PROJECT NAME: Analyse product data for an online sports retail company to optimize revenue

Project Description

Sports clothing is a booming sector!

In this notebook, I will use my SQL skills to analyse product data for an online sports retail company.

In this project, I'll will work with **numeric**, **string**, **and timestamp data** on pricing and revenue, ratings, reviews, descriptions, and website traffic. And I will be using techniques such as **aggregation**, **cleaning**, **labelling**, **Common Table Expressions(CTE)**, and **correlation** to produce recommendations on how the company can maximize revenue.

Task List

- 1. Count the total number of products, along with the number of nonmissing values in description, listing price, and last visited
- 2. Find out how listing price varies between Adidas and Nike products.
- 3. Create labels for products grouped by price range and brand.
- 4. Calculate the correlation between reviews and revenue.
- 5. Split description into bins in increments of one hundred characters, and calculate average rating by for each bin.
- 6. Count the number of reviews per brand per month.
- 7. Create the footwear CTE, then calculate the number of products and average revenue from these items.
- 8. Copy the code used to create footwear then use a filter to return only products that are not in the CTE.

The sports clothing and athleisure industry is a massive market, valued at approximately \$193 billion in 2021, with strong growth projections for the coming decade. In this notebook, I will be playing the role of a product analyst for an online sports clothing company. The company is specifically interested in how it can improve revenue. I will dive into **product data** such as pricing, reviews, descriptions, and ratings, as well as revenue and website traffic, to produce recommendations for its **marketing and sales teams**.

The database provided to us, sports, contains **five tables**, with **product_id** being the primary key for all of them:

info

description	data type	column
Name of the product	varchar	product_name
Unique ID for product	varchar	product_id
Description of the product	varchar	description

finance

description	data type	column
Unique ID for product	varchar	product_id
Listing price for product	float	listing_price
Price of the product when on sale	float	sale_price
Discount, as a decimal, applied to the sale price	float	discount
Amount of revenue generated by each product, in US dollars	float	revenue

reviews

description	data type	column
Name of the product	varchar	product_name
Unique ID for product	varchar	product_id
Product rating, scored from 1.0 to 5.0	float	rating
Number of reviews for the product	float	reviews

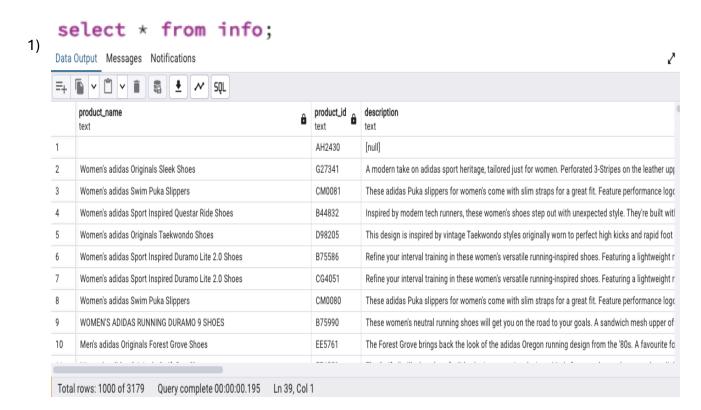
traffic

description	data type	column
Unique ID for product	varchar	product_id
Date and time the product was last viewed on the website	timestamp	last_visited

brands

column	data type	description
product_id	varchar	Unique ID for product

The tables have following columns and rows when the "select" command is executed.



select * from finance;

2)

Data Output Messages Notifications SQL. =+ discount product_id listing_price sale_price revenue text numeric numeric numeric numeric 1 AH2430 [null] [null] [null] [null] 75.99 2 1641.17 G27341 37.99 0.5 CM0081 398.93 3 9.99 5.99 0.4 4 B44832 69.99 34.99 0.5 2204.37 D98205 5 79.99 39.99 0.5 5182.7 6 B75586 47.99 19.2 0.6 1555.2 7 CG4051 23.99 0.5 86.36 47.99 8 CM0080 9.99 5.99 75.47 0.4 B75990 27.99 9 55.99 0.5 806.11 65.99 2779.22 10 EE5761 39.59 0.4 0.4 11 EE4553 75.99 45.59 2954.23 Total rows: 1000 of 3179 Query complete 00:00:00.270 Ln 24, Col 1

select * from reviews;

