## Objective

The purpose of the assignment is to learn about concepts related to unsupervised learning techniques, specifically kMeans Clustering. The data to apply clustering involves the historical transactions of customers for a shopping mart. The assignment also provides valuable hands-on experience with the customer segmentation.

### Description of Raw Data

For this assignment 2 separate datasets made available which represents 2 different branches of the shopping mart. Dataset named "sales211" representing the branch with code 211 is used for further analysis. Above mentioned dataset consisted of 1,726,083 rows and 17 columns. The meta data for the features are as follows:

**TRANSACTION\_RK:** Unique transaction id for each item purchased

**CALENDAR\_DT:** Date of transaction formatted as "dMthyyyy"

**DATE:** Date of transaction formatted as "yyyy-mm-dd" **TIME:** Time of transaction formatted as "hh:mm:ss"

TRANSACTION TM: Time of transaction formatted as "hh:mm:ss"

ITEM\_SK: Unique id for a given product

**RETAIL\_OUTLET\_LOCATION\_SK:** Branch code of given location

POS\_TERMINAL\_NO: Pos Payment Related Information

**CASHIER NO:** Unique code of the cashier handled the transaction

**ITEM\_QTY:** Quantity of items purchased **ITEM\_WEIGHT:** Weight of items purchased

**SALES UOM CD:** Unit of measure for the product

**SELLING\_RETAIL\_AMT:** Cost of given product/s purchased **PROMO\_SALES\_IND\_CD:** Promotion code for the product **STAPLE\_ITEM\_FLG:** Whether the product staple flagged

**REGION\_CD:** Region code of the store **CUSTOMER\_SK:** Unique customer id

# **Exporting Data**

The raw transaction data living on the SMU Server is queried for unique customers and unique products. By doing so, 2 separate tables are created for top 2000 customers and products generating the maximum revenue. The queries for product and customer tables for clustering analysis can be found in the appendix section of the report

### **Appendix**

#### **PRODUCT QUERY**

USE ca\_irfanoglu;
CREATE TABLE productCluster AS
SELECT ITEM\_SK AS ITEM\_ID,
SUM(SELLING\_RETAIL\_AMT) AS TOTAL\_REVENUE,
COUNT(ITEM\_SK) AS VISITS\_PURCHASED,
COUNT(DISTINCT( CUSTOMER\_SK)) AS DISTINCT\_CUSTOMERS,
AVG( SELLING\_RETAIL\_AMT / ITEM\_QTY) AS AVG\_PRICE
FROM dataset01.sales211
WHERE CUSTOMER\_SK > 1 AND ITEM\_QTY > 0 AND SELLING\_RETAIL\_AMT > 0
GROUP BY ITEM\_SK
ORDER BY TOTAL\_REVENUE DESC
LIMIT 2000;

#### **CUSTOMER QUERY**

USE ca\_irfanoglu;
CREATE TABLE customerCluster AS
SELECT CUSTOMER\_SK,
SUM(SELLING\_RETAIL\_AMT) AS TOTAL\_REVENUE,
SUM(ITEM\_QTY) AS TOTAL\_ITEMS,
COUNT(DISTINCT(ITEM\_SK)) AS DISTINCT\_ITEMS,
COUNT(DISTINCT(TRANSACTION\_RK)) AS TIMES\_VISITED,
MAX(DATE) as MOST\_RECENT\_VISIT,
SUM(SELLING\_RETAIL\_AMT) / NULLIF(COUNT(DISTINCT(TRANSACTION\_RK)),0) AS
AVG\_SPENT\_PER\_VISIT
FROM dataset01.sales211
WHERE CUSTOMER\_SK > 1 AND ITEM\_QTY > 0 AND SELLING\_RETAIL\_AMT > 0
GROUP BY CUSTOMER\_SK
ORDER BY TOTAL\_REVENUE DESC
LIMIT 2000;