# D212: Data Mining II – Task 3 Association Rules & Lift Analysis

## Part I: Research Question

#### A1. Research Question

What items are most frequently purchased together according to the provided telecommunications dataset?

#### A2. Defined Goal

The goal of the market basket analysis is to identify the relationships between items that are frequently bought together. The results found can help the executive team decide whether to offer discounts on the items of interest such as bundling options and improve product placement to reduce customer churn.

#### Part II: Market Basket Justification

### **B1.** Market Basket Explanation

According to Dr. Kamara's video, market basket analysis (MBA) is a data mining technique used to reveal any patterns and trends between pairs of items frequently bought together.

The expected outcome of MBA is to analyze the transaction datasets to determine which items are frequently bought together and how strong their relationships are to each other, allowing the company to utilize these findings to retain customers by offering bundling discounts and to optimize product placement on websites, in stores, or in promotions based on the item's popularity.

#### **B2.** Transaction Example

An example of a transaction from the dataset:

Apple Lightning to Digital AV Adapter, TP-Link AC1750 Smart WiFi Router, Apple Pencil

### **B3.** Market Basket Assumption

An assumption of market basket analysis is there is some type of relationship between a pair of items in the large dataset. This once again refers to the "if-then" statement as mentioned above in part B2.

#### Part III: Data Preparation & Analysis

## C1. <u>Transforming the Dataset</u>

A copy of the cleaned data set has been attached as "df cleanl.csv".

In [1]: a Japort the necessary packages & (thrortes import pands as pd from pands import DataFrame import numpy as np import smborn as ns import matplottlb.pyplot as plt imatplottlb inline from mixtend.frequent\_patterns import apriori from mixtend.frequent\_patterns import apriori from mixtend.frequent\_patterns import association\_rules from mixtend.preprocessing import TransactionIncoder

# Ignore warning code import warnings warnings.filterwarnings('ignore')

# Load the data set into the pandas data frame by using read\_csv command data = pd.read\_csv(r'C:\Users\ashle\Downloads\D212\teleco\_market\_basket.csv') data.head(5)

Out[1]:	Item01	Item02	Item03	Item04	Item05	Item06	Item07	Item08	Item09	Item10	Item11	Item12	Item 13	Item14	Item15	Item16	Item17	Item18	Item19	Item20
-	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
,	Logitech M510 Wireless mouse			nonda USB C to USB Adapter				Cleaning Gel Universal Dust Cleaner	Micro Center 32GB Memory card	YUNSONG 3pack 6ft Nylon Lightning Cable	TopMate C5 Laptop Cooler pad		HyperX Cloud Stinger Headset	TONOR USB Gaming Microphone	Dust-Off Compressed Gas 2 pack	3A USB Type C Cable 3 pack 6FT	HOVAMP iPhone charger	SanDisk Ultra 128GB card	FEEL2NICE 5 pack 10ft Lighning cable	FEIYOLD Blue light Blocking Glasses
	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
:	Apple Lightning to Digital AV Adapter	TP-Link AC1750 Smart WiFi Router	Apple Pencil	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN

In [2]: data.shape

Out[2]: (15002, 20)

In [3]: # Remove blank rows from the dataset
data = data[data['Iten01'].notna()]
data.shape

Out[3]: (7501, 20)

In [4]: # Dataframe converted to list of Lists
rows = []
for i in range (0, 7501):
 rows.append([str(data.values[i,j])
for j in range(0,20])

In [S]: # Create an array
DE = TransactionEncoder()
array = DE.fit(rows).transform(rows)

# Convert the array to dataframe transaction = pd.DataFrame(array, columns = DE.columns\_)

In  $[\,6\,]:$  # Display the transaction (True means there is a missing value) transaction

