# **Data Analytics in Business and Industry**

**Group Project** 

Group 4

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#### **Project Background**

Market segmentation is a marketing technique which allows companies to understand customer needs and preferences and thus design strategies which can best satisfy these. Companies recognise that they can neither address all customers uniformly nor approach them uniquely. Customer segmentation sits between these two extremes. The approach facilitates consumer interaction by identifying groups with common characteristics or behaviours which can then inform marketing campaigns, pricing strategies and general consumer communication (Fonseca, 2011).

In this project we analyse customer interaction with a discount offer campaign and monitor its impact on their spending habits. Our aim is to identify groups of customers who are more or less reactive to the campaign, with the motivation that our segmentation will allow for more targeted, efficient campaigns in future.

### **Data Description**

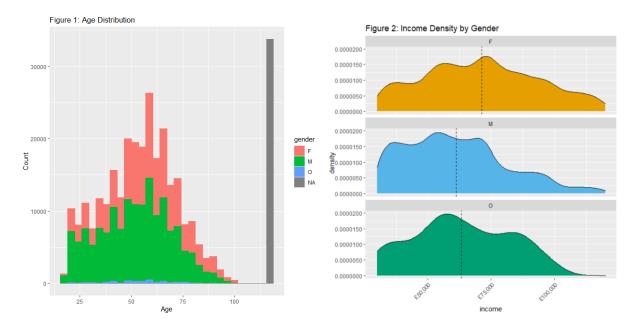
Three datasets were provided for the project:

- portfolio describes the offers available
- profile describes the customers
- transactions provides the customer transactions over the promotional period

We combined them into a single customer level set which contained a combination of customer profile information, such as age and income, with their aggregated transaction data, e.g. average transaction value, frequency of transactions. All categorical variables were converted into dummy variables. We used this combined dataset as a base for our exploratory data analysis. Appendix table 1 displays the fields in our dataset.

# **Exploratory Data Analysis**

Figures 1 and 2 display the customer income and age distributions. As is evident from figure 1,  $\sim$ 13% of the dataset are assigned no gender and an age of 118 – the missing data from these customers mean that they cannot be used within any customer segmentation based on these characteristics. The mean age of females is higher than the mean of males. Regarding income, the higher concentration of males is below £70,000, while females reach a peak of density at £75,000. There is a higher concentration of females with salaries greater than £75,000 compared to males and gender 'other'.



An initial analysis of the relationships between transaction value, transaction frequency, age and income in a simple, universe-wide linear context found several statistically significant relationships. Namely, the average transaction value increases with income and age, while the transaction frequency decreases with income and age. Controlling for income and age, the average transaction value increases with transaction frequency. These relationships can be viewed in table 1, which displays the coefficients, t-statistics and p-values from a simple linear regression, where ATV refers to the average transaction value and TF the transaction frequency:

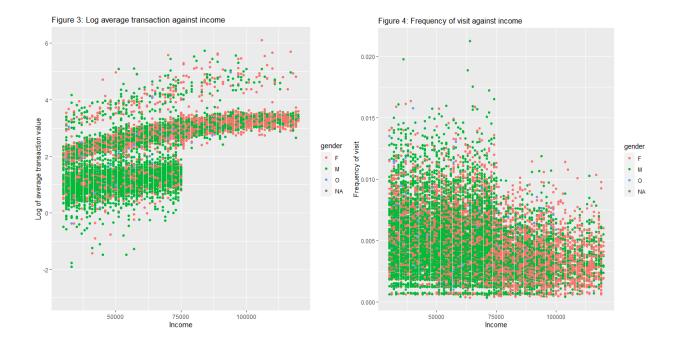
$$Log(ATV) = \alpha + \beta_1 Income + \beta_2 Age + \beta_3 Gender + \beta_4 TF + \varepsilon$$

**Table 1:** Log average transaction value regressed onto customer characteristics

Coefficients	Estimate	Std. Error	t-stat	Pr(> t )
(Intercept)	0.316	0.03	10.10	<2e-16***
TF	27.070	2.37	11.42	<2e-16***
Income	0.000	0.00	92.88	<2e-16***
Age	0.005	0.00	14.20	<2e-16***
Gender: Male	-0.294	0.01	-24.24	<2e-16***
Gender: Other	0.016	0.05	0.32	0.746

Significance codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Figure 3 plots the natural log of each customer's average transaction value against their income, displaying the aforementioned positive relationship between these variables. Figure 4 plots the frequency of each customer's average transaction value against their income, displaying a negative relationship between these variables.



Regarding the relationship between age, income and the percentage of offers completed by a customer, an initial inspection suggests that both characteristics are positively correlated with a propensity to take up discount offers. This relationship can be viewed in figures 5 and 6, which plot the offer completion ratio against age and income respectively.



#### **Next steps**

After the exploratory analysis we have a deeper understanding of the variables that might correlate with a customer's decisions whether to join the promotion. In the next phase of the project we will choose variables, potentially using feature selection, followed by the application of clustering algorithms to find natural groups of customers.

