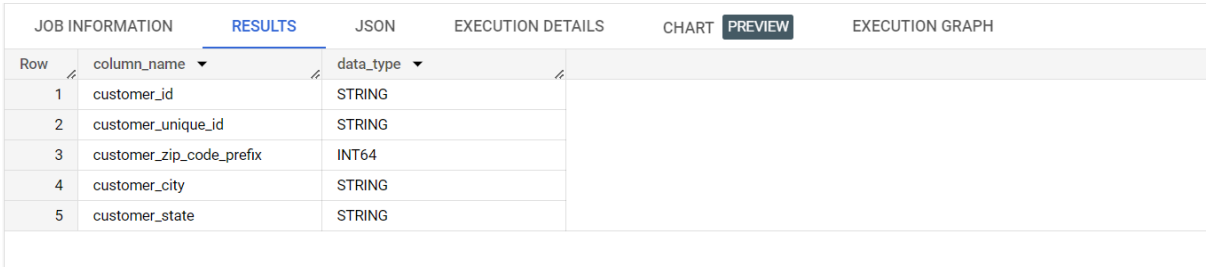
Import the dataset and do usual exploratory analysis steps like checking the structure & characteristics of the dataset:

1. Data type of all columns in the "customers" table.

```
select column_name,data_type from `target_sql`.INFORMATION_SCHEMA.COLUMNS
where table_name='customers'
```



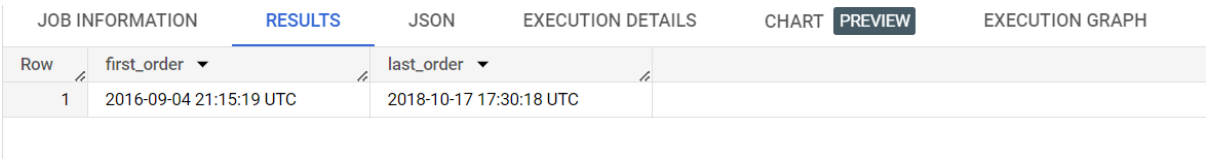
JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	column_name	data_type
1	customer_id	STRING
2	customer_unique_id	STRING
3	customer_zip_code_prefix	INT64
4	customer_city	STRING
5	customer_state	STRING

Inference: Information on data type of the columns in the customers table.

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

```
select min(order_purchase_timestamp) first_order,max(order_purchase_timestamp)
last_order from `target_sql.orders`
```



JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	first_order	last_order
1	2016-09-04 21:15:19 UTC	2018-10-17 17:30:18 UTC

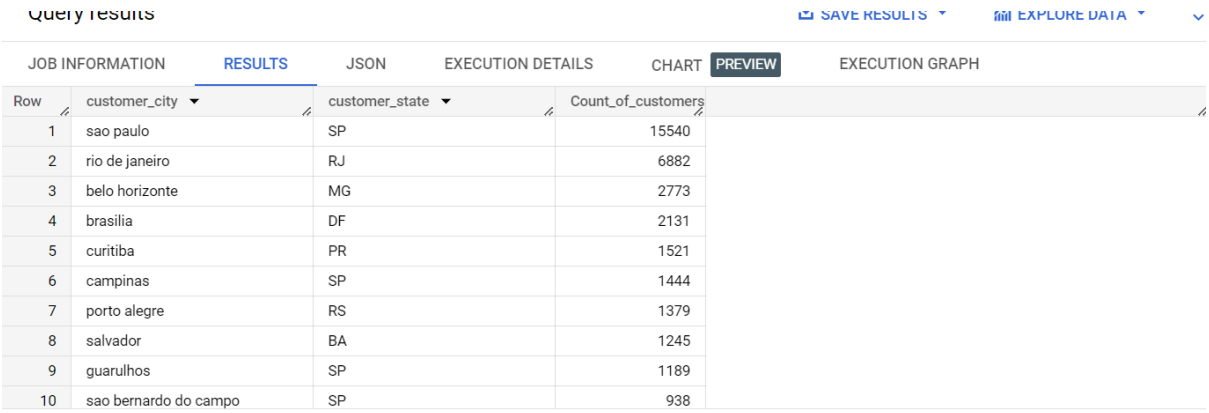
Inference: Date range between which orders were placed i.e. first order was placed on 4th September 2016 and the last order was placed on 17th October 2018.

