

TARGET BUSINESS CASE

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

1.1Data type of all columns in the "customers" table.

```
2
3 SELECT COLUMN_NAME,DATA_TYPE
4 FROM `Target_Dataset.INFORMATION_SCHEMA.COLUMNS`
5 WHERE TABLE_NAME = 'customers'
```

Query results [SAVE RESULTS](#)

JOB INFORMATION	RESULTS	CHART	JSON	EXECUTION DETAILS	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			

Query: `SELECT COLUMN_NAME,DATA_TYPE
FROM `Target_Dataset.INFORMATION_SCHEMA.COLUMNS`
WHERE TABLE_NAME = 'customers'`

INSIGHTS: We can see in above Output Customer_id, customer_unique_id ,customer_city, customer_state is in **String** :

Means we can conclude that these columns data are comprising of numbers & Text so that's why we are refering it as an String

customer_zip_code_prefix is in Integer form: By this we can conclude that customer addresses's zip code would be an integer e.g. 45666

Recommendation: Since question is asking for Data type therefore, no recommendation here;

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

```

3 SELECT min(order_purchase_timestamp) as Begin_Time,
4        max(order_purchase_timestamp) as End_Time
5 FROM `Target_Dataset.orders`;

```

Query results

JOB INFORMATION		RESULTS	CHART	JSON
Row	Begin_Time	End_Time		
1	2016-09-04 21:15:19 UTC	2018-10-17 17:30:18 UTC		

Query : `SELECT min(order_purchase_timestamp) as Begin_Time,`
`max(order_purchase_timestamp) as End_Time`
`FROM `Target_Dataset.orders`;`

Insights : By this output we can see that Order has been placed between above mentioned range

Recommendation: Target can incur more expenses on marketing, so that people get aware about these products & purchase more

By doing this Number of Orders & Time range of purchasing products both can be increase.

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

```

1
2 SELECT
3     COUNT(DISTINCT c.customer_city) AS unique_cities,
4     COUNT(DISTINCT c.customer_state) AS unique_states
5 FROM `Target_Dataset.customers` c
6 JOIN `Target_Dataset.orders` o
7 ON c.customer_id = o.customer_id
8 WHERE EXTRACT(YEAR FROM TIMESTAMP(o.order_purchase_timestamp)) BETWEEN 2016 AND 2018;
9
10
11

```

Query results

[SAVE RESULTS](#)

JOB INFORMATION		RESULTS	CHART	JSON	EXECUTION DETAILS	EXECUTION GRAPH
Row	unique_cities	unique_states				
1	4119	27				

Query: `SELECT`
`COUNT(DISTINCT c.customer_city) AS unique_cities,`
`COUNT(DISTINCT c.customer_state) AS unique_states`
`FROM `Target_Dataset.customers` c`
`JOIN `Target_Dataset.orders` o ON c.customer_id = o.customer_id`

```
WHERE EXTRACT(YEAR FROM TIMESTAMP(o.order_purchase_timestamp)) BETWEEN 2016 AND 2018;
```

Insights: By the above output we can conclude;

27 States where products have been ordered by customers in which 4,119 No of cities

Recommendation: These No of orders can be increased if Target launch their posters like an advertisement at Bus stand, Metro Station, Movies halls etc where number of people gathered more.

2.In-depth Exploration:

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

```
3 ✓SELECT COUNT(DISTINCT(order_id))as No_of_order,
4 | | | | EXTRACT(year from order_purchase_timestamp) as year
5 FROM`Target_Dataset.orders`
6 GROUP BY 2
7 ORDER BY 2 asc;
```

Query results

JOB INFORMATION		RESULTS	CHART	JSON	EXECUTION DETAILS
Row	No_of_order	year			
1	329	2016			
2	45101	2017			
3	54011	2018			

Query:

```
SELECT COUNT(DISTINCT(order_id))as No_of_order,
      EXTRACT(year from order_purchase_timestamp) as year
FROM`Target_Dataset.orders`
GROUP BY 2
ORDER BY 2 asc;
```

Insights : By the above Output we can conclude that:

In Year 2017 : 44772 No of order has increased as compared to 2016 (growth of 13,609%)

In Year 2018 : 8910 No of order has increased from year 2017 (growth of 16% only) & 53682 from year 2016 (growth of 16317%)

Recommendations: Since % of order purchase ratio has been decreased in Year 2018 ,we would suggest that try to maintain the customers trust by improving your products quality , keep checking customer reviews on each products .

