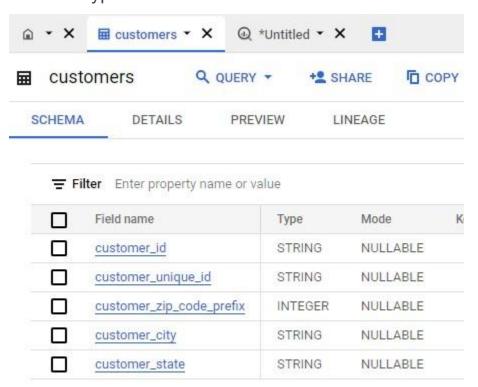
Analyze the given dataset provided by one of the leading Brazilian Clients to extract valuable insights and provide actionable recommendations.

- 1. Imported the dataset and did the usual exploratory analysis steps like checking the structure & characteristics of the dataset:
 - 1.1. The data type of all columns in the "customers" table.



^{*}Note - The screenshot contains only the top 10 results as in several queries I analyzed the trends and the output could be more than what some of the screenshot shows.

1.2. Got the time range between which the orders were placed.

SELECT min(order_purchase_timestamp) Orders_From,
max(order_purchase_timestamp) Orders_Till
FROM `brazilian-client.dataset.orders`



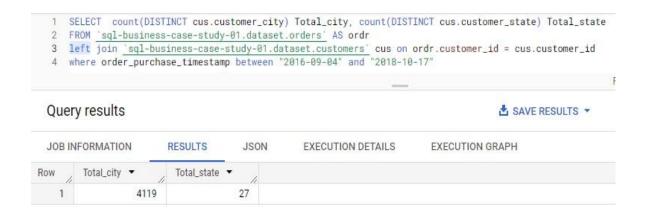
Comments For 1.2 - This dataset contains orders from 09 Sep 2016 to 17 Oct 2018

1.3. Counted the Cities and states of customers who ordered during the given period.

SELECT count(DISTINCT cus.customer_city) Total_city, count(DISTINCT cus.customer_state) Total_state

FROM `brazilian-client.dataset.orders` AS ordr

Left join `brazilian-client.dataset.customers` cus on ordr.customer_id = cus.customer_id where order_purchase_timestamp between "2016-09-04" and "2018-10-17"



Comments For 1.3 - There are a total of 27 states and 4119 cities included.

2. In-depth Exploration:

2.1. I found out if there is a growing trend in the number of orders placed over the past years.

SELECT

EXTRACT(year FROM order_purchase_timestamp) AS Year,

EXTRACT(month FROM order_purchase_timestamp) AS Month,

count(distinct order_id)Number_Of_Orders

FROM `brazilian-client.dataset.orders`

group by Year, Month

order by Year, Month

EXECUTION	JSON	RESULTS	ORMATION	JOB IN
ber_Of_Orders	Nun	Month ▼	Year ▼	Row
4	9		2016	1
324	10		2016	2
1	12		2016	3
800	1		2017	4
1780	2		2017	5
2682	3		2017	6
2404	4		2017	7
3700	5		2017	8
3245	6		2017	9
4026	7.		2017	10

*Comments For 2.1 - There is a general increase in the number of order trends up until September and October month of the year 2018.

2.2. I found out if there is some kind of monthly seasonality in terms of the number of orders being placed.

I would like to answer this in two parts.

Maximum Order Trend in each year for given Data

SELECT

FORMAT_DATE('%B', order_purchase_timestamp) AS Month, count(distinct order_id)Number_Of_Orders,

FROM 'brazilian-client.dataset.orders'

group by Month

order by Number_Of_Orders desc

Row	Month ▼	Number_Of_Orders
1	August	10843
2	May	10573
3	July	10318
4	March	9893
5	June	9412
6	April	9343
7	February	8508
8	January	8069
9	November	7544
10	December	5674

Minimum Order Trend in each year for given Data

SELECT

```
FORMAT_DATE('%B', order_purchase_timestamp) AS Month, count(distinct order_id)Number_Of_Orders, FROM `brazilian-client.dataset.orders` group by Month order by Number_Of_Orders
```

Row	Month ▼	Number_Of_Orders
1	September	4305
2	October	4959
3	December	5674
4	November	7544
5	January	8069
6	February	8508
7	April	9343
8	June	9412
9	March	9893
10	July	10318

^{*}Comments For 2.2 -

When commenting on seasonality in terms of the number of orders, it is crucial to look at the highest-order trends and lowest-order trends.

So mid months of August, May, and July generally see an increase in the number of orders

Whereas the months of September, October, and December see a general trend of a lower number of orders.

