### **GCP** Connection

Database: Olist\_data

### **TABLES**

#### **DDL** Command for the table

### $Table\ olist\_customers\_dataset$

```
CREATE TABLE customers (
customer_id VARCHAR(255) NOT NULL,
customer_zip_code_prefix INT,
customer_city VARCHAR,
customer_state VARCHAR
PRIMARY KEY (customer_id)
);
```

### **CREATE TABLE geolocation (**

```
geolocation_zip_code_prefix INT NOT NULL,
geolocation_lat DECIMAL(9,6),
geolocation_lng DECIMAL(9,6),
geolocation_city VARCHAR(100),
geolocation_state VARCHAR(2)
);
```

### **CREATE TABLE olist\_orders\_dataset** (

```
order_id VARCHAR(255) NOT NULL PRIMARY KEY, customer_id VARCHAR(255), order_status VARCHAR(50), order_purchase_timestamp TIMESTAMP, order_approved_at TIMESTAMP, order_delivered_carrier_date TIMESTAMP, order_delivered_customer_date TIMESTAMP, order_delivered_customer_date TIMESTAMP, order_estimated_delivery_date TIMESTAMP);
```

### Table olist order items dataset

```
CREATE TABLE order_items (
order_id VARCHAR(255) NOT NULL,
order_item_id INT,
product_id VARCHAR(255),
seller_id VARCHAR(255),
shipping_limit_date DATETIME,
price FLOAT,
freight_value FLOAT,
```

```
PRIMARY KEY (order id, order item id)
);
Table order payments dataset
CREATE TABLE order payments (
 order_id VARCHAR(255) NOT NULL,
 payment sequential INT NOT NULL,
 payment type VARCHAR,
 payment installments INT,
 payment value DECIMAL,
 PRIMARY KEY (order id, payment sequential)
);
Table olist order reviews dataset
CREATE TABLE order reviews (
 review id VARCHAR(255) NOT NULL,
 order id VARCHAR(255) NOT NULL,
review score INT,
 review comment title VARCHAR,
 review comment message VARCHAR,
 review creation date TIMESTAMP,
 review answer timestamp TIMESTAMP
 PRIMARY KEY (review id)
);
Table olist products dataset
CREATE TABLE products (
product_id VARCHAR(255) NOT NULL,
product category name VARCHAR(255),
seller id VARCHAR(255),
product name lenght INT,
product description lenght INT,
 product photos qty INT,
 product weight g INT,
 product length cm INT,
 product height cm INT,
 product width cm INT,
 PRIMARY KEY (product id)
);
Table olist sellers dataset
CREATE TABLE sellers (
 seller id VARCHAR(255) NOT NULL,
 seller city VARCHAR(255),
 seller_state VARCHAR(255),
```

```
seller_state VARCHAR,
PRIMARY KEY (seller_id)
);
```

### Table product\_category\_name\_translation

```
CREATE TABLE product_category_name_translation (
   product_category_name VARCHAR(255) NOT NULL,
   product_category_name_english VARCHAR(255),
   PRIMARY KEY (product_category_name)
);
```

### **ROWS COUNT**

### **ADVANCED Query:**

### **Query 1 – Average Freight and Price by Seller (Join + Group By)**

```
sseller_id,
s.seller_city,
COUNT(oi.order_id) AS total_orders,
AVG(oi.price) AS avg_price,
AVG(oi.freight_value) AS avg_freight
FROM olist_order_items_dataset oi
JOIN olist_products_dataset p ON oi.product_id = p.product_id
JOIN olist_sellers_dataset s ON p.seller_id = s.seller_id
GROUP BY s.seller_id, s.seller_city;
Limit15;
```

