CS 422

Data Mining

HOMEWORK ASSIGNMENT 2 PRACTICUM QUESTIONS 1 AND 2

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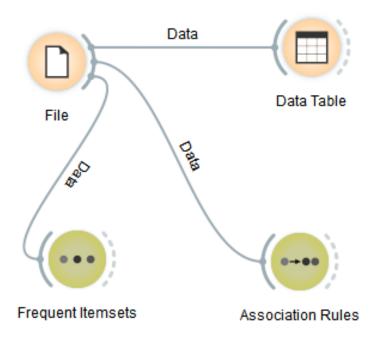
1 Question 1:

Load the *market-basket* sample dataset into the Orange application, and run both frequent itemset as well as association rule modules. Set the *support threshold* to 10% and observe the *antecedent* in the rules with the highest lift. What item is observed to be there, and what is its support? Is this a valuable association rule? Why or why not?

Answer:

• Creating appropriate workflow with the *market-basket.tab* dataset.

Figure 1: Workflow



• Figure 2. shows the market-basket dataset tabulated in *Orange*.

Figure 2: Market Basket Dataset



• Setting up the support threshold to 10% and confidence to 90% in both the Frequent Itemset generation as well as Association Rule generation, we get results as shown in Figure 3 and 4.

Figure 3: Frequent Itemsets in Market-basket

*** Frequent Itemsets

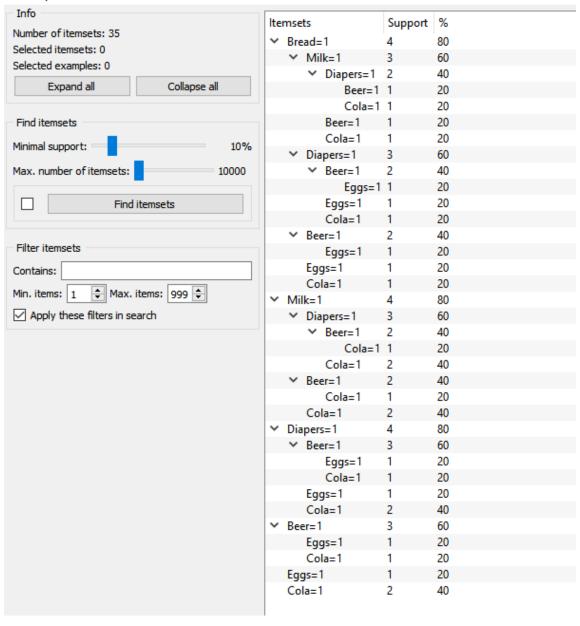
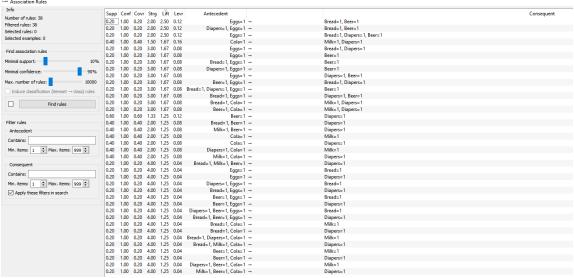


Figure 4: Association Rules in Market-Basket



• the table in Figure 4. has been sorted based on the descending order of the *lift* metric.

1.1 Observations:

- From Figure 4. the association rules with the highest lift are:

```
\{eggs \rightarrow \{bread, beer\}\}\
\{\{diapers, eggs\} \rightarrow \{bread, beer\}\}\
\{Eggs \rightarrow \{bread, diapers, beer\}\}\
```

- All of the above have a lift of textit2.50.
- The two items in the *antecedent*, **eggs** and **diapers**, have a support of **20%** and **80%** respectively.
- As for the association rules, all of them have a support of 20%.

1.2 Inference:

The Lift metric (also known as Interest), is the measure of how many times items X and Y occur together than expected if they were statistically independent.

$$Lift(X \to Y) = \frac{Conf(X \to Y)}{Supp(Y)} = \frac{P(X \to Y)}{P(X)P(Y)}$$
 (1)

From figure 4 it is observed that all associations have a confidence of 1. From the above equation, lift is inversely proportional to the support of the antecedent. Thus, antecedents with a low support count can produce high lift values. This usually happens when the dataset is small like this one and one rare item occurs a minimum of once along with other common items. In this dataset, Eggs appears in only one transaction with 3 other items and results in an enormous lift.

Thus, these association rules are **not valuable at all**.

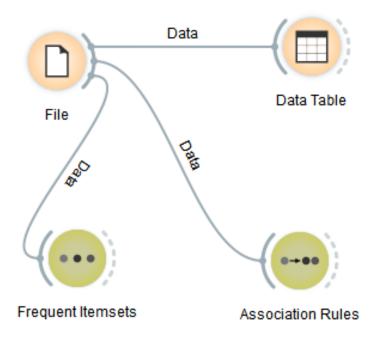
2 Question 2:

Load the Extended Bakery dataset (75000-out2-final.csv) into the Orange application, and run both frequent itemset as well as association rule modules. Set the support threshold to 1% and the confidence threshold to 90%. Observe the association rules containing the Cherry Tart item within the antecedent. What other item appears with it? When the confidence threshold is lowered to 45%, does the Cherry Tart item now appear without another item in the antecedent? Is the same consequent observed in both cases? How did lowering the confidence threshold lead to this change? Hint: Reference the Simpsons Paradox section of the text.

