In the Database Design markdown or pdf, provide the Data Definition Language (DDL) commands you used to create each of these tables in the database. Here's the syntax of the CREATE TABLE DDL command: CREATE TABLE table_name (column1 datatype, column2 datatype, column3 datatype,...);

Table Customers

```
CREATE TABLE customers (
customer_id VARCHAR PRIMARY KEY,
customer_zip_code_prefix INT,
customer_city VARCHAR,
customer_state VARCHAR
);
```

Table Geolocation

```
CREATE TABLE geolocation (
geolocation_zip_code_prefix INT,
geolocation_lat DECIMAL,
geolocation_lng DECIMAL,
geolocation_city VARCHAR,
geolocation_state VARCHAR
):
```

Table order_orders

```
CREATE TABLE orders (
order_id VARCHAR PRIMARY KEY,
customer_id VARCHAR,
order_status VARCHAR,
order_purchase_timestamp TIMESTAMP,
order_approved_at TIMESTAMP,
order_delivered_carrier_date TIMESTAMP,
order_delivered_customer_date TIMESTAMP,
order_estimated_delivery_date TIMESTAMP);
```

Table order items

```
CREATE TABLE order_items (
order_id VARCHAR,
order_item_id INT,
product_id VARCHAR,
seller_id VARCHAR,
shipping_limit_date TIMESTAMP,
price DECIMAL,
```

```
freight_value DECIMAL,
 PRIMARY KEY (order_id, order_item_id)
);
Table order_payments
CREATE TABLE order payments (
 order_id VARCHAR,
 payment sequential INT,
 payment_type VARCHAR,
 payment installments INT,
 payment_value DECIMAL,
 PRIMARY KEY (order_id, payment_sequential)
);
Table order_reviews
CREATE TABLE order reviews (
 review_id VARCHAR PRIMARY KEY,
 order id VARCHAR,
 review_score INT,
 review_comment_title VARCHAR,
 review comment message VARCHAR,
 review_creation_date TIMESTAMP,
 review_answer_timestamp TIMESTAMP
);
Table Products
CREATE TABLE products (
 product_id VARCHAR PRIMARY KEY,
 product_category_name VARCHAR,
 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
);
Table Sellers
CREATE TABLE sellers (
 seller_id VARCHAR PRIMARY KEY,
 seller_zip_code_prefix INT,
 seller_city VARCHAR,
 seller state VARCHAR
);
```

Table product_category_name_translation

```
CREATE TABLE product_category_name_translation ( product_category_name VARCHAR PRIMARY KEY, product_category_name_english VARCHAR );
```

ROWS COUNT

```
| table name
                                                   | row_count
| olist_customers_dataset
| olist_geolocation_dataset
| olist_order_items_dataset
| olist_order_payments_dataset
| olist_order_reviews_dataset
                                                          99442
                                                  | 1000164
                                                       102426
                                                       112650
                                                          99224
| olist orders dataset
                                                          99442
| olist products dataset
                                                          32329
| olist sellers dataset
                                                           3096 |
| product category name translation |
                                                             72 |
9 rows in set (4.21 sec)
```

ADVANCED Query:

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

```
SELECT
s.seller_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	seller_city	total_orders	avg_price	avg_freight
0015a82c2db000af6aaaf3ae2ecb0532	santo andre	3	895	21.020000457763672
001cca7ae9ae17fb1caed9dfb1094831	cariacica	240	105.1626248995463	36.99858363866806
001e6ad469a905060d959994f1b41e4f	sao goncalo	1	250	17.940000534057617
002100f778ceb8431b7a1020ff7ab48f	franca	55	22.445454198663885	14.43018214485862
003554e2dce176b5555353e4f3555ac8	goiania	1	120	19.3799991607666
004c9cd9d87a3c30c522c48c4fc07416	ibitinga	170	115.95711768655217	20.889588170893052
00720abe85ba0859807595bbf045a33b	guarulhos	26	38.750000073359566	12.153076777091393
00ab3eff1b5192e5f1a63bcecfee11c8	sao paulo	1	98	12.079999923706055
00d8b143d12632bad99c0ad66ad52825	belo horizonte	1	86	51.099998474121094
00ee68308b45bc5e2660cd833c3f81cc	sao paulo	319	110.4404068471496	17.52097165734043
00fc707aaaad2d31347cf883cd2dfe10	maringa	848	84.6035264483038	15.271910347325623
010543a62bd80aa422851e79a3bc7540	sao paulo	2	708	15.97499942779541
010da0602d7774602cd1b3f5fb7b709e	sao bernardo do campo	5	169.89999389648438	46.04999923706055
011b0eaba87386a2ae96a7d32bb531d1	pompeia	2	49.9900016784668	14.59000015258789
01266d4c46afa519678d16a8b683d325	curitiba	3	30.083333651224773	15.743333498636881

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
TM	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;
```