seller_id   se	eller_city	total_orders	avg_price	avg_freight
0015a82c2db000af6aaaf3ae2ecb0532   sa	anto andre	3	895	21.020000457763672
001cca7ae9ae17fb1caed9dfb1094831   ca	ariacica	240	105.1626248995463	36.99858363866806
001e6ad469a905060d959994f1b41e4f   sa	ao goncalo	1	250	17.940000534057617
002100f778ceb8431b7a1020ff7ab48f   fr	ranca	55	22.445454198663885	14.43018214485862
003554e2dce176b5555353e4f3555ac8   gc	oiania	1	120	19.3799991607666
004c9cd9d87a3c30c522c48c4fc07416   ik	bitinga	170	115.95711768655217	20.889588170893052
00720abe85ba0859807595bbf045a33b   gu	uarulhos	26	38.750000073359566	12.153076777091393
00ab3eff1b5192e5f1a63bcecfee11c8   sa	ao paulo	1	98	12.079999923706055
00d8b143d12632bad99c0ad66ad52825   be	elo horizonte	1	86	51.099998474121094
00ee68308b45bc5e2660cd833c3f81cc   sa	ao paulo	319	110.4404068471496	17.52097165734043
00fc707aaaad2d31347cf883cd2dfe10   ma	aringa	848	84.6035264483038	15.271910347325623
010543a62bd80aa422851e79a3bc7540   sa	ao paulo	2	708	15.97499942779541
010da0602d7774602cd1b3f5fb7b709e   sa	ao bernardo do campo	5	169.89999389648438	46.04999923706055
011b0eaba87386a2ae96a7d32bb531d1   pc	ompeia	2	49.9900016784668	14.59000015258789
01266d4c46afa519678d16a8b683d325   cu	uritiba	3	30.083333651224773	15.743333498636881

# Query 2: Top States by Average Review Score (JOINs, GROUP BY, ORDER BY, and aggregation) WORKING

```
SELECT
```

c.customer\_state,
COUNT(r.review\_id) AS total\_reviews,
ROUND(AVG(r.review\_score), 2) AS avg\_review\_score
FROM olist\_order\_reviews\_dataset r
JOIN olist\_orders\_dataset o ON r.order\_id = o.order\_id
JOIN olist\_customers\_dataset c ON o.customer\_id = c.customer\_id
GROUP BY c.customer\_state
ORDER BY avg\_review\_score DESC
LIMIT 15;

customer_state	total_reviews	avg_review_score
AP	67	4.19
AM	147	4.18
PR	5038	4.18
SP	41689	4.17
MG	11625	4.14
RS	5483	4.13
MS	724	4.12
RN	482	4.11
TO	279	4.10
MT	903	4.10
SC	3623	4.07
DF	2148	4.06
AC	81	4.05
RO	252	4.05
ES	2016	4.04

### **Query 3: Sellers with Low Freight Costs (Subquery)**

This query finds sellers whose average freight value is greater than \$50, using a subquery in the WHERE clause:

```
SELECT
s.seller_id,
s.seller_city,
s.seller_state
FROM olist_sellers_dataset s
WHERE s.seller_id IN (
SELECT p.seller_id
FROM olist_order_items_dataset oi
JOIN olist_products_dataset p ON oi.product_id = p.product_id
GROUP BY p.seller_id
HAVING AVG(oi.freight_value) < 10
)
LIMIT 15;
```

# Query 4: Compare Products in High vs Low Freight Brackets (SET OPERATOR: UNION)

```
SELECT
    product_id,
    price,
    freight_value,
    'High Freight' AS freight bracket
  FROM olist_order_items_dataset
  WHERE freight_value > 100
)
UNION
(
  SELECT
    product_id,
    price,
    freight_value,
    'Low Freight' AS freight_bracket
  FROM olist_order_items_dataset
  WHERE freight value < 5
```

product_id	price	freight_value	freight_bracket
43cc8e4d981bc04b9d78b12e8a908d41	1240	102.63	High Freight
a233df9a388d27dbdfd31731d4236db0	2649.99	134.17	High Freight
8d4e92265a16e69a1e1d76e67e46d72f	1350	294.76	High Freight
63c4a70e0a12b4bd9475fca9e9937e76	122.99	164.98	High Freight
e303dfa61ada1f0823b4775f192606b3	148.5	165.32	High Freight
660422061e06da17ca6101e9d6b7aae8	649	169.12	High Freight
ab495f166205a883ffe5ab0b5b55f867	629.9	117.35	High Freight
90c1b4e040d1d1c45897ec2dad4a809d	839.99	174.49	High Freight
ef854c7d98d5eba672287b0a9d37075b	1990	125.05	High Freight
72d0a38fe43ba7087d71e245d1b76c9e	229	106.31	High Freight
5640c59a8f6a08b3758272590693eec3	238.47	119.56	High Freight
67a7c7243c0585ccc44471b5a5c115e9	322	186.38	High Freight
65841ad29fc48cd40902e03da7511e05	849	174.95	High Freight
1945afae0c93166dce1f186d00125695	517.99	107.87	High Freight
8a443635fdf9759915c9be5be2e3b862	99.9	112.44	High Freight