Answer:

• Creating appropriate workflow with the 75000-out2-final.csv dataset.

Figure 5: Workflow



 \bullet Figure 6. shows some of the market-basket dataset tabulated in Orange.

Figure 6: Extended Bakery Dataset



• First setting up the support threshold to 1% and confidence to 90% in both the Frequent Itemset generation as well as Association Rule generation, we get results as shown in Figure 7 and 8.

Figure 7: Frequent Itemsets in Extended Bakery%

*** Frequent Itemsets

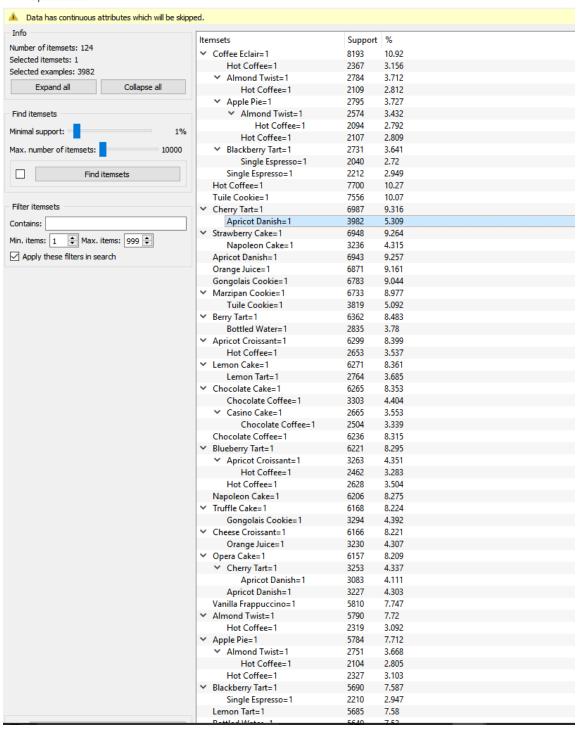
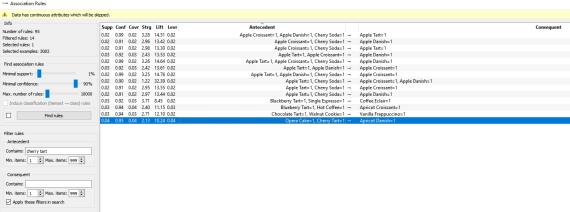


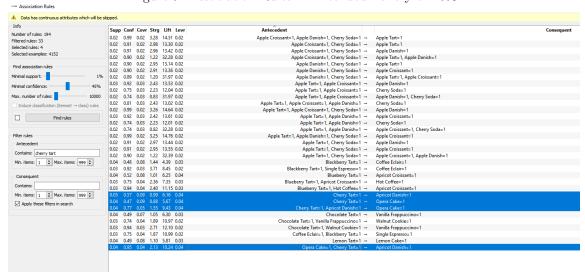
Figure 8: Association Rules in Extended Bakery c=90%



In Figure 8. the antecendents have been filtered by the "Cherry Tart" variable and highlighted.

• Now, we reduce the confidence level and observe the changes as shown in figure 9.

Figure 9: Association Rules in Extended Bakery c=45%



In Figure 10. the antecendents have been filtered by the "Cherry Tart" variable and highlighted.

2.1 Observations:

- For the 90% confidence threshold, the antecedent Cherry Tart appears with **Opera Cake** and the consequent is **Apricot Danish** and has a confidence of 95%.
- For the 45% confidence threshold, the association rules and their confidences are:

```
\{\{Opera\ Cake,\ Cherry\ Tart\} \rightarrow \{Apricot\ Danish\}\} = 0.95
\{\{Cherry\ Tart,\ Apricot\ Danish\} \rightarrow \{Opera\ Cake\}\} = 0.77
\{\{Cherry\ Tart\} \rightarrow \{Apricot\ Danish\}\} = 0.57
\{\{Cherry\ Tart\} \rightarrow \{Opera\ Cake\}\} = 0.47
```

- On lowering the Confidence Threshold, *Cherry Tart* appears without another item in the *antecedent*.
- The item Apricot Danish appears as consequent in both cases.

2.2 Inference:

Opera Cake becomes a consequent if the confidence threshold is reduced. Let's calculate the odds ratio from the contingency tables given below.

Table 1: Cherry Tart and Apricot Danish

	У	\overline{y}
X	3982	3005
\overline{x}	2961	65052

Table 2: Cherry Tart and Opera Cake

	У	\overline{y}
X	3253	3734
\overline{x}	2904	65109

Table 3: Opera Cake and Apricot Danish

	У	\overline{y}
X	3227	2930
\overline{x}	3716	65127

The odds ratio for the above contingency tables is 29, 19 and 19. This means that if we look for the odds of finding *Opera Cake* in transactions containing either *Apricot Danish* or *Cherry Tart*, the odds are the same. The probability of buying *Apricot Danish* when a customer purchased *Cherry Tart* is high when *Opera Cake* is included in the transaction. The confidence goes lower when its removed. Thus the item *Opera Cake* behaves as a **confounding variable** and skews the inference from the association leading to **Simpson's Paradox**.