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Batch: 09012021

ASSOCIATION RULES

Problem Statement: - A Mobile Phone manufacturing company wants to launch its three brand new phone into the market, but before going with its traditional marketing approach this time it want to analyze the data of its previous model sales in different regions and you have been hired as an Data Scientist to help them out, use the Association rules concept and provide your insights to the company's marketing team to improve its sales.

4.) myphonedata.csv

Importing Libraries

```
In [1]:
```

```
import pandas as pd
from mlxtend.frequent_patterns import apriori, association_rules
```

Loading the dataset

In [2]:

```
phone = pd.read_csv(r"F:\360\associationrules\myphonedata.csv")
phone
```

Out[2]:

	V1	V2	V3	red	white	green	yellow	orange	blue
0	red	white	green	1	1	1	0	0	0
1	white	orange	NaN	0	1	0	0	1	0
2	white	blue	NaN	0	1	0	0	0	1
3	red	white	orange	1	1	0	0	1	0
4	red	blue	NaN	1	0	0	0	0	1
5	white	blue	NaN	0	1	0	0	0	1
6	red	blue	NaN	1	0	0	0	0	1
7	red	white	blue	1	1	0	0	0	1
8	green	NaN	NaN	0	0	1	0	0	0
9	red	white	blue	1	1	0	0	0	1
10	yellow	NaN	NaN	0	0	0	1	0	0

Removing the unwanted columns

In [3]:

```
phone.drop(phone.iloc[: , 0:3], inplace = True, axis = 1)
phone
```

Out[3]:

	red	white	green	yellow	orange	blue
0	1	1	1	0	0	0
1	0	1	0	0	1	0
2	0	1	0	0	0	1
3	1	1	0	0	1	0
4	1	0	0	0	0	1
5	0	1	0	0	0	1
6	1	0	0	0	0	1
7	1	1	0	0	0	1
8	0	0	1	0	0	0
9	1	1	0	0	0	1
10	0	0	0	1	0	0

Finding the frequent items

In [4]:

```
frequent_itemsets = apriori(phone,min_support=0.005, max_len=3,use_colnames = True)
frequent_itemsets.shape
```

Out[4]:

(16, 2)

Most Frequent item sets based on support

In [5]:

```
frequent_itemsets.sort_values('support',ascending = False,inplace=True)
```

In [6]:

```
import matplotlib.pylab as plt
```

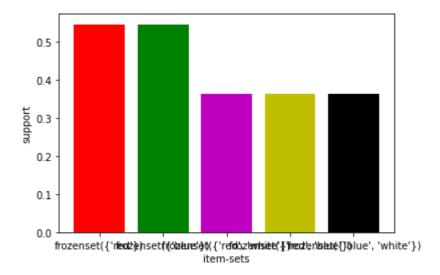
In [7]:

```
plt.bar(x = list(range(1,6)),height = frequent_itemsets.support[1:6],color='rgmyk')
plt.xticks(list(range(1,6)),frequent_itemsets.itemsets[1:6])
plt.xlabel('item-sets')
plt.ylabel('support')
```

<ipython-input-7-86a3c60dc1e3>:1: MatplotlibDeprecationWarning: Using a stri
ng of single character colors as a color sequence is deprecated since 3.2 an
d will be removed two minor releases later. Use an explicit list instead.
 plt.bar(x = list(range(1,6)),height = frequent_itemsets.support[1:6],color
='rgmyk')

Out[7]:

Text(0, 0.5, 'support')



In [8]:

```
rules = association_rules(frequent_itemsets, metric="lift", min_threshold=1)
rules.shape
```

Out[8]:

(18, 9)

In [9]:

```
rules.head(20)
rules.sort_values('lift',ascending = False,inplace=True)
```

In [10]:

```
# To eliminate Redudancy in Rules
```

```
In [11]:
```

```
def to_list(i):
    return (sorted(list(i)))
```

In [12]:

```
ma_X = rules.antecedents.apply(to_list)+rules.consequents.apply(to_list)
```

In [13]:

```
ma_X = ma_X.apply(sorted)
```

In [14]:

```
rules_sets = list(ma_X)
```

In [15]:

```
unique_rules_sets = [list(m) for m in set(tuple(i) for i in rules_sets)]
index_rules = []
for i in unique_rules_sets:
   index_rules.append(rules_sets.index(i))
```

