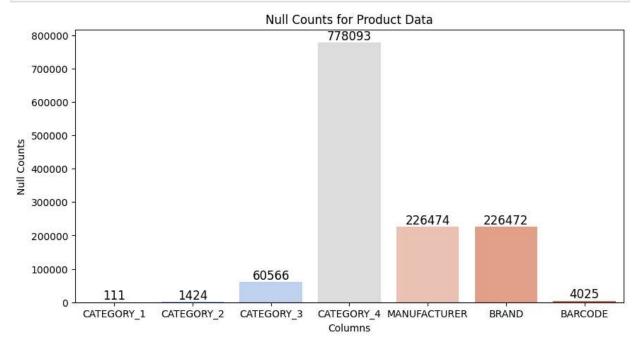
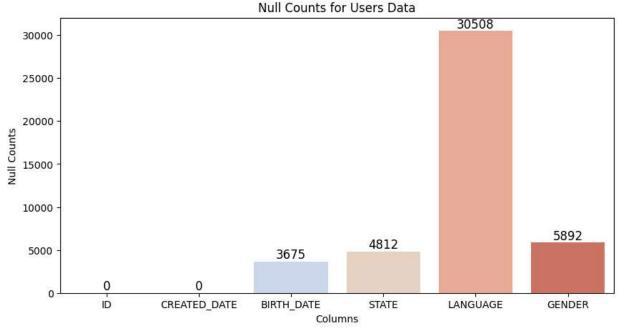
## **Take Home Data Analysis Assignment**

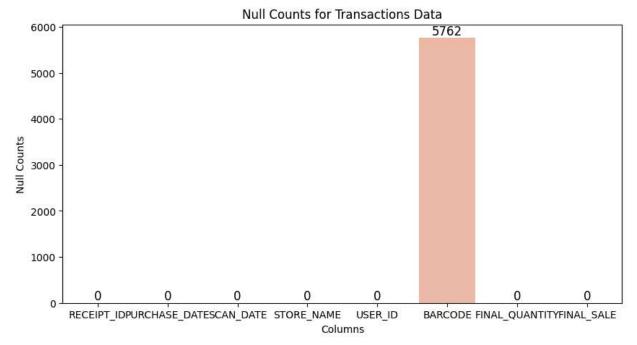
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```
In [1]: import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
        products = pd.read_csv("PRODUCTS_TAKEHOME.csv")
In [ ]:
         transactions = pd.read_csv("TRANSACTION_TAKEHOME.csv")
        users = pd.read csv("USER TAKEHOME.csv")
In [4]: # Identify Nulls in each data set
         p null counts = products.isnull().sum()
         u null counts = users.isnull().sum()
         t null counts = transactions.isnull().sum()
        # print(p_null_counts, u_null_counts, t_null_counts, sep='\n')
In [5]: plt.figure(figsize=(10, 5))
         ax p = sns.barplot(
                 x=p_null_counts.index,
                y=p_null_counts.values,
                 palette="coolwarm"
             )
         ax_p.bar_label(ax_p.containers[0], fmt="%.0f", fontsize=12, color="black")
         plt.xlabel("Columns")
         plt.ylabel("Null Counts")
         plt.title("Null Counts for Product Data")
         plt.show()
         plt.figure(figsize=(10, 5))
         ax u = sns.barplot(
                 x=u_null_counts.index,
                 y=u_null_counts.values,
                 palette="coolwarm"
         ax_u.bar_label(ax_u.containers[0], fmt="%.0f", fontsize=12, color="black")
         plt.xlabel("Columns")
         plt.ylabel("Null Counts")
         plt.title("Null Counts for Users Data")
         plt.show()
         plt.figure(figsize=(10, 5))
         ax_t = sns.barplot(
                 x=t_null_counts.index,
                 y=t_null_counts.values,
                 palette="coolwarm"
```

```
ax_t.bar_label(ax_t.containers[0], fmt="%.0f", fontsize=12, color="black")
plt.xlabel("Columns")
plt.ylabel("Null Counts")
plt.title("Null Counts for Transactions Data")
plt.show()
```







```
In [16]: # Mismatch of data types within columns
    condition = (transactions['FINAL_QUANTITY']=='zero')
    print("numeric column with string value 'zero' record count: ",transactions[condition]
    numeric column with string value 'zero' record count: 12500

