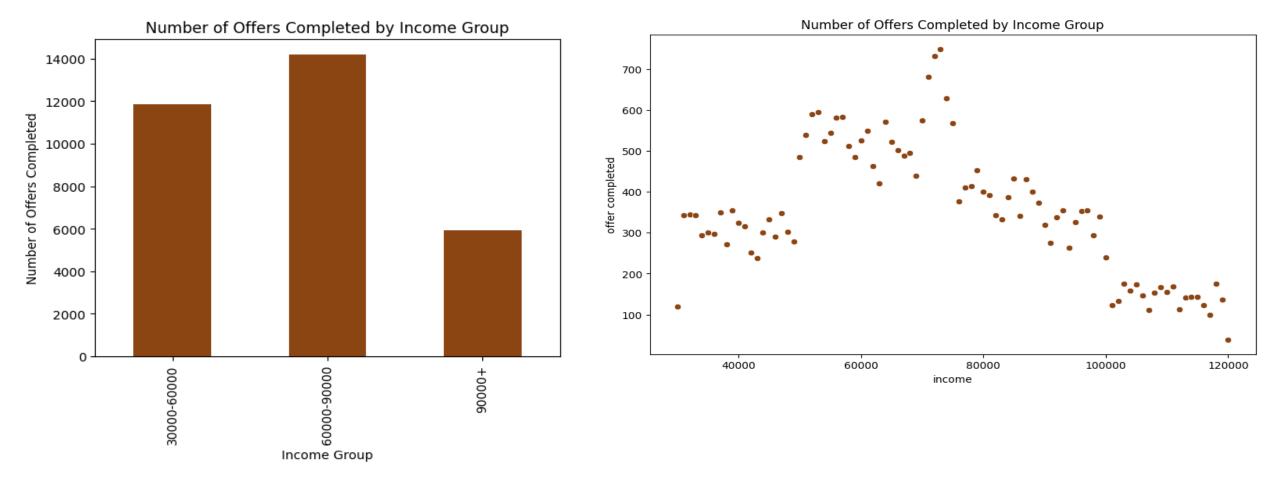
- What is the age distribution for each offer type?
- How does gender affect the success of offers (e.g., number of offers completed for each gender)?
- What are the most popular offers based on number of offers completed?
- Is there a relationship between income and the success of offers (offers completed)?
- What channels are most effective for promoting offers?



