

Olist Store Analysis

Presented By Group 2



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Data Set

- The Olist Store Analysis project focuses on understanding customer buying behaviors and payment statistics on the Olist e-commerce platform.
- It analyzes key performance indicators(KPI's) like weekday vs weekend sales, payment statistics, delivery times, and customer behavior.
- The analysis is based on nine CSV files, which are cleaned and manipulated to extract valuable insights.

Name	Date modified	Type	Size
 Joins	07-03-2024 14:06	PNG File	112 KB
 olist_customers_dataset	07-03-2024 14:06	Microsoft Excel Com...	8,459 KB
 olist_geolocation_dataset	07-03-2024 14:06	Microsoft Excel Com...	59,838 KB
 olist_order_items_dataset	07-03-2024 14:06	Microsoft Excel Com...	15,077 KB
 olist_order_payments_dataset	07-03-2024 14:06	Microsoft Excel Com...	5,642 KB
 olist_order_reviews_dataset	07-03-2024 14:06	Microsoft Excel Com...	14,113 KB
 olist_orders_dataset	07-03-2024 14:06	Microsoft Excel Com...	17,242 KB
 olist_products_dataset	07-03-2024 14:07	Microsoft Excel Com...	2,324 KB
 olist_sellers_dataset	07-03-2024 14:07	Microsoft Excel Com...	171 KB
 product_category_name_translation	07-03-2024 14:07	Microsoft Excel Com...	3 KB
 Project Bootcamp - E Commerce	26-05-2024 10:59	Microsoft PowerPoint...	40 KB

KPI-1

Weekday vs weekend payment statistics

*Weekday Vs Weekend (order_purchase_timestamp)
Payment Statistics*

- Helps to understand the customer buying behaviour.
- Reveals which days have the highest sales.
- Allows Olist to enhance weekend sales and plan targeted promotions.

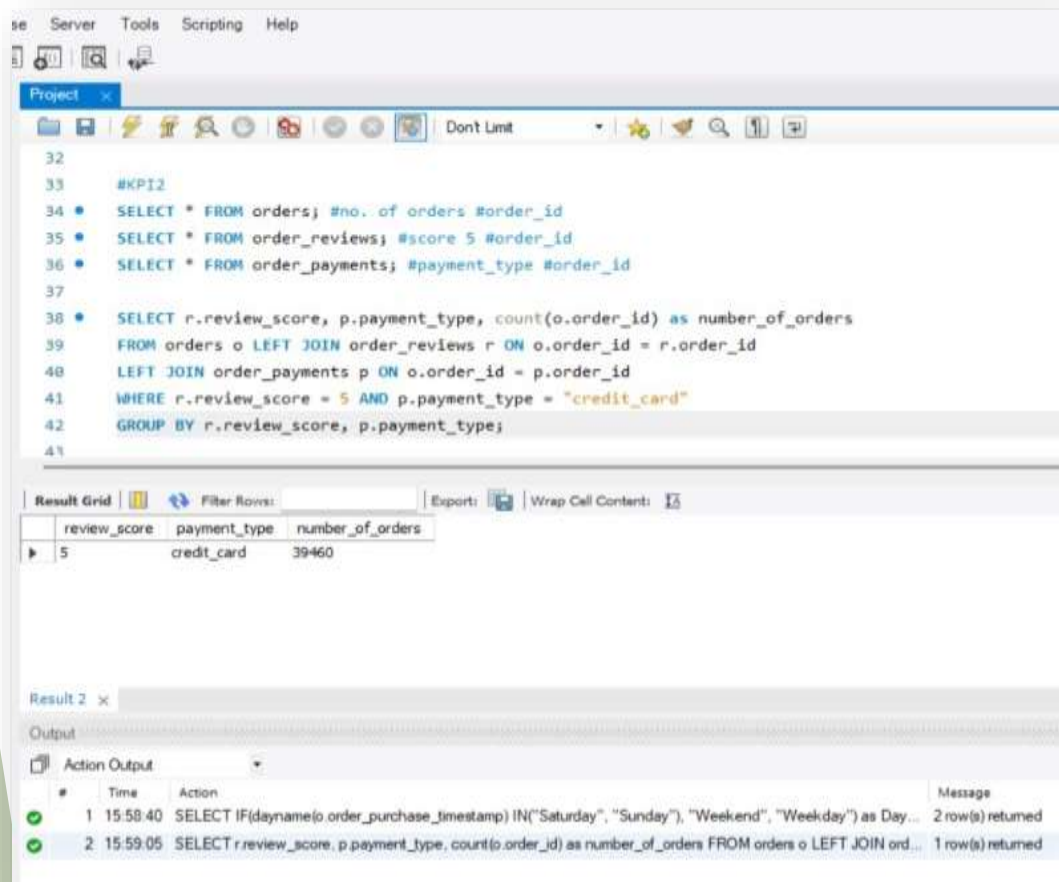
```
23 #KPI1
24 • SELECT * FROM orders;
25 • SELECT * FROM order_payments;
26
27 • SELECT Day_type, total_payments,
28   sum(total_payments) Over() as total,
29   concat(round((total_payments*100/sum(total_payments) OVER())), "%") as payment_percentage
30 FROM
31   (SELECT IF(dayname(o.order_purchase_timestamp) IN("Saturday", "Sunday"), "Weekend", "Weekday") as Day_type,
32    round(sum(payment_value)) as total_payments
33   FROM orders o LEFT JOIN order_payments od ON o.order_id = od.order_id
34   GROUP BY Day_type) as a;
```

Result Grid | Filter Rows: | Export: | Wrap Cell Content: |

Day_type	total_payments	total	payment_percentage
Weekend	3242864	14255235	23%
Weekday	11012371	14255235	77%

KPI-2

Payment type with review score 5



The screenshot shows a SQL IDE with a query editor and a results grid. The query is as follows:

```

32
33 #KPI2
34 SELECT * FROM orders; #no. of orders #order_id
35 SELECT * FROM order_reviews; #score 5 #order_id
36 SELECT * FROM order_payments; #payment_type #order_id
37
38 SELECT r.review_score, p.payment_type, count(o.order_id) as number_of_orders
39 FROM orders o LEFT JOIN order_reviews r ON o.order_id = r.order_id
40 LEFT JOIN order_payments p ON o.order_id = p.order_id
41 WHERE r.review_score = 5 AND p.payment_type = "credit_card"
42 GROUP BY r.review_score, p.payment_type;
43

