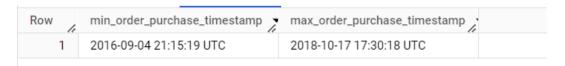
- 1. Import the dataset and do usual exploratory analysis steps like checking the structure & characteristics of the dataset:
- 1. Data type of all columns in the "customers" table.

| Field name | Туре | Mode | Key | Collation | Default value | Policy tags ? | Description |
|--------------------------|---------|----------|-----|-----------|---------------|---------------|-------------|
| customer_id | STRING | NULLABLE | | | | | |
| customer_unique_id | STRING | NULLABLE | | | | | |
| customer_zip_code_prefix | INTEGER | NULLABLE | | | | | |
| customer_city | STRING | NULLABLE | | | | | |
| customer_state | STRING | NULLABLE | | | | | |

2. Get the time range between which the orders were placed.

select

min(order_purchase_timestamp) as min_order_purchase_timestamp,
max(order_purchase_timestamp) as max_order_purchase_timestamp,
from Target_Buz_case.orders;



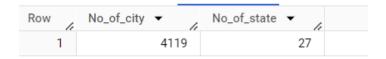
Insight:

Here, first order is placed at 2016-09-04 and the is placed at 2018-10-17, This is basically purchasing start 2016-09-04 and the purchasing ended at 2018-10-17 orders were placed.

3. Count the Cities & States of customers who ordered during the given period.

select

count(distinct c.customer_city) as No_of_city,
count(distinct c.customer_state) as No_of_state
from Target_Buz_case.customers c join Target_Buz_case.orders
using(customer_id)



Insight:

Here the Number of cities are 4119 and The Number of States are 27. This are the count of customer who ordered during the given period

2. In-depth Exploration:

1. Is there a growing trend in the no. of orders placed over the past years

```
select count(order_id) as count_order,
extract(year from order_purchase_timestamp) as year
from Target_Buz_case.orders
group by 2
order by 2
```

| Row | count_order ▼ | year ▼ | lı. |
|-----|---------------|--------|------|
| 1 | 329 | | 2016 |
| 2 | 45101 | | 2017 |
| 3 | 54011 | | 2018 |

Insight:

Here we can see that the sudden orders are increase in 2017 and gradually Increase as well in 2018

2. Can we see some kind of monthly seasonality in terms of the no. of orders being placed?

```
select
extract(month from order_purchase_timestamp) as month,
count(order_id) as count_order
from Target_Buz_case.orders
group by 1
order by 1;
```

| Row | month ▼ | 10 | count_order ▼ |
|-----|---------|----|---------------|
| 1 | | 1 | 8069 |
| 2 | | 2 | 8508 |
| 3 | | 3 | 9893 |
| 4 | | 4 | 9343 |
| 5 | | 5 | 10573 |
| 6 | | 6 | 9412 |
| 7 | | 7 | 10318 |
| 8 | | 8 | 10843 |
| 9 | | 9 | 4305 |
| 10 | | 10 | 4959 |

Insight:

Slightly higher count in may, July and August.

3. During what time of the day, do the Brazilian customers mostly place their orders? (Dawn, Morning, Afternoon or Night)

a. 0-6 hrs: Dawnb. 7-12 hrs: Morningsc. 13-18 hrs: Afternoond. 19-23 hrs: Night

select

sum(case when extract(hour from order_purchase_timestamp) between 0 and 6 then 1 else 0 end) as DAWN,

sum(case when extract(hour from order_purchase_timestamp) between 7 and 12 then 1 else 0 end) as MORNING,

sum(case when extract(hour from order_purchase_timestamp) between 13 and 18 then 1 else 0 end) as AFTERNOON,

sum(case when extract(hour from order_purchase_timestamp) between 19 and 23 then 1 else 0 end) as NIGHT from Target_Buz_case.orders



Insight:

Here we can easily see that Brazilian mostly purchase between 13:00 to 18:0 i.e. afternoon

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

1. Get the month-on-month no. of orders placed in each state.

```
select
extract(year from order_purchase_timestamp) as year,
extract(month from order_purchase_timestamp) as month,
customer_state,
count(order_id) as cnt_orders
from Target_Buz_case.orders inner join
Target_Buz_case.customers using(customer_id)
group by 1,2,3
order by 1,2
limit 10;
```

| Row | year ▼ | month ▼ | customer_state ▼ | cnt_orders ▼ |
|-----|--------|---------|------------------|--------------|
| 1 | 2016 | 9 | RS | 1 |
| 2 | 2016 | 9 | RR | 1 |
| 3 | 2016 | 9 | SP | 2 |
| 4 | 2016 | 10 | SP | 113 |
| 5 | 2016 | 10 | MG | 40 |
| 6 | 2016 | 10 | GO | 9 |
| 7 | 2016 | 10 | CE | 8 |
| 8 | 2016 | 10 | SC | 11 |
| 9 | 2016 | 10 | RJ | 56 |
| 10 | 2016 | 10 | RS | 24 |

