

# Chip Analysis

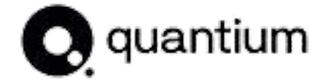
Understand customer purchasing behaviour on Chips Sales

Imam Ahmad Qusyairi - Retail Analytics Team

## **Virtual Internship Program**

About virtual internship program

- Currently i'm participating in Quantum Data Analytics
   Virtual Experience Program by <u>theForage.com</u>
- I've given task that designed to replicate life in the Retail Analytics and Strategy team at Quantium



## **Background**

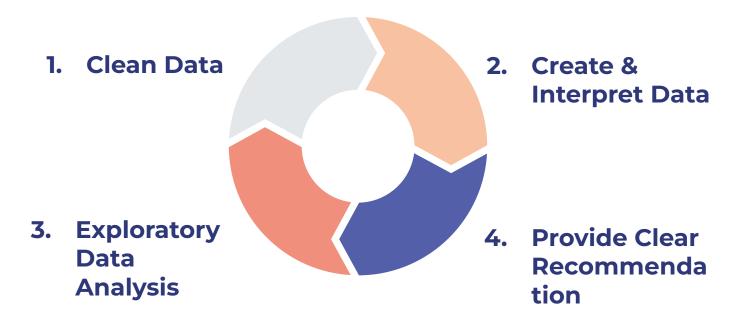
Story behind this project

- As Category Manager, Our client wants to better understand the types of customers who purchase Chips and their purchasing behaviour within the region.
- The insights from your analysis will feed into the supermarket's strategic plan for the chip category in the next half year.



## **OUR TASKS**

Job we need to get done!



## **Table of Contents**

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#### **Data Checks**

Check data quality from missing, duplicates & outliers

**Feature Engineering** 

Modify current & create new features

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**EDA** 

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Conclusion

Summary of this work & recommendation



## **Data Overview**

Things We Analyze

We have 2 dataset files used in this project:

#### **Purchasing Behaviour Data**



Customer Personal information including membership, lifestage, etc.

#### **Transaction Data**



Detailed record of transaction data including product name, quantity, etc.

## **Quality Check**

Make sure the data was ready

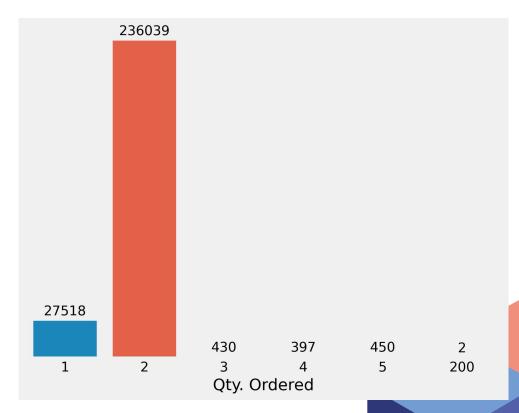
	Purchasing Data	Transaction Data
Missing Values	No	No
Duplicated Values	No	No
Outliers	No	2 Instances

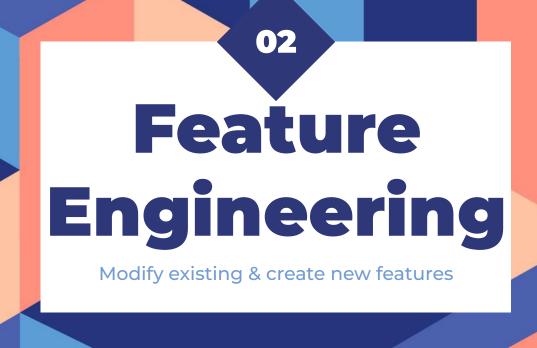
The data was clean except 2 outliers from Trans. data

## **The Outliers**

Verify & Clean the Instances

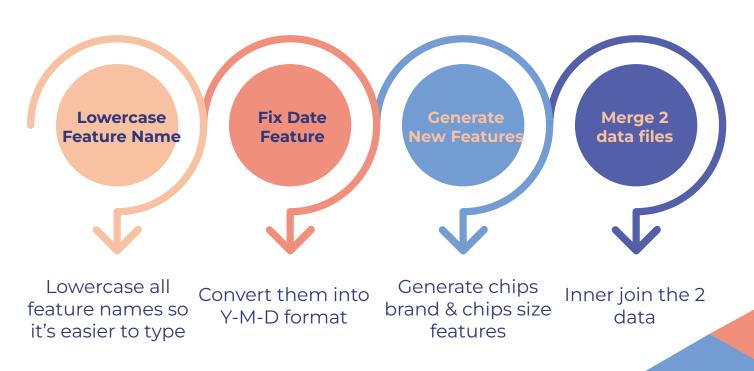
- There are 2 customer that order 200 chips!
- It's possible that they were distributor, not a customer
- We will **exclude** them from analysis





## **Feature Engineering**

What We Will Do:



