# **Business Case: Target SQL**

# **About Target:**

Target is one of the world's most recognized brands and one of America's leading retailers. Target makes itself a preferred shopping destination by offering outstanding value, inspiration, innovation and an exceptional guest experience that no other retailer can deliver.

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

```
Query:
```

```
SELECT DATA_TYPE,

column name

from Target_sql1234.INFORMATION_SCHEMA. COLUMNS

where table_name = 'customers'
```

#### Output:

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

2. Time period for which the data is given

```
Query:
```

```
Select
```

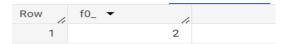
```
date_diff(max, min, year)
```

from

#### (select

```
extract(date from (min(order_purchase_timestamp))) as min1,
extract (date from (max(order_purchase_timestamp))) as max1 from `Target_sql1234.orders`)
x;
```

## Output



3. Cities and States of customers ordered during the given period

## Query:

#### **SELECT**

```
customer_city, customer_state, customer_unique_id
from `Target_sql1234.customers`;
```

## Output

Row	customer_city ▼	customer_state ▼	customer_unique_id ▼
1	acu	RN	fcb003b1bdc0df64b4d065d9b
2	acu	RN	46824822b15da44e983b021d
3	acu	RN	b6108acc674ae5c99e29adc10
4	ico	CE	402cce5c0509000eed9e77fec
5	ico	CE	6ba00666ab7eada5ceec279b2
6	ico	CE	796a0b1a21f597704057184a1
7	ico	CE	05d1d2d9f0161c5f397ce7fc77
8	ico	CE	c34585a0276ecc5e4fb03de75
9	ico	CE	01a4fe5fc00bbdb0b0a4af5a53
10	ico	CE	8f399f3b7ace8e6245422c9e1f

## **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?

```
Query:
SELECT
 Month,
  Year,
 total_orders_made
FROM
 (SELECT
  Extract(MONTH FROM order_purchase_timestamp) AS Month,
  Extract(YEAR FROM order_purchase_timestamp) AS Year,
  COUNT(*) AS total_orders_made
  FROM
  `Target_sql1234.orders`
  GROUP BY
  Year,
 Month
  order by Year, Month) t1;
```

Row	Month ▼	Year ▼	total_orders_made
1	9	2016	4
2	10	2016	324
3	12	2016	1
4	1	2017	800
5	2	2017	1780
6	3	2017	2682
7	4	2017	2404
8	5	2017	3700
9	6	2017	3245
10	7	2017	4026

#### Graph deduced:



## Insights

We can conclude with the help of this graph that there is not an increasing trend in orders placed by Brazilian customers, the trend is rather fluctuating to a point. There are variations in the growing pattern. It includes both increases and decreases (a volatile trend). The trend was at peak once then it took a big dip. Also, there is no seasonality in the trend.

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

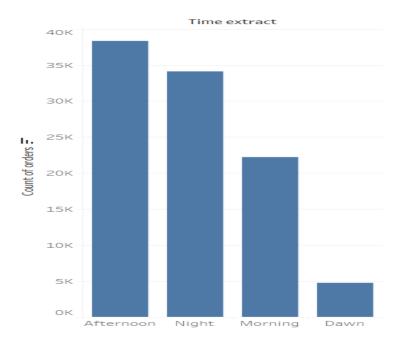
#### **SELECT**

#### 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 'Morning'
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 as Time_of_purchase,
count(*) purchase_count
from `Target_sql1234.orders`
group by Time_of_purchase
order by purchase_count desc;
```

Row	Time_of_purchase ▼	purchase_count ▼
1	Afternoon	38135
2	Night	28331
3	Morning	27733
4	Dawn	5242

## Graph Deduced:



## Insight

As we can see from the graph and the data extracted via the query that Brazilian customers tend to buy more during the afternoon followed by night. There could be many possible reasons for this trend to happen. People could have more free time in afternoon and then in Night after work. We can also see that there are very few numbers in dawn because of obvious reasons i.e limited availability, these are usually the sleeping hours and many more.

## **Evolution of E-commerce orders in the Brazil region:**

1. Get month on month orders by states

## Query:

