

# E-commerce Data Analysis Report

*This report summarizes the data analysis performed on an e-commerce dataset, covering basic, intermediate, and advanced business questions. The analysis uses MySQL for data querying and manipulation, and Python (Pandas, Matplotlib, Seaborn) for data presentation and visualization.*

## Basic Queries :

### Customer Geographic Distribution

The customer base shows a high degree of geographic diversity and a strong concentration in a single state.

- The total number of **unique cities** where customers are located is **4,119**.
- When grouping customers by state, the state with the highest number of customers is **SP (São Paulo)**, with **41,746** customers.

### Order Volume and Trends

Order volume in 2017 provides a measure of annual activity.

- The total number of orders placed in the year **2017** was **45,101**.

### Sales and Payment Analysis

#### Sales by Product Category

Total sales revenue was aggregated and calculated for all 74 product categories.

- The category generating the **highest total sales** is **BED TABLE BATH**, with revenue of **\$1,712,553.67**.
- Other high-revenue categories include **FURNITURE DECORATION** (\$1,430,176.39) and **AUTOMOTIVE** (\$852,294.33).
- Categories with the lowest sales include **FASHION CHILDREN'S CLOTHING** (\$785.67) and **INSURANCE AND SERVICES** (\$324.51).

### Payment Method

The preference for instalment payments is nearly universal in the dataset.

- The percentage of orders that were paid using one or more instalments (`payment_installments >= 1`) is **99.9981%**. This suggests that instalment payment options are the default or heavily preferred method for transactions.

## Intermediate Queries :

### Sales and Product Metrics

#### 1. Orders Per Month in 2018

The total number of orders placed was calculated for each month in 2018.

- The month with the **highest order count** was **August**, with **6,608** orders.
- The data covers orders placed from **January** through **October**.

- Order counts generally trended upward through the summer months, a trend easily observable in the generated bar chart.

## 2. Average Products Per Order by City

The analysis determined the average number of products contained in each order, grouped by the customer's city.

- This was achieved by first creating a **Common Table Expression (CTE)** to count the items in every individual order, and then calculating the average of these counts for each city.
- For the city of **Sao Paulo**, the average was **1.16 products per order**.
- Some smaller cities, like **buriti**, showed a much higher average of **3.00 products per order**.

## Revenue and Correlation Analysis

### 3. Percentage of Total Revenue by Product Category

The contribution of each product category to the total overall revenue was calculated.

- The top-contributing categories are:
  - **BED TABLE BATH** with **10.70%**.
  - **HEALTH BEAUTY** with **10.35%**.
  - **COMPUTER ACCESSORIES** with **9.90%**.
- The least contributing categories, at **0.00%**, are **FASHION CHILDREN'S CLOTHING** and **INSURANCE AND SERVICES**.

### 4. Correlation: Price vs. Purchase Frequency

A correlation analysis was performed to identify the relationship between the average price of a product category and the total number of times products in that category were purchased.

- The calculated **correlation coefficient is -0.106** (rounded).
- This indicates a **weak negative correlation**, suggesting that as the average price for a product category increases, the number of times it's purchased slightly decreases.

### 5. Total Revenue and Rank by Seller

The total revenue generated by each seller was calculated, and sellers were then ranked by this metric.

- The analysis used the **DENSE\_RANK()** window function on the total revenue grouped by seller\_id to determine the ranking.
- The **Top-Ranked Seller** (Rank 1) generated **\$227,092.36** in total revenue.
- The total revenue for the top 5 sellers is presented in the bar chart, highlighting the concentration of sales among the leading sellers.

## Advance Queries :

### Time-Series and Trend Analysis

#### 1. Moving Average of Order Values

The **moving average** of order payment values was calculated for each customer over their order history.

- The analysis used a **window function** (**AVG()** **OVER(...)**) that partitioned the data by customer\_id and ordered it by order\_purchase\_timestamp.
- The window size was set to the **current row and the 2 preceding rows** (a 3-order moving average).
- This metric is used to **smooth out short-term fluctuations** in individual customer spending to identify underlying trends in their monetary value over time. The output shows the calculated moving average for each sequential order by a customer.

