

# DMPM Lab 12

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Roll no.: 17

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
```

```
In [2]: df = pd.read_csv("D:/TY sem6/DMPM LAB/lab 12/MBO.csv", header=None)
```

```
In [3]: df.head()
```

Out[3]:

|   | 0                | 1         | 2             | 3                   | 4               | 5                      | 6    | 7                 | 8               | 9               | 10                   | 11           | 12    | 13    | 14               | 15     |        |
|---|------------------|-----------|---------------|---------------------|-----------------|------------------------|------|-------------------|-----------------|-----------------|----------------------|--------------|-------|-------|------------------|--------|--------|
| 0 | shrimp           | almonds   | avocado       | vegetables<br>mix   | green<br>grapes | whole<br>weat<br>flour | yams | cottage<br>cheese | energy<br>drink | tomato<br>juice | low<br>fat<br>yogurt | green<br>tea | honey | salad | mineral<br>water | salmon | antiox |
| 1 | burgers          | meatballs | eggs          | NaN                 | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             | NaN                  | NaN          | NaN   | NaN   | NaN              | NaN    | NaN    |
| 2 | chutney          | NaN       | NaN           | NaN                 | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             | NaN                  | NaN          | NaN   | NaN   | NaN              | NaN    | NaN    |
| 3 | turkey           | avocado   | NaN           | NaN                 | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             | NaN                  | NaN          | NaN   | NaN   | NaN              | NaN    | NaN    |
| 4 | mineral<br>water | milk      | energy<br>bar | whole<br>wheat rice | green<br>tea    | NaN                    | NaN  | NaN               | NaN             | NaN             | NaN                  | NaN          | NaN   | NaN   | NaN              | NaN    | NaN    |

```
In [4]: df.shape
```

Out[4]: (7501, 20)

```
In [5]: transactions = df.values.reshape(-1).tolist()
```

```
In [6]: transactions
```

```
Out[6]: ['shrimp',  
         'almonds',  
         'avocado',  
         'vegetables mix',  
         'green grapes',  
         'whole weat flour',  
         'yams',  
         'cottage cheese',  
         'energy drink',  
         'tomato juice',  
         'low fat yogurt',  
         'green tea',  
         'honey',  
         'salad',  
         'mineral water',  
         'salmon',  
         'antioxydant juice',  
         'frozen smoothie',  
         'spinach',  
         ...]
```

```
In [7]: df2 = pd.DataFrame(transactions)  
df2['Count']=1  
df2.head()
```

```
Out[7]:
```

|   | 0              | Count |
|---|----------------|-------|
| 0 | shrimp         | 1     |
| 1 | almonds        | 1     |
| 2 | avocado        | 1     |
| 3 | vegetables mix | 1     |
| 4 | green grapes   | 1     |

```
In [8]: df2=df2.groupby(by=[0], as_index=False).count().sort_values(by=['Count'], ascending=True) # count
df2['Percentage'] = (df2['Count'] / df2['Count'].sum()) # percentage
df2=df2.rename(columns={0 : 'Item'})
```