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

```
select distinct c.customer_city,c.customer_state,count(c.customer_id) Count_of_customers
from `target_sql.customers` c

join `target_sql.orders` o on o.customer_id=c.customer_id

group by c.customer_state,c.customer_city
order by 3 desc
```



Query results SAVE RESULTS EXPLORE DATA

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	customer_city	customer_state	Count_of_customers
1	sao paulo	SP	15540
2	rio de janeiro	RJ	6882
3	belo horizonte	MG	2773
4	brasilia	DF	2131
5	curitiba	PR	1521
6	campinas	SP	1444
7	porto alegre	RS	1379
8	salvador	BA	1245
9	guarulhos	SP	1189
10	sao bernardo do campo	SP	938

Inference: Sao Paulo city from SP state has the highest number of orders.

In-depth Exploration:

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

```
select extract(year from o.order_purchase_timestamp) as year,
count(distinct o.order_id) as No_of_orders
 from `target_sql.orders` o
join `target_sql.customers` c on c.customer_id=o.customer_id
group by year
order by 1
```

Query results

SAVE RESULTS EXPLOR

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	year	No_of_orders
1	2016	329
2	2017	45101
3	2018	54011

Inference: There has been approximately 137 times increase in orders from 2016-2017.

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

```
select extract(month from o.order_purchase_timestamp) as month,
count(distinct o.order_id) as No_of_orders
from `target_sql.orders` o
join `target_sql.customers` c on c.customer_id=o.customer_id
group by month
order by 1
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	month	No_of_orders
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
11	11	7544

Inference: The count of orders increases from March to August with fluctuations in between.

In September we can see a massive dip in orders count.

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

```
select
case when extract(hour from o.order_purchase_timestamp)  between 0 and 6 then 'Dawn'
     when extract(hour from o.order_purchase_timestamp)  between 7 and 12 then 'Mornings'
     when extract(hour from o.order_purchase_timestamp)  between 13 and 18 then 'Afternoon'
     when extract(hour from o.order_purchase_timestamp)  between 19 and 23 then 'Night'
end as hour, count(o.order_id) as order_placed
from `target_sql.orders` o
join `target_sql.customers` c on c.customer_id=o.customer_id
group by hour
order by 2 desc
```

Query results

SAVE RESULTS | E

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	hour	order_placed
1	Afternoon	38135
2	Night	28331
3	Mornings	27733
4	Dawn	5242

Inference: During the afternoon session most of the orders were placed.

Evolution of E-commerce orders in the Brazil region:

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

```
select c.customer_state,extract(month from o.order_purchase_timestamp) as month,
count(o.order_id) order_count
from `target_sql.orders` o
join `target_sql.customers` c on c.customer_id=o.customer_id
group by 1,2
order by 1,2
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	customer_state	month	order_count
1	AC	1	8
2	AC	2	6
3	AC	3	4
4	AC	4	9
5	AC	5	10
6	AC	6	7
7	AC	7	9
8	AC	8	7
9	AC	9	5
10	AC	10	6

Inference: We can infer that SP state has highest number of orders than any other states.

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

```
Select c.customer_state,COUNT(c.customer_id) as customers_count
From `target_sql.customers` c
group by c.customer_state
order by customers_count desc
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	customer_state	customers_count
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

Results per page: 50 1 – 27 of 27

Inference: We can infer that highest distribution of customers is in SP -Sao Paulo state.

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).**

```
with cte1 as
(
select extract(year from order_purchase_timestamp) year,sum(p.payment_value)
curr_month
from `target_sql.orders` o
join `target_sql.payments` p on p.order_id=o.order_id
where  extract(month from order_purchase_timestamp) between 1 and 8
group by 1
order by 1
),

cte2 as
(
select *, lead(curr_month,1) over (order by year) next_month
from cte1
)

select cte2.year,(next_month-curr_month)/curr_month*100 as percent_increase from
cte2
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	year	percent_increase
1	2017	136.9768716466...
2	2018	null

Inference: We can infer that there was 137 percent increase in payment values from 2017 to 2018(including only months between January-August)

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

```
select c.customer_state,round(sum(oi.price),2) Total_price,round(avg(oi.price),2)
Avg_price
from `target_sql.customers` c
join `target_sql.orders` o on o.customer_id=c.customer_id
join `target_sql.order_items` oi on oi.order_id=o.order_id
group by c.customer_state
order by 2 desc,3 asc
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	customer_state	Total_price	Avg_price
1	SP	5202955.05	109.65
2	RJ	1824092.67	125.12
3	MG	1585308.03	120.75
4	RS	750304.02	120.34
5	PR	683083.76	119.0
6	SC	520553.34	124.65
7	BA	511349.99	134.6
8	DF	302603.94	125.77
9	GO	294591.95	126.27
10	ES	275037.31	121.91

Inference: SP state has highest total price value of all the states but also has the lowest average price value.

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