#### **Feature Engineering**

## **Lowercase Feature Name**

Make feature names easier to typed

Before					After						
	DATE	STORE_NBR	LYLTY_CARD_NBR	TXN_ID	PROD_NBR		date	store_nbr	lylty_card_nbr	txn_id	prod_nbr
0	43390	1	1000	1	5	0	2018-10-17	1	1000	1	5
1	43599	1	1307	348	66	1	2019-05-14	1	1307	348	66
2	43605	1	1343	383	61	2	2019-05-20	1	1343	383	61
3	43329	2	2373	974	69	3	2018-08-17	2	2373	974	69
4	43330	2	2426	1038	108	4	2018-08-18	2	2426	1038	108

After lowercase, the feature names will be easier to typed

# Feature Engineering Fix Date Feature

Revise date into correct format date

	Before	After
0	43390	0 2018-10-17
1	43599	1 2019-05-14
2	43605	2 2019-05-20
3	43329	3 2018-08-17
4	43330	4 2018-08-18
5	43604	5 2019-05-19
6	43601	6 2019-05-16
7	43601	7 2019-05-16
8	43332	8 2018-08-20
9	43330	9 2018-08-18
Nam	ne: date, dtype: int64	Name: date, dtype: object

- Previously, date feature was integer value of day after 'date origins'
- We revise them into correct date by counting days from the integer value
- **Date Origins**: 30-12-1899

#### **Feature Engineering**

## **Generate New Feature: Brand**

Create New Feature from Existing Data

**Product Name** Feature Examples

Natural Chip Compny SeaSalt175g
CCs Nacho Cheese 175g
Smiths Crinkle Cut Chips Chicken 170g

- We found that for every first word from each instances of product name feature, indicate a brand name
- We will extract brand name form product name feature by splitting string values by whitespace and take the first splitting into brand name
- We also fix several mispelled brand name so it shows only one distinct brand

#### **Feature Engineering**

## **Merge 2 Files Data**

Unite 2 different files into 1 main dataset

Purchasing Transaction **Behaviour Data** Data **Main Sales Data** 

03

# Exploratory Data Analysis

Deep-dive analysis toward sales data

## **Sales Performance**

Descriptively Analyze Sales Performance Data

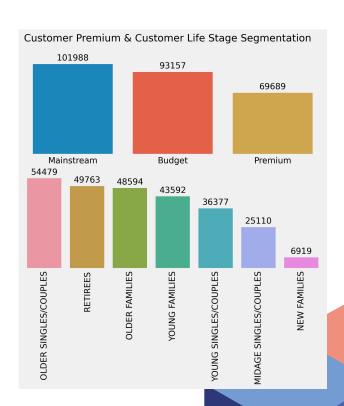
	Total	Mean	Median	Min	Max
Product (Qty)	504,724	1.905	2.0	1.0	5.0
Revenue (\$)	1,933,114	7.299	7.4	1.5	29.5

- Sales performance in 1 Year (July 2018 July 2019) reach total product sales
   504,724 units and total revenue \$1,933,114
- Product sales has mean values of 1.905. Meaning for every chips purchasing, customer buy at least 2 chips
- Revenue has mean values of \$7.299. Meaning for every chips purchase,
   \$7.3 dollars collected

## **Customer Segmentation**

Customer Group based on Membership & Life Stage

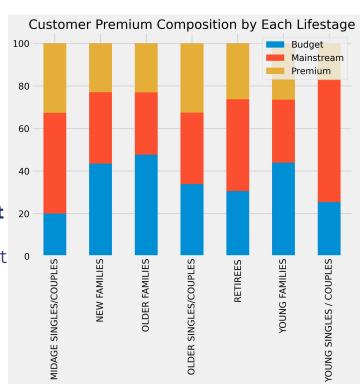
- Most customer who buy chips was not a member / regular customer.
   Where Customer who buy the premium only reach ~30% of total customer who buy chips
- Older singles/couples founded as a most people who buy chips. Young customer founded relatively low compared to elderly for buying chips



## **Membership Composition**

Membership Proportion from Customer Life Stage

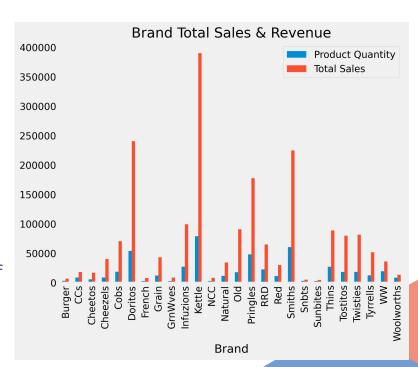
- Middle age customer tend to have higher membership propotion compared to younger and elder age customer
- Young singles / couples have lowest membership proportion. Should be the main concern if we want to target the younger customer market