# **Query 5: Products That Were Reviewed the Most with Highest Average Ratings (Multiple Joins, GROUP BY, HAVING with aggregation)**

```
p.product_id,
pct.product_category_name_english AS category,
COUNT(r.review_id) AS total_reviews,
ROUND(AVG(r.review_score), 2) AS avg_score
FROM olist_order_reviews_dataset r
JOIN olist_orders_dataset o ON r.order_id = o.order_id
JOIN olist_order_items_dataset oi ON o.order_id = oi.order_id
JOIN olist_products_dataset p ON oi.product_id = p.product_id
JOIN product_category_name_translation pct
ON p.product_category_name = pct.product_category_name
GROUP BY p.product_id, pct.product_category_name_english
HAVING total_reviews > 5 AND avg_score >= 4.5
ORDER BY total_reviews DESC, avg_score DESC
```

## Part 2 Indexing: **QUERY 1**

```
| >> Limit: 15 row(s) (actual time=9424.3424 rows=15 loops=1)
| >> Table scan on temporaryy (actual time=9424.3424 rows=15 loops=1)
| >> Apgregate using temporaryy table (actual time=03424.3424 rows=2986 loops=1)
| -> Apgregate using temporary table (actual time=03424.3424 rows=2986 loops=1)
| -> Nested loop inner join (cost=99795 rows=111760) (actual time=150.2589 rows=111024 loops=1)
| -> Table scan on oi (cost=19079 rows=111760) (actual time=65.1.1246 rows=112651 loops=1)
| -> Filter: (p.seller_id is not null) (cost=0.688 rows=1) (actual time=0.0113..016 rows=0.986 loops=12651)
| -> Single-row index lookup on p using PRIMARY (groduc-id=0.1,rows=0.0168 rows=1) (actual time=0.0133..0.0133 rows=0.986 loops=112651)
| -> Single-row index lookup on s using PRIMARY (seller_id=0.produc-id=0.1) (actual time=0.00320..0.00331 rows=1 loops=111024)
```

### **Index Choice: 2**

We chose Index Choice 2 because it improved query performance by speeding up joins and average calculations. The indexes reduced full table scans and enabled faster data access, especially on large datasets.

### 1. CREATE INDEX idx\_oi\_price\_freight ON olist\_order\_items\_dataset(price, freight\_value);

This index is designed to optimize the aggregation operations in the query by allowing the database to quickly access the 'price' and 'freight\_value' columns used in the AVG functions. By having these columns indexed together, the system can reduce full table scans and speed up the computation of averages, especially when working with a large dataset.

### 2. CREATE INDEX idx\_p\_seller\_id ON olist\_products\_dataset(seller\_id); JOINED ATTRIBUTE

```
| -> Limit: 15 row(s) (actual time=2282..2282 rows=15 loops=1)
-> Table scan on <temporary> (actual time=2282..2282 rows=15 loops=1)
-> Aggregate using temporary table (actual time=2282..2282 rows=2996 loops=1)
-> Nested loop inner join (cost=173840 rows=111760) (actual time=170..1857 rows=111024 loops=1)
-> Nested loop inner join (cost=59904 rows=111760) (actual time=109..1392 rows=111024 loops=1)
-> Table scan on oi (cost=17384 rows=111760) (actual time=109..1392 rows=111024 loops=1)
-> Table scan on oi (cost=17384 rows=111760) (actual time=109..888 rows=112651 loops=1)
-> Filter: (p.selier_id is not null) (cost=070..25 rows=1) (actual time=0.00418...0.0042 rows=0.986 loops=112651)
-> Single=row index lookup on p using FRIMARY (product id=0.iproduct id) (cost=0.25 rows=1) (actual time=0.00394..0.00397 rows=1 loops=11024)
-> Single=row index lookup on s using FRIMARY (seller_id=p.seller_id) (cost=1 rows=1) (actual time=0.00394..0.00397 rows=1 loops=111024)
```

Index on olist\_products\_dataset(seller\_id): This index improves the efficiency of the join between the products and sellers datasets. By indexing the seller\_id in the products table, the database engine can rapidly match products to their corresponding sellers, reducing lookup time during the join process and ultimately enhancing the overall query performance.

### 3. CREATE INDEX idx\_oi\_product\_id ON olist\_order\_items\_dataset(product\_id);

Index on olist\_order\_items\_dataset(product\_id): This index is established to speed up the join between order items and products. With the product\_id column indexed in the order items

dataset, the database can quickly locate matching records for each product, which minimizes the data scanned during the join operation and improves the execution speed of the query.