```

The results grid shows the following data:

review_score	payment_type	number_of_orders
5	credit_card	39460

The bottom section of the IDE shows the 'Action Output' with two rows of execution logs:

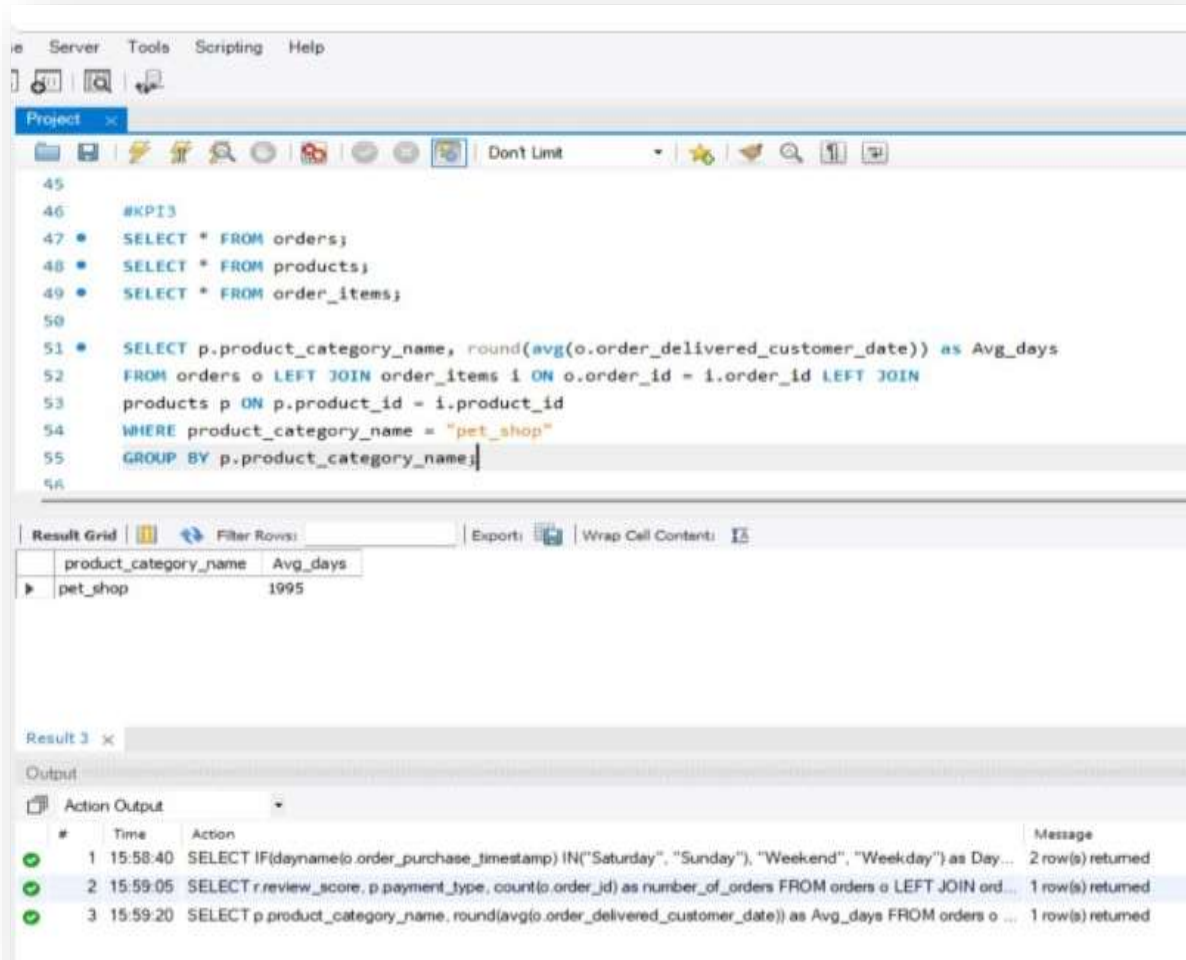
#	Time	Action	Message
1	15:58:40	SELECT IF(dayname(o.order_purchase_timestamp) IN("Saturday", "Sunday"), "Weekend", "Weekday") as Day...	2 row(s) returned
2	15:59:05	SELECT r.review_score, p.payment_type, count(o.order_id) as number_of_orders FROM orders o LEFT JOIN ord...	1 row(s) returned

Number of Orders with Review score as 5 And payment type as Credit card

- Helps to understand customer satisfaction and payment preferences
- Allows the store to target satisfied customers for repeat purchases.

KPI-3

Average number of delivery days taken for pet shop



The screenshot shows a SQL IDE interface with a query editor and a results grid. The query is as follows:

```

45
46 #KPI3
47 SELECT * FROM orders;
48 SELECT * FROM products;
49 SELECT * FROM order_items;
50
51 SELECT p.product_category_name, round(avg(o.order_delivered_customer_date)) as Avg_days
52 FROM orders o LEFT JOIN order_items i ON o.order_id = i.order_id LEFT JOIN
53 products p ON p.product_id = i.product_id
54 WHERE product_category_name = "pet_shop"
55 GROUP BY p.product_category_name;
56

```

The results grid shows the following data:

product_category_name	Avg_days
pet_shop	1995

The output pane shows the execution log with three actions:

