# **BUSINESS CASE STUDY**

**Target SQL** 

- Q.1) Import the dataset and do usual exploratory analysis steps like checking the structure & characteristics of the dataset.
- 1. Data type of columns in a table

Ans:

Row	table_schema	table_name	column_name	data_type
1	Target_Dataset	order_items	order_id	STRING
2	Target_Dataset	order_items	order_item_id	INT64
3	Target_Dataset	order_items	product_id	STRING
4	Target_Dataset	order_items	seller_id	STRING
5	Target_Dataset	order_items	shipping_limit_date	TIMESTAMP
6	Target_Dataset	order_items	price	FLOAT64
7	Target_Dataset	order_items	freight_value	FLOAT64
8	Target_Dataset	sellers	seller_id	STRING
9	Target_Dataset	sellers	seller_zip_code_prefix	INT64
10	Target_Dataset	sellers	seller_city	STRING

2. Time period for which the data is given

```
SELECT Min(Extract (date FROM order_purchase_timestamp)) AS
    Orders_Start_Time_Period,
    Max(Extract (date FROM order_purchase_timestamp)) AS
    Orders_End_Time_Period
FROM `target_dataset.orders`;
```

Row	Orders_Star	Orders_End
1	2016-09-04	2018-10-17

### 3. Cities and States covered in the dataset

Ans:

#### **Distinct Cities**

```
SELECT seller_city AS city
FROM `Target_Dataset.sellers`
UNION DISTINCT
SELECT customer_city AS city
FROM `Target_Dataset.customers`;
```

Row	11	city
1		acu
2		ico
3		ipe
4		ipu
5		ita
6		itu
7		jau
8		luz
9		poa
10		uba

#### **Distinct States**

```
SELECT seller_state AS state
FROM `target_dataset.sellers`
UNION DISTINCT
SELECT customer_state AS state
FROM `target_dataset.customers`;
```

Row	state //
1	AC
2	AM
3	BA
4	CE
5	DF
6	ES
7	GO
8	MA
9	MG
10	MS

# Q.2) In-depth Exploration:

**1.** Is there a growing trend on e-commerce in Brazil? How can we describe a complete scenario? Can we see some seasonality with peaks at specific months?

```
        Row
        purchase_y...
        total_no_of_...

        1
        2018
        54011

        2
        2017
        45101

        3
        2016
        329
```

Row	purchase_month	total_no_of
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

2. What time do Brazilian customers tend to buy (Dawn, Morning, Afternoon or Night)?

Ans:

SELECT

```
CASE
WHEN Extract(hour FROM order_purchase_timestamp) >=00 AND Extract(ho
ur FROM order_purchase_timestamp) <=05 THEN "dawn"
WHEN Extract(hour FROM order_purchase_timestamp) >=06 AND Extract(ho
ur FROM order_purchase_timestamp) <=11 THEN "morning"
WHEN Extract(hour FROM order_purchase_timestamp) >=12 AND Extract(ho
ur FROM order_purchase_timestamp) <=17 THEN "afternoon"
WHEN Extract(hour FROM order_purchase_timestamp) >=18 AND Extract(ho
ur FROM order_purchase_timestamp) <=23 THEN "night"
end AS buy_time,
Count(order purchase timestamp) AS order count</pre>
```

 Row
 buy\_time
 order\_count

 1
 Afternoon
 38361

 2
 Night
 34100

 3
 Morning
 22240

 4
 Dawn
 4740

FROM `target dataset.orders`

ORDER BY order count DESC;

GROUP BY buy time

- Q.3) Evolution of E-commerce orders in the Brazil region:
- 1. Get month on month orders by region, states

```
select distinct
customer_state,
format_datetime("%B", order_purchase_timestamp) as purchase_month,
count(order_id) over(partition by customer_state order by extract(mo
nth from order_purchase_timestamp)) as order_count
from Target_Dataset.orders as o
join Target Dataset.customers as c on o.customer id=c.customer id;
```