### **Query2: Before Indexing Performance:**

```
| -> Limit: 15 row(s) (actual time=3530..3530 rows=15 loops=1)
-> Sort: avg_review_score DESC, limit input to 15 row(s) per chunk (actual time=3530..3530 rows=15 loops=1)
-> Table scan on freemporary (actual time=3530..3530 rows=28 loops=1)
-> Aggregate using temporary table (actual time=3530..3530 rows=28 loops=1)
-> Nested loop inner join (cost=104e46 rows=94e46) (actual time=1059..1832 rows=99224 loops=1)
-> Filter: (o.order_id = r.order_id) (cost=94.4e46 rows=94e46) (actual time=1059..1832 rows=99224 loops=1)
-> Inner hash join (chash>(o.order_id) ~chash>(o.order_id) (cost=94.4e66 rows=94e46) (actual time=1059..1832 rows=99224 loops=1)
-> Filter: (o.oustomer_id is not null) (cost=4.89 rows=9615) (actual time=49.6..655 rows=9642 loops=1)
-> Table scan on o (cost=10043 rows=95150) (actual time=49.6..614 rows=99422 loops=1)
-> Table scan on r (cost=10043 rows=95722) (actual time=89..916 rows=99224 loops=1)
-> Single-row index lookup on c using FRIMARY (customer_id=0.customer_id) (cost=10.4e-6 rows=1) (actual time=0.0154..0.0154 rows=1 loops=99224)
```

### **Index Choice: 3**

We chose Index Choice 3 because it significantly improved join performance between reviews, orders, and customers. Indexing order\_id in both the reviews and orders datasets sped up matching operations, while indexing customer\_state allowed quicker filtering. These indexes reduced full table scans and improved overall query efficiency, especially on large datasets.

### 1. CREATE INDEX idx reviews order id ON olist order reviews dataset(order id);

This index is designed to speed up the join between reviews and orders by allowing the database to quickly locate matching order\_id values in the reviews table. This minimizes the need for full table scans during the join operation, significantly enhancing performance when working with large datasets.

### 2. CREATE INDEX idx orders order id ON olist orders dataset(order id);

```
| -> Limit: 15 row(s) (actual time=11232.11232 cons=15 loops=1)
-> Sort: avg_review_sorce RSSC, limit input to 15 row(s) per chunk (actual time=11232.11232 rows=15 loops=1)
-> Table scan on temporary (actual time=11231.11231 rows=28 loops=1)
-> Aggregate using recorder recorder (actual time=11231.11231 rows=28 loops=1)
-> Nested loop inner join (cost=016007 rows=95722) (actual time=155.11047 rows=99224 loops=1)
-> Nested loop inner join (cost=01638 rows=95722) (actual time=155.11047 rows=99224 loops=1)
-> Nested loop inner join (cost=01638 rows=95722) (actual time=60.8.913 rows=99224 loops=1)
-> Filter: (r.order id is not null) (cost=01063 rows=95722) (actual time=60.8.913 rows=99224 loops=1)
-> Table scan on r (cost=10043 rows=95722) (actual time=60.8.900 rows=99224 loops=1)
-> Filter: (o.customer id is not null) (cost=0.451 rows=1) (actual time=0.0755 rows=1 loops=99224)
-> Index lookup on ousing did orders order id (order_ider.order_id) (cost=0.451 rows=1) (actual time=0.0758.0.0756 rows=1 loops=99224)
-> Single-row index lookup on c using FRIMARY (customer_id=0.customer_id) (cost=1 rows=1) (actual time=0.0257.0.0257 rows=1 loops=99224)
```

The index on order\_id in the orders dataset optimizes the join process by efficiently retrieving corresponding order records for each review. By using this index, the database can quickly match order\_id values between the reviews and orders tables, reducing lookup time and improving overall query performance.

### **3.CREATE INDEX idx\_customers\_state ON olist\_customers\_dataset(customer\_state)**;

Indexing customer\_state in the customers dataset is intended to improve the performance of the grouping and aggregation operations in the query. Since the query groups results by customer\_state and computes aggregate functions like count and average, this index helps the database quickly sort and group rows by state, leading to faster execution even if the column has relatively few unique values.

