

Task 3: Association and Lift Analysis

Data Preparation

```
In [1]: # import libraries and packages

import pandas as pd
import numpy as np
from mlxtend.preprocessing import TransactionEncoder
from mlxtend.frequent_patterns import apriori, association_rules

In [2]: # load data set

df = pd.read_csv('teleco_market_basket.csv')

In [3]: df.shape

Out[3]: (15002, 20)

In [4]: # inspect first 5 rows

df.head()

Out[4]:
```

	Item01	Item02	Item03	Item04	Item05	Item06	Item07	Item08	Item09	Item10	Item11	Item12	Item13	Item14
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1	Logitech M510 Wireless mouse	HP 63 Ink	HP 65 ink	nonda USB C to USB Adapter	10ft iPhone Charger Cable	HP 902XL ink	Creative Pebble 2.0 Speakers	Cleaning Gel Universal Dust Cleaner	Micro Center 32GB Memory card	YUNSONG 3pack 6ft Nylon Lightning Cable	TopMate C5 Laptop Cooler pad	Apple USB-C Charger cable	HyperX Cloud Stinger Headset	TONOR USB Gaming Microphone
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	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
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN

```
In [5]: # look for null values

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15002 entries, 0 to 15001
Data columns (total 20 columns):
 #   Column      Non-Null Count  Dtype
---  --
 0   Item01      7501 non-null    object
 1   Item02      5747 non-null    object
 2   Item03      4389 non-null    object
 3   Item04      3345 non-null    object
 4   Item05      2529 non-null    object
 5   Item06      1864 non-null    object
 6   Item07      1369 non-null    object
 7   Item08      981 non-null     object
 8   Item09      654 non-null     object
 9   Item10      395 non-null     object
10  Item11      256 non-null     object
11  Item12      154 non-null     object
12  Item13      87 non-null      object
13  Item14      47 non-null      object
14  Item15      25 non-null      object
15  Item16      8 non-null       object
16  Item17      4 non-null       object
17  Item18      4 non-null       object
18  Item19      3 non-null       object
19  Item20      1 non-null       object
dtypes: object(20)
memory usage: 2.3+ MB

In [6]: # Drop records with all missing values

df.dropna(how = 'all', inplace = True)
df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 7501 entries, 1 to 15001
Data columns (total 20 columns):
 #   Column      Non-Null Count  Dtype
---  --
 0   Item01      7501 non-null    object
 1   Item02      5747 non-null    object
 2   Item03      4389 non-null    object
 3   Item04      3345 non-null    object
 4   Item05      2529 non-null    object
 5   Item06      1864 non-null    object
 6   Item07      1369 non-null    object
 7   Item08      981 non-null     object
 8   Item09      654 non-null     object
 9   Item10      395 non-null     object
10  Item11      256 non-null     object
11  Item12      154 non-null     object
12  Item13      87 non-null      object
13  Item14      47 non-null      object
14  Item15      25 non-null      object
15  Item16      8 non-null       object
16  Item17      4 non-null       object
17  Item18      4 non-null       object
18  Item19      3 non-null       object
19  Item20      1 non-null       object
dtypes: object(20)
memory usage: 1.2+ MB

In [7]: df.describe()

Out[7]:
```

	Item01	Item02	Item03	Item04	Item05	Item06	Item07	Item08	Item09	Item10	Item11	Item12	Item13	Item14
count	7501	5747	4389	3345	2529	1864	1369	981	654	395	256	154	87	47
unique	115	117	115	114	110	106	102	97	88	80	66	50	47	47
top	Dust-Off Compressed Gas 2 pack	Dust-Off Compressed Gas 2 pack	Dust-Off Compressed Gas 2 pack	Dust-Off Compressed Gas 2 pack	Dust-Off Compressed Gas 2 pack	Apple USB-C Charger cable	USB 2.0 Printer cable	Apple USB-C Charger cable	Apple USB-C Charger cable	Apple USB-C Charger cable	Apple USB-C Charger cable	TopMate C5 Laptop Cooler pad	Apple USB-C Charger cable	TONOR USB Gaming Microphone
freq	577	484	375	201	153	107	96	67	57	31	22	15	15	15

```
In [8]: # replace NaN with 0

df.fillna(0, inplace = True)
df.head()

Out[8]:
```