```
In [9]: df2 = df2.reset_index().drop('index',axis=1)
```

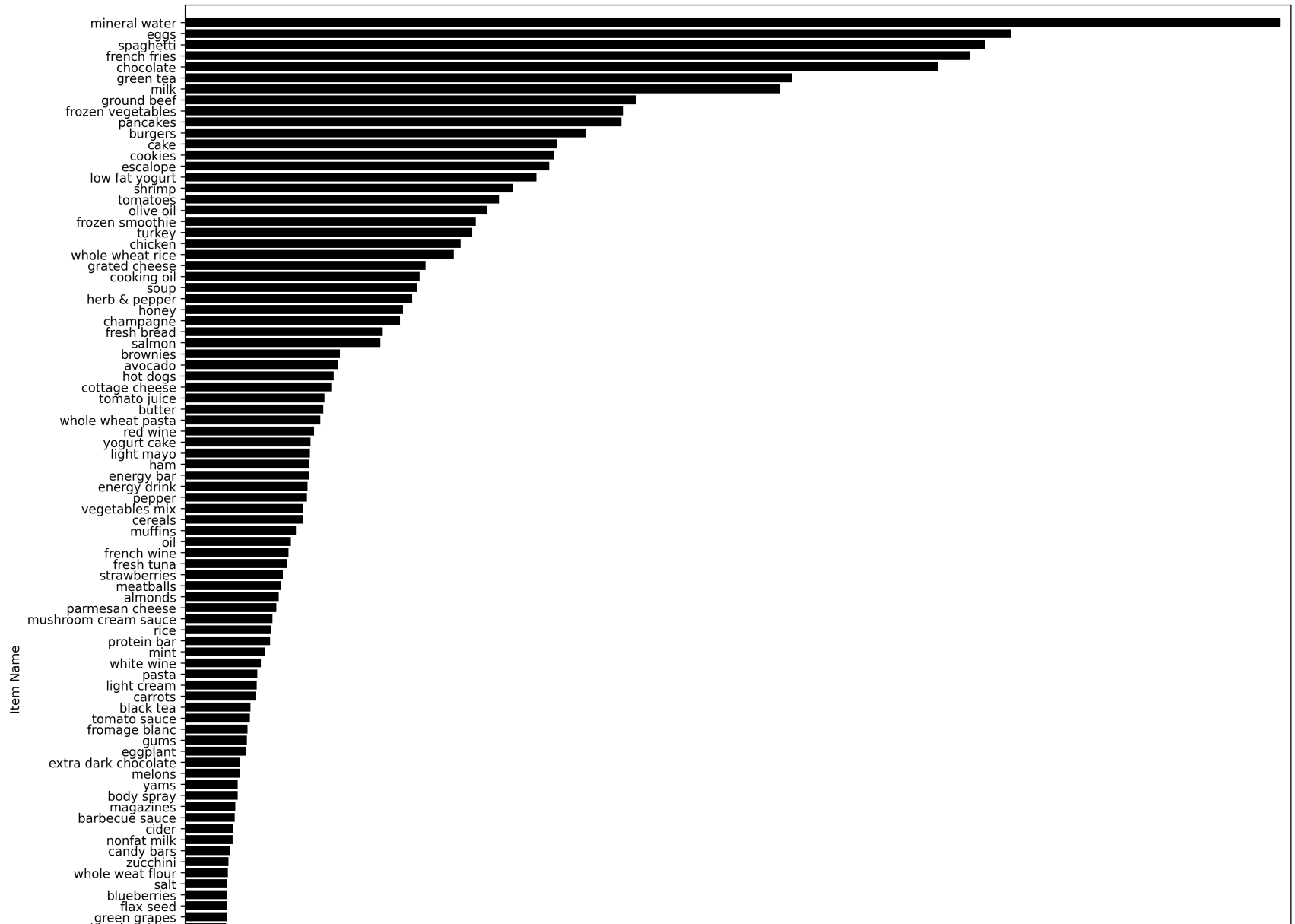
```
In [10]: df2
```

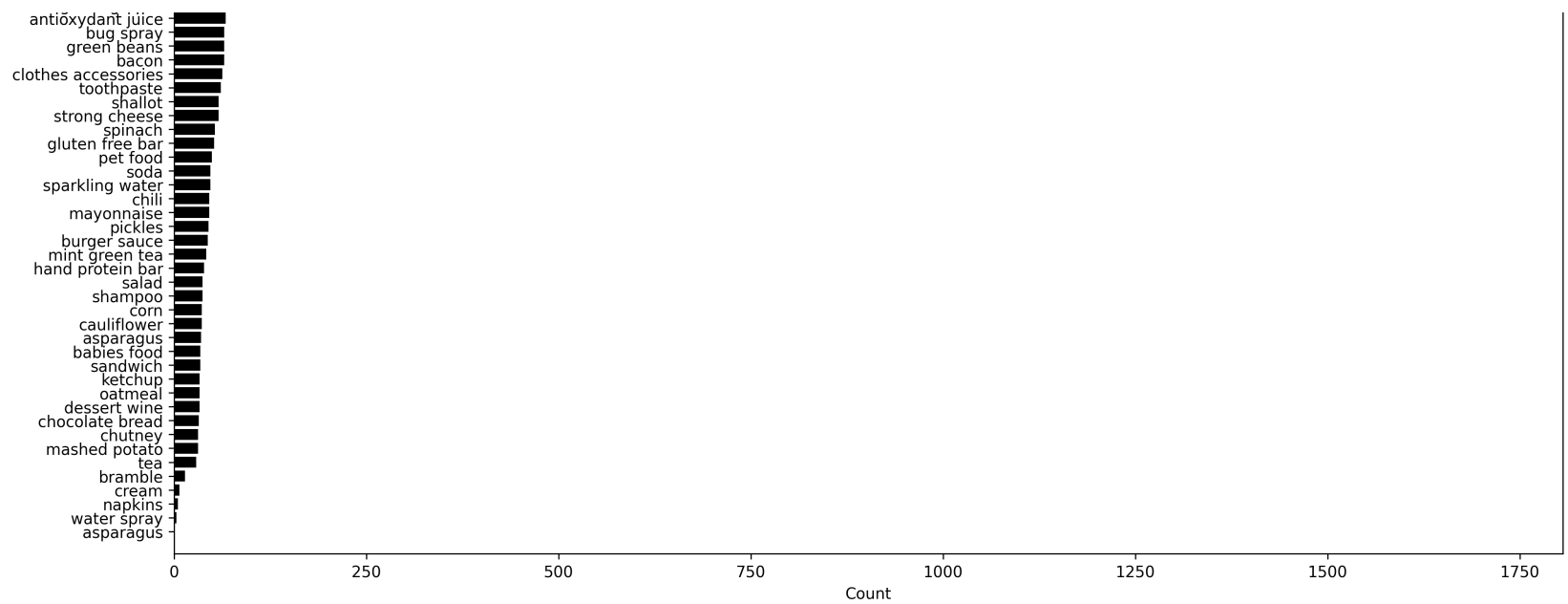
Out[10]:

|     | Item          | Count | Percentage |
|-----|---------------|-------|------------|
| 0   | asparagus     | 1     | 0.000034   |
| 1   | water spray   | 3     | 0.000102   |
| 2   | napkins       | 5     | 0.000170   |
| 3   | cream         | 7     | 0.000238   |
| 4   | bramble       | 14    | 0.000477   |
| ... | ...           | ...   | ...        |
| 115 | chocolate     | 1230  | 0.041889   |
| 116 | french fries  | 1282  | 0.043660   |
| 117 | spaghetti     | 1306  | 0.044478   |
| 118 | eggs          | 1348  | 0.045908   |
| 119 | mineral water | 1788  | 0.060893   |

120 rows × 3 columns