2.3. I found out the time of the day, do the Brazilian customers mostly place their orders? (Dawn, Morning, Afternoon, or Night)

```
2.3.1.
        0-6 hrs: Dawn
2.3.2. 7-12 hrs: Mornings
2.3.3. 13-18 hrs: Afternoon
2.3.4. 19-23 hrs: Night
        with cte as (SELECT distinct order id Orders,
        case
        when EXTRACT(hour FROM order_purchase_timestamp) between 0 and 6 then
        "Dawn"
        when EXTRACT(hour FROM order_purchase_timestamp) between 7 and 12 then
        "Mornings"
        when EXTRACT(hour FROM order_purchase_timestamp) between 13 and 18
        then "Afternoon"
        when EXTRACT(hour FROM order purchase timestamp) between 19 and 23
        then "Night"
        end time_of_day
         FROM 'brazilian-client.dataset.orders')
         select time of day, count(Orders) Number Of Orders
         from cte
         group by time of day
         order by Number_Of_Orders desc
```

Row /	time_of_day ▼	Number_Of_Orders
1	Afternoon	38135
2	Night	28331
3	Mornings	27733
4	Dawn	5242

Comments For 2.3 - As per the results, it could be safely inferred that the highest number of orders were placed in the afternoon.

3. Evolution of E-commerce orders in the Brazil region:

3.1. I extracted the month-on-month number of orders placed in each state.

SELECT

```
cus.customer_state State,

EXTRACT(month FROM order_purchase_timestamp) AS Month_Number,

FORMAT_DATE("%b", order_purchase_timestamp) AS Month,

count(distinct order_id)Number_Of_Orders

FROM `brazilian-client.dataset.orders` odr

Left join `brazilian-client.dataset.customers` cus on odr.customer_id = 
cus.customer_id

group by State, Month_Number, Month

order by State, Month_Number
```

Row	State ▼	Month_Number ▼	Month ▼	Number_Of_Orders
1	AC	1	Jan	8
2	AC	2	Feb	6
3	AC	3	Mar	4
4	AC	4	Apr	9
5	AC	5	May	10
6	AC	6	Jun	7
7	AC	7	Jul	9
8	AC	8	Aug	7
9	AC	9	Sep	5
10	AC	10	Oct	6

Comments For 3.1 - Analysis done as per the question asked by the client.

3.2. I found out the customer distribution across all the states.

SELECT

customer_state State,
count(distinct customer_id) Number_Of_Customers
FROM `brazilian-client.dataset.customers`
group by State
order by Number_Of_Customers desc

Row /	State ▼	Number_Of_Custom
1	SP	41746
2	RJ	12852
3	MG	11635
4	RS	5466
5	PR	5045
6	SC	3637
7	BA	3380
8	DF	2140
9	ES	2033
10	GO	2020

^{*}Comments For 3.2 - The top 3 states in terms of the number of unique customers are SP, RJ, MG

- 4. Impact on the Economy: Analyze the money movement by e-commerce by looking at order prices, freight, and others.
 - 4.1. Analyzed the % increase in the cost of orders from 2017 to 2018 (including the months between Jan to Aug only).

```
With cte as (Select
SUM(CASE WHEN EXTRACT(year FROM order_purchase_timestamp) =2017 and
EXTRACT(month FROM order_purchase_timestamp) between 1 and 8 THEN
payment_value ELSE 0 END) For_2017,
SUM(CASE WHEN EXTRACT(year FROM order_purchase_timestamp) =2018 and
EXTRACT(month FROM order_purchase_timestamp) between 1 and 8 THEN
payment_value ELSE 0 END) For_2018,
FROM `brazilian-client.dataset.orders` odr
join `brazilian-client.dataset.payments` p on odr.order_id = p.order_id)
select (((For_2018-For_2017)/For_2017)*100) AS Percent_Growth
from cte
```

Comments For 4.1 - There has been a sweet 136.97% growth in the cost of orders.