Row	customer_state	purchase_month	order_count
1	AC	January	8
2	AC	February	14
3	AC	March	18
4	AC	April	27
5	AC	May	37
6	AC	June	44
7	AC	July	53
8	AC	August	60
9	AC	September	65
10	AC	October	71

#### 2. How are customers distributed in Brazil

```
SELECT c1.customer_state, Count(o1.order_id) AS order_count
FROM `Target_dataset.orders` AS o1
INNER JOIN `Target_dataset.customers` AS c1
ON o1.customer_id = c1.customer_id
GROUP BY c1.customer_state
ORDER BY order count DESC;
```

Row	customer_state	order_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

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

```
WITH
cost 2017 AS
(SELECT
EXTRACT (MONTH FROM of order_purchase_timestamp) AS purchase_month_2017,
ROUND (SUM (oil.price), 2) AS sum_of_price_by_month_2017,
FROM `Target Dataset.orders` AS o1
INNER JOIN `Target_Dataset.order_items` AS oil
ON ol.order_id = oil.order_id
WHERE EXTRACT (YEAR FROM ol order purchase timestamp) = 2017 AND EXTRACT (MON
TH FROM ol.order purchase timestamp) <= 8
GROUP BY purchase month 2017),
cost 2018 AS
(SELECT
EXTRACT (MONTH FROM ol order purchase timestamp) AS purchase month 2018,
ROUND (SUM (oil.price), 2) AS sum of price by month 2018,
FROM `Target Dataset.orders` AS o1
INNER JOIN `Target_Dataset.order_items` AS oil
ON ol.order_id = oil.order_id
WHERE EXTRACT (YEAR FROM ol order purchase timestamp) = 2018 AND EXTRACT (MON
TH FROM ol.order purchase timestamp) <= 8
GROUP BY purchase month 2018)
SELECT *
ROUND((cost_2018.sum_of_price_by_month_2018 -
 cost 2017.sum of price by month 2017), 2) AS cost diff,
ROUND((((cost_2018.sum_of_price_by_month_2018
cost_2017.sum_of_price_by_month_2017) / cost_2017.sum_of_price_by_month_20
17) * 100), 2) AS percentage_increase
FROM cost_2017
INNER JOIN cost_2018
ON cost 2017 purchase month 2017 = cost 2018 purchase month 2018
ORDER BY percentage_increase DESC;
```

Row	purchase_month_2017	sum_of_price_by_month_2017	purchase_month_2018	sum_of_price_by_month_2018	cost_diff	percentage_increase
1	1	120312.87	1	950030.36	829717.49	689.63
2	2	247303.02	2	844178.71	596875.69	241.35
3	4	359927.23	4	996647.75	636720.52	176.9
4	3	374344.3	3	983213.44	608869.14	162.65
5	6	433038.6	6	865124.31	432085.71	99.78
6	5	506071.14	5	996517.68	490446.54	96.91
7	7	498031.48	7	895507.22	397475.74	79.81
8	8	573971.68	8	854686.33	280714.65	48.91

2. Mean & Sum of price and freight value by customer state

Ans:

Row	customer_state //	average_price //	sum_price	average_freight_value	sum_freight_value
1	MT	148.3	156453.53	28.17	29715.43
2	MA	145.2	119648.22	38.26	31523.77
3	AL	180.89	80314.81	35.84	15914.59
4	SP	109.65	5202955.05	15.15	718723.07
5	MG	120.75	1585308.03	20.63	270853.46
6	PE	145.51	262788.03	32.92	59449.66
7	RJ	125.12	1824092.67	20.96	305589.31
8	DF	125.77	302603.94	21.04	50625.5
9	RS	120.34	750304.02	21.74	135522.74
10	SE	153.04	58920.85	36.65	14111.47