- Offer (bogo) used 15100
- Offer (discount) used 16970

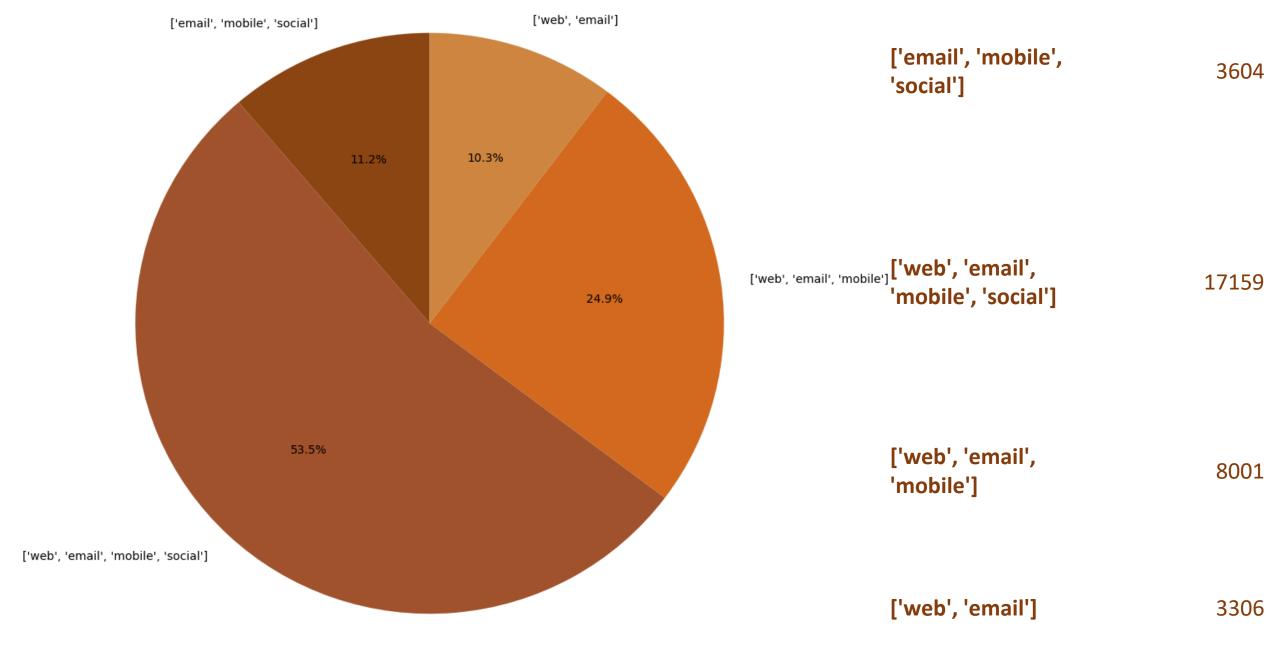


 Male and Female used discount more than bogo.



- We note that those with a low income of 30,000-60,000 have a good number of completed offers 11,838
- We also note that those with a high income of 90,000+ have a low number of completed offers 5,924
- But the largest percentage of completed offers was in the medium income category of 60,000-90,000 and they have **14,188** completed offers



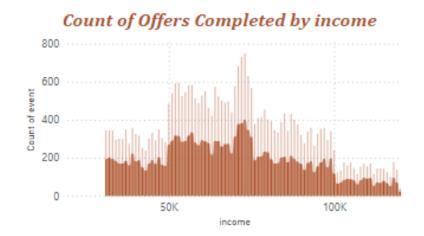


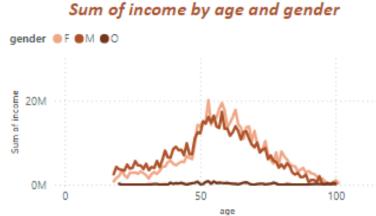


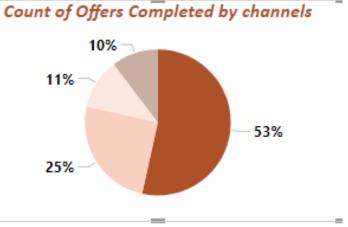


17 **N K** 

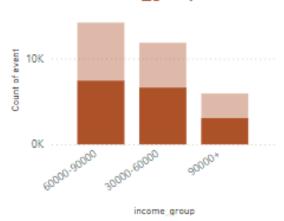




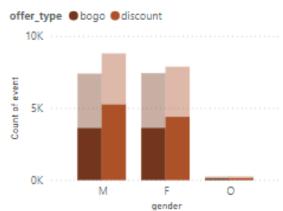




Count of Offers Completed by income\_group



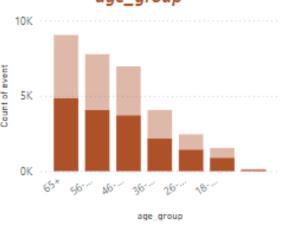
Count of offers completed by gender and offer\_type