4.2. Calculated the total and average value of the order price for each state.

SELECT

```
cus.customer_state State,
round(sum(price),3) Total_value,
round(avg(price),3) Average_value
FROM `sql-business-case-study-01.dataset.order_items` oi
join `sql-business-case-study-01.dataset.orders` odr on oi.order_id = odr.order_id
Left join `sql-business-case-study-01.dataset.customers` cus on odr.customer_id =
cus.customer_id
group by state
order by Average_value desc
```

Row /	State ▼	Total_value ▼	Average_value ▼
1	PB	115268.08	191.475
2	AL	80314.81	180.889
3	AC	15982.95	173.728
4	RO	46140.64	165.974
5	PA	178947.81	165.692
6	AP	13474.3	164.321
7	PI	86914.08	160.358
8	то	49621.74	157.529
9	RN	83034.98	156.966
10	CE	227254.71	153.758

^{*}Comments For 4.2 - As per the above results, the best in terms of average order price is PB

4.3. Calculated the Total and average value of order freight for each state.

SELECT

```
cus.customer_state State,
sum(freight_value) Total_Freight_Value,
avg(freight_value) Average_Freight_Value
FROM `brazilian-client.dataset.orders` odr
join `brazilian-client.dataset.order_items` oit on odr.order_id = oit.order_id
Left join `brazilian-client.dataset.customers` cus on odr.customer_id = cus.customer_id
group by state
order by Average_Freight_Value asc
```

Row	State ▼	Total_Freight_Value	Average_Freight_Value ▼
1	SP	718723.0699999	15.147275390419248
2	PR	117851.6800000	20.531651567944248
3	MG	270853.4600000	20.630166806306541
4	RJ	305589.3100000	20.96092393168248
5	DF	50625.49999999	21.041354945968383
6	SC	89660.26000000	21.470368773946436
7	RS	135522.7400000	21.735804330392945
8	ES	49764.59999999	22.058776595744682
9	G0	53114.979999999	22.766815259322794
10	MS	19144.03000000	23.374884004884006

^{*}Comments For 4.3 - Based on the above data, the state with the least average freight value is SP

- 5. Analysis based on sales, freight, and delivery time.
 - 5.1. Extracted the no. of days taken to deliver each order from the order's purchase date as delivery time.
 - Also, calculate the difference (in days) between the estimated & actual delivery date of an order.
 - Calculated the delivery time and the difference between the estimated & actual delivery date.

SELECT

```
order_id,
date_diff(order_delivered_customer_date,
order_purchase_timestamp,day)time_to_deliver,
date_diff(order_estimated_delivery_date,
order_delivered_customer_date,day) diff_estimated_delivery
FROM `brazilian-client.dataset.orders`
where order_delivered_customer_date is not null and order_status
="delivered"
```

Row	order_id ▼	time_to_deliver ▼	diff_estimated_delive
1	1950d777989f6a877539f5379	30	-12
2	2c45c33d2f9cb8ff8b1c86cc28	30	28
3	65d1e226dfaeb8cdc42f66542	35	16
4	635c894d068ac37e6e03dc54e	30	1
5	3b97562c3aee8bdedcb5c2e45	32	0
6	68f47f50f04c4cb6774570cfde	29	1
7	276e9ec344d3bf029ff83a161c	43	-4
8	54e1a3c2b97fb0809da548a59	40	-4
9	fd04fa4105ee8045f6a0139ca5	37	-1
10	302bb8109d097a9fc6e9cefc5	33	-5

5.2. Extracted the top 5 states with the highest & lowest average freight value. with bottom5 as (SELECT

```
c.customer_state,
avg(freight_value) Fright_value,
FROM `brazilian-client.dataset.orders` o
join `brazilian-client.dataset.customers` c on o.customer_id = c.customer_id
```

```
join `brazilian-client.dataset.order_items` oi on o.order_id = oi.order_id
group by c.customer_state
order by Fright_value limit 5),
bottom5cte as (select *, row_number() over(order by bottom5.Fright_value) as NU
from bottom5),
top5 as (
 SELECT
c.customer_state,
avg(freight_value) Fright_value
FROM 'brazilian-client.dataset.orders' o
join `brazilian-client.dataset.customers` c on o.customer id = c.customer id
join `brazilian-client.dataset.order_items` oi on o.order_id = oi.order_id
group by c.customer_state
order by Fright_value desc limit 5
),
top5cte as (select top5.customer_state, top5.Fright_value,
row_number() over (order by top5.Fright_value) as NU
from top5
order by top5.Fright_value)
select
top5cte.customer_state Top_5_States,
top5cte.Fright_value Freight_Value,
bottom5cte.customer_state Bottom_5_States,
bottom5cte.Fright_value Freight_Value
from
```

bottom5cte join top5cte using (NU)

Row	Top_5_States ▼	Freight_Value ▼	Bottom_5_States ▼	Freight_Value_1 ▼
1	PI	39.147970479704767	SP	15.147275390419248
2	AC	40.073369565217405	PR	20.531651567944248
3	RO	41.069712230215842	MG	20.630166806306541
4	PB	42.723803986710941	RJ	20.96092393168248
5	RR	42.984423076923093	DF	21.041354945968383