	Item01	Item02	Item03	Item04	Item05	Item06	Item07	Item08	Item09	Item10	Item11	Item12	Item13	Item14
1	Logitech M510 Wireless mouse	HP 63 Ink	HP 65 ink	nonda USB C to USB Adapter	10ft iPhone Charger Cable	HP 902XL ink	Creative Pebble 2.0 Speakers	Cleaning Gel Universal Dust Cleaner	Micro Center 32GB Memory card	YUNSONG 3pack 6ft Nylon Lightning Cable	TopMate C5 Laptop Cooler pad	Apple USB-C Charger cable	HyperX Cloud Stinger Headset	TONOR USB Gaming Microphone
3	Apple Lightning to Digital AV Adapter	TP-Link AC1750 Smart WiFi Router	Apple Pencil	0	0	0	0	0	0	0	0	0	0	0
5	UNEN Mfi Certified 5-pack Lightning Cable	0	0	0	0	0	0	0	0	0	0	0	0	0
7	Cat8 Ethernet Cable	HP 65 ink	0	0	0	0	0	0	0	0	0	0	0	0
9	Dust-Off Compressed Gas 2 pack	Screen Mom Screen Cleaner kit	Moread HDMI to VGA Adapter	HP 62XL Tri-Color ink	Apple USB-C Charger cable	0	0	0	0	0	0	0	0	0

```
In [9]: # find new shape after changes

df.shape

Out[9]: (7501, 20)
```

Code from http://rasbt.github.io/mlxtend/user_guide/preprocessing/TransactionEncoder/

```
In [10]: # convert dataset into list form

telelist = []
for i in range(0, 7501):
    telelist.append([str(df.values[i, j]) for j in range(0, 20)])

In [11]: # convert lists into one-hot encoded array
# convert boolean to int for memory efficiency
encoder = TransactionEncoder()
encoded_array = encoder.fit(telelist).transform(telelist)
encoded_array.astype('int')

Out[11]: array([[0, 1, 0, ..., 0, 1, 0],
 [1, 0, 0, ..., 0, 0, 0],
 [1, 0, 0, ..., 0, 0, 0],
 ...,
 [1, 0, 0, ..., 0, 0, 0],
 [1, 0, 0, ..., 0, 0, 0],
 [1, 0, 0, ..., 0, 0, 0]])

In [12]: # check list form is correct

telelist[:1]

Out[12]: [['Logitech M510 Wireless mouse',
 'HP 63 Ink',
 'HP 65 ink',
 'nonda USB C to USB Adapter',
 '10ft iPhone Charger Cable',
 'HP 902XL ink',
 'Creative Pebble 2.0 Speakers',
 'Cleaning Gel Universal Dust Cleaner',
 'Micro Center 32GB Memory card',
 'YUNSONG 3pack 6ft Nylon Lightning Cable',
 'TopMate C5 Laptop Cooler pad',
 'Apple USB-C Charger cable',
 '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',
 'FEIYOLD NICE 5 pack 10ft Lightning cable',
 'FEIYOLD Blue light Blocking Glasses']]

In [13]: # turn array into DataFrame

tele_df = pd.DataFrame(encoded_array, columns = encoder.columns_)
tele_df

Out[13]:
```

	0	10ft iPhone Charger Cable	10ft iPhone Charger Cable 2 Pack	3 pack Nylon Braided Lightning Cable	3A USB Type C Cable 3 pack 6FT	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	...	hP 65 Tri-color ink	iFixit Pro Tech Toolkit	iPhone 11 case	iPhone 12 Charger cable
0	False	True	False	False	True	False	False	False	False	False	...	False	False	False	False
1	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
2	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
3	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
4	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
...
7496	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
7497	True	False	False	False	False	False	True	False	False	False	...	False	False	False	False
7498	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
7499	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False
7500	True	False	False	False	False	False	False	False	False	False	...	False	False	False	False

7501 rows x 120 columns

```
In [14]: # remove unneeded column

tele_df = tele_df.drop(['0'], axis = 1)

In [15]: df.to_csv('association_rules_prepared_data.csv')
```

Data Analysis

```
In [16]: # run apriori rule on data set

frequent_items = apriori(tele_df, min_support = 0.01, max_len = 3, use_colnames= True)
frequent_items.head()

Out[16]:
```

	support	itemsets
0	0.050527	(10ft iPhone Charger Cable 2 Pack)
1	0.042528	(3A USB Type C Cable 3 pack 6FT)
2	0.019064	(5pack Nylon Braided USB C cables)
3	0.010932	(ARRIS SURFboard SB8200 Cable Modem)
4	0.029463	(Anker 2-in-1 USB Card Reader)

```
In [17]: # print number of rules created

association_results = list(frequent_items)
print(len(association_results))

257

In [18]: # sort frequent items by support value

frequent_items.sort_values(by = 'support', ascending = False, inplace = True)
frequent_items.head()

Out[18]:
```

	support	itemsets
18	0.238368	(Dust-Off Compressed Gas 2 pack)
8	0.179709	(Apple Pencil)
66	0.174110	(VIVO Dual LCD Monitor Desk mount)
63	0.170911	(USB 2.0 Printer cable)
23	0.163845	(HP 61 ink)

```
In [19]: # generate rules for lift in descending order

lift_rules = association_rules(frequent_items, metric = 'lift', min_threshold = 1)
lift_rules.sort_values(by = 'lift', ascending = False, inplace = True)
lift_rules

Out[19]:
```