Insight:

Here we can see that the count of orders respective to their state and year and month

2. How are the customers distributed across all the states?

```
select customer_state,
count(distinct customer_id) distinct_customer
from Target_Buz_case.customers
group by 1
order by 2
```

| Row | customer_state ▼ | distinct_customer |
|-----|------------------|-------------------|
| 1 | RR | 46 |
| 2 | AP | 68 |
| 3 | AC | 81 |
| 4 | AM | 148 |
| 5 | RO | 253 |
| 6 | TO | 280 |
| 7 | SE | 350 |
| 8 | AL | 413 |
| 9 | RN | 485 |
| 10 | PI | 495 |

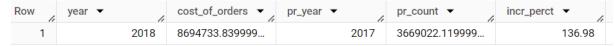
Insight:

Here Each state contain distinct customer and distinct_customer contain the count of customer.

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

 You can use the "payment_value" column in the payments table to get the cost of orders.

```
with cte as(
select sum(p.payment_value) as cost_of_orders,extract(year from
order_purchase_timestamp) as year
from Target_Buz_case.orders o join Target_Buz_case.payments p
on o.order_id=p.order_id
where extract(month from order_purchase_timestamp) between 1 and 8
group by extract(year from order_purchase_timestamp)
)
select *,round(((cost_of_orders-pr_count)/pr_count)*100,2) as incr_perct from(
select year,cost_of_orders,
lag(year) over(order by cte.year) as pr_year,
lead(cost_of_orders) over(order by cost_of_orders desc) as pr_count
from cte) temp_t
where temp_t.pr_count is not null
order by cost_of_orders desc;
```



Here the percentage increase is nearly 137 %

2. Calculate the Total & Average value of order price for each state

select

```
customer_state,
round(sum(price),0) as sum_price_value,
round(avg(price),0) as avg_price_value
from Target_Buz_case.orders o
join Target_Buz_case.order_items oi using(order_id)
join Target_Buz_case.customers c on o.customer_id=c.customer_id
group by 1
limit 10;
```

| Row | customer_state ▼ | sum_price_value 🔻 | avg_price_value 🔻 |
|-----|------------------|-------------------|-------------------|
| 1 | MT | 156454.0 | 148.0 |
| 2 | MA | 119648.0 | 145.0 |
| 3 | AL | 80315.0 | 181.0 |
| 4 | SP | 5202955.0 | 110.0 |
| 5 | MG | 1585308.0 | 121.0 |
| 6 | PE | 262788.0 | 146.0 |
| 7 | RJ | 1824093.0 | 125.0 |
| 8 | DF | 302604.0 | 126.0 |
| 9 | RS | 750304.0 | 120.0 |
| 10 | SE | 58921.0 | 153.0 |

Insight:

Here we can see that, sum of price and average of price According to each state and 'Sp' price value is in comparison with other is low and the Highest average price is 'Al'

3. Calculate the Total & Average value of order freight for each state.

select

```
customer_state,
round(sum(freight_value),0) as sum_freight_value,
round(avg(freight_value),0) as avg_freight_value
from Target_Buz_case.orders o
join Target_Buz_case.order_items oi using(order_id)
join Target_Buz_case.customers c on o.customer_id=c.customer_id
group by 1
```

| Row / | customer_state ▼ | sum_freight_value | avg_freight_value |
|-------|------------------|-------------------|-------------------|
| 1 | MT | 29715.0 | 28.0 |
| 2 | MA | 31524.0 | 38.0 |
| 3 | AL | 15915.0 | 36.0 |
| 4 | SP | 718723.0 | 15.0 |
| 5 | MG | 270853.0 | 21.0 |
| 6 | PE | 59450.0 | 33.0 |
| 7 | RJ | 305589.0 | 21.0 |
| 8 | DF | 50625.0 | 21.0 |
| 9 | RS | 135523.0 | 22.0 |
| 10 | SE | 14111.0 | 37.0 |

Insight:

Here we can see that, sum of freight value and average of freight value According to each state

- 5. Analysis based on sales, freight and delivery time.
- 1. Find 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.

Do this in a single query.

