

# Helen FitzGerald

Case Study - InstaCart







## **BACKGROUND**

### **OBJECTIVES**

#### CONTEXT

The grocery delivery app Instacart wants to dig deeper into their customer profiles and behaviour.

This exploratory analysis uncovers valuable customer insights and suggests strategies for better segmentation based on the provided criteria.

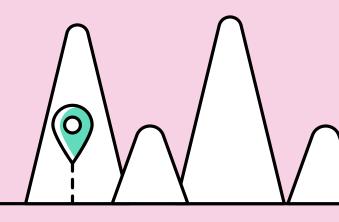
#### **PROJECT GOALS**

Instacart already has very good sales, but they want to uncover more information about their sales patterns.

In this project, I perform an initial data and exploratory analysis of some of their data in order to derive insights and suggest strategies for better segmentation based on the provided criteria.

There are a number of questions that Instacart would like to dive into...





## KEY BUSINESS QUESTIONS



- What are the busiest days of the week and hours of the day?
- Are there are particular times of the day when people spend the most money?
- Could price ranges be simplified?
- Are there certain types of products that are more popular than others?
- Which departments have the highest frequency of product orders?
- What are the different types of customers in their system and how their ordering behaviors differ?

## **DATA**

#### **DATA USED**

Customer data: provided by Instacart

Data dictionary: provided by Instacart

Python repository available here



### **TECHNIQUES**

Python

Data wrangling

Data merging

Deriving variables

Grouping data

Aggregating data

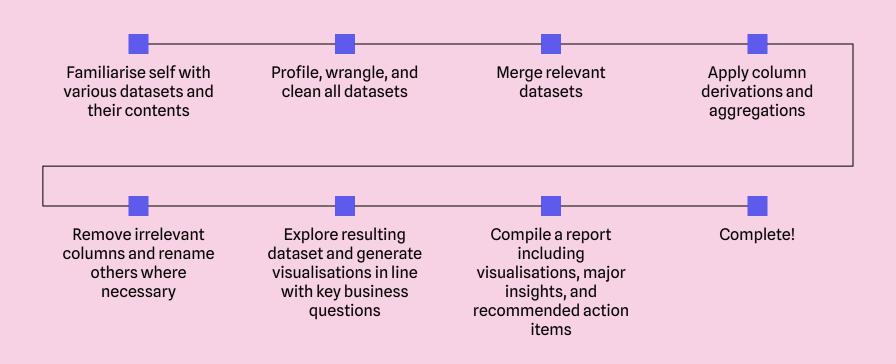
Reporting in Excel

Population flows



## **PROCESS**

### **PROJECT PROCESS**



### **CHALLENGES**

### **DATA SIZE**

The extensive size of the datasets involved in this project resulted, at times, in extremely slow processing speeds.

To address this, various methods of RAM optimisation were applied.

### **DATA LIMITS**

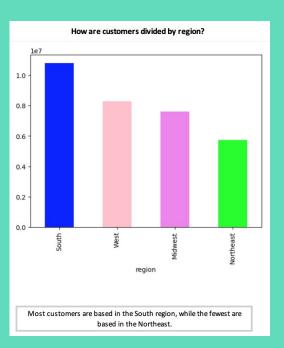
Although exhaustive in some respects, the datasets often did not provide key information, such as as the *quantities* of a particular item per order.

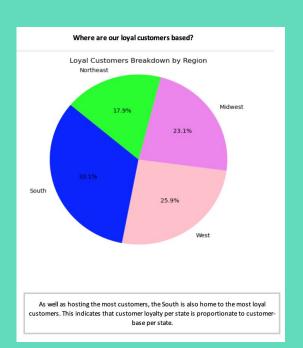


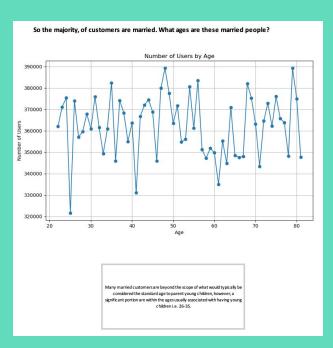


## **INSIGHTS**

## **FINDINGS**

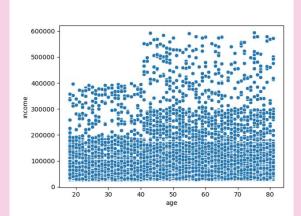




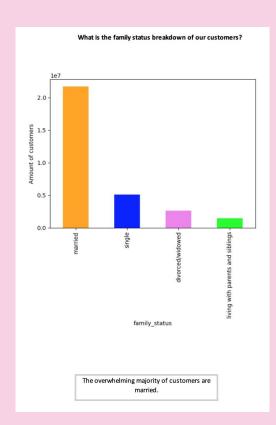


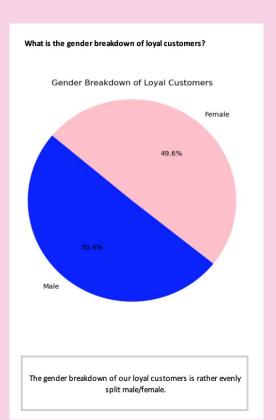
## **FINDINGS**

#### What's the relationship between customer age and income?



Answer: The highest density of users (across age ranges) sit beneath 200,000 in income. The next dear clustering occurs in the age range of 40-80, for whom their income is between 200,000-300,000. The income band of 300,000-400,000 is spread across all age ranges, though less densely. And finally, the income band of 400,000-600,000 is reserved for those aged between 40-80 in this sample.





## RECOMMENDATIONS

Top performers: 'Produce' and 'Dairy/Egg' Lowest performers: 'Bulk' and 'Other'.

Most customers are married and likely sharing the home, the **benefits of bulk-buying** versatile staples should be emphasised.

Regional differences **do not** impact customer behaviour -> employ a **wide marketing stance** without catering to specific geographic factors.

AD SCHEDULING

**DEPARTMENTS** 

INVENTORY

REGIONAL ADJUSTMENTS

**INCENTIVES** 

Quiet days: **Wednesday** and **Thursday** -> additional promotional push-notifications and time-sensitive offers.

Quiet hours: 1 am - 5 am -> niche, targeted advertising.

Pricing has been divided into **low, mid,** and **high** range.

Most products are **mid-range**. -> introduce more **high-range** products to appeal to different demographic.

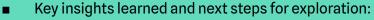
There is a large amount of 'New' customers -> incentivise them to return via a **loyalty system** (this also benefits 'Regular' customers)

Explore **subscription** model for frequently repurchased goods.



## **SUMMARY**

### RETROSPECTIVE



- Wednesday and Thursday are the quietest order days.
  - Introduce niche, targeted advertising during quiet days.
- o 'Produce' and 'Dairy/Egg' are the most popular departments, 'Bulk' is the least popular.
  - Integrate popular departments and unpopular i.e. allow bulk produce purchases.
- o Geographic location does not impact customer behaviours.
  - Do not focus on this for marketing strategy.
- Most products available are within the 'mid price' range.
  - Introduce more 'high price' items.
- There are a lot of new customers but no incentive for them to return.
  - Consider a subscription or loyalty system.

#### Challenges and suggested improvements:

- Data size
  - In future, I would adopt RAM saving techniques early in the exploration process to reduce wait times.
- Curiosity
  - A personal curiosity, at times, led to rabbit-hole explorations that were not explicitly in line with stakeholder questions. In future I would remind myself more frequently to stick to the brief.
- o Data limits
  - Only after formulating research questions did I realise that the data could not provide the necessary information required to answer these queries. In future I would more thoroughly familiarise myself with the available data before stating the insights I intend on providing.