In [17]: # Mismatch in data types within Final_sale
    empty_sale = transactions['FINAL_SALE']==' '
    print(transactions[empty_sale]['FINAL_QUANTITY'].value_counts().sum())
    12500
```

Intresting Fact to note that the number of records with incorect data in columns FINAL\_QUANTITY and FINAL\_SALE are equal in number (12500) but are not the same record, leaving us with 25000 records which have incorect data

• Assuming to drop the records with null barcodes in Transactions as we cannot identify the relation between the transaction and the product details

```
# drop records which do not have barcodes
In [18]:
          transactions = transactions.dropna(subset=['BARCODE'])
          print(transactions.isnull().sum())
         RECEIPT ID
                            0
         PURCHASE_DATE
                            0
         SCAN DATE
                            0
         STORE NAME
                            0
         USER_ID
                            0
         BARCODE
                            0
         FINAL_QUANTITY
                            0
         FINAL SALE
         dtype: int64
```

 Correcting the Incorrect values to make sure all the numeric data columns have only numeric data

```
In [19]: transactions['FINAL_QUANTITY'] = transactions['FINAL_QUANTITY'].replace('zero','0.00')
    transactions['FINAL_SALE'] = transactions['FINAL_SALE'].replace(' ','0.00').astype('f]
```

#### Section 1

## Top 5 brands by receipts scanned by users 21 and over

```
with temp as
    SELECT
    brand,
    count(receipt_id) as c
    FROM (
        SELECT
            t.receipt id,
            p.brand,
        FROM TRANSACTIONS t
        JOIN PRODUCTS p ON t.barcode = p.barcode
        JOIN USERS u ON u.id = t.user id
        WHERE t.barcode IS NOT NULL
        AND u.birth date IS NOT NULL
        AND p.brand IS NOT NULL
        and datediff('year', birth_date, CURRENT_TIMESTAMP) >= 21
    ) subquery
    group by brand
select
    brand
from temp
order by c desc
limit 5;
  BRAND
  NERDS CANDY
  DOVE
  SOUR PATCH KIDS
  COCA-COLA
  TRIDENT
```

# What are the top 5 brands by sales among users that have had their account for at least six months

```
brand,
sum(final_sale::float) over(partition by brand) as sale
```

```
FROM TRANSACTIONS t
JOIN PRODUCTS p ON t.barcode = p.barcode
JOIN USERS u ON u.id = t.user_id
WHERE t.barcode IS NOT NULL
   AND p.brand IS NOT NULL
   and datediff('month', u.created_date, current_timestamp) > 6
order by 2 desc
limit 5
;

BRAND
  CVS
  TRIDENT
  DOVE
  COORS LIGHT
  QUAKER
```

#### Section 2

### At what percent has Fetch grown year over year?

- considering only customer growth as the growth factor because we have sales only for one year(2024)
- assumptions
  - 1. Only user table is considered
  - 2. assume that all the signed up users do not drop over time

```
with base as (
    select
        year(created date) as year
        ,count(id) as total_customers
    from users
    group by 1
    order by 1
select
    round(
    coalesce(((total_customers - lag(total_customers) over(order by year
))/ lag(total_customers) over(order by year ))*100,0) , 1) as "Y-O-Y growth
%"
from base
  YEAR
           TOTAL_CUSTOMERS Y-O-Y growth %
  2014
           30
                       0.0
   2015
           51
                       70.0
   2016
           70
                       37.3
```

644

820.0

2017

2018	2168	236.6
2019	7093	227.2
2020	16883	138.0
2021	19159	13.5
2022	26807	39.9
2023	15464	-42.3
2024	11631	-24.8

#### Section 3

## **Email message for business**

Hi Team,

Analysing the dataset provided. I was able to come up with the following Observations.

- 1. The dataset by itself has a lot of Null values in the columns that are used to identify products and transactions forcing me to exclude all the records reducing the sample size by 10%.
- 2. The numeric columns used for calculating metrics have mix of data types leading us to lot of ambiguity while using data.
- 3. One Interesting finding was that the top 5 brands for our users were all related to daily consumer goods and medicine
- 4. Also another point to note is that the overall new customer enrollment has decreased over the past 2 years

I did have few requests to be made inorder to pull better insights

- 1. If we can have the all the transactions over the last five years would be great to analyse the revenue growth
- 2. Emphasizing on validating the data type for each column must be put in place before inserting records into tables to avaid un-necessary data type conversions and incorrect data in columns
- 3. Adding column descriptions would really help in identifying specific columns based on business requirements .

Thanks and Regards