seller_id	seller_city	seller_state	
seller_id 3442f8959a84dea7ee197c632cb2df15 e9e446d01bd10a97a8ffcfc4a3a20cb2 ec2e006556300a79a5a91e4876ab3a56 05ca864204d09595ae591b93ea9cf93d 0692610d8abe24f287e9fae90ea0bbee ea00f977a203ff88adf7057cb7806998 ffcfefa19b08742c5d315f2791395ee5 c1dde11f12d05c478f5de2d7319ad3b2 bee36b4f9a2b9fdcaff6ec05df202ed0 baf15155e37ef5492731459bdc05be8a 71593c7413973a1e160057b80d4958f6 f12d3c2a14729ae461b920c11fe20fdc f0ec6a2adb05c62655a26dd347b8dede	campinas sao paulo sao paulo barueri sao paulo santos curitiba sao paulo sao paulo sao paulo sao paulo	+	

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

```
SELECT
    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

```
product_id: 639e019a

product_id: 7317d2c

product_id: 7717d2c

product_id: 7717d2c

product_id: 7717d2c

product_id: 555040a

product_id: 555040a

product_id: 555040a

product_id: 6365107

product_id: 6365108

product_id: 63652444

product_id: 63652444
```

Part 2 Indexing:

QUERY 1

Index Choice:

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.

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 scen on ttemporary (actual time=2282..2282 rows=15 loops=1)
-> Agpregate using temporary table (actual time=2282..2282 rows=2396 loops=1)
-> Nested loop inner join (cost=17386 rows=111760) (actual time=120..1857 rows=111024 loops=1)
-> Nested loop inner join (cost=17386 rows=111760) (actual time=109..1852 rows=111024 loops=1)
-> Table scen on oil (cost=1738 rows=111760) (actual time=109..1852 rows=111024 loops=1)
-> Table scen on oil (cost=1738 rows=111760) (actual time=109..1852 rows=11024 loops=1)
-> Table scen on oil (cost=1738 rows=111760) (actual time=0.001815..0.09427 rows=0.986 loops=112651)
-> Single-row index lookup on p using PRIMANY (scaler_id=0.product_id=0.product_id) (cost=0.25 rows=1) (actual time=0.00394..0.00397 rows=1 loops=112651)
-> Single-row index lookup on s using PRIMANY (scaler_id=p.seller_id) (cost=0.25 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.

CREATE INDEX idx_oi_product_id ON olist_order_items_dataset(product_id);