#### References

Fonseca J.R. (2011) Why does Segmentation Matter? Using Mixed Methodology to Identify Market Segments. In: Morschett D., Foscht T., Rudolph T., Schnedlitz P., Schramm-Klein H., Swoboda B. (eds) European Retail Research. European Retail Research. Gabler Verlag, Wiesbaden. https://doi.org/10.1007/978-3-8349-6235-5\_1

## **Appendix**

#### **Appendix table 1:** Dataset fields

Variable	Variable Type	Description	
customer_num	integer	Unique, more readable customer id	
gender	character	Original gender variable	
male	double	Binary - male = 1	
female	double	Binary - female = 1	
other_gender	double	Binary - other = 1	
age_range	character	Age range variable	
income	integer	Original income variable	
income_range	character	A range for the income variable	
month_of_membership	integer	Month when customer joined	
year_of_membership	double	Year when customer joined	
length_of_membership	double	Length or time since customer joined	
amount_transactions	integer	Number of transactions	
transaction_freq	double	Frequency of transactions since joining	
total_transaction_spend	double	Total transaction spend	
average_transaction_spend	double	Average transaction spend	
amount_offer_received	integer	Number of offers received	
amount_offer_viewed	integer	Number of offers viewed	
amount_offer_completed	integer	Number of offers completed	
offer_view_ratio	double	Ratio of offers viewed to recevied	
offer_complete_ratio	double	Ratio of offers completed to received	
total_completed_reward	integer	Total value of rewards from completed offers	
average_completed_reward	double	Average value of rewards from completed offers	
email_offer_view_rate	double	Email view rate	
email_offer_complete_rate	double	Email completion rate	
web_offer_view_rate	double	Web view rate	
web_offer_complete_rate	double	Web completion rate	

mobile_offer_view_rate	double	Mobile view rate	
mobile_offer_complete_rate	double	Mobile completion rate	
social offer_view_rate	double	Social view rate	
social_offer_complete_rate	double	Social complete rate	
bogo_received	double	Bogo offers received	
bogo_view_rate	double	Bogo view rate	
bogo_complete_rate	double	Bogo complete rate	
discount_received	double	Discount offers received	
discount_view_rate	double	Discount view rate	
discount_complete_rate	double	Discount complete rate	
info_received	double	Info offers received	
info_view_rate	double	Info view rate	
info_complete_rate	double	Info complete rate	
rec_bogo_diff10_rew10_dur07	double	Received Bogo_Diff10_Rew10 offers dur07	
view_rate_bogo_diff10_rew10_dur07	double	View rate	
comp_rate_bogo_diff10_rew10_dur07	double	Comp rate	
rec_bogo_diff10_rew10_dur05	double	Received bogo_diff10_rew10_dur07	
view_rate_bogo_diff10_rew10_dur05	double	View rate bogo_diff10_rew10_dur07	
comp_rate_bogo_diff10_rew10_dur05	double	Comp rate bogo_diff10_rew10_dur07	
rec_info_diff00_rew00_dur04	double	Received info_diff00_rew00_dur04	
view_rate_info_diff00_rew00_dur04	double	View rate info_diff00_rew00_dur04	
comp_rate_info_diff00_rew00_dur04	double	Comp rate info_diff00_rew00_dur04	
rec_bogo_diff05_rew05_dur07	double	Received bogo_diff05_rew05_dur07	
view_rate_bogo_diff05_rew05_dur07	double	View rate bogo_diff05_rew05_dur07	
comp_rate_bogo_diff05_rew05_dur07	double	Comp rate bogo_diff05_rew05_dur07	
rec_disc_diff20_rew05_dur10	double	Received disc_diff20_rew05_dur10	
view_rate_disc_diff20_rew05_dur10	double	View rate disc_diff20_rew05_dur10	
comp_rate_disc_diff20_rew05_dur10	double	Comp rate disc_diff20_rew05_dur10	
rec_disc_diff07_rew03_dur07	double	Received disc_diff07_rew03_dur07	
view_rate_disc_diff07_rew03_dur07	double	View rate disc_diff07_rew03_dur07	
comp_rate_disc_diff07_rew03_dur07	double	Comp rate disc_diff07_rew03_dur07	
rec_disc_diff10_rew02_dur10	double	Received disc_diff10_rew02_dur10	
view_rate_disc_diff10_rew02_dur10	double	View rate disc_diff10_rew02_dur10	
comp_rate_disc_diff10_rew02_dur10	double	Comp rate disc_diff10_rew02_dur10	
rec_info_diff00_rew00_dur03	double	Received info_diff00_rew00_dur03	
view_rate_info_diff00_rew00_dur03	double	View rate info_diff00_rew00_dur03	
comp_rate_info_diff00_rew00_dur03	double	Comp rate info_diff00_rew00_dur03	
rec_bogo_diff05_rew05_dur05	double	Received bogo_diff05_rew05_dur05	
view_rate_bogo_diff05_rew05_dur05	double	View rate bogo_diff05_rew05_dur05	
comp_rate_bogo_diff05_rew05_dur05	double	Comp rate bogo_diff05_rew05_dur05	
rec_disc_diff10_rew02_dur07	double	Received disc_diff10_rew02_dur07	
view_rate_disc_diff10_rew02_dur07	double	View rate disc_diff10_rew02_dur07	
comp_rate_disc_diff10_rew02_dur07	double	Comp rate disc_diff10_rew02_dur07	