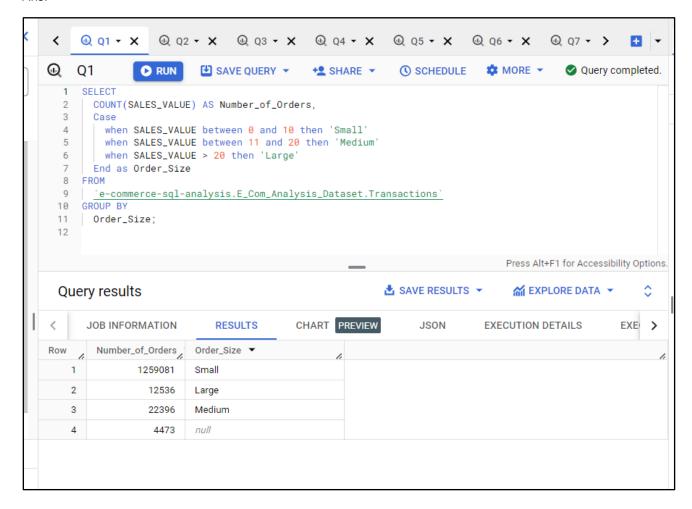
E-commerce SQL Analysis

Problem Statement:

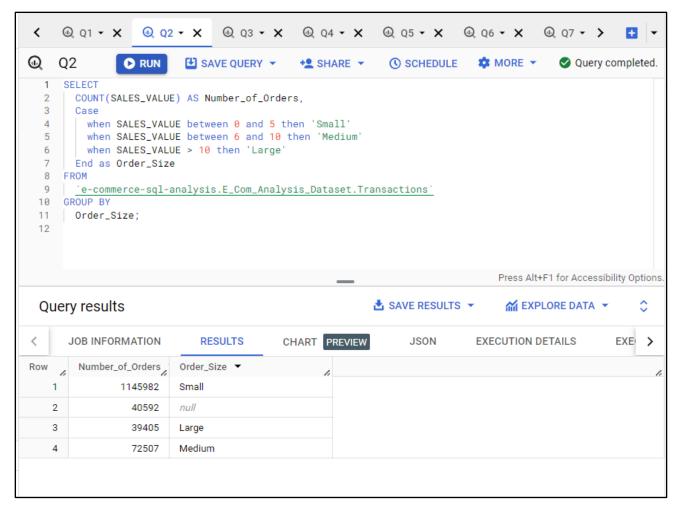
Analysing the sales, product, and customer data for an e-commerce company. Getting various insights and calculating various KPI and data with SQL in Big Query.

1. Find the number of orders that have small, medium, or large order value (small:0-10 dollars, medium:10-20 dollars, large:20+)

Ans:



2. Find the number of orders that are small, medium, or large order value (small:0-5 dollars, medium:5-10 dollars, large:10+)



3. Find top 3 stores with highest foot traffic for each week (Foot traffic: number of customers transacting)

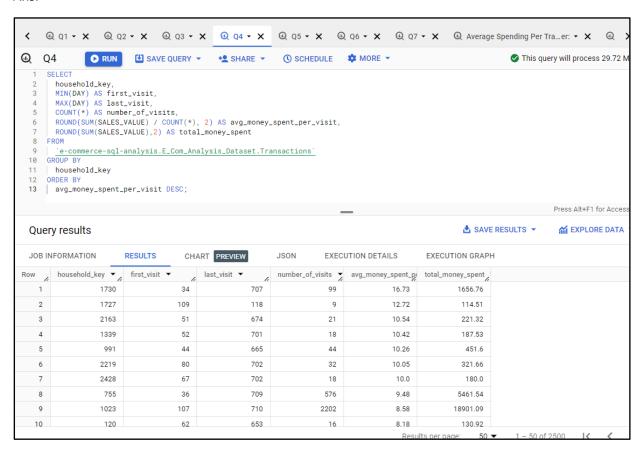
```
① Q4 - X ① Q5 - X ① Q6 - X
<
     @ Q1 ~ X
                  ⊕ Q2 - X
                               Q Q3 → X
                                                                                    ⊕ Q7 - X
Œ.
    Q3
                         MORE 🕶
               RUN

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                                                          +⊈ SHARE ▼
                                                                        ( SCHEDULE
     WITH
  1
       RankedStores AS (
  2
  3
       SELECT
  4
         week_no,
  5
         store_id,
  6
         transaction_count,
         ROW_NUMBER() OVER (PARTITION BY week_no ORDER BY transaction_count DESC) AS row_num
  8
       FROM (
  9
         SELECT
 10
           week_no,
 11
          store_id.
 12
          COUNT(DISTINCT Basket_id) AS transaction_count
 13
 14
            `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Transactions`
         GROUP BY
 15
 16
           week_no,
 17
           store_id ) AS StoreCounts )
 18
     SELECT
 19
       week_no,
 20
       store_id,
 21
       transaction_count
 22
     FROM
 23
     RankedStores
 24
     WHERE
 25
     row_num <= 3;
```