```sql
select c.customer_state,round(sum(oi.freight_value),2)
Total_freight,round(avg(oi.freight_value),2) Avg_freight
from `target_sql.customers` c
join `target_sql.orders` o on o.customer_id=c.customer_id
join `target_sql.order_items` oi on oi.order_id=o.order_id
group by c.customer_state
order by 2 desc,3 asc
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	customer_state	Total_freight	Avg_freight
1	SP	718723.07	15.15
2	RJ	305589.31	20.96
3	MG	270853.46	20.63
4	RS	135522.74	21.74
5	PR	117851.68	20.53
6	BA	100156.68	26.36
7	SC	89660.26	21.47
8	PE	59449.66	32.92
9	GO	53114.98	22.77
10	DF	50625.5	21.04
11	ES	49764.6	22.06

Results per page: 50 1 – 27 of 27

Inference: SP state has the highest total freight value and also lowest average freight value.

Analysis based on the payments:

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

```sql
SELECT p.payment_type,EXTRACT(month from o.order_purchase_timestamp)
month,COUNT(distinct o.order_id)  order_count
FROM `target_sql.orders` o
JOIN `target_sql.payments` p on o.order_id = p.order_id
GROUP BY 1, 2
ORDER BY 1, 2
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	payment_type	month	order_count
1	UPI	1	1715
2	UPI	2	1723
3	UPI	3	1942
4	UPI	4	1783
5	UPI	5	2035
6	UPI	6	1807
7	UPI	7	2074
8	UPI	8	2077
9	UPI	9	903

Inference: We can infer from the data that Credit card payment mode is highly used and popular than others.

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

```
SELECT p.payment_installments,COUNT(o.order_id) order_count
FROM `target_sql.orders` o
join `target_sql.payments` p on o.order_id = p.order_id
where o.order_status != 'canceled'
GROUP BY 1
ORDER BY 2 desc
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	payment_installment	order_count
1	1	52184
2	2	12353
3	3	10392
4	4	7056
5	10	5292
6	5	5209
7	8	4239
8	6	3898
9	7	1620
10	9	638
11	12	133

Results per page: 50 1 – 24 of 24

Inference: It can be inferred that most of the orders (52184) like to pay in 1 instalment and a smaller number of orders i.e. 18 pay in 24 instalments

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:
   - **time_to_deliver = order_delivered_customer_date - order_purchase_timestamp**
   - **diff_estimated_delivery = order_estimated_delivery_date - order_delivered_customer_date**

```
select order_id,
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_sql.orders`
where date_diff(order_delivered_customer_date, order_purchase_timestamp,day) is not null
order by time_to_deliver
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART PREVIEW | EXECUTION GRAPH

Row	order_id	time_to_deliver	diff_estimated_delive
1	e65f1eeee1f52024ad1dcd034...	0	9
2	bb5a519e352b45b714192a02f...	0	25
3	434cecee7d1a65fc65358a632...	0	19
4	d3ca7b82c922817b06e5ca211...	0	11
5	1d893dd7ca5f77ebf5f59f0d20...	0	10
6	d5fbeedc85190ba88580d6f82...	0	7
7	79e324907160caea526fd8b94...	0	8
8	38c1e3d4ed6a13cd0cf612d4c...	0	16
9	8339b608be0d84fca9d8da68b...	0	27
10	f349cdb62f69c3fae5c4d7d3f3...	0	12

Inference: We can infer that time taken to deliver was less than 1 day for 10 orders .

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

--Top 5 states with highest avg freight values

```
select c.customer_state,avg(oi.freight_value) as avg_freight_values from
`target_sql.orders` o
join `target_sql.order_items` oi on oi.order_id=o.order_id
join `target_sql.customers` c on c.customer_id=o.customer_id
group by c.customer_state
order by avg_freight_values desc
limit 5
```

Row	customer_state	avg_freight_values
1	RR	42.98442307692...
2	PB	42.72380398671...
3	RO	41.06971223021...
4	AC	40.07336956521...
5	PI	39.14797047970...

--Bottom 5 states with lowest avg freight values

```
select c.customer_state,avg(oi.freight_value) as avg_freight_values from
`target_sql.orders` o
join `target_sql.order_items` oi on oi.order_id=o.order_id
join `target_sql.customers` c on c.customer_id=o.customer_id
group by c.customer_state
order by avg_freight_values
limit 5
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHART

Row	customer_state	avg_freight_values
1	SP	15.14727539041...
2	PR	20.53165156794...
3	MG	20.63016680630...
4	RJ	20.96092393168...
5	DF	21.04135494596...