Out[6]:		10ft iPHone Charger Cable	10ft iPHone Charger Cable 2 Pack	3 pack Nylon Braided Lightning Cable		5pack Nylon Braided USB C cables	ARRIS SURFboard SB8200 Cable Modem	Anker 2-in-1 USB Card Reader	Anker 4- port USB hub	Anker USB C to HDMI Adapter	Apple Lightning to Digital AV Adapter	iFixit Pro Tech Toolkit	iPhone 11 case	iPhone 12 Charger cable	iPhone 12 iP Pro case	hone 12 case	iPhone Charger Cable Anker 6ft	Phone SE nan	nonda USB C to USB Adapter	seenda Wireless mouse
	0	True	False	False	True	False	False	False	False	False	False	False	False	False	False	False	False	False False	True	False
	1	False	False	False	False	False	False	False	False	False	True	False	False	False	False	False	False	False True	False	False
	2	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False
	3	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False
	4	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False
		-	-	-			-	-	_			-	_	-		_	-		-	
	7496	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False
	7497	False	False	False	False	False	True	False	False	False	True	False	False	False	False	False	False	False True	False	False
	7498	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False
	7499	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False
	7500	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False True	False	False

7501 rows × 120 columns

In [7]: # Display items as columns
for col in transaction.columns:
 print(col)

In Colonia of colonia of colonia of the part (colonia) print(col) and the part (col) and (col)

HP 62XL Ink
HP 63 Int
HP 64 Int
HP 64 Int
HP 64 Int
HP 65 Int
HP 6

NOMER A-2 Air Pump Linear
VASKONA lpack left Nylon Lightning Cable
Ne ST-ri-cable No.
First Two Tech Toolkit
First Two Tech Toolkit
First Two Tech Toolkit
First Two Tech Toolkit
First Lineare cable
Firmone 12 Casse
Firmone 12 Casse
Firmone Chapter Cable Anker 6ft
Firmone Chapter Chapte

In [8]: # Remove NaN columns from the transformed dataset
cleaned\_df = transaction.drop(['nan'], axis = 1)
cleaned\_df.head(7501)

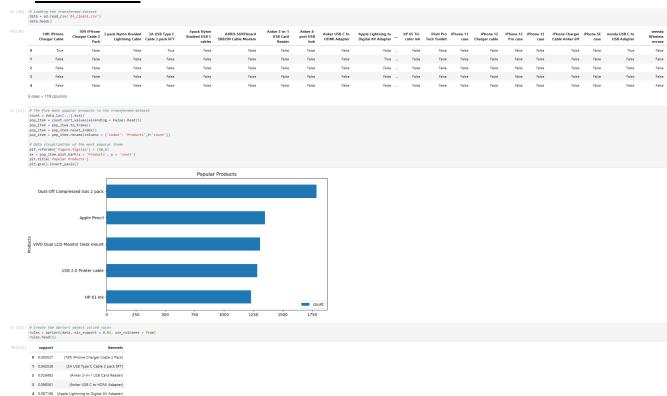
Dut[8]:		10ft iPHone Charger Cable	10ft iPHone Charger Cable 2 Pack	3 pack Nylon Braided Lightning Cable	3A USB Type C Cable 3 pack 6FT	Spack Nylon Braided USB C cables	ARRIS SURFboard SB8200 Cable Modem	Anker 2-in-1 USB Card Reader	Anker 4- port USB hub	Anker USB C to HDMI Adapter	Apple Lightning to Digital AV Adapter	hP 65 Tri- color ink		iPhone 11 case	iPhone 12 Charger cable	iPhone 12 Pro case	iPhone 12 case	iPhone Charger Cable Anker 6ft	iPhone SE case	nonda USB C to USB Adapter	seenda Wireless mouse
	0	True	False	False	True	False	False	False	False	False	False	False	False	False	False	False	False	False	False	True	False
	1	False	False	False	False	False	False	False	False	False	True	False	False	False	False	False	False	False	False	False	False
	2	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False
	3	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False
	4	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False
				-	-	-	-					-	-	-	-	-	-	-	-		
7	496	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False
7	497	False	False	False	False	False	True	False	False	False	True	False	False	False	False	False	False	False	False	False	False
7	498	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False
7	499	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False
7	500	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False	False

7501 rows × 119 columns

In [9]: # Write the transformed dataset to csv cleamed\_df.to\_csv('df\_clean1.csv', index=False) cleamed\_df.columns

hh 65 Ti-color ink', 'Hixit Pro Tech Toolkit', 'HPhone 11 case', 'HPhone 12 Charger cable', 'HPhone 12 Pro case', 'HPhone 12 case', 'HPhone 12 Charger cable Ainer 6\*\*, 'HPhone 54 case', 'Honde USE C to USE Adapter', 'seenda wireless mouse'), drype-'object', 'langthat19'