### **Query 3 – Sellers with Low Freight Costs (Subquery)**

```
| -> Limit: 15 row(s) (cost=314 rows=15) (actual time=5490..5495 rows=15 loops=1)
-> Filter: (in optimizer)(s.seller_id,s.seller_id) in (select_id)) (cost=314 rows=396) (actual time=75490..5495 rows=15 loops=1)
-> Table soon on a (cost=314 rows=396) (actual time=77.4..81.1 rows=275 loops=1)
-> Select_id (s.seller_id) (soot=314 rows=306) (actual time) (soot=316 rows=0.0543 loops=276)
-> Filter: (g.seller_id) (soot=316 rows=3064) (actual time) (s.19.6 rows=0.0543 loops=276)
-> Limit: 1 row(s) (cost=316 rows=3064) (actual time) (s.19.6 rows=0.0543 loops=276)
-> Index lookup on camaterialized subspuery using (sauto distinct_revy (seller_id=s.seller_id) (scutal time=19.6..19.6 rows=0.0543 loops=276)
-> Filter: (avg(oi.freight_value) < 10) (actual time=5400..5412 rows=118 loops=1)
-> Filter: (avg(oi.freight_value) < 10) (actual time=5400..5412 rows=118 loops=1)
-> Aggregate using temporary (actual time=5400..5412 rows=108 loops=1)
-> Aggregate using temporary table (actual time=5400..5490 rows=3066 loops=1)
-> Pilter: (p.product_id is not null) (cost=54155 rows=11865) (actual time=696.6.954 rows=32952 loops=1)
-> Filter: (p.product_id is not null) (cost=54155 rows=31660) (actual time=696.6.954 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974 rows=32952 loops=1)
-> Table soon on (cost=54155 rows=11060) (actual time=696.6.974
```

### **INDEX Choice: 4**

We chose index 4 because it improved the performance of the outer query by speeding up the filtering process. It allowed the database to quickly match seller\_ids from the subquery, reducing lookup time and making the overall query more efficient.

1. CREATE INDEX idx oi product id ON olist order items dataset(product id);

```
| -> Limit: 15 row(s) (cost=327 rows=15) (actual time=4907..4940 rows=15 loops=1)
-> Filter: (in optimizer>(s.seller_id,s.seller_id in (select f2)) (cost=327 rows=3096) (actual time=4907..4940 rows=15 loops=1)
-> Table scan on s (cost=327 rows=3096) (actual time=57.5..90.2 rows=520 loops=1)
-> Select f2 (subjecty in condition: run only once)
-> Filter: ((s.seller_id = 'casterialized gubquery'.seller_id)) (cost=9991..99941 rows=1) (actual time=9.31..9.31 rows=0.0288 loops=521)
-> Limit 1 row(s) (cost=59991..9991 rows=1) (actual time=9.31..9.31 rows=0.0288 loops=521)
-> Limit 1 row(s) (cost=59991..9991 rows=1) (actual time=0.31..9.31 rows=0.0288 loops=521)
-> Limit 1 row(s) (rows=5991 rows=1) (actual time=0.31..9.31 rows=0.0288 loops=521)
-> Materialize with deduplication (cost=99991 rows=1038) (actual time=011 loops=1)
-> Filter: (avg(s).freight_value) (cost=99637 rows=3038) (actual time=011 loops=1)
-> Group aggregate: avg(si.freight_value) (cost=99637 rows=3038) (actual time=0171..4844 rows=2996 loops=1)
-> Nested loop inner join (cost=8845 rows=10222) (actual time=01..4946 rows=11024 loops=1)
-> Covering index scan on p using idx_p seller_id (cost=3559 rows=30269) (actual time=01..3.376 rows=3339 loops=1)
-> Index lookup on oi using idx_p seller_id (product_id=p.product_id) (cost=2.29 rows=3.52) (actual time=0.847..0.135 rows=3.43 loops=32329)
```

