Target Case Study

```
# Data type of all columns in the "customers" table.
# Assuming customers table already exists in the database
SELECT column_name, data_type
FROM target_sales_2016_2018.INFORMATION_SCHEMA.COLUMNS
WHERE table_name = 'customers';
```

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

```
# Get the time range between which the orders were placed.
# Count the Cities & States of customers who ordered during the given period.
SELECT
   MIN(order_purchase_timestamp) AS start_date,
   MAX(order_purchase_timestamp) AS end_date
FROM target_sales_2016_2018.orders;
```

JOB IN	IFORMATION	RESULTS	CHART	JSON	E
Row	start_date ▼		end_date ▼		/
1	2016-09-04 21:15	:19 UTC	2018-10-17 17:	:30:18 UTC	100

```
SELECT
COUNT(DISTINCT customer_city) AS num_cities,
COUNT(DISTINCT customer_state) AS num_states
FROM target_sales_2016_2018.customers
WHERE customer_id IN
(SELECT DISTINCT customer_id
FROM target_sales_2016_2018.orders
);
```

JOB IN	FORMATION	I	RESULTS	CHART
Row	num_cities	• /	num_states	· /
1		4119		27

```
# In-depth Exploration:
# Is there a growing trend in the no. of orders placed over the past years?

SELECT
    EXTRACT(YEAR FROM order_purchase_timestamp) AS order_year,
    COUNT(*) AS num_orders

FROM target_sales_2016_2018.orders

GROUP BY order_year

ORDER BY order_year;
```

•	num_orders	order_year ▼	Row
329		2016	1
45101	4	2017	2
54011		2018	3

Inference: Yes, there has been major growth from 2016-17, then a small growth from 2017-18.

Row	order_month ▼	num_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

Inference: From the results we can infer that peak seasons were mid-years May-Aug, Weak seasons were end of the year Sep-Dec.

```
# 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
    CASE
        WHEN EXTRACT(HOUR FROM order_purchase_timestamp) BETWEEN @ 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'
        ELSE 'Night'
    END AS order_time_of_day,
    COUNT(*) AS num_orders
FROM target_sales_2016_2018.orders
GROUP BY order_time_of_day
ORDER BY num_orders DESC;
```

JOB IN	FORMATION	RESULTS	CHART	JS
Row	order_time_of_day	→	num_orders ▼	1
1	Afternoon		381	35
2	Night		283	31
3	Morning		277	33
4	Dawn		52	42

Inference: Most of the orders were placed during Afternoon time.

```
# Evolution of E-commerce orders in the Brazil region:
# Get the month on month no. of orders placed in each state.
SELECT
    EXTRACT(MONTH FROM o.order_purchase_timestamp) AS order_month,
    c.customer_state,
    COUNT(*) AS num_orders
FROM target_sales_2016_2018.orders o
JOIN target_sales_2016_2018.customers c ON o.customer_id = c.customer_id
GROUP BY order_month, c.customer_state
ORDER BY order_month, c.customer_state;
```

Row	order_month ▼	customer_state ▼	num_orders ▼
1	1	AC	8
2	1	AL	39
3	1	AM	12
4	1	AP	11
5	1	BA	264
6	1	CE	99
7	1	DF	151
8	1	ES	159
9	1	GO	164
10	1	MA	66

$\mbox{\#}\mbox{\ How\ are\ the\ customers\ distributed\ across\ all\ the\ states?}$

Row /	customer_state ▼	num_customers 🕶
1	SP	41746
2	RJ	12852
3	MG	11635
4	RS	5466
5	PR	5045
6	SC	3637
7	ВА	3380
8	DF	2140
9	ES	2033
10	GO	2020

Inference: State São Paulo (SP) has the most customers, while state Roraima (RR) has the least customer base.

```
# Impact on Economy: Analyze the money movement by e-commerce by looking at order prices,
freight and others.
# Get the % increase in the cost of orders from year 2017 to 2018 (include months between
Jan to Aug only).
WITH order_costs_2017 AS (
    SELECT EXTRACT(YEAR FROM o.order_purchase_timestamp) AS order_year,
           EXTRACT(MONTH FROM o.order_purchase_timestamp) AS order_month,
           SUM(p.payment_value) AS total_payment_value
    FROM target_sales_2016_2018.orders o
    JOIN target_sales_2016_2018.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
   GROUP BY order_year, order_month
), order_costs_2018 AS (
    SELECT EXTRACT(YEAR FROM o.order purchase timestamp) AS order year,
           EXTRACT(MONTH FROM o.order_purchase_timestamp) AS order_month,
           SUM(p.payment_value) AS total_payment_value
    FROM target sales 2016 2018.orders o
    JOIN target_sales_2016_2018.payments p ON o.order_id = p.order_id
   WHERE EXTRACT(YEAR FROM o.order_purchase_timestamp)=2018
     AND EXTRACT(MONTH FROM o.order_purchase_timestamp) BETWEEN 1 AND 8
   GROUP BY order_year, order_month
)
SELECT
 order_costs_2017.order_month as month,
 (order_costs_2018.total_payment_value - order_costs_2017.total_payment_value) /
order_costs_2017.total_payment_value * 100 AS percentage_increase
FROM order costs 2017
JOIN order_costs_2018 ON
order_costs_2017.order_month = order_costs_2018.order_month
ORDER BY month;
```