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

```

SELECT COUNT(DISTINCT(order_id))as No_of_order,
       EXTRACT(month from order_purchase_timestamp) as Month,
       EXTRACT(year from order_purchase_timestamp) as Year
FROM `Target_Dataset.orders`
GROUP BY 2,3
ORDER BY 3,2,1 asc;

```

JOB INFORMATION		RESULTS	CHART	JSON
Row	No_of_order	Month	Year	
1	4	9	2016	
2	324	10	2016	
3	1	12	2016	
4	800	1	2017	
5	1780	2	2017	
6	2682	3	2017	
7	2404	4	2017	
8	3700	5	2017	
9	3245	6	2017	

Row	No_of_order	Month	Year	
10	4026	7	2017	
11	4331	8	2017	
12	4285	9	2017	
13	4631	10	2017	
14	7544	11	2017	
15	5673	12	2017	
16	7269	1	2018	
17	6728	2	2018	

Query : `SELECT COUNT(DISTINCT(order_id))as No_of_order,`
`EXTRACT(month from order_purchase_timestamp) as Month,`
`EXTRACT(year from order_purchase_timestamp) as year`
`FROM `Target_Dataset.orders``
`GROUP BY 2,3`
`ORDER BY 3,2,1 asc;`

Insights : From above output we can see that :

Monthly No of orders has been fluctuating & It doesn't create a good impression of Target among customers .

Recommendation: Launch quality products as per season according, so that people can use them according to month season.

2. 3 During what time of the day, do the Brazilian customers mostly place their orders? (Dawn, Morning, Afternoon or Night)

- 0-6 hrs : Dawn
- 7-12 hrs : Mornings
- 13-18 hrs : Afternoon
- 19-23 hrs : Night

```
SELECT COUNT(DISTINCT(order_id)) as No_of_order,
CASE
| WHEN Hour BETWEEN 0 AND 6 THEN 'Dawn'
| WHEN Hour BETWEEN 7 AND 12 THEN 'Mornings'
| WHEN Hour BETWEEN 13 AND 18 THEN 'Afternoon'
| WHEN Hour BETWEEN 19 AND 23 THEN 'Night'
END AS Order_Time
FROM
(SELECT DISTINCT(order_id),
| | | | EXTRACT(hour from order_purchase_timestamp) as Hour
From `Target_Dataset.orders`) a
GROUP BY Order_Time
```

Row	No_of_order	Order_Time
1	27733	Mornings
2	5242	Dawn
3	38135	Afternoon
4	28331	Night

Query: SELECT COUNT(DISTINCT(order_id)) as No_of_order,

```
CASE
| WHEN Hour BETWEEN 0 AND 6 THEN 'Dawn'
| WHEN Hour BETWEEN 7 AND 12 THEN 'Mornings'
| WHEN Hour BETWEEN 13 AND 18 THEN 'Afternoon'
| WHEN Hour BETWEEN 19 AND 23 THEN 'Night'
END AS Order_Time
FROM
(SELECT DISTINCT(order_id),
| EXTRACT(hour from order_purchase_timestamp) as Hour
From `Target_Dataset.orders`) a
GROUP BY Order_Time
```

Insights: By the above output we can say that:

In Dawn time No of orders are less as compared to others time due to various factors such as customers wake up late & No of orders 38135 reflects customers usually prefer afternoon time for buying the product

Recommendations: Target can launch any impressive offers for Dwan time range, so that people can get attracted & purchase more

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

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

```
SELECT c.customer_state,
       COUNT(DISTINCT(o.order_id))as No_of_order,
       EXTRACT(month from o.order_delivered_customer_date) as Month,
       EXTRACT(year from o.order_delivered_customer_date) as year
FROM `Target_Dataset.orders` o
INNER JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id
WHERE o.order_status = 'delivered'AND o.order_delivered_customer_date IS NOT NULL
GROUP BY 1,3,4
```