## **Brand Sales Performance**

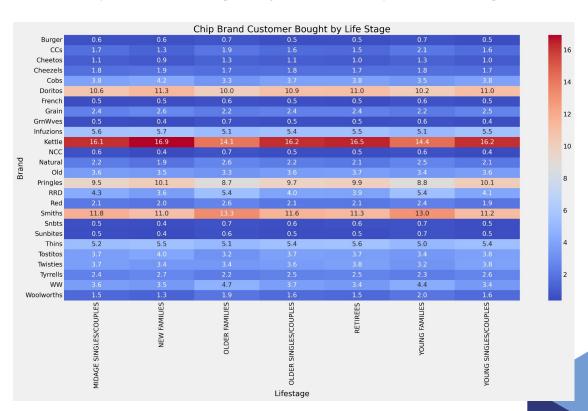
Quantity Product Sold and Revenue from Chip Brands

- We found **Kettle** as the most **success brand** in. They have highest quantity of product sold and highest revenue
- Doritos have higher revenue compared to Smiths although Doritos have smaller quantity of product saled to Smiths



## **Customer Preferred Chips Brands (i)**

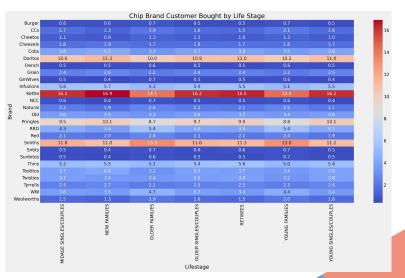
Chip Brand Bought by Customer per Life Stage



## **Customer Preferred Chips Brands (ii)**

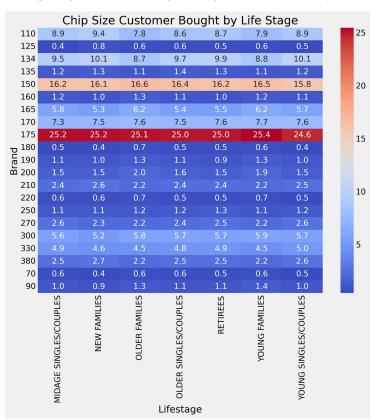
Chip Brand Bought by Customer per Life Stage

- Consistent with other customer life stage, Kettle became the most purchased brand compared
- We can assume that between different customer lifestage, they have relatively similar preferred brands



## **Customer Preferred Chips Size (i)**

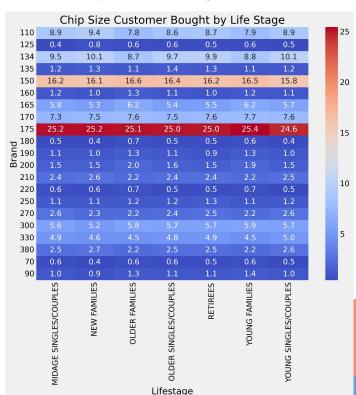
Chip Packaging Size Bought by Customer per Life Stage



## **Customer Preferred Chips Size (ii)**

Chip Packaging Size Bought by Customer per Life Stage

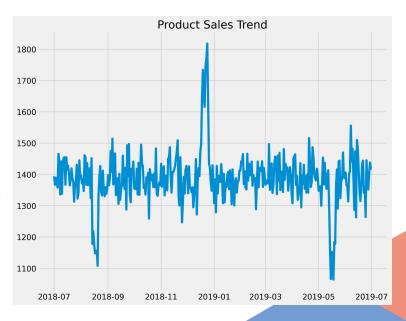
- Chips packed in 175 g founded most purchased chips pack size between different customer lifestage
- Similar with previous findings, there
  is relatively similar preferences for
  chips packaging size between
  different customer lifestage



## **Product Sales Trend**

Chip Product Sales Trend Accross the Year

- There is drastically increase of in December 2018 followed then in January 2018.
- Maybe caused by the increasing demand of chips for Christmas
- There is drastically decrease of quantity product saled in June
   2019. We need to inspect why this happen in the future work





# Conclusion & Recommendation Conclusion

What we found from this project

- Elderly customer have higher number of buying chips compared to younger customer
- Young customer has lower number of buying premium membership compared to elder customer
- Most sucess chips brands in terms of quantity chips saled and revenue collected was Kettle brand
- There is no significant difference of chip brands and chips package size between customer lifestage



#### **Conclusion & Recommendation**

## Recommendation

What We Could Improve Based on Data

- Set marketing srategy targetting on Younger customer. Since younger customer have huge potential and we should exploit them
- Perform Collaboration with Kettle brand to attract existing customer and new customer
- Perform Clustering analysis using unsupervised learning in the future



## **Thanks**

Have Any Questions? Contact me! imam.aqusairi@gmail.com

Click This Link To See the Code

### **Github Link**

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## Jupyter Notebook

Virtual Environment for Python



## **Google Slides**

Create & Modify
Presentation
Slides



## **Obsidian Notes**

Note Taking System