# Q.5) Analysis on sales, freight and delivery time

1. Calculate days between purchasing, delivering and estimated delivery

```
select
order_id,
datetime_diff(order_delivered_customer_date,order_purchase_timestamp
, day) as purchase_to_delivery_days,
datetime_diff(order_estimated_delivery_date,order_delivered_customer
_date, day) as delivery_to_est_delivery_days,
datetime_diff(order_estimated_delivery_date,order_purchase_timestamp
, day) as purchase_to_est_delivery_days
from Target_Dataset.orders
where order_delivered_customer_date is not null;
```

Row	order_id	purchase_to_delivery_days	delivery_to_est_delivery_days	purchase_to_est_delivery_days
1	1950d777989f6a877539f5379	30	-12	17
2	2c45c33d2f9cb8ff8b1c86cc28	30	28	59
3	65d1e226dfaeb8cdc42f66542	35	16	52
4	635c894d068ac37e6e03dc54e	30	1	32
5	3b97562c3aee8bdedcb5c2e45	32	0	33
6	68f47f50f04c4cb6774570cfde	29	1	31
7	276e9ec344d3bf029ff83a161c	43	-4	39
8	54e1a3c2b97fb0809da548a59	40	-4	36
9	fd04fa4105ee8045f6a0139ca5	37	-1	35
10	302bb8109d097a9fc6e9cefc5	33	-5	28

#### 2. Create columns:

- time\_to\_delivery = order\_purchase\_timestamporder\_delivered\_customer\_date
- diff\_estimated\_delivery = order\_estimated\_delivery\_dateorder\_delivered\_customer\_date

```
SELECT Datetime_diff(order_delivered_customer_date, order_purchase_t
imestamp,
          hour) AS
          time_to_delivery,
          Datetime_diff(order_estimated_delivery_date,
               order_delivered_customer_date, hour
          )
          AS diff_estimated_delivery
FROM target_dataset.orders
WHERE order_delivered_customer_date_IS_NOT_NULL;
```

Row	time_to_delivery //	diff_estimated_delivery //
1	168	1088
2	722	-310
3	743	681
4	181	1065
5	262	989
6	853	397
7	565	228
8	311	-133
9	309	298
10	173	24

3. Group data by state, take mean of freight\_value, time\_to\_delivery, diff\_estimated\_delivery

```
WITH cte_o_oi_c
     AS (SELECT c.customer state,
                 oi.freight value,
                 Datetime diff(o.order delivered customer date,
                 o.order_purchase_timestamp, hour)
                       AS time to delivery
                 Datetime diff(o.order estimated delivery date,
                 o.order_delivered_customer_date,
                 hour) AS diff_estimated_delivery
         FROM
                 target_dataset.orders AS o
                 JOIN target_dataset.order_items AS oi
                   ON o.order_id = oi.order_id
                 JOIN target_dataset.customers AS c
                   ON o.customer_id = c.customer_id
         WHERE order_delivered_customer_date IS NOT NULL)
SELECT customer_state,
       Round(Avg(freight_value), 2)
                                                 AS mean_freight_value,
       Round (Avg (freight_value), 2)

AS mean_freight_value,

Round (Avg (time_to_delivery), 2)

AS mean_time_to_delivery,
       Round(Avg(diff_estimated_delivery), 2) AS mean_diff_estimated_delive
ry
FROM
       cte_o_oi_c
GROUP BY customer_state;
```

		_		
Row	customer_state	mean_freight_value //	mean_time_to_delivery	mean_diff_estimated_delivery
1	RJ	20.91	363.06	271.04
2	MG	20.63	287.11	302.91
3	SC	21.51	359.53	260.55
4	SP	15.11	208.87	251.89
5	GO	22.56	369.18	277.89
6	RS	21.61	364.03	321.95
7	BA	26.49	461.45	246.57
8	MT	28.0	430.56	333.06
9	SE	36.57	514.72	223.46
10	PE	32.69	438.19	305.96