In [16]:

```
# getting rules without any redudancy
```

In [17]:

```
rules_no_redudancy = rules.iloc[index_rules,:]
```

In [18]:

```
# Sorting them with respect to list and getting top 10 rules
```

In [19]:

```
rules_no_redudancy.sort_values('lift',ascending=False).head(10)
```

Out[19]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	levera
9	(green, white)	(red)	0.090909	0.545455	0.090909	1.000000	1.833333	0.0413
6	(white)	(orange)	0.636364	0.181818	0.181818	0.285714	1.571429	0.0661
16	(white)	(red, orange)	0.636364	0.090909	0.090909	0.142857	1.571429	0.0330
3	(blue)	(red)	0.545455	0.545455	0.363636	0.666667	1.222222	0.0661
5	(white)	(blue)	0.636364	0.545455	0.363636	0.571429	1.047619	0.0165
1	(white)	(red)	0.636364	0.545455	0.363636	0.571429	1.047619	0.0165
4								•

In [20]:

Perform algorithm for different support, connfidence value and max length

In [21]:

```
frequent_itemsets1 = apriori(phone, min_support=0.007, max_len=4,use_colnames = True)
```

In [22]:

```
# Most Frequent item sets based on support
```

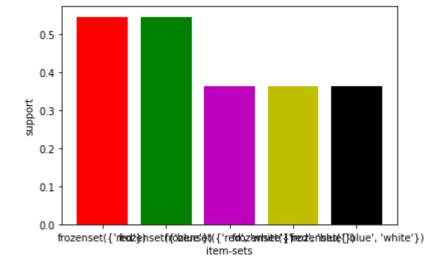
In [23]:

```
frequent_itemsets1.sort_values('support',ascending = False,inplace=True)
plt.bar(x = list(range(1,6)),height = frequent_itemsets1.support[1:6],color='rgmyk')
plt.xticks(list(range(1,6)),frequent_itemsets1.itemsets[1:6])
plt.xlabel('item-sets')
plt.ylabel('support')
```

<ipython-input-23-6296cb9947a0>:2: MatplotlibDeprecationWarning: Using a str
ing of single character colors as a color sequence is deprecated since 3.2 a
nd will be removed two minor releases later. Use an explicit list instead.
 plt.bar(x = list(range(1,6)),height = frequent_itemsets1.support[1:6],colo
r='rgmyk')

Out[23]:

Text(0, 0.5, 'support')



In [24]:

```
rules1 = association_rules(frequent_itemsets1, metric="lift", min_threshold=1)
rules1.head(20)
rules1.sort_values('lift',ascending = False,inplace=True)
```

In [25]:

```
frequent_itemsets2 = apriori(phone, min_support=0.009, max_len=7,use_colnames = True)
```

In [26]:

```
# Most Frequent item sets based on support
```

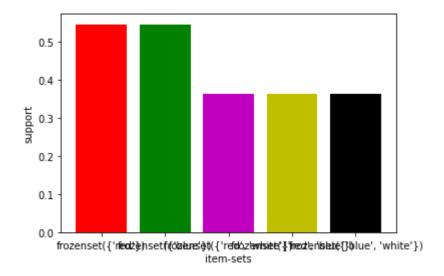
In [27]:

```
frequent_itemsets2.sort_values('support',ascending = False,inplace=True)
plt.bar(x = list(range(1,6)),height = frequent_itemsets2.support[1:6],color='rgmyk')
plt.xticks(list(range(1,6)),frequent_itemsets2.itemsets[1:6])
plt.xlabel('item-sets')
plt.ylabel('support')
```

<ipython-input-27-5c5f88d162be>:2: MatplotlibDeprecationWarning: Using a str
ing of single character colors as a color sequence is deprecated since 3.2 a
nd will be removed two minor releases later. Use an explicit list instead.
 plt.bar(x = list(range(1,6)),height = frequent_itemsets2.support[1:6],colo
r='rgmyk')

Out[27]:

Text(0, 0.5, 'support')



In [28]:

```
rules2 = association_rules(frequent_itemsets2, metric="lift", min_threshold=1)
rules2.head(20)
rules2.sort_values('lift',ascending = False,inplace=True)
```

##As min lenth value is changing the rules is changing. #rules =18 #rules1=18 #rules2=18