```
|-> Limit: 15 row(s) (actual time=3504..3504 rows=15 loops=1)
-> Table scan on <temporary> (actual time=3504..3504 rows=15 loops=1)
-> Algorigate using temporary> (actual time=5504..3504 rows=15 loops=1)
-> Nested loop inner join (cost=13315 rows=115078) (actual time=75.6..357 rows=311074 loops=1)
-> Nested loop inner join (cost=13315 rows=115078) (actual time=75.6..357 rows=31328 loops=1)
-> Nested loop inner join (cost=13715 rows=115078) (actual time=75.6..357 rows=3338 loops=1)
-> Nested loop inner join (cost=13715 rows=13078) (actual time=6.00ps=1)
-> Nested lookup (cost=2727 rows=3964) (actual time=6.00ps=1)
-> Covering index lookup on p using idx p seller id (seller id=seller id) (cost=2.91 rows=10.6) (actual time=0.109..0.161 rows=10.4 loops=3096)
-> Index lookup on oi using idx oi_product_id (product_id=p.product_id) (cost=2.91 rows=3.52) (actual time=0.05..0.0741 rows=3.43 loops=32329)
| -> Index lookup on oi using idx oi_product_id (product_id=p.product_id) (cost=2.91 rows=3.52) (actual time=0.05..0.0741 rows=3.43 loops=32329)
```

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 <temporaryy (actual time=350..3530 rows=28 loops=1)
-> Aggregate using temporary table (actual time=350..3530 rows=28 loops=1)
-> Rested loop inner join (cost=104es forws=94es) (actual time=1109..3448 rows=99224 loops=1)
-> Filter: (o.order_id * r.order_id) * (cost=94.4es forws=94es) (actual time=1059..1892 rows=99224 loops=1)
-> Inner hash join (chash)(o.order_id) * (cost=94.4es forws=94es) (actual time=1059..1892 rows=99224 loops=1)
-> Filter: (o.customer_id is not null) (cost=4.85 rows=94es) (actual time=49.6..655 rows=99412 loops=1)
-> Table scan on (cost=04.89 rows=98150) (actual time=49.6..61 rows=94es2 loops=1)
-> Hash
-> Table scan on (cost=10043 rows=95722) (actual time=80..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:

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);

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 #2)) (cost=314 rows=3996) (actual time=5490..5495 rows=15 loops=1)
-> Faller scan on s (cost=314 rows=3996) (actual time=74..81.1 rows=375 loops=1)
-> Select #2 (subquery in condition; run only once)
-> Filter: (s.seller_id = "construinized subquery".seller_id) (cost=0.0 rows=0) (actual time=19.6..19.6 rows=0.0543 loops=276)
-> limit to seller_id = "construinized subquery" seller_id) (cost=0.0 rows=0) (actual time=19.6..19.6 rows=0.0543 loops=276)
-> limit to seller_id = "construinized subquery" using custo distinct kept (seller_id=s.seller_id) (actual time=19.6..19.6 rows=0.0543 loops=276)
-> Materializer with deduplication (cost=0.0, rows=0) (actual time=5410..5412 rows=118 loops=1)
-> Filter: (avg(ci.freight_value) < 10) (actual time=5410..5412 rows=108 loops=1)
-> Table scan on temporary (actual time=5400..5412 rows=108 loops=1)
-> Aggregate using temporary table (actual time=5400..5402 rows=3061 loops=1)
-> Plater: (p.product_id is not null) (cost=5258 rows=3266) (actual time=530..545 rows=32952 loops=1)
-> Filter: (p.product_id is not null) (cost=5258 rows=3266) (actual time=59.6..545 rows=32952 loops=1)
-> Finder lookup on of using ids_of product_id (product_id=p.product_id) (cost=1.22 rows=3.76) (actual time=0.0692..0.127 rows=3.42 loops=32952)
```

INDEX Choice:

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 42)) (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 28 (subquery in conditions; run only once)
-> Filter: ((s.seller id = 'cmsterialized subquery'.seller id()) (cost=9941..99941 rows=1) (actual time=9.31..9.31 rows=0.0288 loops=521)
-> Limit: 1 row[s) (cost=59941..99941 rows=1) (actual time=9.31..9.32 rows=0.0288 loops=521)
-> Index lookup on cmsterialized subquery' using (auto distinct key) (seller id=seller id) (actual time=9.3..9.3 rows=0.0288 loops=521)
-> Materialize with deduplication (cost=59941..99941 rows=3003) (actual time=947..6447 rows=116 loops=1)
-> Filter: (avy[oi.freight_value) < 10) (cost=39637 rows=3003) (actual time=3475..4846 rows=116 loops=1)
-> Filter: (avy[oi.freight_value) < 10) (cost=39637 rows=3003) (actual time=3475..4846 rows=116 loops=1)
-> Rest cost of the scale of the
```

1. olist_order_items_dataset(product_id, freight_value)
CREATE INDEX idx_oi_product_freight ON olist_order_items_dataset(product_id, freight_value);

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.

2. olist_products_dataset(product_id, seller_id)