2. olist\_order\_items\_dataset(product\_id, freight\_value)
CREATE INDEX idx\_oi\_product\_freight ON olist\_order\_items\_dataset(product\_id, freight\_value);

```
| >> Limit: 15 cos(n) (cost=327 cows=3) (actual time=867.901 cows=30 loops=1)
| >> Filter: (in optimizer); seller id, seller id in (select 22) (cost=237 cows=3096) (actual time=867.901 cows=15 loops=1)
| >> Table scan on s (cost=327 cows=3096) (actual time=3.50.42.9 cows=520 loops=1)
| >> Select 22 (subquery) in condition; run only once)
| >> Filter: (id.seller id = 'Casterialized subqueryy'.seller id)| (cost=78505.78505 rows=1) (actual time=1.65.1.65 rows=0.0288 loops=521)
| >> Limit: 1 fow(s)| (cost=78505.78505 rows=1) (actual time=1.65.1.65 rows=0.0288 loops=521)
| >> Index lookup on casterialized subqueryy using (auto distinct key) (seller id=seller id) (actual time=1.65.1.65 rows=0.0288 loops=521)
| >> Index lookup on casterialized subqueryy using (auto distinct key) (seller id=seller id) (actual time=1.65.1.65 rows=0.0288 loops=521)
| >> Filter: (avg(oi.freight, value) < 10) (cost=78303) (actual time=525.858 rows=116 loops=1)
| >> Filter: (avg(oi.freight, value) < 10) (cost=78201 rows=308) (actual time=525.858 rows=2964 loops=1)
| >> Filter: (avg(oi.freight, value) < 10) (cost=78201 rows=308) (actual time=625.858 rows=2964 loops=1)
| >> Filter: (avg(oi.freight, value) < 10) (cost=78201 rows=308) (actual time=625.359 rows=3069) (actual time=720.302 rows=3229 loops=1)
| >> Covering index soon on p using idx p seller id (cost=3559 rows=3069) (actual time=720.302 rows=3.51) (actual time=0.0122.0.0147 rows=3.43 loops=32329)
```

This composite index helps the subquery perform both the join on product\_id and the aggregation on freight\_value efficiently. It allows the database to quickly find records based on product\_id while also retrieving freight\_value for computing averages, reducing unnecessary scans. Optimizes join to products + improves AVG computation in HAVING.

### 3. olist products dataset(product id, seller id)

This index speeds up the join between order items and products by enabling fast lookup of seller\_id using product\_id. It makes grouping by seller\_id in the subquery more efficient, which is crucial for filtering based on the average freight value. Helps with the join and allows efficient grouping by seller id.

### 4. olist sellers dataset(seller id)

```
|-> limit: 15 row(s) (cost-326 rows-15) (actual time-1590.1582 fows-15 loops-1)
-> Filter (in pointare; (actual time-1590.1582 fows-15 loops-1)
-> Select 12 (subquery in condition; run only once)
-> Select 12 (subquery in condition; run only once)
-> Filter: ((s.-seller id " casterialized subquery'.seller id)) (cost-0.0 rows-0) (actual time-3.05..3.05 rows-0.0288 loops-521)
-> Limit: 1 row(s) (cost-0.0 rows-0) (actual time-3.05..3.05 rows-0.0288 loops-521)
-> Linder lookup on casterialized subquery' using (actual cimit reps) (seller id=seller id=seller id) (actual time-3.05..3.05 rows-0.0288 loops-521)
-> Thick row(s) (cost-0.0 rows-0) (actual ine-1588.1589 rows-18 loops-1)
-> Filter: (actual time-3.05..3.05 rows-0.0288 loops-521)
-> Filter: (actual time-3.05..3.05 rows-0.0288 loops-521)
-> Filter: (actual time-1588.1589 rows-18 loops-1)
-> Apprend to sing the propary table (actual time-1588.1589 rows-18 loops-1)
-> Nested loop inner join (cost-58105 rows-11760) (actual time-9.0.1265 loops-1)
-> Table san on of (cost-58105 rows-11760) (actual time-9.2.726 rows-11265 loops-1)
-> Table san on of (cost-58105 rows-11760) (actual time-9.2.726 rows-112651 loops-1)
-> Single-row index lookup on p using PRIMENT (product_id=0.jproduct_id) (cost-0.314 rows-1) (actual time-0.00592.0.00595 rows-0.986 loops-112651)
```

This index ensures that the outer query can rapidly match seller records against the list of seller\_ids produced by the subquery. It speeds up the filtering process in the main query by quickly locating the relevant sellers based on their seller\_id. Helps with filtering in the outer query's WHERE ... IN (...).

### **QUERY 4:**

```
| -> Table scan on <union temporary> (cost=30915..31849 rows=74499) (actual time=2625..2625 rows=1270 loops=1)
| -> Union materialize with deduplication (cost=30915..30915 rows=74499) (actual time=2625..2625 rows=1270 loops=1)
| -> Filter: (olist_order_items_dataset.freight_value > 100) (cost=11733 rows=37250) (actual time=15.2..2567 rows=671 loops=1)
| -> Table scan on olist_order_items_dataset (cost=11733 rows=111760) (actual time=4.07..2558 rows=112651 loops=1)
| -> Filter: (olist_order_items_dataset.freight_value < 5) (cost=11733 rows=37250) (actual time=0.434.48.3 rows=1228 loops=1)
| -> Table scan on olist_order_items_dataset (cost=11733 rows=111760) (actual time=0.387..41 rows=122651 loops=1)
| -> Table scan on olist_order_items_dataset (cost=11733 rows=111760) (actual time=0.387..41 rows=112651 loops=1)
```

### **Index Choice: 3**

We chose index 3 because it optimizes both the filtering on freight\_value and access to product\_id, which improves overall query performance by reducing unnecessary lookups.

