Fetch - Data Analyst Take Home

Exercise:

First: explore the data

Review the unstructured csv files and answer the following questions with code that supports your conclusions:

- Are there any data quality issues present?
- Are there any fields that are challenging to understand?

Two step Approach

- Data Ingestion: First step would be to import all three csv files as tables into a database lets call this database dbs. Let's assume our schema name is fetch, let's assume three new tables are created namely; fetch.product, fetch_user, fetch.transactions.
- 2. Data Exploration by each table: SQL Script to Check Data Quality Issues & Data Types
- 2.1) Products table : fetch.products , primary key barcode (underlying assumption , each barcode several columns have missing data, which may impact analysis and decision-making:

1. Products.csv Data

Data Quality Issues:

- CATEGORY_4: 92% missing (only 67,459 non-null out of 845,552).
- CATEGORY 3: 7.1% missing (784,986 non-null).
- MANUFACTURER & BRAND: 27% missing (~619,000 non-null).
- BARCODE: ~4,000 missing values (841,527 non-null).
- Potential Issue: Missing CATEGORY_4 values might mean this field is optional or inconsistently filled. Missing MANUFACTURER and BRAND data could affect product identification and grouping.
- BARCODE Field Issue: The BARCODE column is stored as float64, which is incorrect
 for a barcode. Floating-point storage may introduce rounding issues or convert long
 numeric values into scientific notation (e.g., 8.068109e+11). Some barcodes might have

- non-numeric characters or inconsistent lengths. Potential Fix: Convert BARCODE to string (VARCHAR) in SQL to prevent rounding issues.
- Duplicates & Data Integrity Issues: If a duplicate check returns multiple records with
 the same BARCODE, CATEGORY_1, CATEGORY_2, etc., then:Either duplicates
 exist(bad data entry), OR Some fields are not unique enough to differentiate
 products.Potential Fix: Ensure BARCODE is a unique identifier if available. Otherwise,
 use a combination of fields (CATEGORY_1, BRAND, etc.) to detect true duplicates.
- Inconsistent Category Hierarchy: CATEGORY_1 → CATEGORY_2 → CATEGORY_3

 → CATEGORY_4 should form a logical hierarchy. If a product has CATEGORY_3 but no
 CATEGORY_2, it violates the hierarchy. Potential Fix: Check if higher-level categories
 always exist when a lower-level one is present.

Challenging Fields:

- CATEGORY_4 Field: The high percentage of missing values makes it unclear whether this column is important. Does this field add value, or should it be ignored if CATEGORY_1-3 are enough?
- BARCODE as an Identifier:Ilf BARCODE is missing or duplicated, what is the alternative unique identifier?Are manufacturer-assigned barcodes included, or are there

2. Transaction.csv Data

Data Quality Issues:

- Missing Values in BARCODE:
- Only 44,238 out of 50,000 transactions have a BARCODE.
- Some products might be missing barcodes, or there could be data entry errors.
- FINAL QUANTITY & FINAL_SALE Have Inconsistent Values:
- FINAL QUANTITY contains values like "zero", which should be numeric.
- Some FINAL SALE entries are empty, possibly missing transaction amounts.
- Timestamp Format in SCAN DATE & PURCHASE DATE:
- Check for time zone mismatches.

Challenging Fields:

- FINAL QUANTITY: Does "zero" mean a failed transaction, refund, or incorrect entry?
- FINAL_SALE: Missing values—are these zero-dollar transactions, or was the sale amount lost in processing?
- BARCODE Missing: How should transactions without barcodes be handled?private-label products without barcodes

SQL Script to Check Data Quality Issues & Data Types

This script will:

- Check data types for each column.
- Find missing values in each column.
- Identify duplicate records.
- Detect inconsistent data formats (e.g., non-numeric BARCODE values).
- Check string length variations for categorical column
- BARCODE Missing: How should transactions without barcodes be handled?

3.User.csv Data

Data Quality Issues:

- Missing Values:
 - BIRTH_DATE: Missing for 3.7% of users.
 - o STATE: 4.8% missing.
 - LANGUAGE: Over 30% missing, making it unreliable for personalization.
 - o GENDER: 5.9% missing, could be due to user preference or system error.
- Date Format Issues:
 - BIRTH_DATE and CREATED_DATE have timestamps- need to check with daat source provider if Birth Date needs stamp
 - Check for users without birth dates but valid accounts.

