

MARKET BASKET INSIGHTS

MEMBER: J.FARHATH NASEEM (922121106014)

PHASE 3 SUBMISSION DOCUMENT: DEVELOPMENT PART 1

PROJECT: Market basket insights



Phase 3: Development Part 1

In this part we will begin building your project by loading and preprocessing the dataset. We start the market basket insights project by loading and preprocessing the transaction data. Load the transaction dataset and preprocess the data for association analysis.

Dataset Link: <https://www.kaggle.com/datasets/aslanahmedov/market-basket-analysis>

About Dataset

Market Basket Analysis

Introduction

Association Rule is most used when you are planning to build association in different objects in a set. It works when you are planning to find frequent patterns in a transaction database. It can tell you what items do customers frequently buy together and it allows retailer to identify relationships between the items.

An Example of Association Rules

Assume there are 100 customers, 10 of them bought Computer Mouth, 9 bought Mat for Mouse and 8 bought both of them.

- bought Computer Mouth => bought Mat for Mouse
- support = $P(\text{Mouth \& Mat}) = 8/100 = 0.08$
- confidence = $\text{support}/P(\text{Mat for Mouse}) = 0.08/0.09 = 0.89$
- lift = $\text{confidence}/P(\text{Computer Mouth}) = 0.89/0.10 = 8.9$

This just simple example. In practice, a rule needs the support of several hundred transactions, before it can be considered statistically significant, and datasets often contain thousands or millions of transactions.

Strategy

- Data Import
- Data Understanding and Exploration
- Transformation of the data – so that is ready to be consumed by the association rules algorithm •

Running association rules

- Exploring the rules generated
- Filtering the generated rules
- Visualization of Rule

Dataset Description

- File name: Assignment-1_Data
- List name: retaildata
- File format: .xlsx
- Number of Row: 522065
- Number of Attributes: 7
- BillNo: 6-digit number assigned to each transaction. Nominal.
- Itemname: Product name. Nominal.
- Quantity: The quantities of each product per transaction. Numeric.
- Date: The day and time when each transaction was generated. Numeric.

	A	B	C	D	E	F	G
1	BillNo	Itemname	Quantity	Date	Price	CustomerID	Country
2	536365	WHITE HANGING HEART-LIGHT HOLDER	6	01.12.2010 08:26	2.55	17850	United Kingdom
3	536365	WHITE METAL LANTERN	6	01.12.2010 08:26	3.39	17850	United Kingdom
4	536365	CREAM CUPID HEARTS COAT HANGER	8	01.12.2010 08:26	2.75	17850	United Kingdom
5	536365	KNITTED UNION FLAG HOT WATER BOTTLE	6	01.12.2010 08:26	3.39	17850	United Kingdom
6	536365	RED WOOLLY HOTTIE WHITE HEART.	6	01.12.2010 08:26	3.39	17850	United Kingdom

- Price: Product price. Numeric.
- CustomerID: 5-digit number assigned to each customer. Nominal.
- Country: Name of the country

where each customer resides. Nominal.

Libraries in R

First, we need to load required libraries. Shortly I describe all libraries.

- arules - Provides the infrastructure for representing, manipulating and analyzing transaction data and patterns (frequent itemsets and association rules).
- arulesViz - Extends package 'arules' with various visualization techniques for association rules and item-sets. The package also includes several interactive visualizations for rule exploration.
- tidyverse - The tidyverse is an opinionated collection of R packages designed for data science. •

- **plyr** - Tools for Splitting, Applying and Combining Data.
- **ggplot2** - A system for 'declaratively' creating graphics, based on "The Grammar of Graphics". You provide the data, tell 'ggplot2' how to map variables to aesthetics, what graphical primitives to use, and it takes care of the details.
- **knitr** - Dynamic Report generation in R.

- **magrittr**- Provides a mechanism for chaining commands with a new forward-pipe operator, `%>%`. This operator will forward a value, or the result of an expression, into the next function call/expression.

- dplyr - A fast, consistent tool for working with data frame like objects, both in memory and out of memory.

- tidyverse - This package is designed to make it easy to install and load multiple 'tidyverse' packages in a single step.