# 4. Sort the data to get the following:

4.1 Top 5 states with highest/lowest average freight value - sort in desc/asc limit 5

Ans:

#### Top 5 States with hightest average freight value

```
WITH mean cte
AS
  (WITH cte o oi c
AS
  (
        SELECT c.customer state,
               oi.freight value,
               datetime_diff(o.order_delivered_customer_date, o.order_p
urchase timestamp, hour)
                             AS time to delivery,
               datetime diff(o.order estimated delivery date, o.order d
elivered customer date, hour) AS diff estimated delivery
        FROM target dataset orders
                             AS o
        JOIN target_dataset.order_items
                             AS oi
               o order id=oi order id
               target dataset customers AS c
               o.customer id=c.customer id
        WHERE order_delivered_customer_date IS NOT NULL)
         customer state,
  SELECT
          round(avg(freight value), 2)
                                                  AS mean freight value,
          round(avg(time to delivery), 2)
                                                 AS mean time to delive
ry,
          round(avg(diff_estimated_delivery), 2) AS mean_diff_estimated
_delivery
 FROM
          cte o oi c
  GROUP BY customer state)
       customer state,
SELECT
        mean freight value
FROM
        mean cte
ORDER BY mean freight value DESC
LIMIT
```

Row	customer_state	mean_freigh
1	PB	43.09
2	RR	43.09
3	RO	41.33
4	AC	40.05
5	PI	39.12

#### Top 5 States with lowest average freight value

```
WITH mean cte
  (WITH cte o oi c
AS
  (
         SELECT c.customer state,
                oi.freight_value,
                datetime diff(o order delivered customer date, o order purc
hase timestamp, hour)
                           AS time_to_delivery
                datetime_diff(o.order_estimated_delivery_date, o.order_deli
vered_customer_date, hour) AS diff_estimated_delivery
         FROM
                target_dataset.orders
                           AS o
                target_dataset.order_items
         JOIN
                           AS oi
         ON
                o.order_id=oi.order_id
         JOIN
                target_dataset.customers AS c
                o.customer id=c.customer id
         WHERE order delivered customer date IS NOT NULL)
  SELECT customer state,
           round(avg(freight value), 2)
                                                  AS mean freight value,
           round(avg(time_to_delivery), 2)
                                                  AS mean_time_to_delivery,
           round(avg(diff_estimated_delivery), 2) AS mean_diff_estimated_de
livery
 FROM
          cte_o_oi c
  GROUP BY customer state)
SELECT customer state,
        mean freight value
FROM
        mean cte
ORDER BY mean freight value
LIMIT
        5;
```

Row	customer_state	mean_freigh
1	SP	15.11
2	PR	20.47
3	MG	20.63
4	RJ	20.91
5	DF	21.07

4.2 Top 5 states with highest/lowest average time to delivery

Ans:

#### Top 5 States with highest average time to delivery in Hrs

```
with mean cte as
(with cte o oi c as
(select
c.customer state,
oi.freight value,
datetime diff(o.order delivered customer date, o.order purchase t
imestamp, hour) as time to delivery in hrs,
datetime diff(o.order estimated delivery date, o.order delivered
customer date, hour) as diff estimated delivery in hrs
from Target Dataset.orders as o
join Target Dataset order items as oi on o order id-oi order id
join Target Dataset customers as c on o customer id-c customer id
where order delivered customer date is not null)
select
customer state,
round(avg(freight value), 2) as mean freight value,
round(avg(time to delivery in hrs), 2) as mean time to delivery i
round(avg(diff_estimated_delivery_in_hrs), 2) as mean_diff_estima
ted delivery in hrs
from cte o oi c
group by customer state)
select
customer_state,
mean_time_to_delivery_in_hrs
from mean cte
order by mean time to delivery in hrs desc
limit 5;
```

