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"I was graduated from Bogor Agricultural University at 2018. I am working as finance and accounting in manufacturing company. I join Rakamin Data Science Bootcamp in January 2022, to enhance my skill in data field and hopefully to switch my career to data field."

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#### Overview



"In this project I analyzed business performance of an e-Commerce that involves reviewing customer growth, products quality, and payment methods using PostgreSQL, and the visualisation using Tableau and Ms. Excel.

Datasets that I used in this project was provided by Rakamin Academy, it was from multiple eCommerce in Brazil that covers around 100.000 datasets from 2016–2018. The datasets provides some metrics such: orders, payments, customers, sellers, locations, products, and reviews. These solutions on solving the problem heavily rely on the collection of e-commerce-related data."

# **Data Preparation**



Before processing the data, first step that should be done is preparing raw data and make it ready to process data. These are the steps on making a new database:

- 1. Creating the new database with all tables for the prepared data, by paying attention on data types for every columns in tables.
- 2. Importing the file into the database.
- 3. Creating the entity relationship diagram (ERD) by paying attention on relationship between every tables.

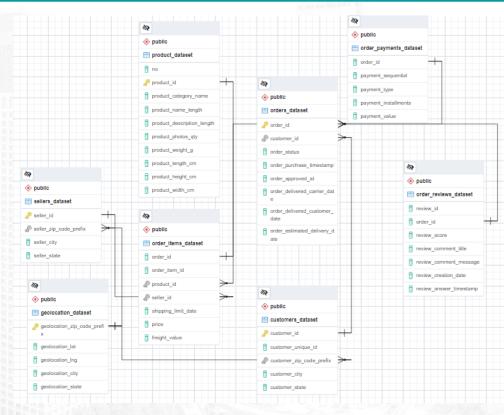


Figure 1. Entity Relationship Diagram (ERD)

# **Annual Customer Activity Growth Analysis**



There are some metrics the needed on analysis customer activity growth, such as monthly active users, new customers, repeat order, and average order frequency per customers (Table 1).

4	year double precision	average_mau numeric	new_customers bigint	repeating_customers bigint	avg_order_freq_per_cust numeric
1	2016	110	326	3	1.01
2	2017	3758	43708	1256	1.03
3	2018	5401	52062	1167	1.02

Table 1. Customer growth metrics

# **Annual Customer Activity Growth Analysis**



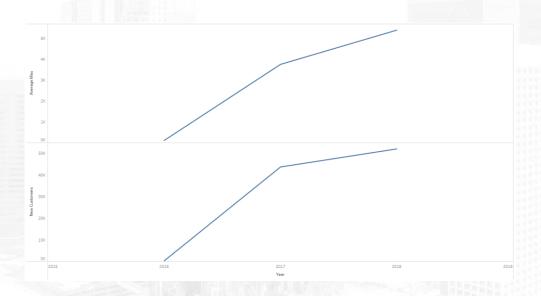


Figure 2. Monthly active users (MAU) vs new customers graphic in 2016-2018

- 1. For both metrics (MAU and new customers), was increasing from 2016-2018
- Both metrics significantly increase in 2017, it happened because of the data in 2016 only provided from September-December

## **Annual Customer Activity Growth Analysis**



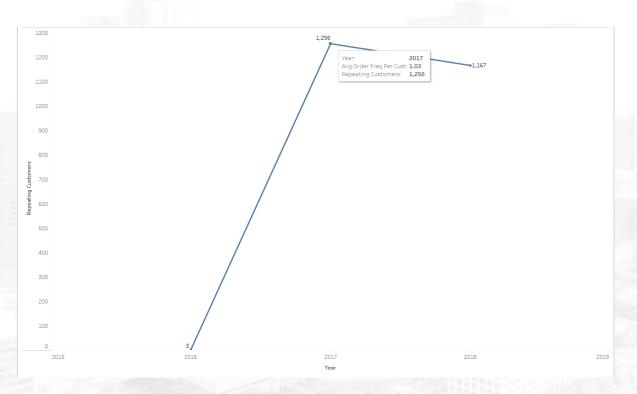


Figure 3. Repeat orders graphic in 2016-2018

- Repeat orders in 2017 has significantly increase because data in 2016 only from September December
- In 2018 repeat orders was decreasing around 7%, it would be happen because the data is only until September 2018