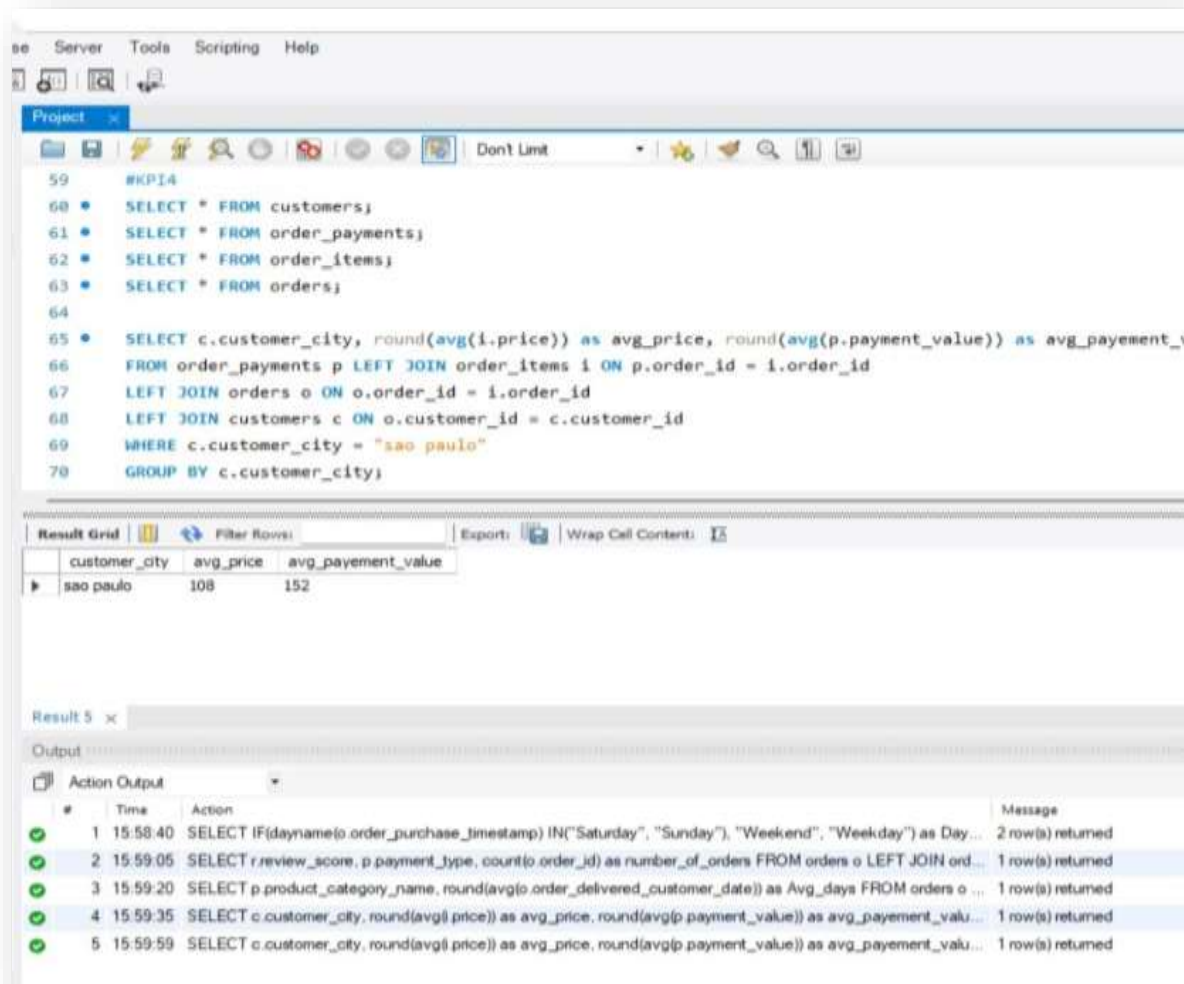
#	Time	Action	Message
1	15:58:40	SELECT IF(dayname(o.order_purchase_timestamp) IN("Saturday", "Sunday"), "Weekend", "Weekday") as Day...	2 row(s) returned
2	15:59:05	SELECT r.review_score, p.payment_type, count(o.order_id) as number_of_orders FROM orders o LEFT JOIN ord...	1 row(s) returned
3	15:59:20	SELECT p.product_category_name, round(avg(o.order_delivered_customer_date)) as Avg_days FROM orders o ...	1 row(s) returned

Average number of days taken for order_delivered_customer_date for pet_shop

- This KPI measures the average delivery time for pet shop orders.
- It helps Olist identify opportunities to improve delivery speed and maintain customer satisfaction.

KPI-4

Average price and payment value of Sao Paulo city.



The screenshot shows a SQL IDE with a query window and a results grid. The query is as follows:

```

59 #KPI4
60 SELECT * FROM customers;
61 SELECT * FROM order_payments;
62 SELECT * FROM order_items;
63 SELECT * FROM orders;
64
65 SELECT c.customer_city, round(avg(i.price)) as avg_price, round(avg(p.payment_value)) as avg_payment_value
66 FROM order_payments p LEFT JOIN order_items i ON p.order_id = i.order_id
67 LEFT JOIN orders o ON o.order_id = i.order_id
68 LEFT JOIN customers c ON o.customer_id = c.customer_id
69 WHERE c.customer_city = "sao paulo"
70 GROUP BY c.customer_city;
  
```

The results grid shows the following data:

customer_city	avg_price	avg_payment_value
sao paulo	108	152

The output window shows the following actions and messages:

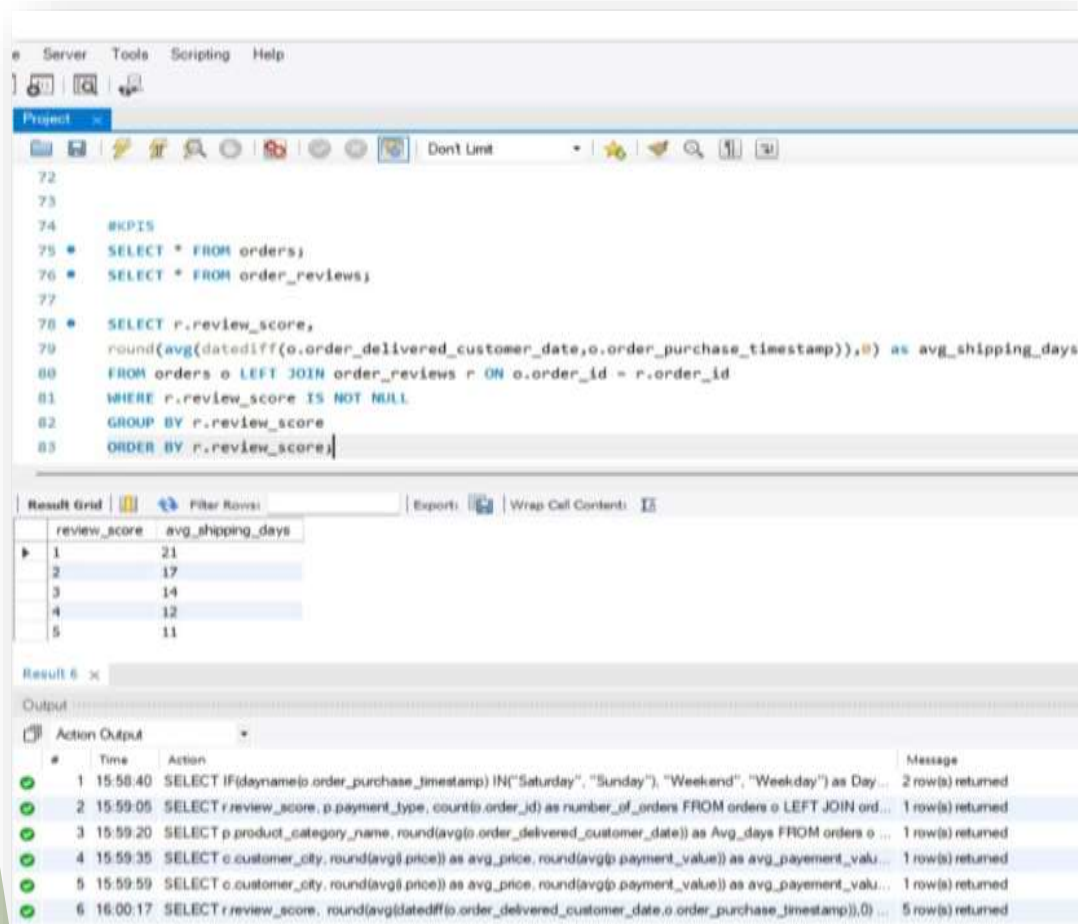
#	Time	Action	Message
1	15:58:40	SELECT IF(dayname(o.order_purchase_timestamp) IN("Saturday", "Sunday"), "Weekend", "Weekday") as Day...	2 row(s) returned
2	15:59:05	SELECT r.review_score, p.payment_type, count(o.order_id) as number_of_orders FROM orders o LEFT JOIN ord...	1 row(s) returned
3	15:59:20	SELECT p.product_category_name, round(avg(o.order_delivered_customer_date)) as Avg_days FROM orders o ...	1 row(s) returned
4	15:59:35	SELECT c.customer_city, round(avg(i.price)) as avg_price, round(avg(p.payment_value)) as avg_payment_valu...	1 row(s) returned
5	15:59:59	SELECT c.customer_city, round(avg(i.price)) as avg_price, round(avg(p.payment_value)) as avg_payment_valu...	1 row(s) returned

Average price and payment values from customers of sao paulo city

- Reveals their spending habits and purchasing power of customers.
- Helps the Olist to identify segments of high-value customers within this demographic.
- Olist can tailor marketing campaigns to effectively target these customers, to increase sales and customer loyalty.

KPI-5

Average shipping days vs review scores.



The screenshot shows a SQL IDE with a query window and a results grid. The query is as follows:

```

72
73
74 #KPI5
75 * SELECT * FROM orders;
76 * SELECT * FROM order_reviews;
77
78 * SELECT r.review_score,
79 round(avg(datediff(o.order_delivered_customer_date,o.order_purchase_timestamp)),0) as avg_shipping_days
80 FROM orders o LEFT JOIN order_reviews r ON o.order_id = r.order_id
81 WHERE r.review_score IS NOT NULL
82 GROUP BY r.review_score
83 ORDER BY r.review_score;

```

The results grid shows the following data:

review_score	avg_shipping_days
1	21
2	17
3	14
4	12
5	11

Below the results grid, there is an 'Output' section showing a log of actions and their messages:

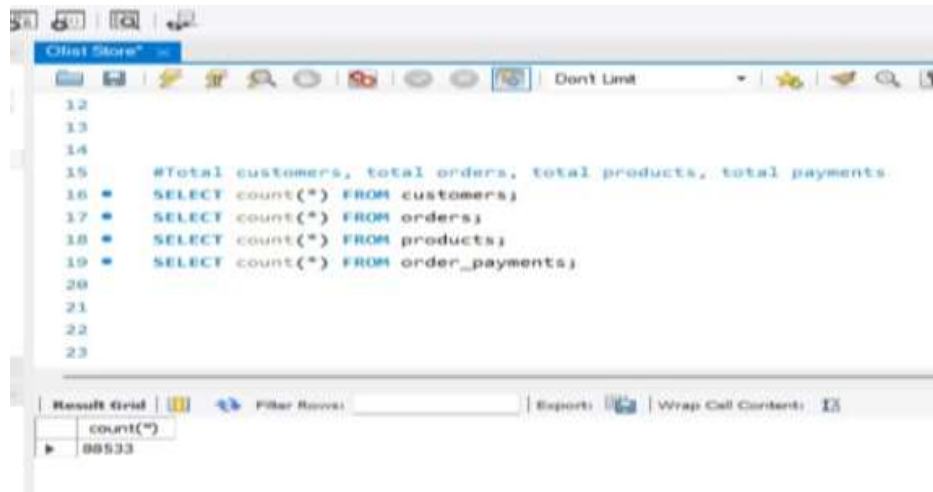
#	Time	Action	Message
1	15:58:40	SELECT IF(dayname(o.order_purchase_timestamp) IN("Saturday", "Sunday", "Weekend", "Weekday")) as Day...	2 row(s) returned
2	15:59:05	SELECT r.review_score, p.payment_type, count(o.order_id) as number_of_orders FROM orders o LEFT JOIN ord...	1 row(s) returned
3	15:59:20	SELECT p.product_category_name, round(avg(o.order_delivered_customer_date)) as Avg_days FROM orders o ...	1 row(s) returned
4	15:59:35	SELECT c.customer_city, round(avg(p.price)) as avg_price, round(avg(p.payment_value)) as avg_payment_valu...	1 row(s) returned
5	15:59:59	SELECT c.customer_city, round(avg(p.price)) as avg_price, round(avg(p.payment_value)) as avg_payment_valu...	1 row(s) returned
6	16:00:17	SELECT r.review_score, round(avg(datediff(o.order_delivered_customer_date,o.order_purchase_timestamp)),0) ...	5 row(s) returned

Relationship between shipping days (order_delivered_customer_date_order_purchase_timestamp) Vs review scores.

