

# Customer Segmentation (BigQuery)

## Step

1. Query data from supermarket data
2. Create customer single view for those with CUST\_CODE
  - 5 features
  - Average day between purchase, Day since last purchase, Number of product per visit, Number of visit per week, Spend per visit
3. Create K-Mean using BigQuery ML
  - Try different value of K (Number of clusters)
  - Choose K that minimize Davies–Bouldin index (WSS/BSS) in order to maximize between sum of square and minimize within sum of square distance
4. See result of centroid value for each features
5. Interpretation and possible action for each cluster

# Query

```
CREATE OR REPLACE MODEL
`elemental-alloy-308203.supermarket.kaj_clusters_7groups`
OPTIONS(model_type='kmeans', num_clusters =7)
AS (
SELECT TOTAL_SPEND/NUMBER_OF_VISIT as SPEND_PER_VISIT, NUMBER_OF_PRODUCT/NUMBER_OF_VISIT as NO_PRODUCT_PER_VISIT,AVG_DAYBTW_PURCHASE ,DAYS_SINCE_LAST_PURCHASE
FROM(
    SELECT
        CUST_CODE,
        COUNT(DISTINCT BASKET_ID) AS NUMBER_OF_VISIT,
        SUM(SPEND) AS TOTAL_SPEND,
        COUNT(DISTINCT PROD_CODE) AS NUMBER_OF_PRODUCT,
        SUM(QUANTITY) as NUMBER_OF_UNIT,
        COUNT(DISTINCT SHOP_WEEK) as NUMBER_OF_WEEK
        FROM `elemental-alloy-308203.supermarket.supermarket`
        WHERE CUST_CODE IS NOT NULL
        GROUP BY CUST_CODE) t1
left join(
    select CUST_CODE,date_diff(PARSE_DATE('%Y%m%d', CAST('20080706' AS STRING)),max(PARSE_DATE('%Y%m%d', CAST(SHOP_DATE AS STRING))),day) as DAYS_SINCE_LAST_PURCHASE
    from `elemental-alloy-308203.supermarket.supermarket`
    where CUST_CODE is not null
    group by CUST_CODE) t2
on t1.CUST_CODE = t2.CUST_CODE
left join(
    select CUST_CODE, ROUND(avg(DAY_BTW_PURCHASE)) AVG_DAYBTW_PURCHASE
    from(
        select CUST_CODE,SHOPDATE,lag(SHOPDATE) over (partition by CUST_CODE order by SHOPDATE asc ),date_diff(SHOPDATE,lag(SHOPDATE) over (partition by CUST_CODE order by SHOPDATE asc ),day) DAY_BTW_PURCHASE
        from(
            select distinct CUST_CODE,PARSE_DATE('%Y%m%d', CAST((SHOP_DATE) AS STRING)) as SHOPDATE
            from `elemental-alloy-308203.supermarket.supermarket`
            where CUST_CODE is not null)
        )
    where DAY_BTW_PURCHASE is not null
    group by CUST_CODE) t3
on t1.CUST_CODE=t3.CUST_CODE)
```

# Try different value of K

K=3

Metrics	
Davies–Bouldin index	1.1668
Mean squared distance	1.8857

K=6

Metrics	
Davies–Bouldin index	1.1227
Mean squared distance	1.093

K=4

Metrics	
Davies–Bouldin index	1.3188
Mean squared distance	1.5714

K=7

Metrics	
Davies–Bouldin index	1.0992
Mean squared distance	0.9745

K=5

Metrics	
Davies–Bouldin index	1.1472
Mean squared distance	1.316

Choose smallest Davies–Bouldin index  
(Minimize WSS and Maximize BSS) but since  
7 might be hard to interpret I will choose 6

# Interpretation of each cluster

Metrics	
Davies–Bouldin index	1.1227
Mean squared distance	1.093

6100 customers



























Centroid ID	Count	AVG_DAYBTW_PURCHASE	DAYS_SINCE_LAST_PURCHASE	NO_PRODUCT_PER_VISIT	SPEND_PER_VISIT
1	431	<div><div></div></div> 434.8144	<div><div></div></div> 93.7633	<div><div></div></div> 2.0240	<div><div></div></div> 5.6256
2	847	<div><div></div></div> 87.3539	<div><div></div></div> 185.1381	<div><div></div></div> 8.8586	<div><div></div></div> 25.7332
3	1,236	<div><div></div></div> 107.4923	<div><div></div></div> 324.1327	<div><div></div></div> 2.1375	<div><div></div></div> 3.6390
4	2,339	<div><div></div></div> 60.1880	<div><div></div></div> 46.4519	<div><div></div></div> 2.3922	<div><div></div></div> 7.7910
5	1,071	<div><div></div></div> 93.6513	<div><div></div></div> 657.0999	<div><div></div></div> 2.6952	<div><div></div></div> 4.7514
6	176	<div><div></div></div> 134.0431	<div><div></div></div> 411.7614	<div><div></div></div> 20.7967	<div><div></div></div> 54.6719

- Group1 – Purchase once a year, low spending
- Group2 – High ticket size with several product per visit
- Group3 – Low ticket size, purchase 3 times a year and have not purchased for almost a year
- Group4 – Frequent purchaser with low ticket size and just purchased no more than 2 months
- Group5 – Idle for almost 2 year but also have very low ticket size
- Group6 – High ticket size, purchase many product but haven't purchased for a year

# Action: Focus on Group2&6

 good

 bad

	Ticket size	Recency	Time to purchase	Variety	ACTION
Group1					Get them to purchase more frequent Since time to purchase is almost a year
Group2					Get them to purchase again and try To upsell to increase ticket size since This group has 2 <sup>nd</sup> highest ticket size
Group3					Try to get them to purchase again since Their last purchase has been a year
Group4					Ignore since very low ticket size and they Purchase regularly
Group5					Try to get them to purchase again since Their last purchase has been 2 year
Group6					Give very special promotion especially Get them to purchase again since this group has highest ticket size but Idle for year