Row	customer_state	No_of_order	Month	year
1	GO	71	5	2017
2	SP	1454	5	2017
3	RS	201	5	2017
4	BA	129	5	2017
5	MG	414	5	2017
6	MT	41	5	2017
7	RJ	520	5	2017
8	SC	155	5	2017
9	SE	16	5	2017
10	PE	23	4	2017

Query:

```
SELECT c.customer_state,
       COUNT(DISTINCT(o.order_id))as No_of_order,
       EXTRACT(month from o.order_delivered_customer_date) as Month,
       EXTRACT(year from o.order_delivered_customer_date) as year
FROM `Target_Dataset.orders` o
INNER JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id
WHERE o.order_status = 'delivered'AND o.order_delivered_customer_date IS NOT NULL
GROUP BY 1,3,4
```

Insights: As we can see from above Output, No. of orders values are different in each state as per Year 2017.It reflects customers demand of products as per categorised states.

Recommendations: For higher sales, Target can aim for those states where No of orders are less

3.2 How are the customers distributed across all the states?

```
SELECT COUNT(DISTINCT(customer_id)) as No_of_customer,  
       customer_state  
FROM `Target_Dataset.customers`  
GROUP BY 2  
ORDER BY 1 asc;
```

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

Query:

```
SELECT COUNT(DISTINCT(customer_id)) as No_of_customer,  
       customer_state  
FROM `Target_Dataset.customers`  
GROUP BY 2  
ORDER BY 1 asc;
```

Insights: We can see from above output No of customers are very less in RR state as compared to other states.

Recommendations : Target can do survey in each State for knowing the customers demand & By doing this they can launch their products as per customer demand in a state where Numbers of customers are less.

4.Impact on Economy: Analyze the money movement by e-commerce by looking at order prices, freight and others.

4.1 Get the % increase in the cost of orders from year 2017 to 2018 (include months between Jan to Aug only).

```
1
2 WITH CTE AS (
3     SELECT
4         ROUND(SUM(p.payment_value),2) as Cost_of_orders,
5         EXTRACT (year from o.order_purchase_timestamp) as Year
6     FROM `Target_Dataset.orders` o
7     INNER JOIN `Target_Dataset.payments` p
8     ON o.order_id = p.order_id
9     WHERE EXTRACT (year from o.order_purchase_timestamp) IN (2017,2018)
10    AND EXTRACT (month from o.order_purchase_timestamp) BETWEEN 1 AND 8
11    GROUP BY 2)
12 SELECT
13     ((SUM(CASE
14         WHEN CTE.year = 2018 THEN CTE.Cost_of_orders ELSE 0
15     END) -
16     SUM(CASE
17         WHEN CTE.year = 2017 THEN CTE.Cost_of_orders ELSE 0
18     END)) / SUM(CASE
19         WHEN CTE.year = 2017 THEN CTE.Cost_of_orders ELSE 0
20     END)) * 100 AS year_on_year_increase_Percent
21 FROM CTE;
```

Query results

JOB INFORMATION		RESULTS	CHART	JSON	EXECUTION DETAILS
Row	year_on_year_increas				
1	136.9768716466...				

Query: WITH CTE AS (

```
SELECT
    ROUND(SUM(p.payment_value),2) as Cost_of_orders,
    EXTRACT (year from o.order_purchase_timestamp) as Year
FROM `Target_Dataset.orders` o
INNER JOIN `Target_Dataset.payments` p
ON o.order_id = p.order_id
WHERE EXTRACT (year from o.order_purchase_timestamp) IN (2017,2018)
AND EXTRACT (month from o.order_purchase_timestamp) BETWEEN 1 AND 8
GROUP BY 2)
SELECT
    ((SUM(CASE
        WHEN CTE.year = 2018 THEN CTE.Cost_of_orders ELSE 0
    END) -
    SUM(CASE
        WHEN CTE.year = 2017 THEN CTE.Cost_of_orders ELSE 0
    END)) / SUM(CASE
        WHEN CTE.year = 2017 THEN CTE.Cost_of_orders ELSE 0
    END)) * 100 AS year_on_year_increase_Percent
```



```
END)) * 100 AS year_on_year_increase
FROM CTE;
```

Insights: From above output we can see % No of order increament from year 2017 to 2018 is 136. 97, No of order increament could be happen for many reasons like enhance the quality of product, incurred expenses on branding.