- This KPI measures how shipping days, time taken between order placement and delivery, correlate with review scores.
- This helps how delivery speed influences customer satisfaction levels.

Additional Insights

Total Orders



SQL query results for Total Orders:

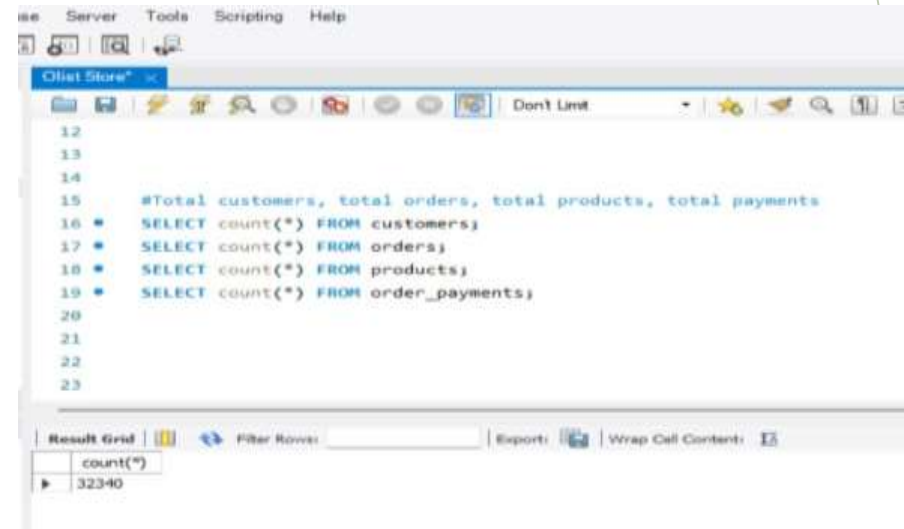
```

12
13
14
15 #Total customers, total orders, total products, total payments
16 * SELECT count(*) FROM customers;
17 * SELECT count(*) FROM orders;
18 * SELECT count(*) FROM products;
19 * SELECT count(*) FROM order_payments;
20
21
22
23

```

count(*)
88533

Total Products



SQL query results for Total Products:

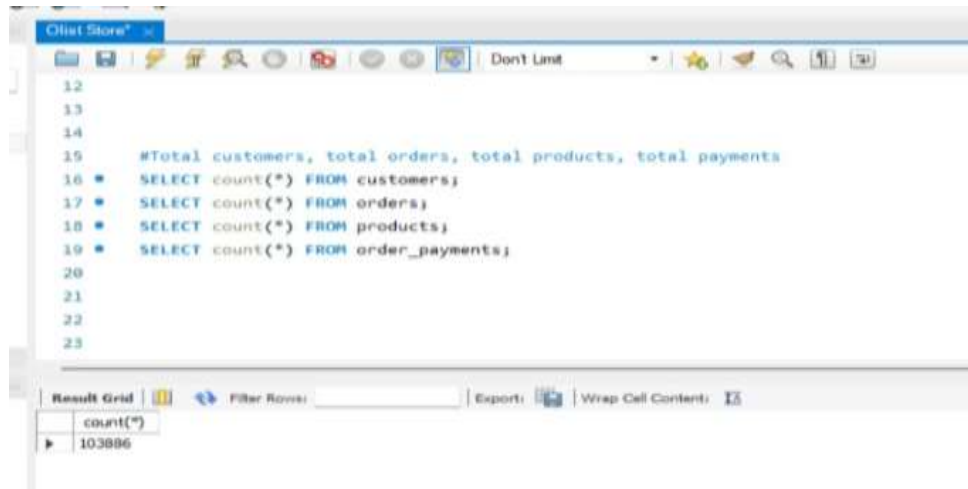
```

12
13
14
15 #Total customers, total orders, total products, total payments
16 * SELECT count(*) FROM customers;
17 * SELECT count(*) FROM orders;
18 * SELECT count(*) FROM products;
19 * SELECT count(*) FROM order_payments;
20
21
22
23

```

count(*)
32340

Total Payments



SQL query results for Total Payments:

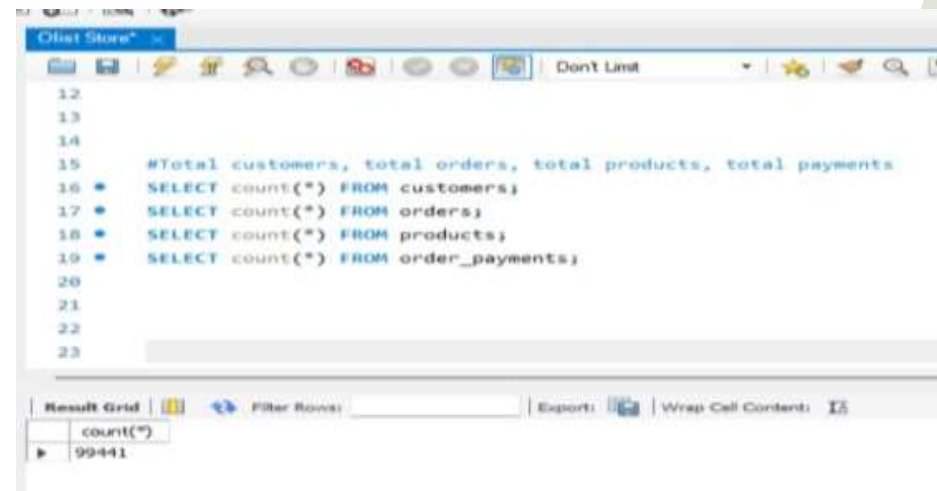
```

12
13
14
15 #Total customers, total orders, total products, total payments
16 * SELECT count(*) FROM customers;
17 * SELECT count(*) FROM orders;
18 * SELECT count(*) FROM products;
19 * SELECT count(*) FROM order_payments;
20
21
22
23