Challenging Fields:

- LANGUAGE: If 30% is missing, should it still be used for customer segmentation?
- GENDER: Should missing values be treated as "unknown", or is this a system flaw?
- BIRTH DATE: Missing fields
 - Using SQL to identify Data Quality Issues taking products data as example, similarly this process can be repeated for user and transactions data.
- 1. Check Data Types of Each Column

SELECT COLUMN NAME, DATA TYPE

FROM INFORMATION_SCHEMA.COLUMNS

```
WHERE TABLE_NAME = 'PRODUCTS';
```

2. Check for Missing Values

SELECT

'CATEGORY_1' AS Column_Name, COUNT(*) AS Null_Count FROM PRODUCTS WHERE CATEGORY 1 IS NULL

UNION ALL

SELECT

'CATEGORY_2', COUNT(*) FROM PRODUCTS WHERE CATEGORY_2 IS NULL

UNION ALL

SELECT

'CATEGORY_3', COUNT(*) FROM PRODUCTS WHERE CATEGORY_3 IS NULL

UNION ALL

SELECT

'CATEGORY_4', COUNT(*) FROM PRODUCTS WHERE CATEGORY_4 IS NULL

UNION ALL

SELECT

'MANUFACTURER', COUNT(*) FROM PRODUCTS WHERE MANUFACTURER IS NULL

UNION ALL

SELECT

'BRAND', COUNT(*) FROM PRODUCTS WHERE BRAND IS NULL

UNION ALL

SELECT

'BARCODE', COUNT(*) FROM PRODUCTS WHERE BARCODE IS NULL;

3. Check for Duplicate Records

SELECT CATEGORY_1, CATEGORY_2, CATEGORY_3, CATEGORY_4, MANUFACTURER, BRAND, BARCODE, COUNT(*)

FROM PRODUCTS

GROUP BY CATEGORY_1, CATEGORY_2, CATEGORY_3, CATEGORY_4, MANUFACTURER, BRAND, BARCODE

HAVING COUNT(*) > 1;

4. Check for Non-Numeric BARCODE Values

SELECT BARCODE

FROM PRODUCTS

WHERE TRY_CAST(BARCODE AS BIGINT) IS NULL AND BARCODE IS NOT NULL;

5. Detect String Length Issues in Categorical Columns

SELECT

'CATEGORY_1' AS Column_Name, MIN(LEN(CATEGORY_1)) AS Min_Length, MAX(LEN(CATEGORY_1)) AS Max_Length FROM PRODUCTS

UNION ALL

SELECT

'CATEGORY_2', MIN(LEN(CATEGORY_2)), MAX(LEN(CATEGORY_2)) FROM PRODUCTS

UNION ALL

UNION ALL

SELECT

'CATEGORY_3', MIN(LEN(CATEGORY_3)), MAX(LEN(CATEGORY_3)) FROM PRODUCTS

```
SELECT
```

'CATEGORY_4', MIN(LEN(CATEGORY_4)), MAX(LEN(CATEGORY_4)) FROM PRODUCTS

UNION ALL

SELECT

'MANUFACTURER', MIN(LEN(MANUFACTURER)), MAX(LEN(MANUFACTURER)) FROM PRODUCTS

UNION ALL

SELECT

'BRAND', MIN(LEN(BRAND)), MAX(LEN(BRAND)) FROM PRODUCTS;

Second: provide SQL queries

Answer three of the following questions with at least one question coming from the closed-ended and one from the open-ended question set. Each question should be answered using one query.

Closed-ended questions:

What are the top 5 brands by receipts scanned among users 21 and over

—-#get all users of age 21 and above , where age is the difference in current year and birth date year. We will use a CTE for code efficiency and to avoid a user subquery on the main query down below.

```
With user_over_21 as

(SELECT distinct user_id

FROM om fetch.user

where birth_date is not null

AND date_diff('year', birth_date :: date, getdate() :: date) >=21)
```

```
SELECT p.brand, count(distinct t.receipt_id) receipts_count
FROM fetch.transactions t
INNER JOIN fetch.products p on t.barcode = p.barcode —#left join to bring in brand from
product table
Inner join user_helper_table uh on t.userid =uh.userid —-# only analyze for users above 21
WHERE p.brand is not null
Group by p.brand
Order by receipts_count desc
Limit 5:
   • What are the top 5 brands by sales among users that have had their account for at least
       six months?
       —get total sales from transactions by each brand, filter it on top 5 brands, further filter it
       for users who purchased at least 6 months prior to today's date
SELECT
  p.brand,
  SUM(t.final sale) AS total sales
FROM fetch.transactions t
JOIN fetch.product as p ON t.barcode=p.barcode
WHERE t.purchase_date <= DATEADD(month, -6, CURRENT_DATE)
GROUP BY
  p.brand
ORDER BY
  total_sales DESC
LIMIT 5;
```

