#getting information of the dataset

<class 'pandas.core.frame.DataFrame'> RangeIndex: 52924 entries, 0 to 52923 Data columns (total 10 columns):

comprehensive market basket analysis.

#loading the packages

import pandas as pd import numpy as np

# Load the datasets

online sales.info()

# Column

0 1

3

5

7

In [5]: #taking subset

subset

2

52919

52920

52921

52922

52923

0

2

4

3203

3204

3205

3206

3207

In [12]:

In [13]:

Out[13]:

Out[7]:

Out[5]:

-----

CustomerID

memory usage: 4.0+ MB

In [11]:

In [2]:

In [3]:

Market Basket Analysis
In this assignment, we will leverage the Apriori algorithm, a powerful technique rooted in associative learning, to conduct a

from mlxtend.frequent patterns import apriori

CustomerID 52924 non-null int64
Transaction\_ID 52924 non-null int64

Transaction Date 52924 non-null int64 Product\_SKU 52924 non-null object
Product\_name 52924 non-null object

Product\_Category 52924 non-null object

Quantity 52924 non-null float64

Delivery\_Charges 52924 non-null float64 Coupon Status 52924 non-null object

20190101

20190101

20190101

20190101

20191231

20191231

20191231

20190915

20190324

20191102

20190622

20190405

20190620

20191023

20190729

20191010

20190915

20190324

20191101

20191102

20190622

20190405

20190620

20191023

20190729

20191010

20191231

dtypes: float64(2), int64(4), object(4)

subset=online sales[columns of interest]

**CustomerID** Transaction Date

17850

17850

17850

17850

17850

14410

14410

14410

14600

14600

CustomerID Transaction\_Date

52924 rows × 3 columns

transaction data

12346

12347

12347

12347

12348

18269

18269

18277

18283

18283

3208 rows × 3 columns

transaction data

0

1

2

3

4

3203

3204

3205

3206

3207

In [14]:

In [17]:

In [18]:

In [19]:

Out[19]:

**CustomerID** Transaction Date

12346

12347

12347

12347

12348

18269

18269

18277

18283

18283

#dropping useless columns

3208 rows × 407 columns

min\_support = 0.01

warnings.warn(

0

1

2

278421

278422

278424

278425

In [20]:

Out[20]:

# Generating association rules

antecedents

(Google 22 oz

Water Bottle)

(1 oz Hand

(1 oz Hand

(1 oz Hand

Sanitizer)

USA)

(Google Kick Ball)

(Nest Cam Indoor

Security Camera -

(Nest Protect

Smoke + CO

White Wired

Alarm-USA)

- USA)

(Nest Thermostat E

(Nest Learning

Thermostat 3rd

(Nest Learning

Thermostat 3rd

antecedents

(1 oz Hand

(1 oz Hand

(1 oz Hand

(1 oz Hand

(20 oz Stainless

Steel Insulated

(Nest Protect

Smoke + CO

(Nest Protect

Smoke + CO

White Battery

(Nest Secure Alarm System

Starter Pack -

(Nest Protect

Smoke + CO

White Battery

(Nest Protect

Smoke + CO

White Battery

64545 rows × 10 columns

Alarm-U...

Alarm-U...

USA, ...

Alarm-U...

Alarm-U...

White Battery Starter Pack - USA,

Sanitizer)

Tumbler)

Sanitizer)

Sanitizer)

Sanitizer)

Gen-USA - Stainl...

278426 rows × 10 columns

filtered\_rules

1

10

12

14

28

278164

278173

278187

278202

278237

Gen-USA - White)

Sanitizer)

Sanitizer)

from mlxtend.frequent patterns import association rules

Non-Null Count Dtype

columns of interest = ['CustomerID', 'Transaction Date', 'Product name']

online sales = pd.read csv('C:\\Users\\sujoydutta\\Desktop\\Data analysis\\Projects\\Marketing insights\\Online

Product\_name

Nest Learning Thermostat 3rd Gen-USA - Stainle...