Recommendation: This % can increased in upcoming years if Target focused on customer demand only that what actually customer desired in product

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

```
SELECT c.customer_state,
       ROUND(SUM(oi.price),2) AS Total_value,
       ROUND(AVG(oi.price),2) AS Average_value
FROM `Target_Dataset.orders` o
JOIN `Target_Dataset.order_items` oi
ON o.order_id = oi.order_id
JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id
GROUP BY 1
ORDER BY Total_value desc,
         Average_value desc;
```

Row	customer_state	Total_value	Average_value
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

Query : SELECT c.customer_state,

```
       ROUND(SUM(oi.price),2) AS Total_value,
       ROUND(AVG(oi.price),2) AS Average_value
FROM `Target_Dataset.orders` o
JOIN `Target_Dataset.order_items` oi
```

```

ON o.order_id = oi.order_id
JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id
GROUP BY 1
ORDER BY Total_value desc,
         Average_value desc;

```

Insights: From the above output we can see Product values keep varying as per States, This could be happen because of different categorization of products

Recommendation : Target can focus more on quality of products and can try to focus on minimize the cost of product, By this no of customers can increase & sales turnover would get high

4. 3 Calculate the Total & Average value of order freight for each state.

```

3 SELECT c.customer_state,
4        ROUND(SUM(freight_value)) AS Total_freight,
5        ROUND(AVG(freight_value)) AS Average_freight
6 FROM `Target_Dataset.orders` o
7 JOIN `Target_Dataset.order_items` oi |
8 ON o.order_id = oi.order_id
9 JOIN `Target_Dataset.customers` c
10 ON o.customer_id = c.customer_id
11 GROUP BY 1
12 ORDER BY Total_freight asc,
13          Average_freight asc;
14

```

Row	customer_state	Total_freight	Average_freight
1	RR	2235.0	43.0
2	AP	2789.0	34.0
3	AC	3687.0	40.0
4	AM	5479.0	33.0
5	RO	11417.0	41.0
6	TO	11733.0	37.0
7	SE	14111.0	37.0
8	AL	15915.0	36.0
9	RN	18860.0	36.0
10	MS	19144.0	23.0

Query : `SELECT c.customer_state,

 ROUND(SUM(freight_value)) AS Total_freight,
 ROUND(AVG(freight_value)) AS Average_freight
 FROM `Target_Dataset.orders` o
 JOIN `Target_Dataset.order_items` oi
 ON o.order_id = oi.order_id
 JOIN `Target_Dataset.customers` c
 ON o.customer_id = c.customer_id
 GROUP BY 1
 ORDER BY Total_freight asc,
 Average_freight asc;`

Insights : From the above Output we can say that Freight value has been changing as per States,
 It could happened because of longest or shortest distance from the warehouse location.

Recommendation: Target can invest to buy different state locations warehouses so that freight value can decreased

5. Analysis based on sales, freight and delivery time

5.1 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_delivered_customer_date - order_estimated_delivery_date`

```
SELECT DISTINCT(order_id),
    date_diff(order_delivered_customer_date,order_purchase_timestamp,day) as Delivery_time,
    date_diff(order_delivered_customer_date,order_estimated_delivery_date,day) as Diff_estimated_delivery
FROM `Target_Dataset.orders`
WHERE order_status = 'delivered'
AND order_delivered_customer_date IS NOT NULL
ORDER BY 3,2 desc ;
```