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction
162	(SanDisk 128GB Ultra microSDXC card)	(SanDisk Ultra 64GB card)	0.049460	0.098254	0.015998	0.323450	3.291994	0.011138	1.332860
163	(SanDisk Ultra 64GB card)	(SanDisk 128GB Ultra microSDXC card)	0.098254	0.049460	0.015998	0.162822	3.291994	0.011138	1.135410
131	(VIVO Dual LCD Monitor Desk mount, Dust-Off Co...	(SanDisk Ultra 64GB card)	0.059725	0.098254	0.017064	0.285714	2.907928	0.011196	1.262445
134	(SanDisk Ultra 64GB card)	(VIVO Dual LCD Monitor Desk mount, Dust-Off Co...	0.098254	0.059725	0.017064	0.173677	2.907928	0.011196	1.137902
387	(FEIYOLD Blue light Blocking Glasses)	(VIVO Dual LCD Monitor Desk mount, Dust-Off Co...	0.065858	0.059725	0.010265	0.155870	2.609786	0.006332	1.113898
...
252	(TopMate C5 Laptop Cooler pad)	(USB 2.0 Printer cable)	0.076523	0.170911	0.013332	0.174216	1.019340	0.000253	1.004003
266	(Apple Pencil)	(FEIYOLD Blue light Blocking Glasses)	0.179709	0.065858	0.011998	0.066766	1.013783	0.000163	1.000973
267	(FEIYOLD Blue light Blocking Glasses)	(Apple Pencil)	0.065858	0.179709	0.011998	0.182186	1.013783	0.000163	1.003029
231	(HP 63XL Ink)	(VIVO Dual LCD Monitor Desk mount)	0.079323	0.174110	0.013998	0.176471	1.013557	0.000187	1.002866
230	(VIVO Dual LCD Monitor Desk mount)	(HP 63XL Ink)	0.174110	0.079323	0.013998	0.080398	1.013557	0.000187	1.001169

406 rows x 9 columns

```
In [20]: # generate confidence rules in descending order

confidence_rules = association_rules(frequent_items, metric = 'confidence', min_threshold = 0.1)
confidence_rules.sort_values(by = 'confidence', ascending = False, inplace = True)
confidence_rules

Out[20]:
```

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction
310	(SanDisk Ultra 64GB card, Apple Pencil)	(Dust-Off Compressed Gas 2 pack)	0.019997	0.238368	0.010132	0.506667	2.125563	0.005365	1.543848
264	(SanDisk Ultra 64GB card, Screen Mom Screen Cl...	(Dust-Off Compressed Gas 2 pack)	0.021997	0.238368	0.011065	0.503030	2.110308	0.005822	1.532552
273	(HP 61 ink, SanDisk Ultra 64GB card)	(Dust-Off Compressed Gas 2 pack)	0.023064	0.238368	0.010932	0.473988	1.988472	0.005434	1.447937
271	(Nylon Braided Lightning to USB cable, Screen ...	(Dust-Off Compressed Gas 2 pack)	0.023597	0.238368	0.011065	0.468927	1.967236	0.005440	1.434136
67	(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
...
56	(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
214	(Screen Mom Screen Cleaner kit)	(TopMate C5 Laptop Cooler pad)	0.129583	0.076523	0.013198	0.101852	1.330994	0.003282	1.028201
218	(Screen Mom Screen Cleaner kit)	(Apple Pencil, Dust-Off Compressed Gas 2 pack)	0.129583	0.050927	0.013065	0.100823	1.979774	0.006466	1.055491
59	(Dust-Off Compressed Gas 2 pack)	(TopMate C5 Laptop Cooler pad)	0.238368	0.076523	0.023997	0.100671	1.315565	0.005756	1.026851
145	(HP 61 ink)	(FEIYOLD Blue light Blocking Glasses)	0.163845	0.065858	0.016398	0.100081	1.519657	0.005607	1.038030

320 rows x 9 columns

```
In [21]: # find top three rules

rules = association_rules(frequent_items, metric = 'lift', min_threshold = 1)
top_three = rules[(rules['support'] > 0.003) & (rules['confidence'] > 0.4) & (rules['lift'] > 1.00)]
top_three.head(3)

Out[21]:
```

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction
8	(SanDisk Ultra 64GB card)	(Dust-Off Compressed Gas 2 pack)	0.098254	0.238368	0.040928	0.416554	1.747522	0.017507	1.305401
39	(FEIYOLD Blue light Blocking Glasses)	(Dust-Off Compressed Gas 2 pack)	0.065858	0.238368	0.027596	0.419028	1.757904	0.011898	1.310962
62	(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

```
In [ ]:

In [ ]:
```