Nest Learning Thermostat 3rd Gen-USA - Stainle...

20190101 Google Men's 100% Cotton Short Sleeve Hero Tee...

Google Laptop and Cell Phone Stickers

Nest Cam Indoor Security Camera - USA

Nest Learning Thermostat 3rd Gen-USA - White

Nest Protect Smoke + CO White Wired Alarm-USA

Android Men's Engineer Short Sleeve Tee Charco...

Four Color Retractable Pen, Red Spiral Google N...

Google Doodle Decal, Google Twill Cap, Windup An...

Nest Learning Thermostat 3rd Gen-USA - White, N...

Android BTTF Cosmos Graphic Tee, Android Men's ...

Nest Learning Thermostat 3rd Gen-USA - Stainle...

Google Leather Perforated Journal, Recycled Pap...

Keyboard DOT Sticker, Google Laptop and Cell Ph...

1 oz

0

0

0

0

0

0

0

0

0

0

Hand

Sanitizer

# Split the Product name column by comma and create dummy variables (one-hot encoding)

onehot data = transaction data['Product name'].str.get dummies(',')

transaction data = pd.concat([transaction data, onehot data], axis=1)

Product\_name

Android Men's **Engineer Short** 

> Sleeve Tee Charco...

Four Color Retractable

Pen,Red Spiral Google N...

Google Doodle Decal, Google

> Cap,Windup An...

**Nest Learning** Thermostat 3rd

> Gen-USA -White, N...

26 oz Double Wall Insulated

Bottle,Google

Android BTTF Cosmos Graphic

> Tee, Android Men's ...

Google Men's

Vintage Tank

Nest Learning Thermostat 3rd

Google Leather Perforated

Journal, Recycled

**Keyboard DOT** Sticker, Google

Laptop and Cell

# Drop the original Product name column if you no longer need it

transaction data.drop(columns=['Product name'], inplace=True)

# Set a minimum support threshold (e.g., 1% of transactions)

consequents

(1 oz Hand

(Google 22 oz

Water Bottle)

(1 oz Hand

Sanitizer)

Stickers)

(Google Kick Ball)

(Google Laptop

and Cell Phone

(Nest Secure

Alarm System

Starter Pack - USA,

Starter Pack - USA,

# Filtering rules based on confidence and lift thresholds

consequents

(Google 22 oz

Water Bottle)

USA)

USA)

(Nest Cam

(Nest Cam Indoor

Security Camera -

**Outdoor Security** 

Camera - USA)

(Nest Learning

Thermostat 3rd

Gen-USA - Stainl...

(Nest Cam Indoor

Security Camera -

(Nest Secure Alarm System

(Nest Cam

Nest ...

**Outdoor Security** 

Camera - USA,

(Nest Protect

Smoke + CO

White Battery

(Nest Cam IQ -

USA, Nest Cam

(Nest Secure

Alarm System

Starter Pack - USA,

**Outdoor Security** 

Alarm-U...

Starter Pack - USA,

Starter Pack - USA,

Starter Pack - USA,

Sanitizer)

# Apply Apriori algorithm to find frequent itemsets

in the future.Please use a DataFrame with bool type

Рар...

Ph...

transaction\_data.drop(columns=['CustomerID','Transaction\_Date'], inplace=True)

rules = association rules(frequent itemsets, metric='lift', min threshold=1.0)

antecedent

support

0.176746

0.036471

0.036471

0.081359

0.036471

0.471633

0.236908

0.125935

0.221945

0.475374

filtered\_rules = rules[(rules['confidence'] >= 0.5) & (rules['lift'] >= 1.0)]

antecedent

support

0.036471

0.036471

0.036471

0.036471

0.033042

0.020262

0.018703

0.019638

0.020262

0.020574

than random chance, and the conviction value of 166.07 suggests a strong belief in the association.