Data Output Messages Notifications

Data	Output Mess	ages Notifi	cations		
=+	- ~ - ~		* [~]	SQL	
	product_id text	rating numeric	reviews numeric		
1	AH2430	[null]	[null]		
2	G27341	3.3	24		
3	CM0081	2.6	37		
4	B44832	4.1	35		
5	D98205	3.5	72		
6	B75586	1	45		
7	CG4051	4.4	2		
8	CM0080	2.8	7		
9	B75990	4.5	16		
10	EE5761	4	39		
11	EE4553	2.7	36		
Tota	al rows: 1000 of	3179 Que	ery complete	00:00:00.300	Ln 32, Col 1

select * from traffic;

Data Output Messages Notifications

=+	□ ∨ □ 、		✓ SQL
	product_id text	last_visited_date date	last_visited_time time with time zone
1	AH2430	2018-05-19	15:13:00+05:30
2	G27341	2018-11-29	16:16:00+05:30
3	CM0081	2018-02-01	10:27:00+05:30
4	B44832	2018-09-07	20:06:00+05:30
5	D98205	2019-07-18	15:26:00+05:30
6	B75586	2019-01-30	12:09:00+05:30
7	CG4051	2019-03-22	16:36:00+05:30
8	CM0080	2019-03-10	01:46:00+05:30
9	B75990	2018-05-29	08:16:00+05:30
10	EE5761	2019-11-29	17:22:00+05:30
11	EE4553	2018-06-26	23:34:00+05:30
Tota	al rows: 1000 o	f 3179 Query cor	mplete 00:00:00.251

select * from brand; 5) Data Output Messages Notifications 霜 SQL product_id brand character varying text 1 AH2430 [null] 2 G27341 Adidas CM0081 Adidas 3 4 B44832 Adidas D98205 Adidas 5 B75586 Adidas 6 7 CG4051 Adidas CM0080 Adidas 8 B75990 Adidas 9 10 EE5761 Adidas EE4553 Adidas 11

Query complete 00:00:00.259

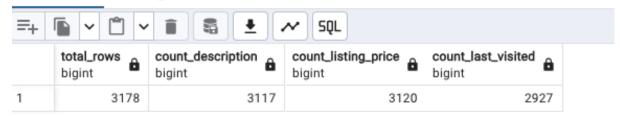
Ln 53, Col 21

Now let's start working on the "Task Lists".

Total rows: 1000 of 3179

1. Counting missing values

```
--1)
SELECT COUNT(info) as total_rows,
COUNT(info.description) as count_description, COUNT(finance.listing_price) as count_listing_price,
COUNT(traffic.last_visited_date) as count_last_visited
FROM info
INNER JOIN traffic
ON traffic.product_id = info.product_id
INNER JOIN finance
ON finance.product_id = info.product_id;
```



- 1) **'total_rows'**: The total number of rows or entries in the analysed dataset is 3,179 rows.
- 2) 'count_description': The total number of entries in the description column that have non-empty values is 3,117 entries.
- 3) `count_listing_price`: The total number of entries in the listing_price column that have non-empty values is 3,120 entries.
- 4) `count_last_visited`: The total number of entries in the last_visited column that have non-empty values is 2,928 entries.

2. Nike vs Adidas pricing

We can see the database contains 3,179 products in total. Of the columns we previewed, only one **last_visited** is missing more than five percent of its values. Now let's turn our attention to pricing.