What is the percentage of sales in the Health & Wellness category by generation?

```
—classify user age into generations using birth year i am creating a 4 category based
generation mapping to birth year
WITH UserGenerations AS (
  SELECT user id,
Case when date trunc('year', birth date') :: date BETWEEN '1946-01-01' AND
'1964-01-01' then 'Baby Boomer'
     when date_trunc('year', birth_date') :: date BETWEEN '1965-01-01" AND
'1980-01-01' then 'Gen X'
      When date_trunc('year', birth_date') :: date BETWEEN '1981-01-01' AND
'1996-01-01' then 'Millennial'
      When date trunc('year', birth date') :: date '1997-01-01' AND '2012-01-01' then
'Gen Z'
      ELSE 'Alpha' END AS Generation FROM fetch.user),
SalesByGeneration AS (
  SELECT U.Generation,
    SUM(case when p.CATEGORY_1 = 'Health & Wellness' then
CAST(T.FINAL SALE AS FLOAT) ELSE 0 END) AS HealthWellnessSales,
    SUM(CAST(T.FINAL SALE AS FLOAT)) AS TotalSales
  FROM fetch.transaction as T
  Inner join UserGenerations as U ON T.USER_ID = U.USER_ID
  Join fetch.products P ON T.BARCODE = P.BARCODE
  Where T.FINAL SALE IS NOT NULL AND T.FINAL SALE != "
  GROUP BY U.Generation
SELECT
```

Generation,

(HealthWellnessSales * 100.0 / NULLIF(TotalSales, 0)) AS HealthWellnessPercentage

FROM SalesByGeneration;

Open-ended questions: for these, make assumptions and clearly state them when answering the question.

1. Who are Fetch's power users?

Assumptions:Definition of Power Users: Users who frequently scan receipts and have high total spending.Measurement Criteria:

- Number of transactions per user.
- Total sales per user.
- Frequency of purchases over time.

Top power users key metrics:

- Power users are those with the highest number of transactions and total sales.
- The top user (USER_ID: 64e62de5ca929250373e6cf5) has made 22 transactions, totaling \$57.65 in sales.
- Another notable user (USER_ID: 60a5363facc00d347abadc8e) has fewer transactions (14) but the highest spending (\$101.97).
- Conclusion: Fetch's power users are those who either:
- Frequently make purchases (high transaction count)
- Spend significantly on each transaction (high total sales)

2. Which is the leading brand in the Dips & Salsa category?

Assumption: We determine the leading brand by total sales revenue rather than the number of transactions or units sold.

• Finding: TOSTITOS is the top-selling brand in the Dips & Salsa category, with total sales of \$103,354.84.

3. At what percent has Fetch grown year over year?

Assumptions: Four assumptions - Sales Data is Complete and Accurate , Fetch's Business Model Remains Consistent , Excluding Invalid or Zero Sales Transactions are excluded, Yearly Sales Fluctuations Reflect Real Growth and Sales Growth is the Primary Metric of Business Growth.

```
WITH SalesByYear AS (
  SELECT left(purchase_date,4) as year,
    SUM(CAST(final_sale AS FLOAT)) AS Total_Sales
  FROM fetch.transactions
  Where FINAL SALE IS NOT NULL AND FINAL SALE != "
  Group by year ),
YoY_Growth AS (
  SELECT
   year,
    Total Sales,
    LAG(Total Sales) OVER (ORDER BYyear) AS Previous Year Sales,
    ((Total_Sales - LAG(Total_Sales) OVER (ORDER BY Year)) * 100.0 /
    NULLIF(LAG(Total_Sales) OVER (ORDER BY Year), 0)) AS YoY_Growth_Percent
  FROM SalesByYear
)
SELECT * FROM YoY_Growth ORDER BY Year DESC;
```

Third: communicate with stakeholders

Construct an email or slack message that is understandable to a product or business leader who is not familiar with your day-to-day work. Summarize the results of your investigation. Include:

- Key data quality issues and outstanding questions about the data
- One interesting trend in the data
 - Use a finding from part 2 or come up with a new insight
- Request for action: explain what additional help, info, etc. you need to make sense of the data and resolve any outstanding issues

Email:

Hi Person X(Business/Product leader)

I've been reviewing Fetch's sales and user data, and wanted to share some key findings, along with a few areas where I need clarification to ensure we're making the most of our insights.

Insights and key trends:

- Our most active users fall into two groups:
- 1. Those who make <u>frequent purchases</u> (up to 22 transactions).
- 2. Those who <u>spend more per transaction</u> (one top user made only 14 purchases but spent over **\$100**).

This suggests we may have different types of high-value users—one driven by frequency, the other by order size. Understanding this split could help us tailor engagement strategies.

 Additionally, <u>Tostitos</u> is by far the best-selling brand in the <u>Dips & Salsa category</u>, with \$103K+ in sales—far ahead of competitors. It would be helpful to know whether this is due to special promotions, brand partnerships, or organic customer preference.

Next Steps & Support Needed

To refine these insights and ensure accuracy, I'd appreciate your inputs on:

- 1. **Handling missing/zero-dollar sales**—We found some zero dollar sales, could you please help share business insights on what type of sale is a zero dollar sale, this will help us decide if they should be excluded or investigated further?
- 2. **Purchase date inconsistencies**—To investigate if this is a known data issue, could you please direct me to the 'Purchase' team manager.
- 3. **Brand partnerships or promotions**—is there any additional context behind Tostitos' strong performance?

Would love your thoughts on these, and let me know if there's someone else I should check in with for further details. Happy to discuss!

Thanks and Regards, Dakshika Chauhan