Count of offers completed and Count of gender by offer\_type

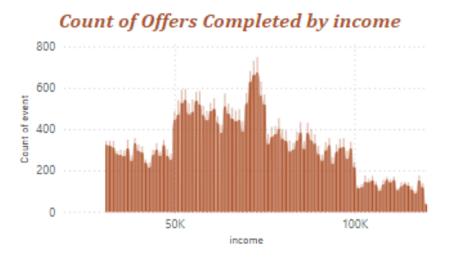


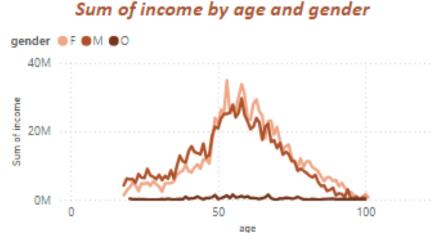
Count of offers completed by age\_group



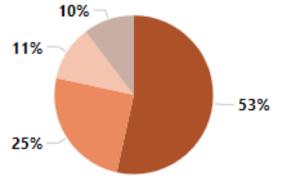


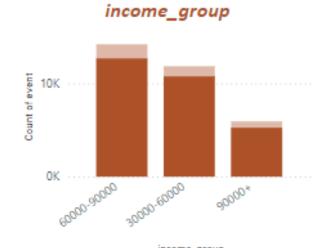
28.65K



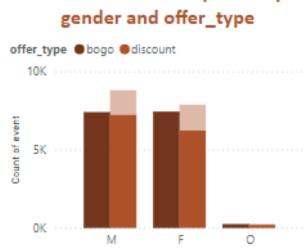


### Count of Offers Completed by channels

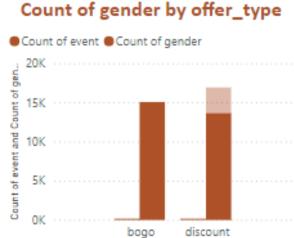




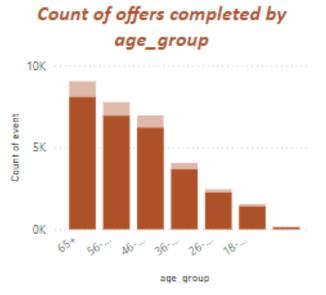
Count of Offers Completed by



Count of offers completed by

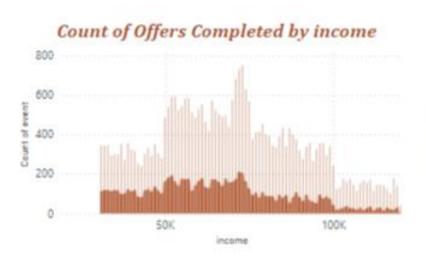


Count of offers completed and





Sum Of offers completed **8788** 



Sum of income by age and gender

gender M

20M

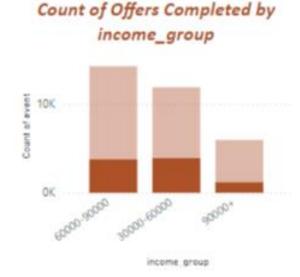
10M

0

50

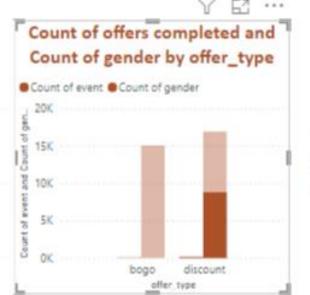
100

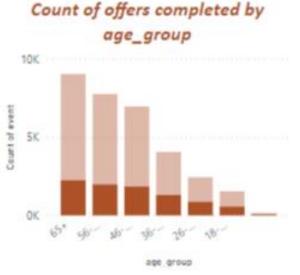






Count of offers completed by



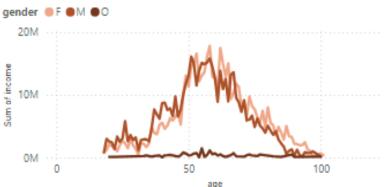




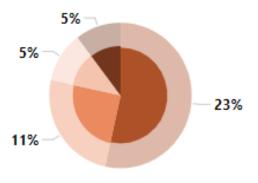
Sum Of offers completed