Inference: We can infer from above data as the top most average freight value is of state- RR and the lowest avg freight value is of state -SP

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

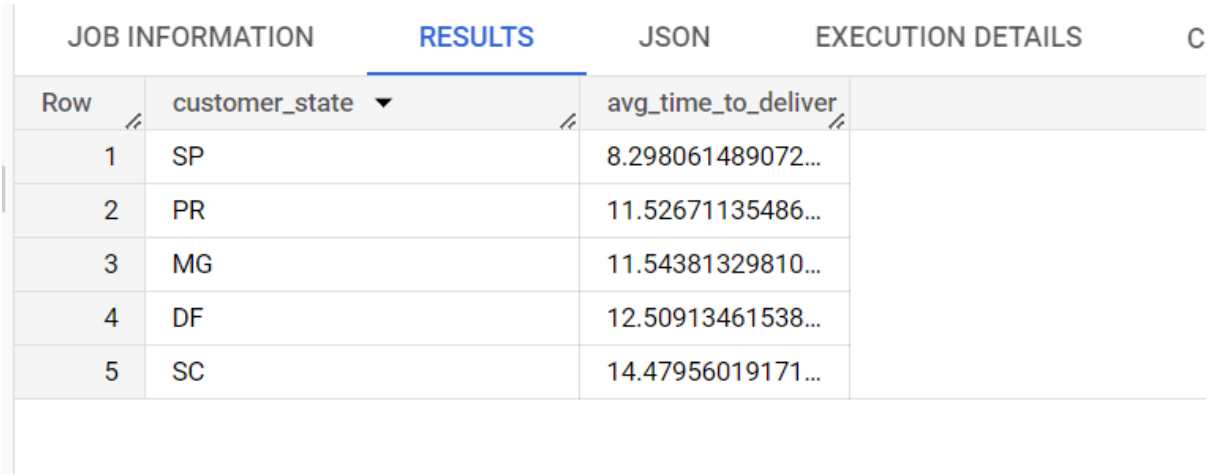
--top 5 states ordered by delivery time

```
select c.customer_state,
avg(date_diff(order_delivered_customer_date, order_purchase_timestamp,day)) as
avg_time_to_deliver
from `target_sql.orders` o
join `target_sql.customers` c on o.customer_id = c.customer_id
group by c.customer_state
order by avg_time_to_deliver desc
limit 5
```

Row	customer_state	avg_time_to_deliver
1	RR	28.97560975609...
2	AP	26.73134328358...
3	AM	25.98620689655...
4	AL	24.04030226700...
5	PA	23.31606765327...

----bottom 5 states ordered by delivery time

```
select c.customer_state,
avg(date_diff(order_delivered_customer_date, order_purchase_timestamp,day)) as
avg_time_to_deliver
from `target_sql.orders` o
join `target_sql.customers` c on o.customer_id = c.customer_id
group by c.customer_state
order by avg_time_to_deliver
limit 5
```



JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | C

Row	customer_state	avg_time_to_deliver
1	SP	8.298061489072...
2	PR	11.52671135486...
3	MG	11.54381329810...
4	DF	12.50913461538...
5	SC	14.47956019171...

Inference: We can infer that SP state has lowest delivery time and RR state takes lot of time to deliver.

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

```
select c.customer_state,
avg(date_diff(order_estimated_delivery_date, order_delivered_customer_date, day)) as
avg_diff_estimated_delivery
from `target_sql.orders` o
join `target_sql.customers` c on o.customer_id = c.customer_id
group by c.customer_state
order by avg_diff_estimated_delivery desc
limit 5
```

JOB INFORMATION | RESULTS | JSON | EXECUTION DETAIL

Row	customer_state	avg_diff_estimated_d
1	AC	19.76250000000...
2	RO	19.13168724279...
3	AP	18.73134328358...
4	AM	18.60689655172...
5	RR	16.41463414634...

Inference: We can infer fastest delivery as compared to estimated delivery is in AC state.