You can calculate the delivery time and the difference between the estimated & actual delivery date using the given formula:

- o time_to_deliver = order_delivered_customer_date order_purchase_timestamp
- diff_estimated_delivery = order_estimated_delivery_date order_delivered_customer_date

select order_id,
order_purchase_timestamp,
order_delivered_customer_date,
order_estimated_delivery_date,
date_diff(order_delivered_customer_date,order_purchase_timestamp,day) as
time_to_deliver,
date_diff(order_estimated_delivery_date,order_delivered_customer_date,day)as
diff_estimated_delivery
from Target_Buz_case.orders

| Row | order_id ▼ | order_purchase_timestamp ▼ | order_delivered_customer_date | order_estimated_delivery_date 🔻 | time_to_deliver ▼ | diff_estimated_delive |
|-----|----------------------------|----------------------------|-------------------------------|---------------------------------|-------------------|-----------------------|
| 1 | 1950d777989f6a877539f5379 | 2018-02-19 19:48:52 UTC | 2018-03-21 22:03:51 UTC | 2018-03-09 00:00:00 UTC | 30 | -12 |
| 2 | 2c45c33d2f9cb8ff8b1c86cc28 | 2016-10-09 15:39:56 UTC | 2016-11-09 14:53:50 UTC | 2016-12-08 00:00:00 UTC | 30 | 28 |
| 3 | 65d1e226dfaeb8cdc42f66542 | 2016-10-03 21:01:41 UTC | 2016-11-08 10:58:34 UTC | 2016-11-25 00:00:00 UTC | 35 | 16 |
| 4 | 635c894d068ac37e6e03dc54e | 2017-04-15 15:37:38 UTC | 2017-05-16 14:49:55 UTC | 2017-05-18 00:00:00 UTC | 30 | 1 |
| 5 | 3b97562c3aee8bdedcb5c2e45 | 2017-04-14 22:21:54 UTC | 2017-05-17 10:52:15 UTC | 2017-05-18 00:00:00 UTC | 32 | 0 |
| 6 | 68f47f50f04c4cb6774570cfde | 2017-04-16 14:56:13 UTC | 2017-05-16 09:07:47 UTC | 2017-05-18 00:00:00 UTC | 29 | 1 |
| 7 | 276e9ec344d3bf029ff83a161c | 2017-04-08 21:20:24 UTC | 2017-05-22 14:11:31 UTC | 2017-05-18 00:00:00 UTC | 43 | -4 |
| 8 | 54e1a3c2b97fb0809da548a59 | 2017-04-11 19:49:45 UTC | 2017-05-22 16:18:42 UTC | 2017-05-18 00:00:00 UTC | 40 | -4 |
| 9 | fd04fa4105ee8045f6a0139ca5 | 2017-04-12 12:17:08 UTC | 2017-05-19 13:44:52 UTC | 2017-05-18 00:00:00 UTC | 37 | -1 |
| 10 | 302bb8109d097a9fc6e9cefc5 | 2017-04-19 22:52:59 UTC | 2017-05-23 14:19:48 UTC | 2017-05-18 00:00:00 UTC | 33 | -5 |

Here, time_to_deliver column shows the difference that taken to deliver each order from the order's purchase date as delivery time.

```
In diff_estimated_delivery, we can see some negative,
```

That mean of -ve difference shows that the customer received order late as per expected date,

Where as +ve difference shows that the customer received order before as per expected date.

2. Find out the top 5 states with the highest & lowest average freight value.

```
with cte as(
select
customer_state,
round(avg(freight_value),2)as average
from Target_Buz_case.orders o
join Target_Buz_case.order_items oi using(order_id)
join Target_Buz_case.customers c on o.customer_id=c.customer id
group by 1
),
cte2 as(
select customer_state,average,
dense rank()over(order by average desc ) as rnk top 5,
dense_rank()over(order by average asc ) as rnk_lowest_5
from cte)
select distinct a.customer state, a.average, a.rnk top 5,
          b.customer_state,b.average,b.rnk_lowest_5
from cte2 a join cte2 b
on a.rnk_top_5=b.rnk_lowest_5
where a.rnk_top_5<=5 and b.rnk_lowest_5<=5
order by a.rnk_top_5,b.rnk_lowest_5;
```