### **C2.** Code Execution



# C3. Association Rules Table

P	Create the rules table (lift > 1, ules_table = association_rules(rulules_table.head(10)		I)							
13]:	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric
0	(10ft iPHone Charger Cable 2 Pack)	(Dust-Off Compressed Gas 2 pack)	0.050527	0.238368	0.023064	0.456464	1.914955	0.011020	1.401255	0.503221
1	(Dust-Off Compressed Gas 2 pack)	(10ft iPHone Charger Cable 2 Pack)	0.238368	0.050527	0.023064	0.096756	1.914955	0.011020	1.051182	0.627330
2	(Anker USB C to HDMI Adapter)	(Dust-Off Compressed Gas 2 pack)	0.068391	0.238368	0.024397	0.356725	1.496530	0.008095	1.183991	0.356144
3	(Dust-Off Compressed Gas 2 pack)	(Anker USB C to HDMI Adapter)	0.238368	0.068391	0.024397	0.102349	1.496530	0.008095	1.037830	0.435627
4	(VIVO Dual LCD Monitor Desk mount)	(Anker USB C to HDMI Adapter)	0.174110	0.068391	0.020931	0.120214	1.757755	0.009023	1.058905	0.521973
5	(Anker USB C to HDMI Adapter)	(VIVO Dual LCD Monitor Desk mount)	0.068391	0.174110	0.020931	0.306043	1.757755	0.009023	1.190117	0.462740
6	(Apple Lightning to Digital AV Adapter)	(Apple Pencil)	0.087188	0.179709	0.028796	0.330275	1.837830	0.013128	1.224818	0.499424
7	(Apple Pencil)	(Apple Lightning to Digital AV Adapter)	0.179709	0.087188	0.028796	0.160237	1.837830	0.013128	1.086988	0.555754
8	(Apple Lightning to Digital AV Adapter)	(Dust-Off Compressed Gas 2 pack)	0.087188	0.238368	0.024397	0.279817	1.173883	0.003614	1.057552	0.162275
9	(Dust-Off Compressed Gas 2 pack)	(Apple Lightning to Digital AV Adapter)	0.238368	0.087188	0.024397	0.102349	1.173883	0.003614	1.016889	0.194486

# C4. Top Three Rules

# A screenshot of the top three relevant rules can be found below.

			-								
t	top_	ort the rules by the confidence three_rules = rules_table.sort_ three_rules		False).head(3)							
14]:		antecedents	consequents as	ntecedent support	onsequent support	support co	onfidence	lift	leverage	conviction 2	rhangs_metric
	0	(10ft iPHone Charger Cable 2 Pack)	(Dust-Off Compressed Gas 2 pack)	0.050527	0.238368 (	.023064	0.456464 1	.914955	0.011020	1.401255	0.503221
	36	(FEIYOLD Blue light Blocking Glasses)	(Dust-Off Compressed Gas 2 pack)	0.065858	0.238368 (	.027596	0.419028 1	.757904	0.011898	1.310962	0.461536
5	52	(SanDisk Ultra 64GB card)	(Dust-Off Compressed Gas 2 pack)	0.098254	0.238368 (	.040928	0.416554	.747522	0.017507	1.305401	0.474369
t	top_	ort the rules by the lift metric three_rules = rules_table.sort_ three_rules		head(3)							
		antecedents	consequents	antecedent suppor	t consequent suppor	t support	confidence	e li	ft leverag	e convictio	n zhangs_me
8	85	(VIVO Dual LCD Monitor Desk mount)	(SanDisk Ultra 64GB card)	0.17411	0.09825	4 0.039195	0.22511	5 2.29116	0.02208	8 1.16371	6 0.682
8	84	(SanDisk Ultra 64GB card)	(VIVO Dual LCD Monitor Desk mount)	0.09825	4 0.17411	0.039195	0.39891	5 2.29116	0.02208	8 1.37399	7 0.624
•	64	(FEIYOLD Blue light Blocking Glasses)	(VIVO Dual LCD Monitor Desk mount)	0.06585	8 0.17411	0.022930	0.34817	8 1.99975	68 0.01146	4 1.26704	8 0.535
t	top_	ort the rules by the support met _three_rules = rules_table.sort_ _three_rules		se).head(3)							
16]:		antecedents	consequents	antecedent suppor	t consequent suppor	t support	confidenc	e li	ft leverag	e convictio	n zhangs_me
•	62	(VIVO Dual LCD Monitor Desk mount)	(Dust-Off Compressed Gas 2 pack)	0.17411	0.23836	8 0.059725	0.34303	2 1.43908	5 0.01822	3 1.15931	4 0.369
6	63	(Dust-Off Compressed Gas 2 pack)	(VIVO Dual LCD Monitor Desk mount)	0.23836	8 0.17411	0.059725	0.25055	9 1.43908	5 0.01822	3 1.10200	8 0.400
4	41	(Dust-Off Compressed Gas 2 pack)	(HP 61 ink)	0.23836	8 0.16384	0.052660	0.22091	7 1.34833	2 0.01360	4 1.07325	6 0.339