1. Single column index on freight value

Best for optimizing the WHERE condition.

olist order items dataset(freight value);

```
1.3 Table non on Conton temperary (contribute, 1222 near-1109) pertual time-1621, 1628 rener-1270 loopes).

2.3 Table non on Conton temperary (contribute, 1222 near-1109) pertual time-1621, 1628 rener-1270 loopes).

3.5 Table non on Conton temperary (contribute, 1222 near-1109) (conton time-1621, 1628 rener-1270 loopes).

3.5 Table non on Conton temperary (contribute, 1222 near-1109) (conton time-1622, 1222 near-1109) (conton time-1622, 1222 near-1270 loopes).

3.5 Table non on Conton temperary (contribute, 1222 near-1109) (conton time-1622, 1222 near-1270 loopes).

3.5 Table non on Conton temperary (contribute, 1222 near-1109) (conton time-1622, 1222 near-1222) (conton time-1622) (conton time-16222) (conton time-16222) (conton time-162222) (conton time-1622222) (conton time-162222
```

This single-column index is designed to accelerate the filtering process in both parts of the UNION query, where freight\_value is compared against 100 and 5. By indexing freight\_value, the database engine can rapidly locate rows that meet the high or low freight conditions, thus minimizing full table scans. However, since the query also retrieves product\_id and price, additional lookups may be necessary once the matching rows are identified.

2. Composite index on (freight value, price)

Improves performance if you're analyzing price ranges or doing additional filtering/sorting by price.

olist\_order\_items\_dataset(freight\_value, price);

```
| > Table scan on conton temporary (conv-1984..2012 row-vil99) (actual time-vil91.1751 row-vil79) logowil (actual time-vil91..1751 row-vil99) (actual time-vil91..1751 row-vil99) logowil (actual time-vil99) logowil (act
```

This composite index enhances performance by starting with the freight\_value column, which is used for filtering, and then including the price column, which is part of the SELECT list. The combination allows the engine to not only quickly identify rows with the desired freight\_value criteria but also to more efficiently access the price data directly from the index. While this can reduce the number of additional lookups compared to a single-column index, the product\_id column is not included, so the engine may still need to fetch that value separately.

3. Composite index on (freight value, product id)

Useful if you later expand this query to group by or join on product id.

olist\_order\_items\_dataset(freight\_value, product id);

This composite index is optimized by pairing freight\_value with product\_id. It efficiently supports the WHERE clause by quickly filtering based on freight\_value, while simultaneously speeding up access to product\_id, which is crucial for the query's output. Although this index improves lookup performance for the filtering and one of the selected columns, the price column is omitted from the index, potentially necessitating extra lookups to retrieve its value.

### **QUERY5**

**Index Choice: none** 

The reason why indexes 2 and 3, as well as the version without indexing, perform the same is likely because the dataset is already small enough to fit in memory, allowing the database engine to perform full table scans efficiently. Additionally, the joins involved may already use primary key lookups, which are optimized by default. As a result, adding secondary indexes doesn't provide a noticeable performance improvement in this specific query.

1. Index for joining reviews to orders:

```
olist_order_reviews_dataset(order_id);
```

Speeds up JOIN r ON r.order\_id = o.order\_id

```
| >> Sort: total reviews DESC, says score DESC (actual time=51493.51499) row=997 loops=1)
| >> Filter: (Total reviews > 5) and (say gone > 4.5)) (actual time=514965.51495 rows=987 loops=1)
| >> Totale scan on temporary (centual time=514940.51488 rows=23271 loops=1)
| >> Apprenate using temporary table (actual time=514938..513498 rows=23271 loops=1)
| >> Nested top inner join (cost=578195 row=113498), (actual time=5484..498418 rows=117499 loops=1)
| >> Nested top inner join (cost=578195 row=113223) (actual time=51..4695 rows=113410 loops=1)
| >> Nested top inner join (cost=58820 row=113223) (actual time=65..1.4695 rows=113410 loops=1)
| >> Nested top inner join (cost=58820 row=113223) (actual time=65..1.4695 rows=11324 loops=1)
| >> Filter: (p.product_category_name (sost=5886 row=32069) (actual time=65..1.4695 row=32329 loops=1)
| >> Single-row index lookup on pct using PRIMANY (croduct_category_name) (cost=1 row=2) (actual time=6.313.0.0192 row=1 loops=33329)
| >> Single-row index lookup on pct using PRIMANY (croduct_category_name) (cost=1 row=2) (actual time=6.313.0.0192 row=1 loops=33329)
| >> Single-row index lookup on octual primany (cost=1 row=1) (actual time=6.31..131 row=100ps=111024)
| >> Index lookup on r using idx_reviews_order_id (croder_id=0..crder_id) (cost=1 row=1) (actual time=6.31..131 row=10ops=111024)
| >> Index lookup on r using idx_reviews_order_id (croder_id=0..crder_id) (cost=1 row=1) (actual time=6.31..131 row=10ops=111024)
```