Row	customer_state	10	mean_time_to_delivery_in_hrs
1	RR		676.98
2	AP		676.46
3	AM		632.85
4	AL		587.23
5	PA		569.6

```
with mean cte as
(with cte_o_oi_c as
(select
c.customer state,
oi.freight_value,
datetime diff(o.order delivered customer date, o.order purchase t
imestamp, hour) as time to delivery in hrs,
datetime diff(o order estimated delivery date, o order delivered
customer date, hour) as diff estimated delivery in hrs
from Target Dataset.orders as o
join Target Dataset order items as oi on o order id=oi order id
join Target Dataset customers as c on o customer id-c customer id
where order delivered customer date is not null)
select
customer state,
round(avg(freight value), 2) as mean freight value,
round(avg(time to delivery in hrs), 2) as mean time to delivery i
n hrs,
round(avg(diff estimated delivery in hrs), 2) as mean diff estima
ted delivery in hrs
from cte o oi c
group by customer state)
select
customer state,
mean time to delivery in hrs
from mean cte
order by mean time to delivery in hrs
limit 5;
```

Row	customer_state	11	mean_time_to_delivery_in_hrs
1	SP		208.87
2	PR		286.24
3	MG		287.11
4	DF		310.52
5	SC		359.53

4.3 Top 5 states where delivery is really fast/ not so fast compared to estimated date

Ans:

## Top 5 States where delivery is really fast compared to estimated date

```
WITH mean cte
  (WITH cte_o_oi_c
AS
         SELECT c.customer state,
               oi.freight value,
               datetime_diff(o.order_delivered_customer_date, o.order_purc
hase timestamp, hour)
                          AS time to delivery in hrs,
               datetime diff(o.order estimated delivery date, o.order deli
vered customer date, hour) AS diff estimated delivery in hrs
         FROM target_dataset.orders
                          AS o
         JOIN target_dataset.order_items
                          AS oi
         ON
               o.order_id=oi.order_id
               target_dataset.customers AS c
         JOIN
               o.customer_id=c.customer_id
        WHERE order delivered_customer_date IS NOT NULL)
  SELECT customer state,
          round(avg(freight value), 2)
                                                        AS mean freight va
lue,
          round(avg(time to delivery in hrs), 2)
                                                        AS mean time to de
livery_in_hrs,
          round(avg(diff estimated delivery in hrs), 2) AS mean diff estim
ated_delivery_in_hrs
 FROM cte_o_oi_c
 GROUP BY customer state)
SELECT customer state,
        mean diff estimated delivery in hrs
FROM
       mean cte
ORDER BY mean diff estimated delivery in hrs DESC
LIMIT
```

Row	customer_state	mean_diff_estimated_delivery_in_hrs
1	AC	487.59
2	RO	463.77
3	AM	460.97
4	AP	426.01
5	RR	422.43

## Q.6) Payment type analysis:

1. Month over Month count of orders for different payment types

Ans:

```
SELECT
DISTINCT EXTRACT(MONTH FROM o1.order_purchase_timestamp) AS purchase
_month,
p1.payment_type,
COUNT(o1.order_id) OVER (PARTITION BY EXTRACT(MONTH FROM o1.order_pu
rchase_timestamp) ORDER BY p1.payment_type) AS order_count

FROM `Target_Ecommerce.orders` AS o1
INNER JOIN `Target_Ecommerce.payments` AS p1
ON o1.order_id = p1.order_id
ORDER BY purchase month, order count DESC;
```

purchase_m	payment_type	order_count
1	voucher	8413
1	debit_card	7936
1	credit_card	7818
1	UPI	1715
2	voucher	8838
2	debit_card	8414
2	credit_card	8332
2	UPI	1723
3	voucher	10349
3	debit_card	9758
	1 1 1 1 2 2 2 2 2 2	1 voucher 1 debit_card 1 credit_card 1 UPI 2 voucher 2 debit_card 2 credit_card 2 UPI 3 voucher