Row	order_id	Delivery_time	Diff_estimated_deliv
1	0607f0efea4b566f1eb8f7d3c2...	3	-146
2	c72727d29cde4cf870d569bf6...	6	-139
3	eec7f369423b033e549c02f3c...	20	-134
4	c2bb89b5c1dd978d507284be...	16	-123
5	40dc2ba6f322a17626aac6244...	7	-108
6	1a695d543b7302aa9446c8d5f...	12	-83
7	39e0115911bf404857e14baa7...	11	-82
8	559eea5a72341a4c82dbce988...	13	-77
9	c5132855100a12d63ed4e8ae0...	12	-77
10	38930f76efb00b138f4d632e4d...	11	-77

Query : `SELECT DISTINCT(order_id),
date_diff(order_delivered_customer_date,order_purchase_timestamp,day) as Delivery_time,
date_diff(order_delivered_customer_date,order_estimated_delivery_date,day) as Diff_estimated_delivery
FROM `Target_Dataset.orders`
WHERE order_status = 'delivered'
AND order_delivered_customer_date IS NOT NULL
ORDER BY 3,2 desc ;`

Insights: From above Output we can say that date of delivery of products are not estimated in a proper way.

Recommendations: Target can do focus on estimation date of delivery of products, It would impact directly to customers as if they will received the products exactly on same date as mentioned & By doing this trust can be achieved from customers.

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

```

With Table_1 as
(
    SELECT c.customer_state,
           ROUND(AVG(oi.freight_value),2) as Avg_Freight_Value,
           dense_rank() over(order by ROUND(AVG(oi.freight_value),2) desc) as h_rnk,
           dense_rank() over (order by ROUND(AVG(oi.freight_value),2) asc) as l_rnk
    FROM `Target_Dataset.orders` o

    INNER JOIN `Target_Dataset.customers` c
    ON o.customer_id = c.customer_id

    INNER JOIN `Target_Dataset.order_items` oi
    ON o.order_id = oi.order_id

    GROUP BY 1)

(
    SELECT customer_state,Avg_Freight_Value,
           'Highest_Top_5_State' as Type
    FROM Table_1
    WHERE h_rnk <=5
    ORDER BY h_rnk)

UNION ALL

```

```

(SELECT customer_state,Avg_Freight_Value,
 'Lowest_Low_5_State' as Type
FROM Table_1
WHERE l_rnk <=5
ORDER BY l_rnk)

ORDER BY Avg_Freight_Value desc ;

```

Row	customer_state	Avg_Freight_Value	Type
1	RR	42.98	Highest_Top_5_State
2	PB	42.72	Highest_Top_5_State
3	RO	41.07	Highest_Top_5_State
4	AC	40.07	Highest_Top_5_State
5	PI	39.15	Highest_Top_5_State
6	DF	21.04	Lowest_Low_5_State
7	RJ	20.96	Lowest_Low_5_State
8	MG	20.63	Lowest_Low_5_State
9	PR	20.53	Lowest_Low_5_State
10	SP	15.15	Lowest_Low_5_State

Query: With Table_1 as

```

(SELECT c.customer_state,
       ROUND(AVG(oi.freight_value),2) as Avg_Freight_Value,
       dense_rank() over(order by ROUND(AVG(oi.freight_value),2) desc) as h_rnk,

```

```

        dense_rank() over (order by ROUND(AVG(oi.freight_value),2) asc) as l_rnk
FROM `Target_Dataset.orders` o

INNER JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id

INNER JOIN `Target_Dataset.order_items` oi
ON o.order_id = oi.order_id

GROUP BY 1)

(SELECT customer_state,Avg_Freight_Value,
'Highest_Top_5_State' as Type
FROM Table_1
WHERE h_rnk <=5
ORDER BY h_rnk)

UNION ALL

(SELECT customer_state,Avg_Freight_Value,
'Lowest_Low_5_State' as Type
FROM Table_1
WHERE l_rnk <=5
ORDER BY l_rnk)

ORDER BY Avg_Freight_Value desc;

```

Insights: By the above output we can see the difference in which state products value are high and where 's low

Recommendation : Target can identify the customer demand and charged the values accordingly