#### Part IV: Data Summary & Implications

### D1. Significance of Support, Life, & Confidence Summary

In this analysis, the significance of support, life, and confidence is that these are the key metrics used to eliminate uncommon associations and focus on the more common, popular associations between items in the dataset. The descriptions of the metrics can be found below:

- Support: The support metric measures how often the item sets appear in the dataset together. A higher number represents a popular item set. According to the analysis, VIVO Dual LCD Monitor Desk Mount and the Dust-Off Compressed Gas 2 pack appear the most frequent together.
- Confidence: The confidence metric measures the strength of an item being purchased with another item, meaning how likely the item combination is to be purchased together. The higher the confidence, the stronger the association between the items. The pair with the highest confidence metric according to the analysis would be the 10ft iPhone Charger Cable 2 Pack and the Dust-Off Compressed Gas 2 pack.
- Lift: The lift metric is a ratio of the actual support to the expected support of the pair of items if they were independent. Essentially, this shows if the pair of items are purchased more together than they are purchased separately. A value greater than 1 represents the items are associated with each other and the first item has some type of influence on the purchase of the second item.

# D2. Practical Significance of Findings

The practical significance of the analysis shows that the relationships between item sets can be found using the market basket analysis. Based on the association rules discussed in D1, the VIVO Dual LCD Monitor Desk Mount and the Dust-Off Compressed Gas 2 pack have a high support value, indicating they are a popular item set, commonly purchased together. The VIVO Dual LCD Monitor Desk mount and SanDisk Ultra 64GB card have a 2.29 lift value, much greater than 1. This indicates the first item has some type of influence on the purchase of the second item. Therefore, I believe the company could benefit from offering bundling discounts on items frequently purchased together and enhancing product placement to retain customers and increase customer satisfaction.

### **D3.** Course of Action

The VIVO Dual LCD Monitor Desk Mount and the Dust-Off Compressed Gas 2 pack appear the most frequent together and are in the top three most popular products. The SanDisk Ultra 64GB card and FEIYOLD Blue Light Blocking Glasses are not as popular but are paired up frequently with the monitor and desk mount. To increase sales for the less popular items, I recommend adding some bundling discounts with the more popular items. For example, if you purchase the mount (full price), you can get the blue light glasses for 30% off. I also recommend arranging the items on websites or in stores to where they are near each other and list it under "frequently bought together" to increase more foot/site traffic.

### **Part V: Attachments**

## E. Panopto Video of Code/Programs

## F. Sources for Third-Party Code

Kamara, K. (n.d.). Market Basket Analysis in Python. WGU Panopto. Retrieved January 2, 2025, from

 $\frac{\text{https://wgu.hosted.panopto.com/Panopto/Pages/Viewer.aspx?id=db85c4f1-0da5-dbde-a1a4-b07c0019d46d.}$ 

## G. Sources

Kamara, K. (n.d.). Market Basket Analysis in Theory. WGU Panopto. Retrieved January 2, 2025, from

 $\frac{https://wgu.hosted.panopto.com/Panopto/Pages/Viewer.aspx?id=9541a29b-2f14-4c5d-9d86-af030005bcf6.$ 

# H. Professional Communication