```
Select
Total_orders,
year,
month,
customer_state
from
(select
Count(o.order_id) as Total_orders,
extract(year from o.order_purchase_timestamp) as year,
extract(month from o.order_purchase_timestamp) as month,
c.customer_state
from `Target_sql1234.orders` o
Join `Target_sql1234.customers` c on o.customer_id = c.customer_id
Group by
              Extract(year from
                                      o.order_purchase_timestamp),
                                                                      extract(month
                                                                                      from
o.order_purchase_timestamp), c.customer_state)
order by year, month;
```

#### Output:

Row	Total_orders ▼	year ▼	month ▼	customer_state ▼
1	1	2016	9	RR
2	1	2016	9	RS
3	2	2016	9	SP
4	113	2016	10	SP
5	24	2016	10	RS
6	56	2016	10	RJ
7	3	2016	10	MT
8	9	2016	10	GO
9	40	2016	10	MG
10	8	2016	10	CE

## Insights

we can see that the state that placed the maximum orders month over month include SP with total order of 654 in month of Feb,2017. The total orders are increasing and decreasing over the months. It keeps on fluctuating. The overall trend of orders made is fluctuating and we can safely conclude that here as well. As we can see, the number of orders increased and then took a dip and then decreased. SP has been constantly placing more orders.

Distribution of customers across the states in Brazil

# Query:

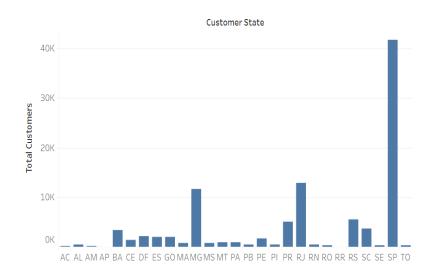
```
select
```

```
customer_state,
count(*) as total_customers
from Target_sql1234.customers
group by customer_state
order by total_customers desc;
```

## Output

Row	customer_state ▼	total_customers 🔻
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

# Graph deduced:



#### Insights

We can see that the maximum orders come from the state of SP in Brazil. It is one of the most populated state in Brazil and hence the customer base is quite large. Also, this state is the economic hub in Brazil thereby making it easier for all economic activities, easy logistics etc.

Impact on Economy: Analyze the money movement 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) - You can use "payment value" column in payments table

```
Query:
SELECT
(ROUND((Sum(Case
When extract(year from o.order_purchase_timestamp) = 2018 and extract(month from
o.order_purchase_timestamp) Between 1 and 8 then p.payment_value
Else 0
End) -
Sum(Case
When Extract(year from o.order_purchase_timestamp) = 2017 and extract(month from
o.order_purchase_timestamp) between 1 and 8 then p.payment_value
Else 0
End)) /
Sum(Case
When Extract(year from o.order_purchase_timestamp) = 2017 and extract (month from
o.order_purchase_timestamp) between 1 and 8 then p.payment_value
ELSE 0
End) * 100),2) as percentage_increase_in_cost
from
`Target_sql1234.orders` o
Join
```

```
`Target_sql1234.payments` p on
o.order_id = p.order_id
where
(Extract
           (Year
                  from
                         o.order purchase timestamp) =
                                                            2017
                                                                   AND
                                                                         Extract(month
                                                                                         from
o.order_purchase_timestamp) between 1 and 8)
OR
(extract(year
                from
                       o.order_purchase_timestamp)
                                                           2018
                                                                        extract(month
                                                                                         from
                                                                  and
o.order_purchase_timestamp) between 1 and 8);
Output
         perc..._field_1 ▼ perc..._field_2

137.0
Row
```

## Insights

We can see that there is about 137% increase in the cost of orders from 2017 to 2018 which is a significant increase in cost during this period. There can be multiple factors contributing and some of those maybe market conditions changing or due to inflation

Mean & Sum of price and freight value by customer state