How do the price points of Nike and Adidas products differ? Answering this question can help us build a picture of the company's stock range and customer market. We will run a query to produce a distribution of the **listing_price** and the count for each price, grouped by **brand**.

```
SELECT brand.brand, CAST(finance.listing_price AS INTEGER), COUNT (finance.product_id)
FROM brand
INNER JOIN finance
ON finance.product_id = brand.product_id
WHERE finance.listing_price > 0
GROUP BY brand.brand, finance.listing_price
ORDER BY finance.listing_price asc;
```

Data Output Messages Notifications

=+		v i s	<u>•</u> ~ 5
	brand text	listing_price integer	count bigint
1	Adidas	9	1
2	Adidas	10	11
3	Adidas	12	1
4	Adidas	13	27
5	Adidas	15	27
6	Adidas	16	4
7	Adidas	18	4
8	Adidas	20	8
9	Adidas	23	1
10	Adidas	25	28
11	Adidas	27	18
12	Adidas	28	38
13	Nike	30	2
14	Adidas	30	37
15	Adidas	33	24
16	Adidas	36	25
17	Adidas	38	24
18	Nike	40	1
19	Adidas	40	81
20	Adidas	43	51
21	Nike	45	3
22	Adidas	45	1
Total	rows: 77 of	77 Query cor	nplete 00:00:

And the result continues till 77th row, as be seen in output of the query.

3. Labelling price ranges

It turns out there are 77 unique prices for the products in our database, which makes the output of our last query quite difficult to analyse.

Let's build on our previous query by assigning labels to different price ranges, grouping by **brand** and **label**. We will also include the total **revenue** for each price range and **brand**.

```
SELECT b.brand, COUNT(f.*), SUM(f.revenue) as total_revenue,
CASE WHEN f.listing_price < 42 THEN 'Budget'
    WHEN f.listing_price >= 42 AND f.listing_price < 74 THEN 'Average'
    WHEN f.listing_price >= 74 AND f.listing_price < 129 THEN 'Expensive'
    ELSE 'Elite' END AS price_category
FROM finance AS f
INNER JOIN brand AS b
    ON f.product_id = b.product_id
WHERE b.brand IS NOT NULL
GROUP BY b.brand, price_category
ORDER BY total_revenue DESC;</pre>
```

Data Output	Messages	Notifications
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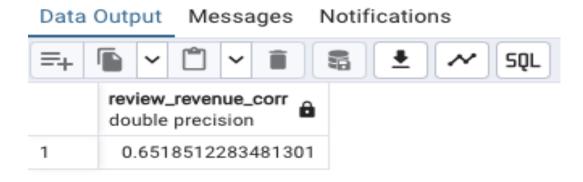
=+	~ <u></u>	v i s		ĮL
	brand text	count bigint	total_revenue numeric	price_category text
1	Adidas	849	4626980.07	Expensive
2	Adidas	1060	3233661.06	Average
3	Adidas	307	3014316.83	Elite
4	Adidas	359	651661.12	Budget
5	Nike	357	595341.02	Budget
6	Nike	82	128475.59	Elite
7	Nike	90	71843.15	Expensive
8	Nike	16	6623.50	Average

The majority of Adidas products fall into the "Average" category, totalling 1060 products. However, the brand also excels in the "Expensive" category, generating a total revenue of \$4,626,980.07. Additionally, Adidas presents luxury products in the "Elite" category, although the quantity is lower (307) products), they contribute significantly to a total revenue of \$3,014,316.83. Affordable products are not neglected, as the brand offers 359 products in the "Budget" category, even though the total revenue from this category is lower. On the other hand, the brand "Nike" also exhibits price variation in its products. Products in the "Budget" category are the primary choice for customers, amounting to 357 products, representing the majority of sales. Despite their large quantity, the total revenue from this category is lower compared to the "Adidas" brand, totalling \$595,341.02. On the flip side, Nike also presents luxury products in the "Elite" category, with a smaller quantity (82 products), yet generating a total revenue of \$128,475.59. Notably, Nike's products in the "Expensive" and "Average" categories also contribute significantly to the total revenue.