consequent

support

0.176746 0.018392

0.471633 0.024002

0.479115 0.023691

0.475374 0.023379

0.471633 0.022756

0.032419 0.010287

0.038653 0.010287

0.035536 0.010287

0.036471 0.010287

0.034289 0.010287

Comment: This table is self explanatory, let us take an example of the first rule: this rule suggests that there is a moderate association between purchasing (1 oz Hand Sanitizer) and (Google 22 oz Water Bottle). Customers who buy (1 oz Hand Sanitizer) are 50.43% more likely to also buy (Google 22 oz Water Bottle) than if they were purchased independently. The lift value of 2.85 indicates that this association is stronger

frequent itemsets = apriori(transaction data, min support=min support, use colnames=True)

Gen-USA -Stainle...

Styl...

Twill

# Concatenate the one-hot encoded data with the original dataset

26 oz Double Wall Insulated Bottle, Google Styl...

20191231 Nest Protect Smoke + CO White Battery Alarm-USA

# Grouping products by CustomerID and Transaction Date, join them with commas

Google Canvas Tote Natural/Navy

Google Zip Hoodie Black

transaction data = subset.groupby(['CustomerID', 'Transaction Date'])['Product name'].apply(','.join).reset indexion of the control of the co

Google Men's Vintage Tank

20 oz

Steel

0

0

0

0

0

0

0

0

0

C:\Users\sujoydutta\anaconda3\lib\site-packages\mlxtend\frequent patterns\fpcommon.py:110: DeprecationWarning: DataFrames with non-bool types result in worse computational performance and their support might be discontinued

consequent

support

0.036471 0.018392

0.176746 0.018392

0.081359 0.010599

0.036471 0.010599

0.010599 0.010287

0.014963 0.010287

0.011534 0.010287

0.011845 0.010287

0.010910 0.010287

support confidence

0.504274

0.658120

0.649573

0.641026

0.688679

22 oz

**Bottle** 

0

0

0

0

0

0

0

0

0

0

support confidence

0.104056 2.853103 0.011945

0.504274 2.853103 0.011945

0.290598 3.571798 0.007631

0.130268 3.571798 0.007631

0.418803 2.086213 0.007953

0.021811 2.057929 0.005288

0.043421 2.901974 0.006742

0.081683 7.082151 0.008834

0.046348 3.912774 0.007658

0.021639 1.983400 0.005100

2.853103 0.011945

1.395405 0.006801

1.355777 0.006217

1.348466 0.006042

1.460200 0.007172

0.507692 15.660355 0.009630

0.550000 14.229032 0.009564

0.523810 14.740184 0.009589

0.507692 13.920316 0.009548

0.500000 14.581818 0.009581

**Android** 

**Stainless** 

Insulated

**Tumbler** 

23 oz

Wide

Mouth

Sport

**Bottle** 

0

0

0

0

0

0

0

0

0

0

22 oz

**Bottle** 

Infuser

0

0

0

0

0

0

0

0

0

YouTube

24 oz

Stripe

**Bottle** 

0

1

0

0

0

0

0

0

0

25L

0

0

0

0

0

0

0

0

0

0

lift leverage conviction zhangs\_metric

1.075435

1.660703

1.294952

1.107846

1.375183

1.011462

1.029750

1.076389

1.036180

1.010966

lift leverage conviction zhangs\_metric

1.660703

1.545472

1.486429

1.461458

1.697178

1.965399

2.136326

2.025374

1.957168

1.931421

0.788947

0.674089

0.747283

0.783798

0.540371

0.972951

0.858883

0.982536

0.956779

0.945084

0.674089

0.294088

0.272348

0.268198

0.325932

0.955505

0.947441

0.950831

0.947358

0.950987

Classic

Rucksack

YouTube

Sergeant

YouTι

Wome

Favoi

Wh

YouTube

Twill

Cap

0

0

0

0

0

0

0

0

0

Product\_name

Market	Basket Analysis	