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

```

With Table_1 as
(
SELECT c.customer_state,
ROUND(AVG(Timestamp_diff(o.order_delivered_customer_date, o.order_purchase_timestamp,day)),2) as AVG_Delivery_Time,
dense_rank() over(order by AVG(Timestamp_diff(o.order_delivered_customer_date,o.order_purchase_timestamp,day)) desc) as h_rnk,
dense_rank() over(order by AVG(Timestamp_diff(o.order_delivered_customer_date,o.order_purchase_timestamp,day)) asc) as l_rnk
FROM `Target_Dataset.orders` o

INNER JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id

INNER JOIN `Target_Dataset.order_items` oi
ON o.order_id = oi.order_id

GROUP BY 1)

(SELECT customer_state,AVG_Delivery_Time,
'Highest_Top_5_State' as Type
FROM Table_1
WHERE h_rnk <=5
ORDER BY h_rnk)

```

```

UNION ALL

(SELECT customer_state,AVG_Delivery_Time,
'Lowest_Low_5_State' as Type
FROM Table_1
WHERE l_rnk <=5
ORDER BY l_rnk)

ORDER BY AVG_Delivery_Time desc ;

```

Row	customer_state ▼	AVG_Delivery_Time	Type ▼
1	RR	27.83	Highest_Top_5_State
2	AP	27.75	Highest_Top_5_State
3	AM	25.96	Highest_Top_5_State
4	AL	23.99	Highest_Top_5_State
5	PA	23.3	Highest_Top_5_State
6	SC	14.52	Lowest_Low_5_State
7	DF	12.5	Lowest_Low_5_State
8	MG	11.52	Lowest_Low_5_State
9	PR	11.48	Lowest_Low_5_State
10	SP	8.26	Lowest_Low_5_State

Query: With Table_1 as

```

(SELECT c.customer_state,
ROUND(AVG(Timestamp_diff(o.order_delivered_customer_date, o.order_purchase_timestamp,day)),2) as
AVG_Delivery_Time,
dense_rank() over(order by
AVG(Timestamp_diff(o.order_delivered_customer_date,o.order_purchase_timestamp,day)) desc) as h_rnk,
dense_rank() over(order by
AVG(Timestamp_diff(o.order_delivered_customer_date,o.order_purchase_timestamp,day)) asc) as l_rnk
FROM `Target_Dataset.orders` o

INNER JOIN `Target_Dataset.customers` c
ON o.customer_id = c.customer_id

INNER JOIN `Target_Dataset.order_items` oi
ON o.order_id = oi.order_id

GROUP BY 1)

(SELECT customer_state,AVG_Delivery_Time,
'Highest_Top_5_State' as Type
FROM Table_1
WHERE h_rnk <=5
ORDER BY h_rnk)

UNION ALL

(SELECT customer_state,AVG_Delivery_Time,
'Lowest_Low_5_State' as Type

```

```
FROM Table_1
WHERE l_rnk <=5
ORDER BY l_rnk)

ORDER BY AVG_Delivery_Time desc ;
```

Insights: By above Output we can see the Top 5 high states where average delivery time are more & 5 Low states where average delivery time are less, This could happened because customers location distance might be too far or short from Target warehouses location.

Recommendations: Target can open warehouses in a state where delivery time are much high ,so that time can be saved or more delivery can be done on any other states

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

```
With Table_1 as
(
SELECT c.customer_state,
ROUND(AVG(Timestamp_diff(o.order_delivered_customer_date,order_estimated_delivery_date,day)),2) as Diff_Estimated_Delivery_date,
dense_rank() over(order by ROUND(AVG(Timestamp_diff(o.order_delivered_customer_date,order_estimated_delivery_date,day)),2) ) as
h_rnk
FROM `Target_Dataset.orders`o

INNER JOIN `Target_Dataset.customers`c
ON o.customer_id = c.customer_id

INNER JOIN `Target_Dataset.order_items`oi
ON o.order_id = oi.order_id

WHERE o.order_delivered_customer_date IS NOT NULL
AND o.order_status = 'delivered'
GROUP BY 1)

SELECT customer_state,
Diff_Estimated_Delivery_date,
FROM Table_1
WHERE h_rnk <= 5
ORDER BY h_rnk
```

Row	customer_state	Diff_Estimated_Deliy
1	AC	-20.01
2	RO	-19.08
3	AM	-18.98
4	AP	-17.44
5	RR	-17.43