2. Distribution of payment installments and count of orders

```
SELECT
DISTINCT pl.payment_installments,
COUNT(pl.order_id) AS order_count

FROM `Target_Ecommerce.payments` AS pl
GROUP BY pl.payment_installments
ORDER BY order count DESC;
```

Row	payment_installments //	order_count //
1	1	52546
2	2	12413
3	3	10461
4	4	7098
5	10	5328
6	5	5239
7	8	4268
8	6	3920
9	7	1626
10	9	644

# **INSIGHTS**

- 1. As per this dataset Target has operations in almost 4196 cities situated in 26 States.
- 2. There is a growing trend on e-commerce in Brazil, as per the analysis on yearly basis the number of orders have grown enormously since 2016 to 2018.
- 3. Highest number of orders are placed in August, followed by May and July, these months fall under winter season.
- 4. Brazilian customers tend to buy in the Afternoon and Night, some customers also buy in Morning but the ratio is not that good compared to Afternoon and Night, customers avoid to buy in the dawn.
- 5. After analysing the data I found out that the count of orders has grown each and every passing month throughout the years 2016, 2017 and 2018 gradually. This depicts that the customers in Brazil are interested in buying on e-commerce platforms.
- 6. 40% of the orders in Brazil are from Sao Paulo, 12% are from Rio de Janeiro and 11% from Minas Gerais. These 3 States combined has 63% orders and rest of the orders are from remaining states.
- 7. Highest increase in cost of orders is in the month January 2018 which is 600% compared to January 2017, followed by February, March and April with more than 150% increase in cost of orders.
- 8. Top 5 States with highest average freight value are State of Paraiba, State of Roraima, State of Rondonia, Acre and Piaui. Top among these is State of Paraiba and Roraima with highest mean freight value of 43.09
- 9. Top 5 States with lowest average freight value are Sau Paulo, State of Parana, Minas Gerais, Rio de Janeiro and Federal District. Top among these is Sau Paulo with lowest mean freight value of 15.11.
- 10. Roraima and Amapa are the states where time taken for delivery of the goods is maximum whereas, Sau Paulo is the state where minimum time is taken for delivery of the goods.
- 11. Maximum number of orders are places using Vouchers followed by Debit cards and Credit cards, seems like UPI payment option is not appreciated by Brazilian customers.

## **RECOMMENDATIONS**

- 1. As the highest number of orders are placed in August, May and July which fall in winter, Target should focus more on these months.
- 2. Brazilian customers buy more in the afternoon and night as compared to morning and dawn, as it is working well for Target, they should start triggering the customers through offers on the products in the Night and Afternoon to increase their sales even to the next level.
- 3. Sau Paulo, Rio de Janeiro and Minas Gerais are the states with maximum number of orders as these are highly populated states and people here are interested in e-commerce but Target should also focus on some moderately populated states as Bahia, Rio Grande do Sul, Parana as this states can be their next big market to capture.
- 4. Looking at the increase in cost of orders from the years 2017 and 2018, seems like 1<sup>st</sup> quarter of year looks good in terms of increasing the cost of orders, Target should continue as it is for the 1<sup>st</sup> quarter months as it is working pretty well for them.
- 5. Customers mostly use vouchers, debit and credit cards as payment options, target should also focus on UPI payment methods as these are more secure and easy to use. Smartphones are pretty common these days even in small cities or towns, Target focusing on UPI payments and giving vouchers or discount offers with UPI payment can attract a lot of customers from small cities as well as small towns.
- 6. Target should also focus on high mean average freight value as it is too much in states like State of Paraiba, State of Roraima, State of Rondonia, Acre and Piaui. Focusing on reducing this cost can save them a lot of money and will also benefit the organisation in increasing the profits even more.
- 7. Target can also focus on reducing the time taken for delivery of goods in the states like Roraima and Amapa. States like Sau Paulo, Rio de Janeiro and Mina Gerais which are pretty big and populated states should be provided with one day delivery option for that logistics network should be strong.