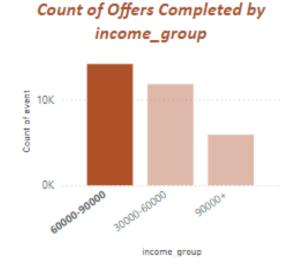


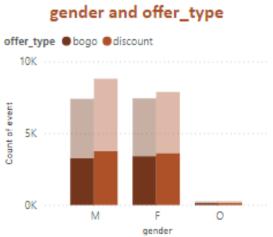
Sum of income by age and gender



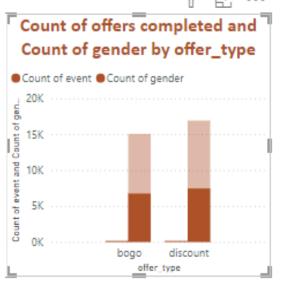
Count of Offers Completed by channels

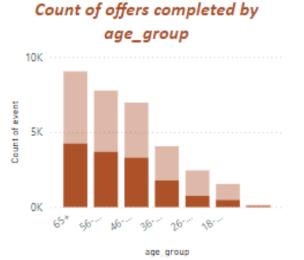






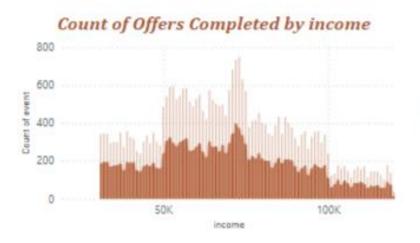
Count of offers completed by



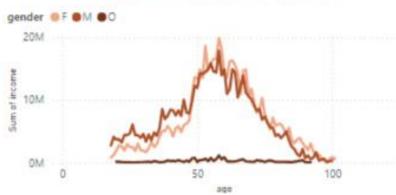




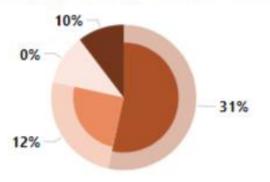
16.91K



Sum of income by age and gender



Count of Offers Completed by channels



income\_group

10K

income group

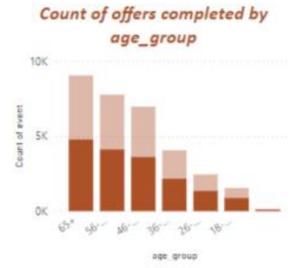
0K

Count of Offers Completed by



Count of offers completed by

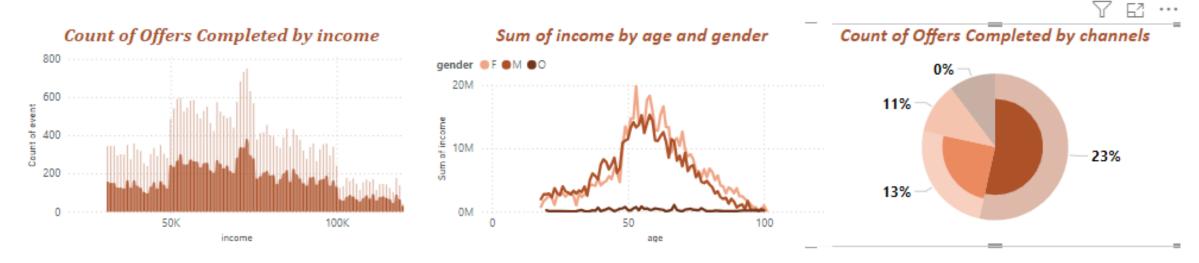


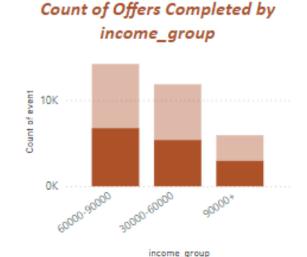


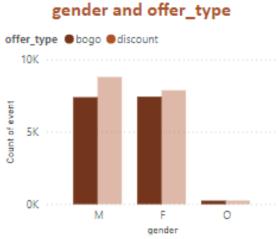


Sum Of offers completed

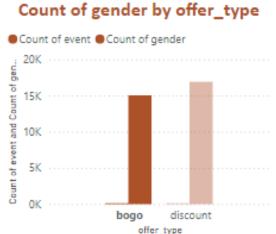
15.04K



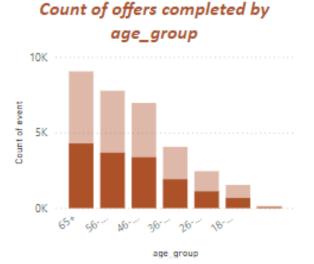




Count of offers completed by

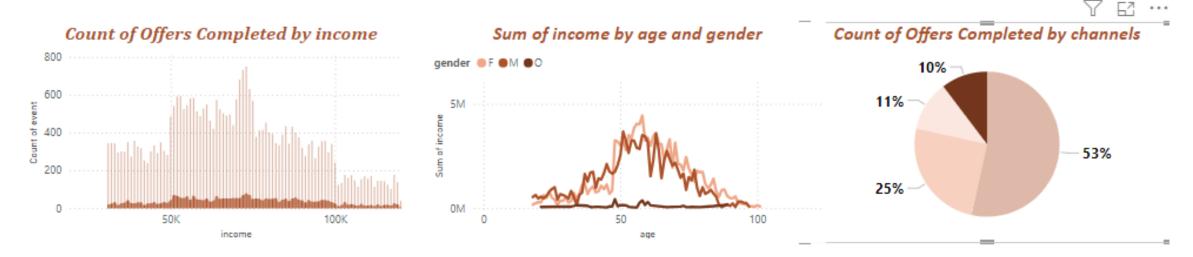