Data Pre-processing

Next, we need to upload

After we will clear our data

are bought together in one invoice will be in one row. Below lines of code will combine all

products from one BillNo and Date and combine all products from that BillNo and Date as one row, with each

it

```
22 transactionData$BillNo <- NULL
23 transactionData$Date <- NULL
24 #will gave the name to column "item"
25 colnames(transactionData) <- c("items")
```

Items			
WHITE HANGING HEART T-LIGHT HOLDER	WHITE METAL LANTERN	CREAM CUPID HEARTS COAT HANGER	KNITTED UNION FLAG HOT WATER BOTTLE
HAND WARMER UNION JACK	HAND WARMER RED POLKA DOT		
ASSORTED COLOUR BIRD ORNAMENT	PORPHY PLAYHOUSE DECKCHAM	PORPHY PLAYHOUSE KITCHEN	FELTCRAFT PRINCESS CHARLOTTE DOLL
JAM MAKING SET WITH JARS	RED COAT RACK PARIS FASHION	YELLOW COAT RACK PARIS FASHION	BLUE COAT RACK PARIS FASHION
BATH BUILDING BLOCK WOOD			
ALARM CLOCK BAKELIKE PINK	ALARM CLOCK BAKELIKE RED	ALARM CLOCK BAKELIKE GREEN	PINKIE AND BUNNIES STICKER SHEET
PAPER CHAIN KIT 80'S CHRISTMAS			
HAND WARMER RED POLKA DOT	HAND WARMER UNION JACK		
WHITE HANGING HEART T-LIGHT HOLDER	WHITE METAL LANTERN	CREAM CUPID HEARTS COAT HANGER	EDWARDIAN PINKSOL RED
VICTORIAN BEWING BOX LARGE			
WHITE HANGING HEART T-LIGHT HOLDER	WHITE METAL LANTERN	CREAM CUPID HEARTS COAT HANGER	EDWARDIAN PINKSOL RED
HOT WATER BOTTLE TEA AND SYMPATHY	RED HANGING HEART T-LIGHT HOLDER		
HAND WARMER RED POLKA DOT	HAND WARMER UNION JACK		
JUMBO BAG PINK POLKA DOT	JUMBO BAG BAROQUE BLACK WHITE	JUMBO BAG CHARLE AND LOULA TOPS	STRAWBERRY CHARLOTTE BAG
JAM MAKING SET PRINTED			
RETROSPOT TEA SET CERAMIC 11 PC	GREY PINK TOOL SET	JUMBO SHOPPER VINTAGE RED PAISLEY	AIRLINE LOUNGE

We don't need BillNo and Date, we will make it as Null.

```
34 transactions <- read.transactions("/Users/asik/Desktop/assignment1/itemslist.csv",
35 format = "basket", sep=',')
```

```
36 summary(transactions)
```

```
41 itemFrequencyPlot(transactions, topN=20, type="absolute",
42 col=brewer.pal(8, "Pastel2"), main="Absolute Item Frequency Plot")
43
```

```
transactions as tibble/tidy in sparse format with
18100 rows (elements/items/transactions) and
7088 columns (items) and a density of 0.00259234

most frequent items:
WHITE HANGING HEART T-LIGHT HOLDER      1718      REGENCY CAKESTAND 3 TIER      1468      JUMBO BAG RED RETROSPOT      1395
PARTY BUNTING      1245      ASSORTED COLOUR BIRD ORNAMENT      1226      (Other)      213843

element (items/transaction) length distribution:
#12861
 0  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27
1346 888 744 743 743 696 642 635 632 568 584 527 691 529 533 588 468 424 468 486 385 387 386 287 232 246 226
28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54
218 215 209 164 155 135 148 131 388 109 86 386 98 86 84 84 65 56 67 59 56 57 48 60 59 59 47
55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81
41 35 27 37 29 26 27 26 24 25 26 27 24 23 13 28 29 13 16 15 11 15 12 6 7 14 13
42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 100 181 182 183 184 185 186 187 188
18 8 8 11 20 13 8 6 5 5 11 9 4 4 3 5 5 2 4 1 4 4 2 2 2 6 3
109 139 112 112 113 114 116 117 118 119 121 122 123 125 126 127 131 132 133 134 140 141 142 143 145 146 147
4 5 2 5 5 1 5 5 5 2 2 5 5 2 2 1 1 2 1 1 2 2 2 1 1 2 1
158 154 157 168 171 177 178 189 182 202 204 228 249 250 285 318 480 419
1 3 2 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

Min. 1st Qu. Median Mean 3rd Qu. Max.
1.00  5.00 15.00 17.64 23.00 419.00

includes extended item information - examples:
1 18 COLOUR SPICEDRY PEN
2 12 COLOURED PARTY BALLOONS
```

This how should look transaction data before we will go to next step.

```
36 if (!require("RColorBrewer")) {install.packages("RColorBrewer")}
37 library(RColorBrewer)
```

The summary gives us som

e useful information:

Interactive Scatter-Plot:

We can have a look for each rule (interactively) and view all quality measures (support, confidence and lift).