## 2. Cumulative Sales Per Month

**Cumulative sales** were calculated for each month across all years in the dataset.

- This was achieved using the **SUM() OVER(ORDER BY year, month)** window function.
- The running total sales show continuous growth, reaching a cumulative total of **\$16,008,872.12** by October 2018.
- For example, total cumulative sales grew from **\$7,309,109.07** at the end of **2017** to **\$8,424,113.25** by **January 2018**.

## 3. Year-over-Year (YoY) Growth Rate of Total Sales

The YoY percentage growth rate of total sales was calculated to assess the business's annual revenue performance.

- The calculation used the **LAG()** window function to retrieve the sales from the previous year for the growth rate formula.
- **2017 Growth:** Sales in 2017 showed a substantial growth of **12,112.70%** over the preceding partial year (2016). This large figure is likely due to the limited number of orders in 2016.
- **2018 Growth:** Sales in 2018 grew by **20.00%** compared to 2017.

## Customer Performance Metrics

### 4. Customer Retention Rate

The **customer retention rate** was defined and calculated as the percentage of customers who make **another purchase within 6 months** of their first purchase.

- The query used **Common Table Expressions (CTEs)** and date arithmetic (**DATE\_ADD(first\_order, INTERVAL 6 MONTH)**) to find customers who placed a second order within the defined 6-month window.
- The query result was **[(None,)]**. This indicates that the specific retention calculation, based on the customer base in the available data, yielded a null or zero result, suggesting **very few or no customers** met the criteria of making a subsequent purchase within 6 months of their initial order.

### 5. Top 3 Customers by Annual Spending

The **top 3 customers** who spent the most money (total payment value) in **each year** were identified and ranked.

- The analysis employed the **DENSE\_RANK()** window function, partitioned by year and ordered by total payment descending, to assign ranks.
- The visualization highlights the spending of these top customers across the years.

# Summarized E-commerce Data Analysis & Business Intelligence Report

## Core Business Profile & Distribution (Basic Findings)

The initial analysis established the basic parameters of the customer base and sales volume:

- **Geographic Diversity:** The customer base is highly diverse, spanning **4,119 unique cities**.
- **Customer Concentration:** Despite the wide reach, customers are heavily concentrated in the state of **SP (São Paulo)**, which accounts for **41,746** customers.
- **Order Volume:** A total of **45,101 orders** were placed in **2017**.
- **Top Sales Category:** The **BED TABLE BATH** category leads in total sales revenue, generating **\$1,712,553.67**.
- **Payment Method:** Instalment payments are the overwhelming preference, used in nearly **99.9981%** of all orders.

## Intermediate Performance Metrics

These findings provide granular detail on monthly performance, product analysis, and seller ranking:

- **Monthly Order Trends (2018):** In 2018, the month with the **highest order count** was **August**, with **6,608** orders.

- **Product Consumption:** The average number of products per order is relatively low, though some smaller cities like **buriti** show higher averages (**3.00**).
- **Revenue Contribution:** The top three categories contributing to total revenue are **BED TABLE BATH (10.70%)**, **HEALTH BEAUTY (10.35%)**, and **COMPUTER ACCESSORIES (9.90%)**.
- **Price-Purchase Correlation:** There is a **weak negative correlation** ( $\approx -0.106$ ) between a product category's average price and its purchase frequency.
- **Top Seller:** The top-ranked seller generated **\$227,092.36** in revenue.

### Advanced Business Intelligence

The advanced analysis revealed critical growth and customer lifetime value metrics:

- **Cumulative Sales:** Cumulative sales show continuous growth, reaching a total of over **\$16 million** by October 2018.
- **Year-over-Year (YoY) Growth:** The YoY sales growth rate from **2017 to 2018 was 20.00%**.
- **Moving Average:** A **3-order moving average** of customer spending was calculated to identify stabilized trends in individual customer monetary value, smoothing out transactional volatility.
- **Top Spenders:** The **top 3 customers** who spent the most money in each year were identified using a dense ranking function.
- **Customer Retention:** The retention rate, defined as the percentage of customers who make a second purchase within 6 months of their first order, was found to be effectively **zero** ( $[(None,)]$ ). This suggests a significant opportunity for strategies focused on increasing repeat purchases and customer loyalty.