CHA	RESULTS	F	FORMATION	JOB IN
increase	percentage_	1	month ▼	Row
4171	705.1266954	1		1
5445	239.991814	2		2
6709	157.778606	3		3
1149	177.840770	4		4
5677	94.6273437	5		5
2456	100.2596912	6		6
3390	80.04245463	7		7
0477	51.60600520	8		8

Inference: Month of January had a huge boost of 700% compared to other months

Row	customer_state ▼	total_order_price 🔻	avg_order_price •
1	AC	19680.62	234.29
2	AL	96962.06	227.08
3	AM	27966.93	181.6
4	AP	16262.8	232.33
5	BA	616645.82	170.82
6	CE	279464.03	199.9
7	DF	355141.08	161.13
8	ES	325967.55	154.71
9	GO	350092.31	165.76
10	MA	152523.02	198.86

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

Row /	customer_state ▼	total_freight_value	avg_freight_value
1	AC	3686.75	40.07
2	AL	15914.59	35.84
3	AM	5478.89	33.21
4	AP	2788.5	34.01
5	BA	100156.68	26.36
6	CE	48351.59	32.71
7	DF	50625.5	21.04
8	ES	49764.6	22.06
9	G0	53114.98	22.77
10	MA	31523.77	38.26

```
# Analysis based on sales, freight and delivery time.
# 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.
WITH delivery_times AS (
    SELECT
        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_sales_2016_2018.orders
   WHERE
        order_delivered_customer_date IS NOT NULL
        AND order_estimated_delivery_date IS NOT NULL
)
SELECT
   order_id,
    delivery_time,
    diff_estimated_delivery
FROM
    delivery_times;
```

ow 1	order_id ▼ 1950d///989f6a8//539f53/9	delivery_time ▼ 30	diff_estimated_delive
2	2c45c33d2f9cb8ff8b1c86cc28	30	-28
3	65d1e226dfaeb8cdc42f66542	35	-16
4	635c894d068ac37e6e03dc54e	30	-1
5	3b97562c3aee8bdedcb5c2e45	32	0
6	68f47f50f04c4cb6774570cfde	29	-1
7	276e9ec344d3bf029ff83a161c	43	4
8	54e1a3c2b97fb0809da548a59	40	4
9	fd04fa4105ee8045f6a0139ca5	37	1
10	302bb8109d097a9fc6e9cefc5	33	5

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

```
WITH state_avg_freight AS (
   SELECT
     c.customer_state,
     AVG(oi.freight_value) AS avg_freight,
    FROM target_sales_2016_2018.orders o
    JOIN target_sales_2016_2018.customers c ON o.customer_id = c.customer_id
    JOIN target_sales_2016_2018.order_items oi ON oi.order_id = o.order_id
   GROUP BY c.customer_state
), state_ranks as (
 SELECT
   customer_state,
    avg_freight,
    ROW_NUMBER() OVER (ORDER BY avg_freight DESC) AS rank_high,
    ROW_NUMBER() OVER (ORDER BY avg_freight ASC) AS rank_low
  FROM state_avg_freight
SELECT
  customer_state,
  avg_freight,
 rank_high
FROM state_ranks
WHERE rank_high<=5
```

JOB IN	FORMATION	RESULTS	CHART	JSON	EXECUTION
Row	customer_state •	1	avg_freight ▼	rank_hig	h • //
1	RR		42.9844230769	2	1
2	PB		42.7238039867	1	2
3	RO		41.0697122302	1	3
4	AC		40.0733695652	1	4
5	PI		39.1479704797	0	5

SELECT

```
customer_state,
avg_freight,
rank_low
FROM state_ranks
WHERE rank_low<=5;</pre>
```

