### Take home assessment procedure followed -

### Step 1 - JSON data structuring process

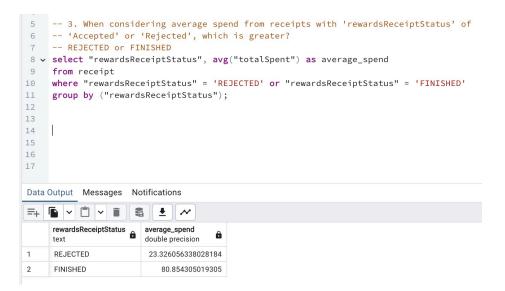
I followed the below process to structure the given json data (Tools & Technologies used - Jupyter Notebook, Python, Stack overflow, Lucid Chart)

- Converted each json file (user, brand and receipt) to pandas dataframe by normalizing the columns. Normalization was required since the json data was nested.
- The receipt data contained a column called rewardsReceiptItemList which was a list of json data and normalizing it within the receipt dataset would not result in a good structure.
- Created another table for rewardsReceiptItemList by normalizing each json within the rewardsReceiptItemList where each row in the table would have details for individual items scanned and the receiptId associated with it.
- Total four tables user, brand, receipt and rewardsitemlist. Created a relational data model for these four tables on Lucid chart
- The four tables were then loaded to the postgresql database by creating a sql engine and establishing connection with postgresql server.

### Step 2 - Queries to answer questions from business stakeholders

Accessed the four tables and ran queries (Tools & Technologies used - Postgresql Admin, PostgreSQL)

 When considering average spend from receipts with 'rewardsReceiptStatus' of 'Accepted' or 'Rejected', which is greater? I did not see any rewardsReceiptStatus with a value of 'Accepted'. Hence, I used 'FINISHED' in place of 'Accepted'. FINISHED rewardsReceiptStatus has a higher average totalSpent from the receipts



 When considering total number of items purchased from receipts with 'rewardsReceiptStatus' of 'Accepted' or 'Rejected', which is greater? I did not see any rewardsReceiptStatus with a value of 'Accepted'. Hence, I used 'FINISHED' in place of 'Accepted'. FINISHED rewardsReceiptStatus has a higher number of items purchased from the receipts



## Step 3 - Evaluate Data Quality issues

Below are some of the data quality issues I identified (Tools & Technologies used - Jupyter Notebook, Python) -

• **Duplicate records** - The users data has 283 duplicate records.

```
#duplicate records
print("-----Users-
print(user_df.shape)
test_user_df = user_df.drop_duplicates()
print(test_user_df.shape)
                      --Brand--
print("------)
print(brand_df.shape)

test_brand_df = brand_df.drop_duplicates()
print(test_brand_df.shape)
print("----Receipt-
print(receipt_df.shape)
test_receipt_df = receipt_df.drop_duplicates()
print(test_receipt_df.shape)
print("------Rewards List-----")
print(rewards_item_list_df.shape)
test_rewards_df = rewards_item_list_df.drop_duplicates()
print(test_rewards_df.shape)
          ---Users----
(495, 7)
(212, 7)
            -Brand----
(1167, 9)
(1167, 9)
          ---Receipt----
(1119, 14)
(1119, 14)
            -Rewards List----
(6941, 35)
(6941, 35)
```

• Null values - What is the nature of these null values? How do we treat them?

```
#null values
print("------Users------")
print(test_user_df.isnull().sum()) #using the user data without duplicates

print("------Brand-------")
print(brand_df.isnull().sum())

print("-------Receipt-------")
print(receipt_df.isnull().sum())

print("-------Rewards List-------")
print(rewards_item_list_df.isnull().sum())
```

Users	
userId	0
active	0
role	0
signUpSource	5
state	6
createdDate	0
lastLogin	40
dtype: int64	
Brand	
brandId	0
barcode	0
category	155
categoryCode	650
name	0
topBrand	612
cpgId	0
cpg_ref	0
brandCode	234
dtype: int64	

Receipt	
receiptId	0
bonusPointsEarned	575
bonusPointsEarnedReason	575
pointsEarned	510
purchasedItemCount	484
rewardsReceiptStatus	0
totalSpent	435
userId	0
createDate	0
dateScanned	0
finishedDate	551
modifyDate	0
pointsAwardedDate	582
purchaseDate	448
dtype: int64	

Rewards List	
receiptId	0
barcode	3851
description	381
finalPrice	174
itemPrice	174
needsFetchReview	6128
partnerItemId	0
preventTargetGapPoints	6583
quantityPurchased	174
userFlaggedBarcode	6604
userFlaggedNewItem	6618
userFlaggedPrice	6642
userFlaggedQuantity	6642
needsFetchReviewReason	6722
pointsNotAwardedReason	6601
pointsPayerId	5674
rewardsGroup	5210
rewardsProductPartnerId	4672
userFlaggedDescription	6736
originalMetaBriteBarcode	6870
originalMetaBriteDescription	6931
brandCode	4341
competitorRewardsGroup	6666
discountedItemPrice	1172
originalReceiptItemText	1181
itemNumber	6788
originalMetaBriteQuantityPurchased	6926
pointsEarned	6014
targetPrice	6563
competitiveProduct	6296
originalFinalPrice	6932
originalMetaBriteItemPrice	6932
deleted	6932
priceAfterCoupon	5985
metabriteCampaignId	6078
dtype: int64	

• **Formatting** - The date format by default were in the below format. I converted them to datetime format for better readability and analysis.

### Before -

createdDate	lastLogin
1609687444800	1.609688e+12
1609687444800	1.609688e+12

# After -

createdDate	lastLogin
2021-01-03 15:24:04.800	2021-01-03 15:25:37.857999872
2021-01-03 15:24:04.800	2021-01-03 15:25:37.857999872

• **Multiple rows** - brandCode has multiple rows in the brand dataset. There are also null brandCodes.

```
select * from brand
    where "brandCode" = 'HUGGIES'
    ;
ata Output Messages Notifications
   category
                                                    categoryCode
text
                                                                            topBrand
boolean
                                                                                                                         brandCode
text
    brandId
                           barcode
                                                                  name
text
                                                                                       cpgld
text
                                                                                                           cpg_ref text
    [PK] text
                           text
                                         text
    5bd2011f90fa074576779a17
                          511111704652
                                         Baby
                                                                  Huggies
                                                                                       550b2565e4b001d5e9e4146f
                                                                                                              Cogs
                                                                             false
                                                                                                                         HUGGIES
    5c7d9cb395144c337a3cbfbb 511111707202
                                                                                       5459429be4b0bfcb1e864082 Cogs
                                         Baby
                                                    BABY
                                                                  Huggies
                                                                            true
                                                                                                                         HUGGIES
```