```
Query:

select

customer_state,

ROUND(AVG(freight_value),2) as mean_freight_value,

ROUND(Avg(Price),2) as mean_price,

ROUND(sum(freight_value),2) as Sum_freight_value,

ROUND(Sum(Price),2) as Sum_price,

From

Target_sql1234.orders o

Join
```

`Target\_sql1234.order\_items` oi

```
on o.order_id = oi.order_id
join
`Target_sql1234.customers` c on
o.customer_id = c.customer_id
group by
customer_state
order by customer_state;
```

Row	customer_state ▼	mean_freight_value	mean_price ▼	Sum_freight_value	Sum_price ▼
1	AC	40.07	173.73	3686.75	15982.95
2	AL	35.84	180.89	15914.59	80314.81
3	AM	33.21	135.5	5478.89	22356.84
4	AP	34.01	164.32	2788.5	13474.3
5	BA	26.36	134.6	100156.68	511349.99
6	CE	32.71	153.76	48351.59	227254.71
7	DF	21.04	125.77	50625.5	302603.94
8	ES	22.06	121.91	49764.6	275037.31
9	GO	22.77	126.27	53114.98	294591.95
10	MA	38.26	145.2	31523.77	119648.22

## Insights

We can see that the maximum contributor is the state SP and the lowest contributor is RR for both Average and sum of price and freight value.

## Analysis on sales, freight and delivery time

1. Calculate days between purchasing, delivering and estimated delivery.

```
Query:

Select

order_id,

date_diff(order_estimated_delivery_date,order_purchase_timestamp,day)as Estimated_delivery,

date_diff(order_delivered_customer_date,order_purchase_timestamp,day) as Actual_delivery

from `target-sql-388013.Target_sql1234.orders`

where order_status= 'delivered';
```

Row	order_id ▼	Estimated_delivery	Actual_delivery ▼
1	635c894d068ac37e6e03dc54e	32	30
2	3b97562c3aee8bdedcb5c2e45	33	32
3	68f47f50f04c4cb6774570cfde	31	29
4	276e9ec344d3bf029ff83a161c	39	43
5	54e1a3c2b97fb0809da548a59	36	40
6	fd04fa4105ee8045f6a0139ca5	35	37
7	302bb8109d097a9fc6e9cefc5	28	33
8	66057d37308e787052a32828	32	38
9	19135c945c554eebfd7576c73	33	36
10	4493e45e7ca1084efcd38ddeb	33	34

## Insights

We can see that for most of the occurrences the actual delivery is done before the estimated delivery. Target has an efficient network of logistics services and that could be the possible reason for the same. We can also see that planning and forecasting has done effectively to for this to happen and for someday when the case is opposite we can contribute that to bad weather or unviability of certain product etc.

- 2. Find time\_to\_delivery & diff\_estimated\_delivery. Formula for the same given below:
- time\_to\_delivery = order\_delivered\_customer\_date-order\_purchase\_timestamp
- diff\_estimated\_delivery = order\_estimated\_delivery\_date-order\_delivered\_customer\_date

#### Query:

```
Select
order_id,
date_diff(order_estimated_delivery_date,order_delivered_customer_date,day)as
diff_estimated_delivery,
date_diff(order_delivered_customer_date,order_purchase_timestamp,day) as time_to_delivery
from `target-sql-388013.Target_sql1234.orders`
where order_status= 'delivered';
Output-
```

Row	order_id ▼	diff_estimated_delive	time_to_delivery
1	635c894d068ac37e6e03dc54e	1	30
2	3b97562c3aee8bdedcb5c2e45	0	32
3	68f47f50f04c4cb6774570cfde	1	29
4	276e9ec344d3bf029ff83a161c	-4	43
5	54e1a3c2b97fb0809da548a59	-4	40
6	fd04fa4105ee8045f6a0139ca5	-1	37
7	302bb8109d097a9fc6e9cefc5	-5	33
8	66057d37308e787052a32828	-6	38
9	19135c945c554eebfd7576c73	-2	36
10	4493e45e7ca1084efcd38ddeb	0	34

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

```
Query:
```

group by c.customer\_state;