Row	customer_state ▼	avg_freight ▼ rank_le	ow ▼
1	SP	15.14727539041	1
2	PR	20.53165156794	2
3	MG	20.63016680630	3
4	RJ	20.96092393168	4
5	DF	21.04135494596	5

```
# Find out the top 5 states with the highest & lowest average delivery time.
WITH state_delivery_time AS (
    SELECT
      c.customer_state,
      AVG(DATE_DIFF(order_delivered_customer_date, order_purchase_timestamp, DAY)) AS
avg_delivery_time,
      ROW_NUMBER() OVER (ORDER BY AVG(DATE_DIFF(order_delivered_customer_date,
order_purchase_timestamp, DAY)) DESC) AS rank_high,
      ROW_NUMBER() OVER (ORDER BY AVG(DATE_DIFF(order_delivered_customer_date,
order_purchase_timestamp, DAY)) ASC) AS rank_low
    FROM target_sales_2016_2018.orders o
    JOIN target_sales_2016_2018.customers c ON o.customer_id = c.customer_id
    WHERE order_delivered_customer_date IS NOT NULL
    GROUP BY c.customer_state
)
SELECT
    customer_state,
    avg_delivery_time,
    rank_high
FROM state_delivery_time
WHERE rank_high <= 5;</pre>
```

Row	customer_state ▼	avg_delivery_time rank_hig	gh ▼
1	RR	28.97560975609	1
2	AP	26.73134328358	2
3	AM	25.98620689655	3
4	AL	24.04030226700	4
5	PA	23.31606765327	5

```
# SELECT
# customer_state,
# avg_delivery_time,
# rank_low
# FROM state_delivery_time
# WHERE rank_low <= 5</pre>
```

Row	customer_state ▼	avg_delivery_time 🔻 1	rank_low 🔻
1	SP	8.298061489072	1
2	PR	11.52671135486	2
3	MG	11.54381329810	3
4	DF	12.50913461538	4
5	SC	14.47956019171	5

```
# 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 state_delivery_speed AS (
    SELECT
        c.customer_state,
        AVG(DATE_DIFF(order_delivered_customer_date, order_estimated_delivery_date, DAY))
AS avg_delivery_speed,
        ROW_NUMBER() OVER (ORDER BY AVG(DATE_DIFF(order_delivered_customer_date,
order_estimated_delivery_date, DAY)) ASC) AS rank_fastest
    FROM target_sales_2016_2018.orders o
    JOIN target_sales_2016_2018.customers c ON o.customer_id = c.customer_id
    WHERE
        order_delivered_customer_date IS NOT NULL
        AND order_estimated_delivery_date IS NOT NULL
    GROUP BY c.customer_state
)
SELECT
    customer_state,
    avg_delivery_speed,
    rank_fastest
FROM state_delivery_speed
WHERE rank_fastest <= 5;</pre>
```

Row	customer_state ▼	avg_delivery_speed ▼	rank_fastest ▼
1	AC	avg_delivery_speed	1
2	RO	-19.13168724279836	2
3	AP	-18.731343283582088	3
4	AM	-18.60689655172413	4
5	RR	-16.414634146341463	5

```
# Analysis based on the payments:
# Find the month on month no. of orders placed using different payment types.

SELECT
    EXTRACT(MONTH FROM o.order_purchase_timestamp) AS order_month,
    p.payment_type,
    COUNT(*) AS num_orders

FROM target_sales_2016_2018.orders o

JOIN target_sales_2016_2018.payments p ON o.order_id = p.order_id

GROUP BY order_month, p.payment_type

ORDER BY order_month, p.payment_type;
```

Row /	order_month ▼	payment_type ▼	num_orders ▼
1	1	UPI	1715
2	1	credit_card	6103
3	1	debit_card	118
4	1	voucher	477
5	2	UPI	1723
6	2	credit_card	6609
7	2	debit_card	82
8	2	voucher	424
9	3	UPI	1942
10	3	credit_card	7707

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

SELECT

```
payment_installments,
    COUNT(*) AS num_orders
FROM target_sales_2016_2018.payments
GROUP BY payment_installments
ORDER BY payment_installments;
```

JOB IN	FORMATION	RESULTS	CHA
ow	payment_installment	num_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	8 4268	
10	9		644

Actionable Insights and recommendations:

- 1. Introducing special offers, discounts, events, etc., in the time of peak season May-Aug.
- 2. Taking special care and measures to maintain the order servers in Afternoon period.
- 3. Designating some states like SP as high value due to large customer base and creating business plans to improve sales push in these regions.
- 4. Upgrading transport and infrastructure to improve the Delivery days taken for states like RR, AP, who has smaller customer base.
- 5. Introducing offers catered to the credit card users due to its popularity as a preferred payment method.