Query results								
<	JOB INFORMATION	RESULTS	CHART PREVIEW	JSON				
Row	week_no ▼	store_id ▼	transaction_count					
1	28	367	65					
2	28	361	60					
3	28	406	53					
4	62	367	68					
5	62	321	52					
6	62	406	52					
7	15	367	63					
8	15	361	59					
9	15	381	51					
10	42	367	66					
11	42	361	58					
12	42	343	53					
13	11	367	59					
1/	11	381	41					

4. Create a basic customer profiling with first, last visit, number of visits, average money spent per visit and total money spent order by highest average money.



5. Do a single customer analysis selecting most spending customer for whom we have demographic information (because not all customers in transaction data are present in demographic table) (show the demographic as well as total spent)

Ans:





6. Find products (product table: SUB_COMMODITY_DESC) which are most frequently bought together and the count of each combination bought together. do not print a combination twice (A-B / B-A)

```
<
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                                                                MORE 🕶
  1 SELECT
     p1.SUB_COMMODITY_DESC AS product_1,
  2
     p2.SUB_COMMODITY_DESC AS product_2,
  3
     COUNT(*) AS combination_count
  4
  5 FROM
  6 `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Transactions` AS t1
  8 `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Transactions` AS t2
 9 ON
     t1.BASKET_ID = t2.BASKET_ID
 11
     AND t1.PRODUCT_ID < t2.PRODUCT_ID
    JOIN
 12
 13
    `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Product` AS p1
 14 ON
 15 t1.PRODUCT_ID = p1.PRODUCT_ID
    JOIN
 16
 17 `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Product` AS p2
 18 ON
 19
    t2.PRODUCT_ID = p2.PRODUCT_ID
 20 GROUP BY
 21
     p1.SUB_COMMODITY_DESC,
     p2.SUB_COMMODITY_DESC
 22
 23 ORDER BY
 24 combination_count DESC;
```

Quei	y results						₫ SAVE RE
JOB IN	NFORMATION	RESULTS	CHART	PREVIEW	JSON	EXECUTION DETAILS	EXECUTION GRAPH
Row	product_1 ▼			product_2 •	,		combination_count
1	YOGURT NOT MU	JLTI-PACKS		YOGURT NOT	MULTI-PACKS		15947
2	BABY FOOD - BE	GINNER		BABY FOOD -	BEGINNER		10080
3	SS ECONOMY EN	NTREES/DINNERS ALL		SS ECONOM	/ ENTREES/DI	NNERS ALL	6633
4	SOFT DRINK PO	WDER POUCHES		SOFT DRINK	POWDER POU	CHES	6375
5	FRZN SS PREMI	UM ENTREES/DNRS/N		FRZN SS PRE	MIUM ENTRE	ES/DNRS/N	6340
6	SFT DRNK 2 LITE	ER BTL CARB INCL		SFT DRNK 2	LITER BTL CAR	B INCL	5459
7	SOFT DRINKS 12	2/18&15PK CAN CAR		SOFT DRINKS	3 12/18&15PK	CAN CAR	5173
8	CANDY BARS (SI	INGLES)(INCLUDING		CANDY BARS	(SINGLES)(IN	CLUDING	4194
9	BABY FOOD JUN	IIOR ALL BRANDS		BABY FOOD	JUNIOR ALL BF	RANDS	3751
10	FLUID MILK WHI	TE ONLY		SOFT DRINKS	3 12/18&15PK	CAN CAR	3580
11	FLUID MILK WHI	TE ONLY		YOGURT NOT	MULTI-PACKS		3520
12	CANNED CAT FO	OD (9 LIVES/FRISK		CANNED CAT	FOOD (9 LIVE	S/FRISK	3477
13	MAINSTREAM W	/HITE BREAD		FLUID MILK \	VHITE ONLY		3158
14	FLUID MILK WHI	TE ONLY		BANANAS			2971