4. Correlation between revenue and reviews

To improve revenue further, the company could try to reduce the amount of discount offered on Adidas products, and monitor sales volume to see if it remains stable. Alternatively, it could try offering a small discount on Nike products. This would reduce average revenue for these products, but may increase revenue overall if there is an increase in the volume of Nike products sold.

```
SELECT CORR(reviews.reviews, revenue) AS review_revenue_corr
FROM reviews
INNER JOIN finance
ON finance.product_id = reviews.product_id;
```



The correlation result between revenue and reviews is approximately 0.651. This indicates a moderate positive relationship between the two variables. With a correlation value exceeding 0.5, it can be concluded that the higher the number of reviews for a product, the tendency is for an increase in the product's revenue as well. However, it's important to note that correlation does not imply causation, meaning it cannot be determined whether more reviews directly cause an increase in revenue or vice versa.

5. Ratings and reviews by product description length

Interestingly, there is a strong positive correlation between **revenue** and **reviews**. This means, potentially, if we can get more reviews on the company's website, it may increase sales of those items with a larger number of reviews.

Perhaps the length of a product's **description** might influence a product's **rating** and **reviews**, if so, the company can produce content guidelines for listing products on their website and test if this influences **revenue**.

```
SELECT TRUNC(LENGTH(i.description), -2) AS description_length,
    ROUND(AVG(r.rating::numeric), 2) AS average_rating
FROM info AS i
INNER JOIN reviews AS r
    ON i.product_id = r.product_id
WHERE i.description IS NOT NULL
GROUP BY description_length
ORDER BY description_length;
```

Data	Output Messages	Notifications
=+		SQL SQL
	description_length numeric	average_rating numeric
1	0	1.87
2	100	3.20
3	200	3.27
4	300	3.29
5	400	3.32
6	500	3.12
7	600	3.65

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The results indicate a relationship between the length of product descriptions (**description_length**) and the average product rating (**average_rating**). It is evident that as the length of product descriptions increases, the average rating tends to rise. For instance, products with a description length of 600 characters have a higher average rating (3.65) compared to products with shorter descriptions.

6. Reviews by month and brand

As we know a correlation exists between **reviews** and **revenue**, one approach the company could take is to run experiments with different sales processes encouraging more reviews from customers about their purchases, such as by offering a small discount on future purchases.

```
SELECT b.brand, DATE_PART('month', t.last_visited_date) AS month, COUNT(r.*) AS num_reviews
FROM brand AS b
INNER JOIN traffic AS t
    ON b.product_id = t.product_id
INNER JOIN reviews AS r
    ON t.product_id = r.product_id
GROUP BY b.brand, month
HAVING b.brand IS NOT NULL
    AND DATE_PART('month', t.last_visited_date) IS NOT NULL
ORDER BY b.brand, month;
```

Data Output Messages Notifications

=+		v i s	✓ SQL
	brand text	month double precision	num_reviews bigint
1	Adidas	1	253
2	Adidas	2	272
3	Adidas	3	269
4	Adidas	4	180
5	Adidas	5	172
6	Adidas	6	159
7	Adidas	7	170
8	Adidas	8	189
9	Adidas	9	181
10	Adidas	10	192
11	Adidas	11	150
Total	rows: 24 of	24 Query comple	ete 00:00:00.248

On the Adidas side, there's an interesting variation in their review counts from month to month:

- Starting from January, Adidas received 253 reviews.
- The review count increased to 272 in February.
- March and April also showed high review counts with 269 and 180 reviews, respectively.
- May and June followed with 172 and 159 review counts.
- Then, July, August, and September recorded 170, 189, and 181 reviews.
- October exhibited a significant increase with 192 reviews, followed by November with 150 reviews.
- The year ended with December having 190 reviews.

On the Nike side, the review trend also captures attention:

- In January and February, Nike received an equal number of reviews, 52 each.
- March showed an increase to 55 reviews.
- Subsequent months like April and May recorded 42 and 41 reviews.
- June displayed a slight increase with 43 reviews.
- July had 37 reviews, while August and September had 29 and 28 reviews, respectively.