Graph - Based Visualization and Group Method:

Graph plots are a great way to visualize rules but tend to become congested as the number of rules increases. So, it is better to visualize a smaller number of rules with graph-based visualizations. We can see as well group method for top 10 items.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import plotly.express as px

import warnings
warnings.filterwarnings('ignore')

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

/kaggle/input/market-basket-analysis/Assignment-1_Data.xls
x
/kaggle/input/market-basket-analysis/Assignment-1_Data.csv
```

[2]:

	BillNo	Itemname	Quantity	Date	Price	CustomerID	Country
--	--------	----------	----------	------	-------	------------	---------

0	536365	WHITE HANGING HEART T LIGHT HOLDER	6	01.12.2010 08:26	2,55	17850.0	United Kingdom
1	536365	WHITE METAL LANTERN	6	01.12.2010 08:26	3,39	17850.0	United Kingdom
2	536365	CREAM CUPID HEARTS COAT HANGER	8	01.12.2010 08:26	2,75	17850.0	United Kingdom
3	536365	KNITTED UNION FLAG HOT WATER BOTTLE	6	01.12.2010 08:26	3,39	17850.0	United Kingdom
4	536365	RED WOOLLY HOTTIE WHITE HEART.	6	01.12.2010 08:26	3,39	17850.0	United Kingdom

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 522064 entries, 0 to 522063
Data columns (total 7 columns):
# Column Non-Null Count Dtype
---  -
0 BillNo 522064 non-null object
1 Itemname 520609 non-null object
2 Quantity 522064 non-null int64
3 Date 522064 non-null object
4 Price 522064 non-null object
5 CustomerID 388023 non-null float64
6 Country 522064 non-null object
dtypes: float64(1), int64(1), object(5)
memory usage: 27.9+ MB

```

Out[4]:

```

BillNo 0
Itemname 1455
Quantity 0
Date 0
Price 0
CustomerID 134041

```

Country 0
dtype: int64

1-2. | Dropping data with negative or zero quantity

In [6]:

```
df=df.loc[df['Quantity']>0]
```

1-3. | Dropping data with zero price

In [8]:

```
df=df.loc[df['Price']>'0']
```

1-4. | Dropping Non-product data.

In [10]:

```
df=df.loc[(df['Itemname']!='POSTAGE')&(df['Itemname']!='DOTCOM  
POSTAGE')&(df['Itemname']!='Adjust bad  
debt')&(df['Itemname']!='Manual')]
```

1-5. | Filling null data

In [12]:

```
df=df.fillna('-')
df.isnull().sum()
```

Out[12]:

```
BillNo 0
Itemname 0
Quantity 0
Date 0
Price 0
CustomerID 0
Country 0
dtype: int64
```

1-6. | Splitting data into year and month

In [13]:

```
df['Year']=df['Date'].apply(lambda x:x.split('.')[2])
df['Year']=df['Year'].apply(lambda x:x.split(' ')[0])
df['Month']=df['Date'].apply(lambda x:x.split('.')[1])
df.head()
```

Out[13]:

	BillNo	Itemname	Quantity	Date	Price	CustomerID	Country	Year	Month
--	--------	----------	----------	------	-------	------------	---------	------	-------

0	536365	WHITE HANGIN G HEART T-LIGHT HOLDER	6	01.12.20 10 08:26	2,55	17850.0	United Kingdom	2010	12
1	536365	WHITE METAL LANTERN	6	01.12.20 10 08:26	3,39	17850.0	United Kingdom	2010	12
2	536365	CREAM CUPID HEARTS COAT HANGER	8	01.12.20 10 08:26	2,75	17850.0	United Kingdom	2010	12
3	536365	KNITTED UNION FLAG HOT WATER BOTTLE	6	01.12.20 10 08:26	3,39	17850.0	United Kingdom	2010	12
4	536365	RED WOOLLY HOTTIE WHITE HEART.	6	01.12.20 10 08:26	3,39	17850.0	United Kingdom	2010	12

1-7. | Creating a Total price column

In [14]:

```
df['Price']=df['Price'].str.replace(',','').astype('float64')
df['Total price']=df.Quantity*df.Price
df.head()
```

Out[14]:

	BillNo	Itemname	Quantity	Date	Price	CustomerID	Country	Year	Month	Total price
0	536365	WHITE HANGIN G HEART T-LIGHT HOLDER	6	01.12.2010 08:26	2.55	17850.0	United Kingdom	2010	12	15.30
1	536365	WHITE METAL LANTERN	6	01.12.2010 08:26	3.39	17850.0	United Kingdom	2010	12	20.34

2	536365	CREAM CUPID HEARTS COAT HANGE R	8	01.12.2 010 08:26	2.75	17850.0	United Kingdo m	2010	12	22.00
3	536365	KNITTE D UNION FLAG HOT WATER BOTTLE	6	01.12.2 010 08:26	3.39	17850.0	United Kingdo m	2010	12	20.34
4	536365	RED WOOLL Y HOTTIE WHITE HEART.	6	01.12.2 010 08:26	3.39	17850.0	United Kingdo m	2010	12	20.34

1-8. | Checking the Total price in each month.

In [15]:

```
df.groupby(['Year', 'Month'])['Total price'].sum()
```

Out[15]:

```
Year Month
2010 12 778386.780
2011 01 648311.120
    02 490058.230
    03 659979.660
    04 507366.971
    05 721789.800
    06 710158.020
    07 642528.481
    08 701411.420
    09 981408.102
    10 1072317.070
    11 1421055.630
    12 606953.650
```

Name: Total price, dtype: float64

It is appropriate to look at 12-month increments to implement data analytics properly, so I'll drop the data for 2020 Dec.

In [16]:

```
df=df.loc[df['Year'] != '2010']
```

2. | Exploratory Data Analysis

2-1. | Sales amount and quantity

2468101200.2M0.4M0.6M0.8M1M1.2M1.4M

CountryAustraliaBelgiumFranceGermanyGreeceHong

KongIcelandIsraelItalyLebanonNetherlandsPolandPortugalSingaporeSpainSwedenSwitzerlandUnited KingdomAustriaJapanNorwaySaudi ArabiaUnited Arab

EmiratesBrazilUSAUnspecifiedBahrainMaltaRSAMonthly sales amount in each country in 2021MonthSales amount

Most of the sales amounts are occupied by the UK.

PortugalItalyHong

KongSingaporeAustriaIsraelPolandUnspecifiedGreeceIcelandUSA01M2M3M4M5M6M7M8

m Sales amount in each country in 2021CountrySales amount

2-2. | Category

Top 10 highest sales amount items

5	ENAMEL BREAD BIN CREAM	6585.93
6	WHITE HANGING HEART T-LIGHT HOLDER	6563.80
7	DOORMAT KEEP CALM AND COME IN	6385.09
8	SPOTTY BUNTING	6262.40
9	RED RETROSPOT CAKE STAND	6035.29

0	REGENCY CAKESTAND 3 TIER	24653.67
1	PARTY BUNTING	9416.13
2	SET OF 3 CAKE TINS PANTRY DESIGN	7621.05
3	CREAM SWEETHEART MINI CHEST	6836.38
4	SET/4 WHITE RETRO STORAGE CUBES	6714.75
5	ENAMEL BREAD BIN CREAM	6585.93
6	WHITE HANGING HEART T-LIGHT HOLDER	6563.80
7	DOORMAT KEEP CALM AND COME IN	6385.09
8	SPOTTY BUNTING	6262.40
9	RED RETROSPOT CAKE STAND	6035.29

Top 10 most purchased items

Out[21]:

	Itemname	Quantity
520583	PAPER CRAFT , LITTLE BIRDIE	80995
59999	MEDIUM CERAMIC TOP STORAGE JAR	74215
405138	WORLD WAR 2 GLIDERS ASSTD DESIGNS	4800
198929	SMALL POPCORN HOLDER	4300
94245	EMPIRE DESIGN ROSETTE	3906
260928	ESSENTIAL BALM 3.5g TIN IN ENVELOPE	3186
51228	FAIRY CAKE FLANNEL ASSORTED COLOUR	3114
154834	FAIRY CAKE FLANNEL ASSORTED COLOUR	3114
416997	SMALL CHINESE STYLE SCISSOR	3000

280572	ASSORTED COLOUR BIRD ORNAMENT	2880
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Top 10 most frequently purchased items