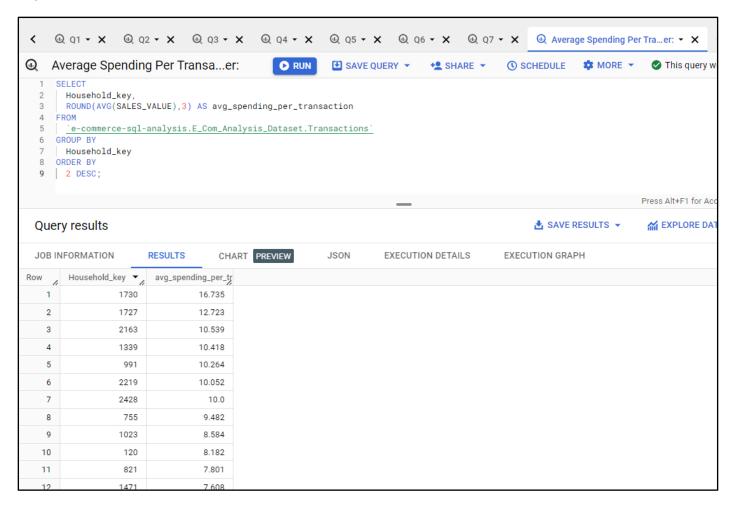
7. Find the weekly change in Revenue Per Account (RPA) (difference in spending by each customer compared to last week) (use lag function).

```
<
                                   ⊕ Q3 - X
      ⊕ Q1 - X
                    ⊕ Q2 - X
                                                 ⊕ Q4 - X
                                                                ⊕ Q5 - X
                                                                               ⊕ Q6 - ×
                                                                                              ⊕ Q7 -
                                                                                                      ×
                                                                                                            Q
     Q7
                            SAVE QUERY ▼
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                                                 +⊈ SHARE ▼
                                                                 () SCHEDULE
                                                                                  MORE -
      WITH
  2
        WeeklySpending AS (
  3
        SELECT
          Household_key,
  5
          WEEK_NO,
          ROUND(SUM(SALES_VALUE), 2) AS weekly_spending
  6
          `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Transactions`
  8
  9
        GROUP BY
  10
          Household_key,
  11
          WEEK_NO
  12
        ORDER BY
  13
          Household_key,
  14
          WEEK_NO )
  15
      SELECT
  16
        Household_key,
        WEEK_NO,
  17
  18
        weekly_spending,
        LAG(weekly_spending) OVER (PARTITION BY Household_key ORDER BY WEEK_NO) AS previous_week_spending,
ROUND(weekly_spending - LAG(weekly_spending) OVER (PARTITION BY Household_key ORDER BY WEEK_NO),2) AS weekly_change_in_RPA
  19
 20
 21
      FROM
 22
        WeeklySpending
 23
      ORDER BY
 24
        Household_key,
 25
        WEEK_NO;
```

JOB INFORMATION			RESULTS CHAR		RT PREVIEW JSON EXEC		CUTION DETAILS EXE	
Row	Household_key	٠,	WEEK_NO ▼	1.	weekly_spending	previous_week_spen	weekly_change_in_R	
1		1	8	3	42.58	nuli	nuli	
2		1	10)	14.01	42.58	-28.57	
3		1	13	3	14.03	14.01	0.02	
4		1	14	1	25.71	14.03	11.68	
5		1	15	5	10.98	25.71	-14.73	
6		1	16	5	9.09	10.98	-1.89	
7		1	17	7	13.98	9.09	4.89	
8		1	19)	47.35	13.98	33.37	
9		1	20)	31.77	47.35	-15.58	
10		1	22	2	38.98	31.77	7.21	
11		1	23	3	26.36	38.98	-12.62	