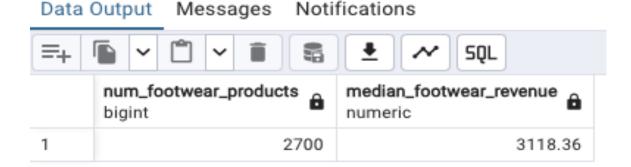
- October showed a drastic increase with 47 reviews.
- November and December then decreased again with 38 and 35 reviews each.

7. Footwear product performance

Looks like product reviews are highest in the first quarter of the calendar year, so there is scope to run experiments aiming to increase the volume of reviews in the other nine months!

So far, we have been primarily analysing Adidas vs Nike products. Now, let's switch our attention to the type of products being sold. As there are no labels for product type, we will create a Common Table Expression (CTE) that filters **description** for keywords, then use the results to find out how much of the company's stock consists of footwear products and the median **revenue** generated by these items.

```
WITH footwear AS
(
    SELECT i.description, f.revenue
    FROM info AS i
    INNER JOIN finance AS f
        ON i.product_id = f.product_id
    WHERE i.description ILIKE '%shoe%'
        OR i.description ILIKE '%trainer%'
        OR i.description ILIKE '%foot%'
        AND i.description IS NOT NULL
)
SELECT COUNT(*) AS num_footwear_products,
        percentile_disc(0.5) WITHIN GROUP (ORDER BY revenue) AS median_footwear_revenue
FROM footwear;
```



The results indicate that there are 2,700 footwear products in the analysed dataset. The median revenue for these footwear products is \$3,118.36. This suggests that half of the footwear products have revenue above \$3,118.36 and the other half have revenue below that figure.

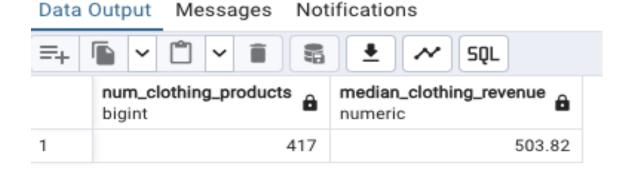
8. Clothing product performance

Recalling from the first task that we found there are 3,117 products without missing values for **description**. Of those, 2,700 are footwear products, which accounts for around 85% of the company's stock. They also generate a median revenue of over \$3000 dollars!

This is interesting, but we have no point of reference for whether footwear's **median_revenue** is good or bad compared to other products. So, for our final task, let's examine how this differs to clothing products. We will reuse **footwear**, adding a filter afterward to count the number of products and **median revenue** of products that are not in **footwear**.

```
WITH footwear AS
(
    SELECT i.description, f.revenue
    FROM info AS i
    INNER JOIN finance AS f
        ON i.product_id = f.product_id
    WHERE i.description ILIKE '%shoe%'
            OR i.description ILIKE '%trainer%'
            OR i.description ILIKE '%foot%'
            AND i.description IS NOT NULL
)

SELECT COUNT(i.*) AS num_clothing_products,
            percentile_disc(0.5) WITHIN GROUP (ORDER BY f.revenue) AS median_clothing_revenue
FROM info AS i
INNER JOIN finance AS f on i.product_id = f.product_id
WHERE i.description NOT IN (SELECT description FROM footwear);
```



The results indicate the presence of 417 clothing products in the analyzed dataset. The median revenue of these clothing products is \$503.82. This means that half of the clothing products have revenue above \$503.82, and the other half have revenue below that figure.

CONCLUSION

- 1. The brand needs to explore opportunities to develop products in the "Expensive" and "Elite" categories that have higher revenue potential.
- 2. Focusing on product quality, customer service, and holistic marketing strategies can help improve reviews and revenue.
- 3. Analysing factors that influence monthly review fluctuations and planning appropriate marketing strategies.
- 4. Continuously monitoring product categories like footwear and clothing and making relevant price adjustments or marketing strategies.
- 5. Using this data as a foundation to design more effective and customeroriented business strategies.
- 6. All of these recommendations can assist the brand in enhancing product performance, increasing revenue, and providing a better experience to customers.