```

count(*)
103886

Total Customers



SQL query results for Total Customers:

```

12
13
14
15 #Total customers, total orders, total products, total payments
16 * SELECT count(*) FROM customers;
17 * SELECT count(*) FROM orders;
18 * SELECT count(*) FROM products;
19 * SELECT count(*) FROM order_payments;
20
21
22
23

```

count(*)
99441

Excel Dashboard



Olist Store Analysis



TOTAL PAYMENTS

BRL 1,64,13,535



TOTAL COST

BRL 1,28,99,322



TOTAL DELIVERED ORDERS

1,04,287



TOTAL CATEGORIES

1,04,294

Year

2016 2017 2018

Quarter

1 2 3

4

Month

January February March

April May June

Day

Monday Tuesday Wednesday

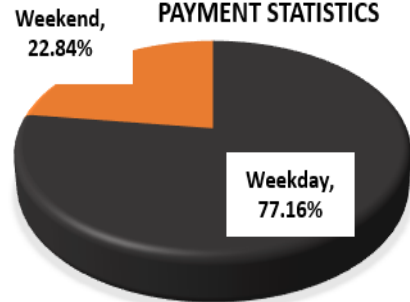
Thursday Friday Saturday

State

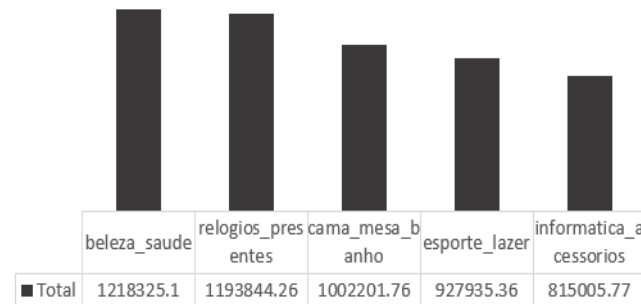
AC AL AM

AP BA CE

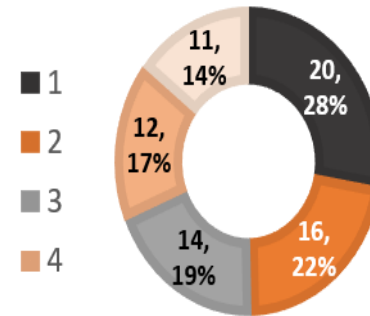
WEEKDAY VS WEEKEND PAYMENT STATISTICS



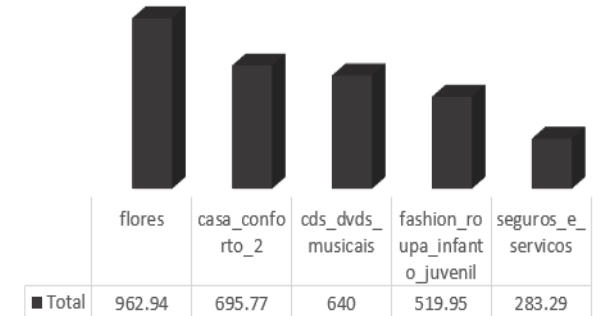
Top 5 Product Category



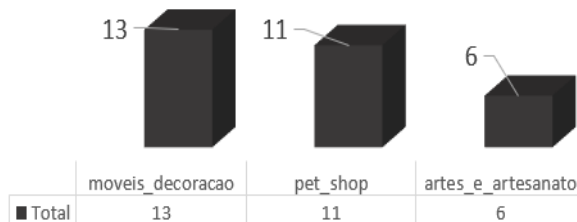
AVERAGE SHIPPING DAYS VS REVIEW SCORES.