```
In [11]: plt.figure(figsize=(16,20), dpi=300)
plt.ylabel('Item Name')
plt.xlabel('Count')
plt.barh(df2['Item'], width=df2['Count'], color='black', height=0.8)
plt.margins(0.01)
plt.show()
```





```
In [12]: from efficient_apriori import apriori
```

In [13]:

```
txns2=df.stack().groupby(level=0).apply(list).tolist()

txns2
```

Out[13]:

```
[['shrimp',
  'almonds',
  'avocado',
  'vegetables mix',
  'green grapes',
  'whole weat flour',
  'yams',
  'cottage cheese',
  'energy drink',
  'tomato juice',
  'low fat yogurt',
  'green tea',
  'honey',
  'salad',
  'mineral water',
  'salmon',
  'antioxydant juice',
  'frozen smoothie',
  'spinach',
  ...], ...]
```

**s= 1, confidence = 30**

```
In [14]: itemsets, rules = apriori(txns2, min_support=0.01, min_confidence=0.3, verbosity=1)
```

```
Generating itemsets.  
Counting itemsets of length 1.  
Found 120 candidate itemsets of length 1.  
Found 75 large itemsets of length 1.  
Counting itemsets of length 2.  
Found 2775 candidate itemsets of length 2.  
Found 165 large itemsets of length 2.  
Counting itemsets of length 3.  
Found 477 candidate itemsets of length 3.  
Found 17 large itemsets of length 3.  
Counting itemsets of length 4.  
Found 2 candidate itemsets of length 4.  
Itemset generation terminated.  
  
Generating rules from itemsets.  
Generating rules of size 2.  
Generating rules of size 3.  
Rule generation terminated.
```

```
In [15]: for item in sorted(rules, key=lambda item: (item.lift), reverse=True)[:5]:  
        print(item)  
  
len(rules)
```

```
{herb & pepper} -> {ground beef} (conf: 0.323, supp: 0.016, lift: 3.292, conv: 1.333)  
{ground beef, mineral water} -> {spaghetti} (conf: 0.417, supp: 0.017, lift: 2.395, conv: 1.416)  
{frozen vegetables, mineral water} -> {milk} (conf: 0.310, supp: 0.011, lift: 2.390, conv: 1.261)  
{soup} -> {milk} (conf: 0.301, supp: 0.015, lift: 2.321, conv: 1.245)  
{ground beef} -> {spaghetti} (conf: 0.399, supp: 0.039, lift: 2.291, conv: 1.374)
```