```
|-> Limit: 15 row(s) (cost=327 rows=15) (actual time=1556..1694 rows=15 loops=1)
-> Filter: (in optimizer)(s.seller_id, s.seller_id in (select f2)) (ocst=27 rows=3096) (actual time=1556..1604 rows=15 loops=1)
-> Table sonn on s (cost=237 rows=5096) (actual time=1.15..47.9 rows=520 loops=1)
-> Sellec sonn on sellec_id, s.seller_id) (sellec_id) (ocst=0.0 rows=0) (actual time=1.580..1604 rows=15 loops=1)
-> Ellec: (is.seller_id = 'casternalized_subquery'.seller_id) (cost=0.0 rows=0) (actual time=2.98..2.98 rows=0.0288 loops=521)
-> Limit: 1 row(s) (cost=0.0 rows=0) (actual time=2.98..2.98 rows=0.0288 loops=521)
-> Index lookup on casterialized_subquery_ using cauto_distinct_keyy (seller_id=s.seller_id) (actual time=2.98..2.98 rows=0.0288 loops=521)
-> Metrialize with deduplication (cost=0.0 rows=0) (actual time=1555..1558 rows=108 loops=1)
-> Filter: (wy(oi.freight_value) < 10) (actual time=1555..1558 rows=108 loops=1)
-> Loopseque using temporary table (actual time=1555..1578 rows=2098 loops=1)
-> Nexted loop inner join (cost=55037 rows=11076) (actual time=110.258 rows=1208 loops=1)
-> Table sonn on (cost=12256 rows=111760) (actual time=110.581 rows=11024 loops=1)
-> Single-row index lookup on p using FEIMARY (product_id=0.product_id) (cost=0.314 rows=1) (actual time=0.00604..0.00607 rows=0.986 loops=112651)
```

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.

3. olist_sellers_dataset(seller_id)

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:

1. Single column index on freight value

Best for optimizing the WHERE condition.

olist order items dataset(freight value);

```
- Table seas on conto temporaryo (controlled .102 power-103) (extent time-103 .103 power-103 loopers)

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- Table seas on conto temporaryo (controlled .103 power-103 loopers)

- Table seas on conto temporaryo (controlled .103 power-103 loopers)

- Table seas on conto temporaryo (controlled .103 power-103 loopers)

- Table seas on conto temporaryo (controlled .103 power-103 loopers)

- Table seas on conto temporaryo (controlled .103 power-103 loopers)

- Table seas on cont
```

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);

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);

```
| > Table scan on conton temporary (contrible..1002 rounties) (actual time=001.003 rounties) (actual time=00.003 rount
```

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

```
| -> Sort: total_reviews DESC, avg_score DESC (actual time=456607.456607 rows=997 loops=1)
-> Filter: ([total_reviews > 5] and (avg_score >= 4.5)) (actual_time=455051.458602 rows=997 loops=1)
-> Filter: ([total_reviews > 5] and (avg_score >= 4.5)) (actual_time=455047.458602 rows=1997 loops=1)
-> Apgregate using temporary stable (cutsal_time=45547.458602 rows=10272) loops=1)
-> Nested loop inner join (cost=290396 rows=112495) (actual_time=12.426430 rows=110791 loops=1)
-> Nested loop inner join (cost=290396 rows=112495) (actual_time=145.372551 rows=112371 loops=1)
-> Nested loop inner join (cost=290396 rows=112495) (actual_time=145.372551 rows=112371 loops=1)
-> Nested loop inner join (cost=19039 rows=112495) (actual_time=145.372551 rows=112371 loops=1)
-> Nested loop inner join (cost=19039 rows=102487 (actual_time=145.372551 rows=112371 loops=1)
-> Nested loop inner join (cost=19039 rows=98842) (actual_time=145.372551 rows=112371 loops=1)
-> Single-row covering index lookup on using RFIMMAY (roter_id=rows=1) (actual_time=0.918.1.05 rows=1.18 loops=99224)
-> Filter: [p_product_catepory_name is not_null] (cost=0.988 rows=1) (actual_time=0.918.0.078 rows=0.986 loops=112371)
-> Single-row index lookup on puting RFIMMAY (product_category_name=p_product_category_name) (cost=1 rows=1) (actual_time=0.015..0.0151 rows=1 loops=110749)
```

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 IRSC, Myg score IRSC (actual time=51493, 51490) row=997 logg=1)
|-> Filter: (Total reviews > 5) and (avg gone > 4.5)) (actual time=514945, 514996 row=997 logg=1)
|-> Fable scan in ottemporary (actual time=514938,.51498 row=2012) (loop=1)
|-> Apgregate using temporary table (actual time=519398,.513498 row=3217) loop=1)
|-> Nested loop inner join (cost=39280 row=11292) (actual time=101.202248 row=110749 loop=1)
|-> Nested loop inner join (cost=39280 row=11292) (actual time=61.34980 row=30290) (actual
```

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.

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.

EXPLAIN ANALYZE

SELECT

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;