8. Average Spending Per Transaction for Each Customer.

Ans:



9. Weekly Change in the Number of Transactions for Each Customer.

```
🔇 🕆 🗶 Q Average Spending Per Tra...er: 🕶 🗶 Weekly Change in the Nu... er: 🕶 🗶 Customer Retention Anal... ek: 🕶 🗶 Customer Lifetim
Weekly Change in the Number...er:
                                                 RUN
                                                             SAVE QUERY ▼
                                                                                +⊈ SHARE ▼
                                                                                               () SCHEDULE
                                                                                                              MORE 🕶
                                                                                                                            This query will
       WeeklyTransactions AS (
       SELECT
         Household_key,
         WEEK_NO,
  6
         COUNT(DISTINCT BASKET_ID) AS weekly_transactions
  8
         \verb|`e-commerce-sql-analysis.E_Com_Analysis_Dataset.Transactions||
  9
       GROUP BY
 10
         Household_kev.
 11
         WEEK NO
 12
       ORDER BY
 13
         Household_key,
 14
         WEEK_NO )
 15
     SELECT
       Household_key,
 16
 17
       WEEK NO.
 18
       weeklv_transactions.
       LAG(weekly_transactions) OVER (PARTITION BY Household_key ORDER BY WEEK_NO) AS previous_week_transactions,
 19
       weekly_transactions - LAG(weekly_transactions) OVER (PARTITION BY Household_key ORDER BY WEEK_NO) AS weekly_change_in_transactions
 20
 21
     FROM
 22
       WeeklyTransactions
 23
     ORDER BY
 24
       Household_key,
       WEEK_NO;
```

JOB INFORMATION		RESULTS CHAR		ART PREVIEW JS0	N EXECUTION DETAILS	EXECUTION GRAPH	
Row /	Household_key	٠,	WEEK_NO ▼	weekly_transactions ▼ //	previous_week_transactions	weekly_change_in_transactions	
1		1	8	1	nuli	null	
2		1	10	1	1	0	
3		1	13	1	1	0	
4		1	14	1	1	0	
5		1	15	1	1	0	
6		1	16	1	1	0	
7		1	17	1	1	0	
8		1	19	1	1	0	
9		1	20	1	1	0	
10		1	22	1	1	0	
11		1	23	1	1	0	
12		1	24	1	1	0	
13		1	25	1	1	0	
14		1	26	1	1	0	
15		1	28	1	1	0	
16		1	30	1	1	0	
17		1	31	1	1	0	

10. Customer Retention Analysis Over Week on Week:

```
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    ⊕ Average Spending Per Tra…er: ▼ X

                                            Q Customer Retention Analysis...ek:
                                              RUN

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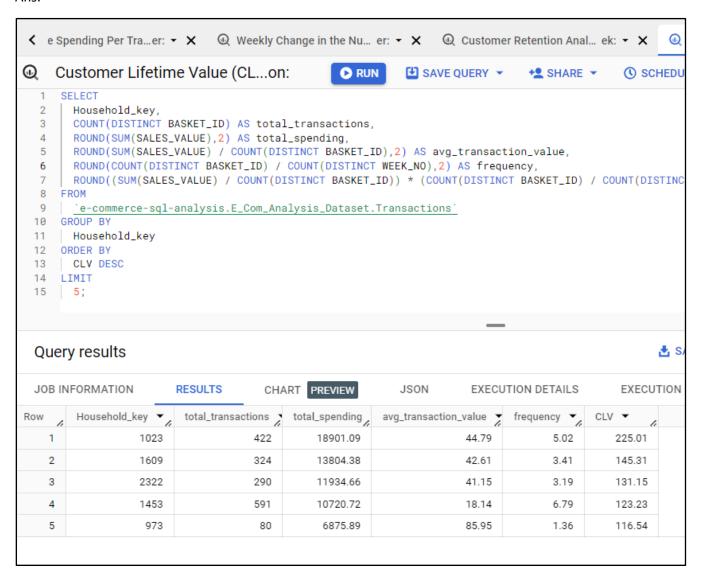
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                                                                                        () SCHEDULE
     WITH
  2
       Retention AS (
       SELECT
        Household_key,
  4
  5
         WEEK_NO,
         COUNT(DISTINCT BASKET_ID) AS weekly_transactions
  6
  8
         `e-commerce-sql-analysis.E_Com_Analysis_Dataset.Transactions`
  9
       GROUP BY
        Household_key,
 10
 11
         WEEK_NO
       ORDER BY
 12
 13
         Household_key,
 14
        WEEK_NO )
    SELECT
 15
 16
       Household_key,
 17
       WEEK_NO,
 18
       weekly_transactions,
 19
       LAG(weekly_transactions, 1, 0) OVER (PARTITION BY Household_key ORDER BY WEEK_NO) AS previous_week_transactions,
 20
       CASE
       | WHEN LAG(weekly_transactions, 1, 0) OVER (PARTITION BY Household_key ORDER BY WEEK_NO) = 0 THEN 0
 21
 22
       ELSE
       (weekly_transactions / LAG(weekly_transactions, 1, 0) OVER (PARTITION BY Household_key ORDER BY WEEK_NO)) * 100
 23
     END
 25
     AS retention_percentage
 26
     FROM
 27
      Retention
    ORDER BY
 28
 29
       Household_key,
 30
       WEEK_NO;
```