This index is created to speed up the join between the reviews and orders datasets by allowing the database to quickly locate matching order\_id values. Since the query starts by joining reviews to orders on order\_id, this index reduces the amount of data scanned and accelerates the first step of the join chain.

2. Index for joining order items with products:

```
olist\_order\_items\_dataset(product\_id);
```

Optimizes JOIN oi ON oi.product id = p.product id

This index improves the join performance between order items and products by making product\_id lookups more efficient. It allows the database to quickly find all order items for a given product, which is important for aggregating review scores and counts by product.

```
| -> Sort: total_reviews EESC, avg_score DESC (actual time=51509...51508) row=997 loops=1)
-> Filter: ([total_review> 5] and (avg_score > 4.5)) (actual time=513131..51508) row=997 loops=1)
-> Filter: ([total_review> 5] and (avg_score > 4.5)) (actual time=513131..51508) row=997 loops=1)
-> Alterian (actual time) (actual tim
```

3. Index for joining products with category translation:

```
olist products dataset(product category name);
```

```
Speeds up JOIN pct ON p.product category name = pct.product category name
```

This index supports the final join in the query, where the products dataset is joined with the category translation table using product\_category\_name. By indexing this column, the database can more efficiently match each product with its English category name, streamlining the grouping and output formatting process.

```
| -> Sort: total reviews DESC, avg_score DESC (actual time=465182.465182 rows=997 loops=1)
| -> Filter: ((total reviews -> 5) and (avg_score >= 4.5)) (actual time=663917.465187 sows=997 loops=1)
| -> Table can on temporaryy (actual time=463912.463612 rows=3917 loops=1)
| -> Aggregate using temporary table (actual time=663912.463912 rows=32172 loops=1)
| -> Nested loop inner join (cost=2097007 rows=112495) (actual time=333.434418 rows=10749 loops=1)
| -> Nested loop inner join (cost=2997007 rows=112495) (actual time=333.434418 rows=10749 loops=1)
| -> Nested loop inner join (cost=2997007 rows=112495) (actual time=333.434418 rows=10749 loops=1)
| -> Nested loop inner join (cost=2997007 rows=112495) (actual time=333.434418 rows=10749 loops=1)
| -> Nested loop inner join (cost=19929 rows=19842) (actual time=111.3092 rows=99224 loops=1)
| -> Nested loop inner join (cost=119929 rows=98842) (actual time=111.3092 rows=99224 loops=1)
| -> Table scan on r (cost=11203 rows=98842) (actual time=111.3092 rows=99224 loops=1)
| -> Single-row covering index lookup on o using PRIMMRY (order_id=-0.955 rows=1.14) (actual time=2.13..213 rows=1 loops=99224)
| -> Filter: (p.product_category_name is not null) (cost=0.5 rows=1) (actual time=0.6.1.6 rows=1.13 loops=99224)
| -> Single-row index lookup on put using PRIMMRY (product_id=0.050-0.050-0.5 rows=1) (actual time=0.488.0.0.488 rows=0.986 loops=112371)
| -> Single-row index lookup on put using PRIMMRY (product_id=0.050-0.050-0.5 rows=1) (actual time=0.488.0.0.488 rows=0.986 loops=112371)
| -> Single-row index lookup on put using PRIMMRY (product_id=0.050-0.050-0.5 rows=1) (actual time=0.488.0.0.488 rows=0.986 loops=112371)
| -> Single-row index lookup on put using PRIMMRY (product_id=0.050-0.050-0.5 rows=1) (actual time=0.488.0.0.488 rows=0.986 loops=112371)
```