```
select
c.customer_state,
ROUND(Avg(oi.freight_value),2) as Avg_freight_value,
ROUND(Avg(date_diff(o.order_delivered_customer_date,o.order_purchase_timestamp, day)),2) as avg_time_to_delivery,
ROUND(Avg(date_DIFF(o.order_estimated_delivery_date,o.order_delivered_customer_date, day)),2) as avg_diff_estimated_delivery
From
Target_sql1234.orders o
Join
Target_sql1234.order_items oi
on o.order_id = oi.order_id
Join
Target_sql1234.customers c
on o.customer_id = c.customer_id
```

Row	customer_state ▼	Avg_freight_value	avg_time_to_delivery	avg_diff_estimated_
1	MT	28.17	17.51	13.64
2	MA	38.26	21.2	9.11
3	AL	35.84	23.99	7.98
4	SP	15.15	8.26	10.27
5	MG	20.63	11.52	12.4
6	PE	32.92	17.79	12.55
7	RJ	20.96	14.69	11.14
8	DF	21.04	12.5	11.27
9	RS	21.74	14.71	13.2
10	SE	36.65	20.98	9.17

## Insights:

We can see that average estimated delivery and delivery time have a positive correlation with average freight. There is a positive correlation between freight value and delivery time

## 4. Sort the data to get the following:

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

```
Query:
Select
c.customer_state,
ROUND(Avg(oi.freight_value),2) as avg_freight_value
from
Target_sql1234.orders o
Join
Target_sql1234.order_items oi
on o.order_id = oi.order_id
Join
```

Target\_sql1234.customers c

```
on o.customer_id = c.customer_id
group by
c.customer_state
order by
avg_freight_value desc
limit 5;
```

Row	customer_state ▼	avg_freight_value
1	RR	42.98
2	PB	42.72
3	RO	41.07
4	AC	40.07
5	PI	39.15

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

```
Query:
```

 $\verb|c.customer_state|$ 

```
Select
c.customer_state,
\label{eq:round_avg} \mbox{ROUND(Avg(oi.freight\_value),2) as avg\_freight\_value)} \mbox{ avg\_freight\_value}
from
Target_sql1234.orders o
Join
Target_sql1234.order_items oi
on o.order_id = oi.order_id
Join
Target_sql1234.customers c
on o.customer_id = c.customer_id
group by
```

```
order by
avg_freight_value
limit 5;
```

Row	customer_state ▼	avg_freight_value
1	SP	15.15
2	PR	20.53
3	MG	20.63
4	RJ	20.96
5	DF	21.04

Top 5 states with highest average time to delivery

```
Query:
```

Output

```
SELECT
c.customer_state,
ROUND(Avg(DATE_DIFF(o.order_delivered_customer_date, o.order_purchase_timestamp, DAY))) as
avg_time_to_delivery
from
Target_sql1234.orders o
join
Target_sql1234.customers c
on o.customer_id = c.customer_id
group by
\texttt{c.customer\_state}
order by
avg_time_to_delivery desc
limit 5;
```

Row	customer_state ▼	11	avg_time_to_delivery
1	RR		29.0
2	AP		27.0
3	AM		26.0
4	AL		24.0
5	PA		23.0

## Insights

We can see that these states have the highest average days that it took to deliver the order there. With RR(Roraima) being on top. This could be possibly because of its remote location. The logistics route may not be as developed as other populated states in Brazil. This could be true for all other 4 states too as these are the least populated states in BR and have poor infrastructure and logistics

Top 5 states with lowest average time to delivery

```
Query:
```

Output

```
SELECT

c.customer_state,

ROUND(Avg(DATE_DIFF(o.order_delivered_customer_date, o.order_purchase_timestamp, DAY))) as avg_time_to_delivery

from

Target_sql1234.orders o

join

Target_sql1234.customers c

on o.customer_id = c.customer_id

group by

c.customer_state

order by

avg_time_to_delivery

limit 5;
```

Row	customer_state ▼	avg_time_to_delivery
1	SP	8.0
2	MG	12.0
3	PR	12.0
4	DF	13.0
5	SC	14.0

Top 5 states where delivery is really fast compared to estimated date

```
Query:
```

#### Select

```
c.customer_state,
```

```
ROUND(AVG(date_DIFF(o.order_estimated_delivery_date, order_delivered_customer_date, day)))
as avg_time_to_delivery
```

#### FROM

Target\_sql1234.orders o

## join