```
Out[15]: 63
```

**s=2% , confidence = 40%**

```
In [16]: itemsets2, rules2 = apriori(txns2, min_support=0.02, min_confidence=0.4, verbosity=1)
```

```
Generating itemsets.  
Counting itemsets of length 1.  
  Found 120 candidate itemsets of length 1.  
  Found 53 large itemsets of length 1.  
Counting itemsets of length 2.  
  Found 1378 candidate itemsets of length 2.  
  Found 50 large itemsets of length 2.  
Counting itemsets of length 3.  
  Found 61 candidate itemsets of length 3.  
Itemset generation terminated.  
  
Generating rules from itemsets.  
  Generating rules of size 2.  
Rule generation terminated.
```

```
In [17]: for item in sorted(rules2, key=lambda item: (item.lift), reverse=True)[:5]:  
        print(item.lhs+item.rhs)  
len(rules2)
```

```
('soup', 'mineral water')  
('olive oil', 'mineral water')  
('ground beef', 'mineral water')
```

```
Out[17]: 3
```

**S= 3%, confidence=50%**



```
In [18]: itemsets3, rules3 = apriori(txns2, min_support=0.03, min_confidence=0.5, verbosity=1)
```

```
Generating itemsets.  
Counting itemsets of length 1.  
  Found 120 candidate itemsets of length 1.  
  Found 36 large itemsets of length 1.  
Counting itemsets of length 2.  
  Found 630 candidate itemsets of length 2.  
  Found 18 large itemsets of length 2.  
Counting itemsets of length 3.  
  Found 14 candidate itemsets of length 3.  
Itemset generation terminated.  
  
Generating rules from itemsets.  
  Generating rules of size 2.  
Rule generation terminated.
```

```
In [19]: for item in sorted(rules3, key=lambda item: (item.lift), reverse=True)[:5]:  
        print(item)  
        len(rules3)
```

```
Out[19]: 0
```

```
In [20]: rules[1]
```

```
Out[20]: {burgers} -> {eggs}
```

```
In [21]: dict = {'Rules':[],
                'Item1':[],
                'Item2':[],
                'Confidence':[],
                'Support':[],
                "Lift":[]
            }

data = pd.DataFrame(dict)

for item in sorted(rules, key=lambda item: (item.lift,item.confiction), reverse=True):
    data.loc[len(data.index)] = [(item.lhs+item.rhs), item.lhs,item.rhs,item.confidence, item.support,item.lift]
```

C:\Users\saniy\Anaconda3\lib\site-packages\numpy\core\\_asarray.py:83: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray  
 return array(a, dtype, copy=False, order=order)