Extracted out the top 5 states with the highest & lowest average delivery time. with bottom5 as (SELECT c.customer_state, avg(date diff(order delivered customer date, order purchase timestamp,day))time to deliver FROM 'brazilian-client.dataset.orders' o join 'brazilian-client.dataset.customers' c on o.customer id = c.customer id group by c.customer_state order by time_to_deliver limit 5), bottom5cte as (select *, row_number() over(order by bottom5.time_to_deliver) as NU from bottom5), top5 as (SELECT c.customer state, avg(date_diff(order_delivered_customer_date, order_purchase_timestamp,day))time_to_deliver FROM 'brazilian-client.dataset.orders' o join `brazilian-client.dataset.customers` c on o.customer_id = c.customer_id group by c.customer_state order by time_to_deliver desc limit 5), top5cte as (select *, row_number() over(order by top5.time_to_deliver) as NU from top5) select top5cte.customer_state Top_5_States, top5cte.time_to_deliver Time_To_Deliver,

top5cte.customer_state lop_5_States, top5cte.time_to_deliver Time_To_Deliver, bottom5cte.customer_state Bottom_5_States, bottom5cte.time_to_deliver Time_To_Deliver from

bottom5cte join top5cte using (NU)

5.3.

Row	Top_5_States ▼	Time_To_Deliver ▼	Bottom_5_States ▼	Time_To_Deliver_1 ▼
1	PA	23.316067653276981	SP	8.2980935447227022
2	AL	24.040302267002513	PR	11.526711354864908
3	AM	25.986206896551728	MG	11.54218777523343
4	AP	26.731343283582085	DF	12.509134615384616
5	RR	28.975609756097562	SC	14.475183305132528

5.4. Extracted the top 5 states where the order delivery is really fast compared to the estimated delivery date.

I used the difference between the averages of actual & estimated delivery dates to figure out how fast the delivery was for each state.

```
with cte as (SELECT
```

```
C.customer_state,

avg(date_diff(order_delivered_customer_date,order_purchase_timestamp, day))

No_of_actual_average_delivery_time,

avg(date_diff(order_estimated_delivery_date,order_purchase_timestamp, day))

No_of_estimated_average_delivery_time,

FROM `brazilian-client.dataset.orders` o

join `brazilian-client.dataset.customers` c on o.customer_id = c.customer_id

where o.order_status = "delivered"

group by c.customer_state

order by c.customer_state)
```

select customer_state,

(No_of_estimated_average_delivery_time - No_of_actual_average_delivery_time) Difference

from cte

order by Difference limit 5

Row	customer_state ▼	Difference ▼
1	AL	8.1687657430730631
2	MA	8.9665271966527413
3	SE	9.4537313432835681
4	ES	9.8882205513784669
5	CE	10.186864738076622

6. Analysis based on the payments:

6.1. Analyzed the month-on-month number of orders placed using different payment types.

SELECT

```
payment_type Payment_Type,

EXTRACT(month FROM order_purchase_timestamp) AS Month_Number,

FORMAT_DATE("%b", order_purchase_timestamp) AS Month,

count(odr.order_id)Number_Of_Orders

FROM `brazilian-client.dataset.orders` odr

Left join `brazilian-client.dataset.payments` pay on odr.order_id = pay.order_id

group by Payment_Type, Month_Number, Month

order by Payment_Type, Month_Number
```

Row	Payment_Type ▼	Month_Number ▼	Month ▼	Number_Of_Orders
1	null	9	Sep	1
2	UPI	1	Jan	1715
3	UPI	2	Feb	1723
4	UPI	3	Mar	1942
5	UPI	4	Apr	1783
6	UPI	5	May	2035
7	UPI	6	Jun	1807
8	UPI	7	Jul	2074
9	UPI	8	Aug	2077
10	UPI	9	Sep	903

Comments For 6.1 - The above data shows the payment distribution by payment type in each state.

However, 1 payment has payment_type = null, we need clarification from the business on how exactly to treat the null value in this case.

6.2. Extracted the number of orders placed on the basis of the payment installments that have been paid.

```
SELECT payment_sequential No_Of_Installments,
count(distinct order_id) No_of_orders
FROM `brazilian-client.dataset.payments`
where payment_sequential>=2
group by payment_sequential
order by payment_sequential desc
```

Row	No_Of_Installments	No_of_orders ▼
1	29	1
2	28	1
3	27	1
4	26	2
5	25	2
6	24	2
7	23	2
8	22	3
9	21	4
10	20	4

Comments For 6.2 - I have calculated only those orders that had payment sequential equal to or greater than 2 so that I can exclude any single payment order that is not on EMI and has been paid in a single go.