```
Target_sql1234.customers c on o.customer_id = c.customer_id
```

## group by

c.customer\_state

## order by

avg\_time\_to\_delivery desc

## limit 5;

## Output:

Row	customer_state ▼	avg_time_to_delivery
1	AC	20.0
2	RO	19.0
3	AM	19.0
4	AP	19.0
5	RR	16.0

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

```
Query:
```

```
Select
```

```
c.customer_state,
ROUND(AVG(date_DIFF(o.order_estimated_delivery_date, order_delivered_customer_date, day)))
as avg_time_to_delivery
FROM
Target_sql1234.orders o
join
Target_sql1234.customers c on o.customer_id = c.customer_id
group by
c.customer_state
order by
avg_time_to_delivery asc
```

## Output

limit 5;

Row	customer_state ▼	avg_time_to_delivery
1	AL	8.0
2	MA	9.0
3	SE	9.0
4	SP	10.0
5	BA	10.0

# 5. Payment type analysis:

Month over Month count of orders for different payment types

## Query:

```
SELECT
extract( year from o.order_purchase_timestamp) as order_year,
extract( month from o.order_purchase_timestamp) as order_month,
payment_type,
COUNT(*) as order_count
from `Target_sql1234.orders`o
join
`Target_sql1234.payments` p on
o.order_id=p.order_id
group by
order_year,
order_month,
payment_type
order by
order_year,
order_month;
```

## Output

Row	order_year ▼	order_month ▼	payment_type ▼	order_count ▼
1	2016	9	credit_card	3
2	2016	10	credit_card	254
3	2016	10	UPI	63
4	2016	10	voucher	23
5	2016	10	debit_card	2
6	2016	12	credit_card	1
7	2017	1	credit_card	583
8	2017	1	UPI	197
9	2017	1	voucher	61
10	2017	1	debit_card	9

Count of orders based on the no. of payment instalments

## Query:

```
select
```

```
payment_installments,
count(*) total_no_of_orders
from `Target_sql1234.payments`
group by payment_installments;
Output
```

Row	payment_installment	total_no_of_orders
1	0	2
2	1	52546
3	2	12413
4	3	10461
5	4	7098
6	5	5239
7	6	3920
8	7	1626
9	8	4268
10	9	644

#### Insights:

As we can see that, most of the people is paying one instalment for their orders which is followed by 2 and 3 instalments. People prefer paying for their orders mostly in 3 maximum payments. While, these higher payment instalments i.e. 22, 23 and 24 are least preferred by people and there are only few orders with that number of instalments.

## **Actionable Insights and Recommendations**

- Target should monitor the market trends in Brazil as the trend is quite fluctuating. They should put more marketing efforts and promote in Brazil, for the customers to buy from there.
- They should also focus on behavioural targeting which means tailoring the needs as per the people of Brazil.
- As Afternoon and night hours are the peak hours for shopping. They should plan
  there marketing/promotional offers around that time. Discounts can also be
  offered during that specific time period of afternoon or night to increase the
  sales.
- There were volatile trends throughout in as we compare the state wise data,
   Target should show the tailor-made products as per the different state and their behavioural preferences.
- Target should identify the top selling products and high revenue products and consider going full throttle on both of those to increase the revenue.

- There were regional sales disparity too, so Target should be making more advertising and marketing efforts towards those state which has lesser order volumes.
- Conducting market research to know the reasons behind the lower order volumes and then taking actions accordingly. They can offer special discounts or promotional offers to get the sales up and running in the states of DE, ES GO etc.
- For the increasing overall cost, we need to identify the factors contributing to
  the increase. Target should think about partnering with local vendors or
  negotiate with existing partners and crack up some beneficial deals to reduce
  the cost.
- For the delivery disparity, they can plan to work with logistics services. Considering to partner with local courier service and setting up statewise/region-wise distribution system to lower the difference between actual and estimated delivery. This disparity in delivery times should be reduced.
- Delivering fast to the states where it takes a lot of time will eventually increase the sales from that state as well.
- Target should promote the flexible payment system and encourage customer for this. They might consider partnering with regional financial institutes.