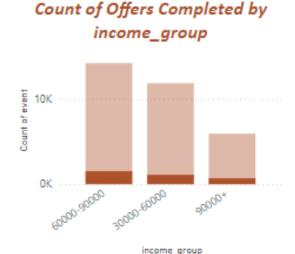
Count of offers completed and

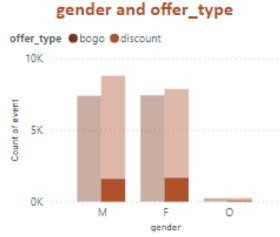




Sum Of offers completed



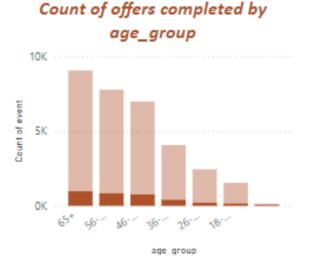




Count of offers completed by



Count of offers completed and

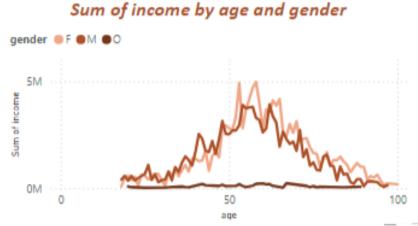


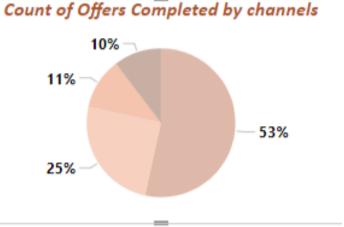


3596

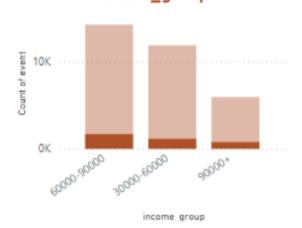




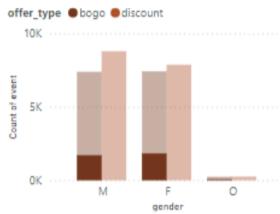




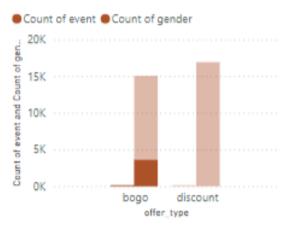
Count of Offers Completed by income\_group



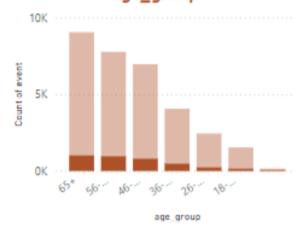
Count of offers completed by gender and offer\_type



Count of offers completed and Count of gender by offer\_type

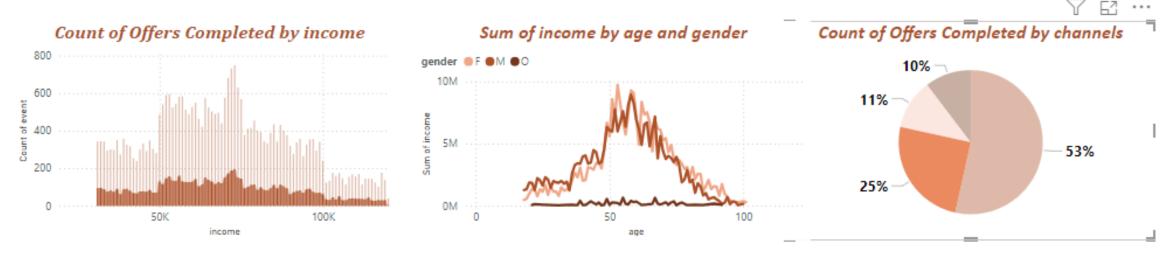


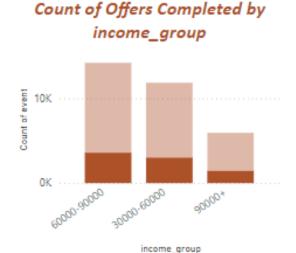
Count of offers completed by age\_group





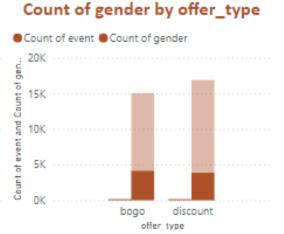
Sum Of offers completed **7071** 







Count of offers completed by



Count of offers completed and