```
In [22]: data.shape
```

```
Out[22]: (63, 6)
```

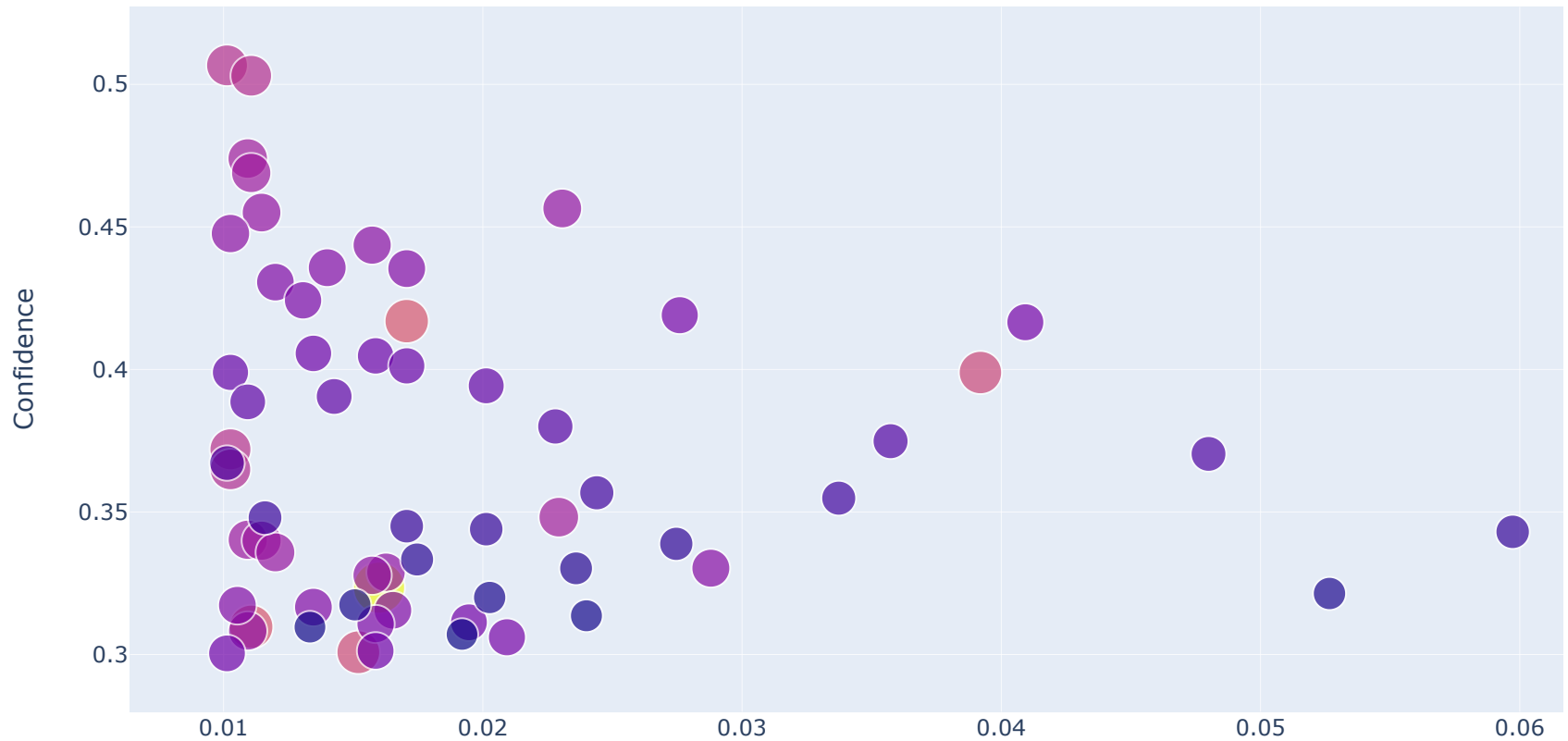
```
In [23]: data.head()
```

```
Out[23]:
```

|   | Rules                                    | Item1                              | Item2          | Confidence | Support  | Lift     |
|---|--|------------------------------------|----------------|------------|----------|----------|
| 0 | (herb & pepper, ground beef)             | (herb & pepper,)                   | (ground beef,) | 0.323450   | 0.015998 | 3.291994 |
| 1 | (ground beef, mineral water, spaghetti)  | (ground beef, mineral water)       | (spaghetti,)   | 0.416938   | 0.017064 | 2.394681 |
| 2 | (frozen vegetables, mineral water, milk) | (frozen vegetables, mineral water) | (milk,)        | 0.309701   | 0.011065 | 2.389991 |
| 3 | (soup, milk)                             | (soup,)                            | (milk,)        | 0.300792   | 0.015198 | 2.321232 |
| 4 | (ground beef, spaghetti)                 | (ground beef,)                     | (spaghetti,)   | 0.398915   | 0.039195 | 2.291162 |