3. Find out the top 5 states with the highest & lowest average delivery time.

```
with cte as (select customer_state,
round(avg(date diff(order delivered customer date, order purchase timestamp
,day)),2) avg_time_to_deliver
from Target_Buz_case.orders join Target_Buz_case.customers
using(customer_id)
group by customer_state),
cte2 as(
 select *,rank()over(order by avg_time_to_deliver desc ) as
top 5 avg time to deliver,
      rank()over(order by avg_time_to_deliver asc ) as
lowest_5_avg_time_to_deliver from cte
 order by 3
select a.customer_state,a.avg_time_to_deliver,a.top_5_avg_time_to_deliver,
    b.customer_state,b.avg_time_to_deliver,b.lowest_5_avg_time_to_deliver
from cte2 a join cte2 b on
a.top_5_avg_time_to_deliver=b.lowest_5_avg_time_to_deliver
where a.top_5_avg_time_to_deliver<=5 and
b.lowest_5_avg_time_to_deliver<=5
order by 3,6
```

| Row | customer_state ▼ | avg_time_to_deliver | top_5_avg_time_to_d | customer_state_1 ▼ | avg_time_to_deliver_ | lowest_5_avg_time_t |
|-----|------------------|---------------------|---------------------|--------------------|----------------------|---------------------|
| 1 | RR | 28.98 | 1 | SP | 8.3 | 1 |
| 2 | AP | 26.73 | 2 | PR | 11.53 | 2 |
| 3 | AM | 25.99 | 3 | MG | 11.54 | 3 |
| 4 | AL | 24.04 | 4 | DF | 12.51 | 4 |
| 5 | PA | 23.32 | 5 | SC | 14.48 | 5 |

Here is the top 5 highest and top 5 lowest state with their average time take to delver.

4. Find out the top 5 states where the order delivery is really fast as compared to the estimated date of delivery.

You can use the difference between the averages of actual &

estimated delivery date to figure out how fast the delivery was for each state.

```
with cte as
(select customer_state,
round(avg(date_diff(order_delivered_customer_date,order_purchase_timestamp)
,day)),2) as avg_delv_actual,
round(avg(date_diff(order_estimated_delivery_date,order_delivered_customer_
date,day)),2) as avg_esti_dele
from Target Buz case.orders join Target Buz case.customers
using(customer_id)
group by customer_state
order by 2 desc ,3 desc
)
select customer_state,avg_delv_actual,avg_esti_dele,round((avg_delv_actual-
avg_esti_dele),2) diff_betweeen_actual_estimate
from cte
order by 4 desc
limit 5;
```

| Row | customer_state ▼ | avg_delv_actual ▼ | avg_esti_dele ▼ | diff_betweeen_actua |
|-----|------------------|-------------------|-----------------|---------------------|
| 1 | AL | 24.04 | 7.95 | 16.09 |
| 2 | RR | 28.98 | 16.41 | 12.57 |
| 3 | MA | 21.12 | 8.77 | 12.35 |
| 4 | SE | 21.03 | 9.17 | 11.86 |
| 5 | CE | 20.82 | 9.96 | 10.86 |

Insight:

Here we can easily see that top 5 states the order delivery is really fast as compared to the estimated date of delivery. The more difference that means the delivery done fast

6.Analysis based on the payments:

1. Find the month on month no. of orders placed using different payment types.

```
select count(order_id) as count_order,payment_type,
extract(year from order_purchase_timestamp) as year,
extract(month from order_purchase_timestamp) as month from
Target_Buz_case.payments join Target_Buz_case.orders using(order_id)
group by 2,3,4
order by 3,4
limit 10;
```

| Row / | count_order ▼ | payment_type ▼ | year ▼ | month ▼ |
|-------|---------------|----------------|--------|---------|
| 1 | 3 | credit_card | 2016 | 9 |
| 2 | 2 | debit_card | 2016 | 10 |
| 3 | 254 | credit_card | 2016 | 10 |
| 4 | 23 | voucher | 2016 | 10 |
| 5 | 63 | UPI | 2016 | 10 |
| 6 | 1 | credit_card | 2016 | 12 |
| 7 | 61 | voucher | 2017 | 1 |
| 8 | 197 | UPI | 2017 | 1 |
| 9 | 583 | credit_card | 2017 | 1 |
| 10 | 9 | debit_card | 2017 | 1 |

Insight:

Above, we can see that count of order using different payment type with month of each year

2. Find the no. of orders placed on the basis of the payment installments that have been paid.

```
select count(order_id) as count_order,
payment_installments from Target_Buz_case.payments join
Target_Buz_case.orders using(order_id)
group by 2
having payment_installments>=1
limit 10;
```

| Row | count_order ▼ | payment_installment |
|-----|---------------|---------------------|
| 1 | 52546 | 1 |
| 2 | 12413 | 2 |
| 3 | 10461 | 3 |
| 4 | 7098 | 4 |
| 5 | 5239 | 5 |
| 6 | 3920 | 6 |
| 7 | 1626 | 7 |
| 8 | 4268 | 8 |
| 9 | 644 | 9 |
| 10 | 5328 | 10 |

Here the count of orders based on the number of installment and the count is slowly getting decrease