4	year double precision	category_most_revenue character varying	most_revenue double precision	total_revenue_per_year double precision	category_most_cancel character varying	num_most_cancel bigint	total_cancel bigint
1	2016	furniture_decor	6899.35	46653.74000000001	toys	3	26
2	2017	bed_bath_table	580949.2000000012	6921535.239999682	sports_leisure	25	265
3	2018	health_beauty	866810.3399999978	8451584.769999918	health_beauty	27	334

#### Table 2. Product category analysis

#### 2016 Recap

- Product category furniture\_decor got highest revenue than other product, which is 6,899.35
- Total revenue is 46,653.74
- This means, furniture\_decor contribute around 14.8% of total revenue in this year
- The most cancel category is toys with 3 cancel orders. Total cancel in this year is 26 orders
- This means toys contribute around 11.5% of total cancel orders in 2016



#### 2017 Recap

- Product category bed\_bath\_table got highest revenue than other product, which is 580,949.20
- Total revenue is 6,921,535.24
- This means, bed\_bath\_table contribute around 8.4% of total revenue in this year
- The most cancel category is sports\_leisure with 25 cancel orders. Total cancel in this year is 265 orders
- This means, sports\_leisure contribute around 9.4% of total cancel in 2017



#### 2018 Recap

- Product category health\_beauty got highest revenue than other product, which is 866,810.34
- Total revenue is 8,451,584.77
- This means, health\_beauty contribute around 10.25% of total revenue in this year
- The most cancel category in this year also health\_beauty with 27 from 334 total cancel in this year
- This means, health\_beauty contribute around 8% of total cancel in 2018



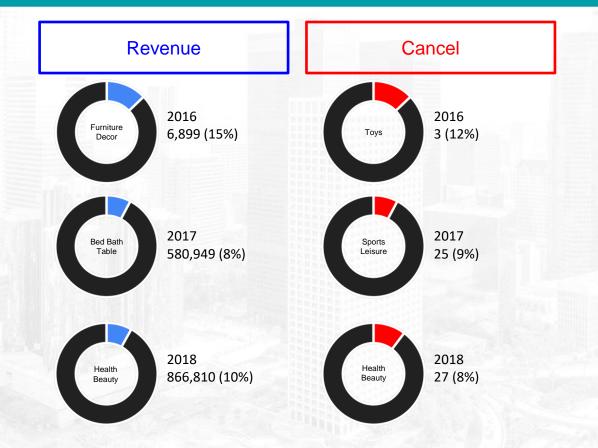


Figure 4. Product category performance 2016-2018

# **Analysis of Annual Payment Type Usage**



4	payment_type character varying	year_2016 numeric	year_2017 numeric	year_2018 numeric	growth_pct numeric
1	not_defined	0	0	3	[null]
2	debit_card	2	422	1105	1.62
3	credit_card	258	34568	41969	0.21
4	boleto	63	9508	10213	0.07
5	voucher	23	3027	2725	-0.1

Table 3. Annual payment type usage

### **Analysis of Annual Payment Type Usage**



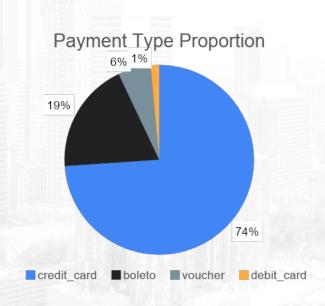


Figure 5. Payment type proportion

- Credit\_card was all time most favorite payment method
- There are 76795 payments using credit\_card (74%)
- Second was boleto with 19784 payments (19%)
- The least is debit\_card besides the not\_defined payment\_type (6%)
- Yearly, credit\_card still the most favorite based on total usage
- The order is still the same as all time

# **Analysis of Annual Payment Type Usage**



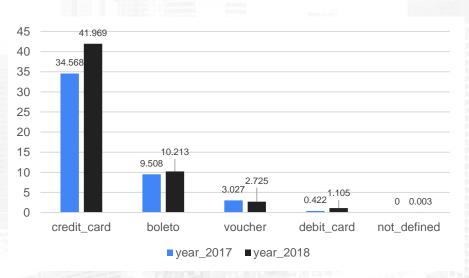


Figure 6. Payment type performance

- From 2017 to 2018, debit\_card payment\_type has the biggest growth than the others
- Voucher has a negative growth which means the usage in 2018 was decreasing from 2017