```
In [24]: import plotly.express as px
```

```
In [25]: fig = px.scatter(data, x="Support", y="Confidence", color="Lift",  
                        size='Lift')  
fig.show()
```



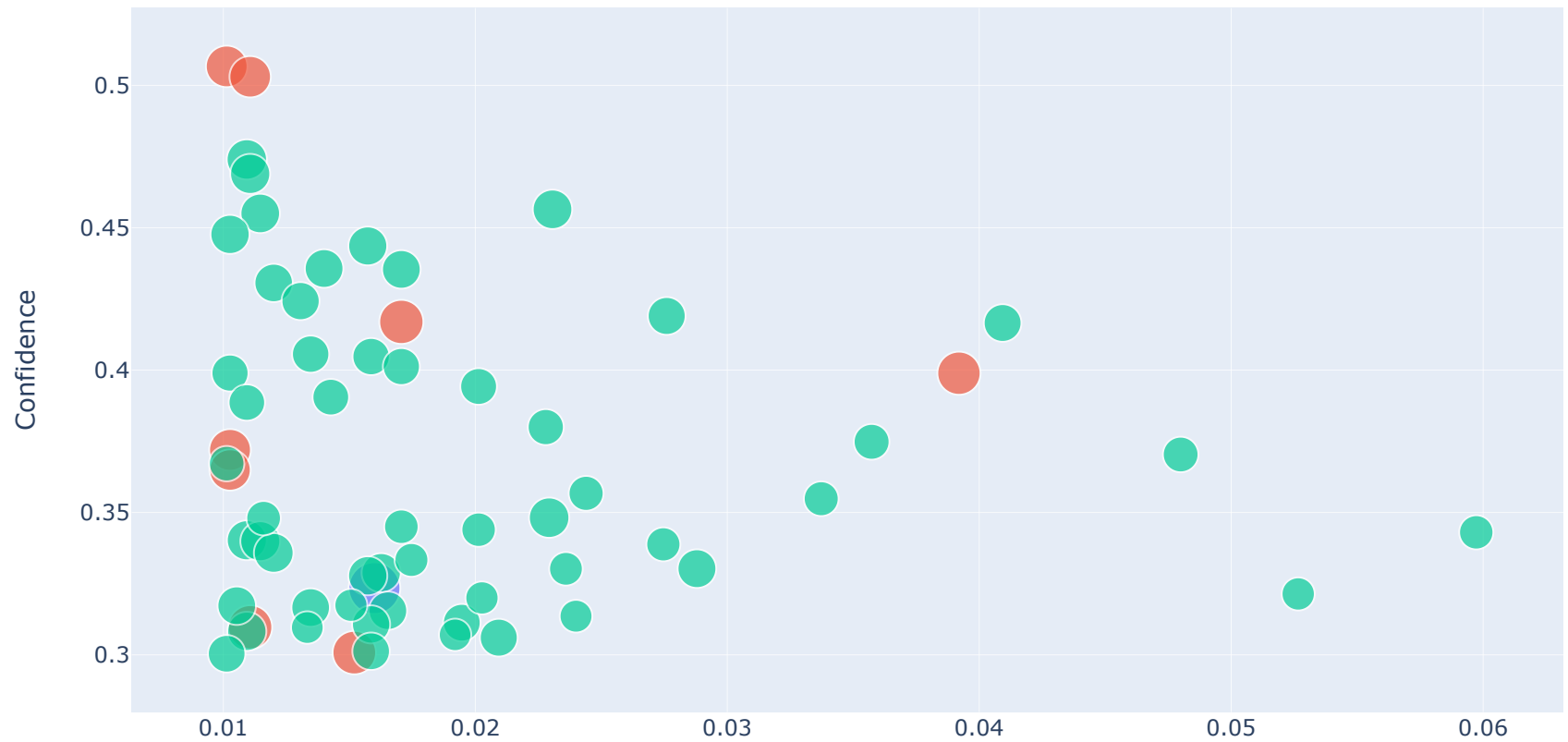
```
In [29]: data['LiftGroups'] = data['Lift'].apply(lambda x: '1-2' if 1<=x<2 else '2-3' if 2<=x<3 else 'Above 3')
```

In [30]: data.head()

Out[30]:

|   | Rules                                    | Item1                              | Item2          | Confidence | Support  | Lift     | LiftGroups |
|---|--|------------------------------------|----------------|------------|----------|----------|------------|
| 0 | (herb & pepper, ground beef)             | (herb & pepper,)                   | (ground beef,) | 0.323450   | 0.015998 | 3.291994 | Above 3    |
| 1 | (ground beef, mineral water, spaghetti)  | (ground beef, mineral water)       | (spaghetti,)   | 0.416938   | 0.017064 | 2.394681 | 2-3        |
| 2 | (frozen vegetables, mineral water, milk) | (frozen vegetables, mineral water) | (milk,)        | 0.309701   | 0.011065 | 2.389991 | 2-3        |
| 3 | (soup, milk)                             | (soup,)                            | (milk,)        | 0.300792   | 0.015198 | 2.321232 | 2-3        |
| 4 | (ground beef, spaghetti)                 | (ground beef,)                     | (spaghetti,)   | 0.398915   | 0.039195 | 2.291162 | 2-3        |

```
In [34]: fig = px.scatter(data, x="Support", y="Confidence", color="LiftGroups",  
                        size='Lift')  
fig.show()
```



```
In [36]: fig = px.bar(data, x="LiftGroups", y="Confidence", color="LiftGroups")  
fig.show()
```