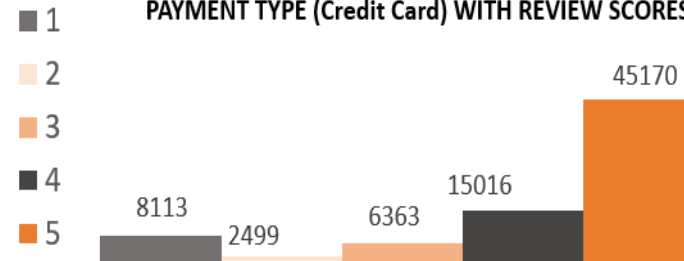
Bottom 5 Product Category



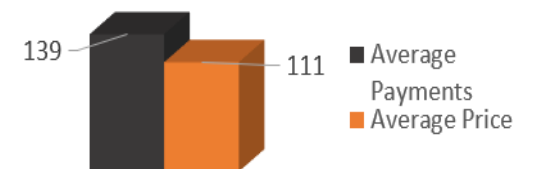
AVERAGE NUMBER OF DELIVERY DAYS



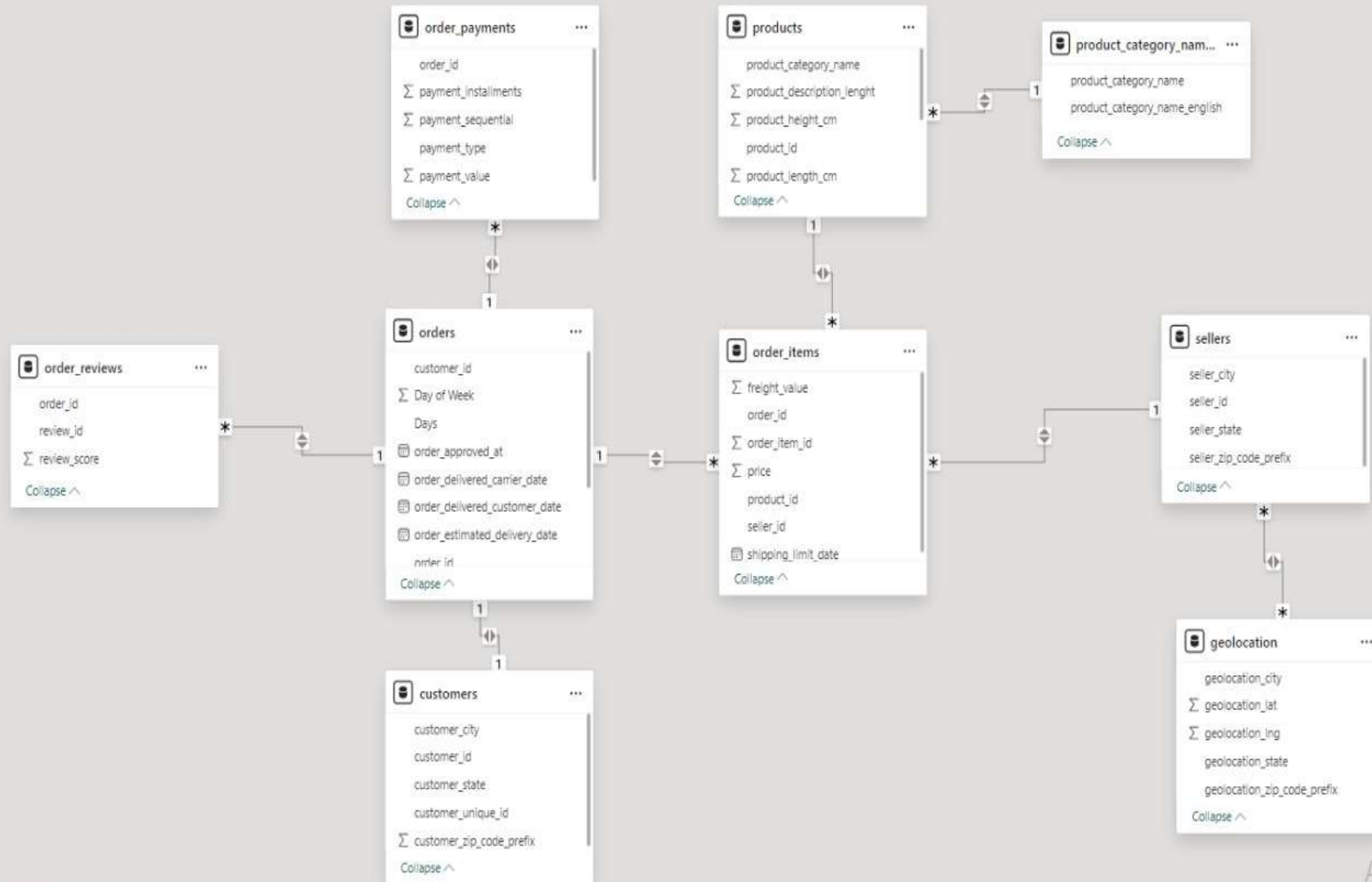
PAYMENT TYPE (Credit Card) WITH REVIEW SCORES



AVERAGE PRICE AND PAYMENT VALUE OF SAO PAULO CITY



Power BI Data Modelling





Olist Store Analysis

Year

All

Month

All

Quarter

All



16.01M

Total sales



3094

Total sellers



2.42M

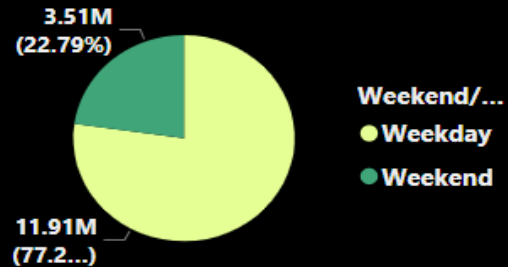
Total profit



96K

Total unique customers

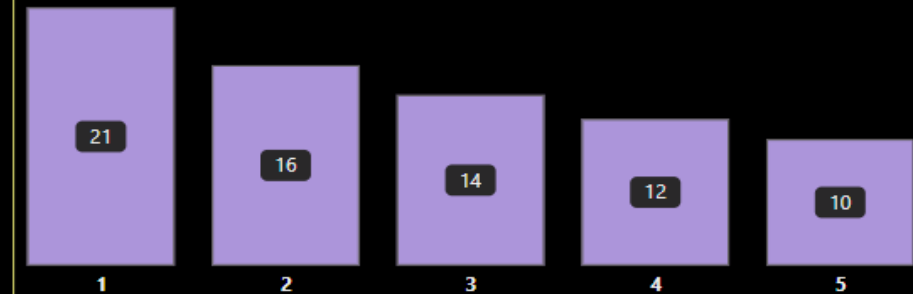
Weekday/Weekend vs payment statistics



Average no. of shipping days taken for pet shop



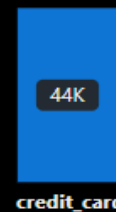
Average of shipping days by review scores



Average price and payment values from customers of Sao Paulo



No. of orders by review score 5 and payment type(CR)



Sum of price and payment by year and month



Tableau Dashboard

olist_order_items_dataset+

olist_order_items_dataset... ▼

Logical Table: olist_order_items_dataset.csv
Double-click this logical table to see its physical table.

olist_order_payments_dat...

olist_orders_dataset.csv

olist_products_dataset.csv

olist_sellers_dataset.csv

olist_customers_dataset.c...

olist_order_reviews_data...

product_category_name_t...

olist_geolocation_dataset....

Tableau Dashboard



Olist Store

Performance Report



Overall Analytics

Key Analytics

Review Score



Customer State

(All)

Payment Type

(All)

Order Status

(All)

Year

(All)