Query: With Table_1 as

```
(SELECT c.customer_state,
ROUND(AVG(Timestamp_diff(o.order_delivered_customer_date,order_estimated_delivery_date,day
)),2) as Diff_Estimated_Delivery_date,
dense_rank() over(order by
ROUND(AVG(Timestamp_diff(o.order_delivered_customer_date,order_estimated_delivery_date,day)),2) ) as
h_rnk
FROM `Target_Dataset.orders`o

INNER JOIN `Target_Dataset.customers`c
ON o.customer_id = c.customer_id

INNER JOIN `Target_Dataset.order_items`oi
ON o.order_id = oi.order_id
```



```
WHERE o.order_delivered_customer_date IS NOT NULL
AND o.order_status = 'delivered'
GROUP BY 1)
```

```
SELECT customer_state,
       Diff_Estimated_Delivery_date,
FROM Table_1
WHERE h_rnk <= 5
ORDER BY h_rnk
```

Insights : From the above Output we can say that estimation of actual date of delivery is not calculated in a proper way or product's delivery might be early of any other reasons

Recommendations : Target can calculate the days of distance of delivery in a proper way, So that customers trust can be achieved.

6. Analysis based on the payments:

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

```
SELECT
    EXTRACT(Year from o.order_purchase_timestamp) as Year,
    EXTRACT(month from o.order_purchase_timestamp) as Month,
    p.payment_type,
    COUNT(DISTINCT(o.order_id)) as No_of_order
FROM `Target_Dataset.orders`o
JOIN `Target_Dataset.payments`p
ON o.order_id = p.order_id
GROUP BY 1,2,3
ORDER BY 1 asc,
        2 asc,
        3 asc;
```

Row	Year	Month	payment_type	No_of_order
1	2016	9	credit_card	3
2	2016	10	UPI	63
3	2016	10	credit_card	253
4	2016	10	debit_card	2
5	2016	10	voucher	11
6	2016	12	credit_card	1
7	2017	1	UPI	197
8	2017	1	credit_card	582
9	2017	1	debit_card	9
10	2017	1	voucher	33

Query : SELECT

```

        EXTRACT(Year from o.order_purchase_timestamp) as Year,
        EXTRACT(month from o.order_purchase_timestamp) as Month,
        p.payment_type,
        COUNT(DISTINCT(o.order_id)) as No_of_order
FROM `Target_Dataset.orders` o
JOIN `Target_Dataset.payments` p
ON o.order_id = p.order_id
GROUP BY 1,2,3
ORDER BY 1 asc,
         2 asc,
         3 asc;

```

Insights: From above Output we can see different type of payment methods customers are using for doing the payment of product.

Recommendation: Target can enable others payments option like NEFT, UPI, QR scanner etc for making payment transactions easy to Customers.

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

```

2
3 | SELECT p.payment_installments,
4 | COUNT(DISTINCT(o.order_id)) as No_of_order
5 FROM `Target_Dataset.orders` o
6 JOIN `Target_Dataset.payments` p
7 ON o.order_id = p.order_id
8 WHERE o.order_status!= 'canceled'
9 GROUP BY 1
10 ORDER BY 1 asc,
11          2 asc;
2

```

Row	payment_installment	No_of_order
1	0	2
2	1	48732
3	2	12329
4	3	10374
5	4	7046
6	5	5204
7	6	3894
8	7	1617
9	8	4224
10	9	638
11	10	5279

Query: SELECT p.payment_installments,
COUNT(DISTINCT(o.order_id)) as No_of_order
FROM `Target_Dataset.orders` o

```
JOIN `Target_Dataset.payments` p
ON o.order_id = p.order_id
WHERE o.order_status!= 'canceled'
GROUP BY 1
ORDER BY 1 asc,
        2 asc;
```

Insights: From the above Output we can see that customers have choosed different Installments options like yearly, monthly, quarterly & doing their payments accordingly.

Recommendations : Target can ace those customers who choose the options on yearly basis & according launch their schemes , So that customers would attract and purchased more product.