JOB INFORMATION		RESULTS	CHART PREVIEW	JSON EXECUTION	ON DETAILS EXI	
Row	Household_key/	WEEK_NO ▼/	weekly_transactions	previous_week_transactions,	retention_percentage	
1	1	8	1	0	0.0	
2	1	10	1	1	100.0	
3	1	13	1	1	100.0	
4	1	14	1	1	100.0	
5	1	15	1	1	100.0	
6	1	16	1	1	100.0	
7	1	17	1	1	100.0	
8	1	19	1	1	100.0	
9	1	20	1	1	100.0	
10	1	22	1	1	100.0	
11	1	23	1	1	100.0	
12	1	24	1	1	100.0	
13	1	25	1	1	100.0	

11. Customer Lifetime Value (CLV) Calculation.



Insights and Recommendations:

1. Leverage Complementary Product Insights:

Identify frequently paired products like yogurt variants or baby foods. Use this knowledge to craft bundled promotions, strategically place related items together, and optimize marketing strategies to encourage cross-selling and enhance customer shopping experiences.

2. Optimize Inventory and Marketing Strategies:

Utilize observed frequent product combinations to refine inventory management, ensuring adequate stock levels for frequently paired items. Tailor marketing campaigns to promote cross-category sales, exploring packaging options and pricing strategies to capitalize on these commonly paired products.

3. Monitor Weekly Spending Trends:

Track household spending week-over-week to identify fluctuations in spending patterns. Analyze changes, especially drastic shifts, to understand potential factors driving increased or decreased spending per household.

4. Customer Spending Stability Assessment:

Assess household spending stability using the calculated weekly change in Revenue Per Account (RPA). Identify customers with consistent spending versus those displaying significant spending variations, aiding in targeted retention or engagement strategies.

5. Identify High-Value Customers:

Highlight customers with significantly higher average spending per transaction. Target these high-value customers for personalized offers, loyalty programs, or exclusive perks to foster continued engagement and increase their lifetime value.

6. Segment Customers by Spending Behaviour:

Segment customers based on their average spending per transaction to tailor marketing strategies. For instance, focus on different messaging or incentives for segments with varying spending levels to optimize conversions and enhance customer satisfaction.

7. Retaining Consistent Transaction Frequency:

Monitor households displaying consistent or increasing transaction frequency week-over-week. Focus on retaining this trend to ensure sustained customer engagement and potentially higher retention rates.

8. Addressing Fluctuating Transaction Patterns:

Identify households with fluctuating transaction frequencies. Investigate and implement targeted strategies to stabilize or improve their engagement, aiming for consistent transaction behaviour to enhance customer loyalty.

9. Leveraging High CLV Customers:

Identify and focus on customers with higher Customer Lifetime Value (CLV). Allocate resources for personalized experiences, loyalty programs, or targeted campaigns to nurture and retain these valuable customers.

10. Improving Engagement for Low-Frequency Customers:

Engage with customers having lower transaction frequency but potential for increased CLV. Implement strategies to increase their transaction frequency, such as personalized offers or improved customer experiences, aiming to enhance their long-term value to the business.