Out[30]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric
0	(ALARM CLOCK BAKELIKE GREEN)	(ALARM CLOCK BAKELIKE RED)	0.05	0.05	0.03	0.64	12.41	0.03	2.64	0.97
1	(ALARM CLOCK BAKELIKE RED)	(ALARM CLOCK BAKELIKE GREEN)	0.05	0.05	0.03	0.59	12.41	0.03	2.32	0.97
2	(GARDENERS KNEELING PAD KEEP CALM)	(GARDENERS KNEELING PAD CUP OF TEA)	0.05	0.05	0.03	0.60	13.23	0.03	2.40	0.98

3	(GARDE NERS KNEELIN G PAD CUP OF TEA)	(GARDE NERS KNEELI NG PAD KEEP CALM)	0.05	0.05	0.03	0.72	13.23	0.03	3.39	0.9 7
4	(PINK REGENC Y TEACUP AND SAUCER)	(GREEN REGEN CY TEAC UP AND SAUCER)	0.04	0.05	0.03	0.82	15.50	0.03	5.25	0.98

3-2. | The top 5 of the highest support value of items(antecedents)

$Support(item) = Transactions\ comprising\ the\ item / Total\ transactions$

Out[32]:

	antecedents	consequents	support
13	frozenset({'JUMBO BAG RED RETROSPOT'})	frozenset({'JUMBO BAG PINK POLKADOT'})	0.05
12	frozenset({'JUMBO BAG PINK POLKADOT'})	frozenset({'JUMBO BAG RED RETROSPOT'})	0.05
16	frozenset({'JUMBO STORAGE BAG SUKI'})	frozenset({'JUMBO BAG RED RETROSPOT'})	0.04
17	frozenset({'JUMBO BAG RED RETROSPOT'})	frozenset({'JUMBO STORAGE BAG SUKI'})	0.04
15	frozenset({'JUMBO SHOPPER VINTAGE RED PAISLEY'})	frozenset({'JUMBO BAG RED RETROSPOT'})	0.04

In the top support value of purchase, it means that "JUMBO BAG PINK RETROSPOT" is present in 5% of all purchases.

3-3. | The top 5 of the highest confidence value of items

$Confidence = Transactions\ comprising\ antecedent\ and\ consequent / Transactions\ comprising\ antecedent$

Out[33]:

	antecedents	consequents	confidence
--	-------------	-------------	------------

4	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	0.82
30	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	0.78
6	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	0.75
7	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	0.73
3	frozenset({'GARDENE RS KNEELING PAD CUP OF TEA'})	frozenset({'GARDENE RS KNEELING PAD KEEP CALM'})	0.72

In the top confidence value of the purchase, it means that 82% of the customers who bought "PINK REGENCY TEACUP AND SAUCER" also bought "GREEN REGENCY TEACUP AND SAUCER".

3-4. | The top 5 of the highest lift value of items

Lift = Confidence (antecedent -> consequent) / Support(antecedent)

In [34]:

```
rules[['antecedents', 'consequents', 'lift']].sort_values('lift', ascending=
F else)[:5].style.background_gradient(cmap=cm).set_precision(2)
```

Out[34]:

	antecedents	consequents	lift
4	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	15.50
5	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	15.50
31	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	14.36
30	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	14.36

6	frozenset({'GREEN	frozenset({'ROSES	13.86
---	-------------------	-------------------	-------

	REGENCY TEACUP AND SAUCER'})	REGENCY TEACUP AND SAUCER'})	
--	------------------------------	------------------------------	--

In the top list value of the purchase, it means that customers are 15.5 times more likely to buy "GREEN REGENCY TEACUP AND SAUCER" if you sell "PINK REGENCY TEACUP AND SAUCER".

3-5. | The best combination of the items

Out[35]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric
4	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	0.04	0.05	0.03	0.82	15.50	0.03	5.25	0.98
30	frozenset({'PINK REGENCY TEACUP AND SAUCER'})	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	0.04	0.05	0.03	0.78	14.36	0.03	4.24	0.97
6	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	0.05	0.05	0.04	0.75	13.86	0.04	3.78	0.98
7	frozenset({'ROSES REGENCY TEACUP AND SAUCER'})	frozenset({'GREEN REGENCY TEACUP AND SAUCER'})	0.05	0.05	0.04	0.73	13.86	0.04	3.55	0.98

3	frozens et({'GA RDENER S KNEELI NG PAD CUP OF TEA'})	frozens et({'GA RDENER S KNEELI NG PAD KEEP CALM'})	0.05	0.05	0.03	0.72	13.23	0.03	3.39	0.97
---	---	--	------	------	------	------	-------	------	------	------