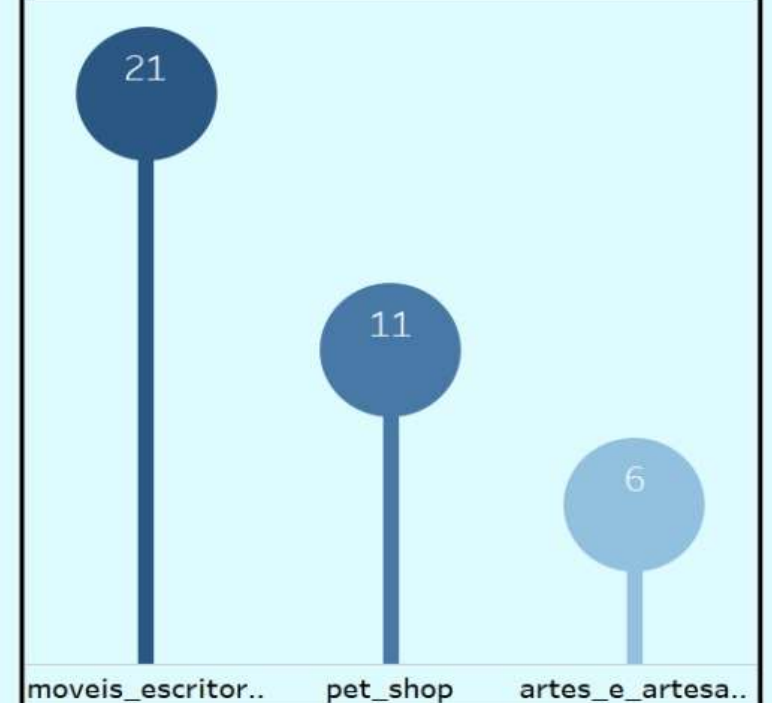
Weekend Vs Weekday Payment Statistics



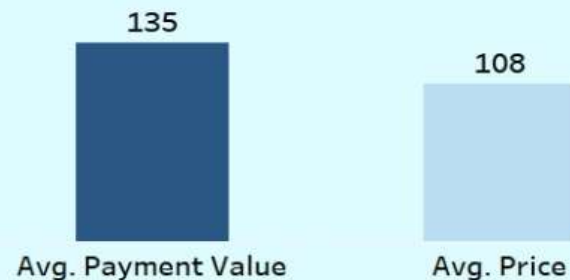
No. of Orders_Payment Type_Review Score = 5



Average No. of Days_Order Delivered to Customer



Average Payment Value_Average Price_Sao Paulo City



Shipping Days Vs Review Score

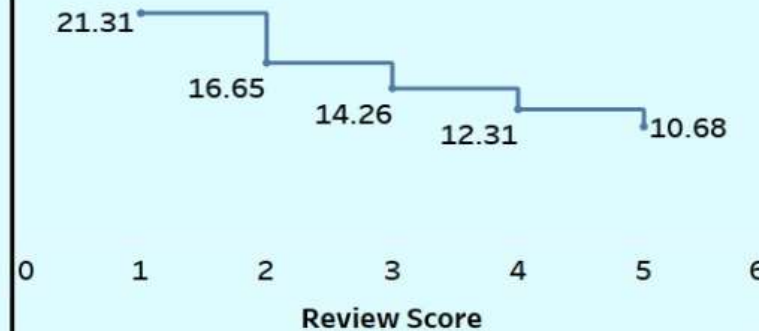


Tableau Dashboard

Olist Store

Performance Report



Key Analytics

Overall Analytics

Review Score



Customer State

(All)

Payment Type

(All)

Order Status

(All)

Year

(All)



98,673
Total Orders



1,11,708
Total Products



4.09
Avg. Review Score



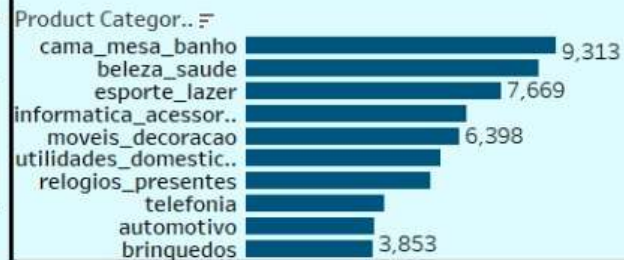
98,673
Total Customers



2.23M
Total Freight

13.47M
Total Price

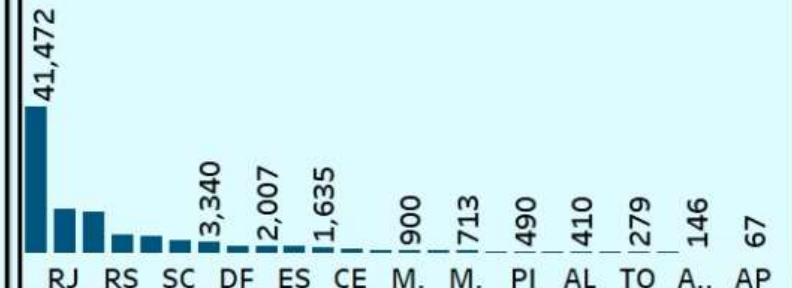
Top 10 Selling Products (Total Order)



Bottom 10 Products (Total Orders)



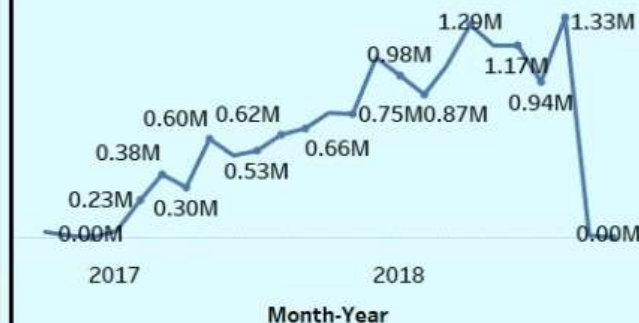
Total Customers_State wise



Order Status

delivered	95,832
shipped	1,032
canceled	605
unavailable	595
invoiced	309
processing	295
created	3
approved	2
Grand Total	98,673

Monthwise Sales Performance



Delivery Performance



Total Sellers_State wise



Insights

- According to the data, Olist E-commerce has about 99,440 orders. With about 89,940 orders being delivered, the company has a 90% delivery success rate.
- Their average product rating is 4.09 stars, with product categories going as high as 4.67 stars and as low as 2.5 stars.
- 1 Star reviews are on third place in the review score distribution ranking which likely indicates that there could be problems with product quality in some product categories.

Summary

- Overall, regular monitoring of customer reviews, real-time tracking of fleet performance, and proper shipment tracking and communication system can improve Olist's delivery and supply chain performance and ensure customers are delighted with the service.

Conclusion

- The Olist Store Analysis project provides valuable insights into customer behavior and payment statistics. The analysis of these KPIs helps Olist in identifying areas of improvement and creating targeted marketing campaigns.
- As a data analyst, We have used Excel and Power BI to clean and manipulate the dataset
- We have Used Power Bi & Tableau to create meaningful visualizations.
- We have used SQL for data validation



Thank you