TP3 : Génération de règles d'association

1 Fouille de données sous Weka

Question 1.1.

L'analyse du fichier weather.nominal.arff nous donne le résultat suivant :

```
Apriori
======
Minimum support: 0.15 (2 instances)
Minimum metric <confidence>: 0.9
Number of cycles performed: 17
Generated sets of large itemsets:
Size of set of large itemsets L(1): 12
Size of set of large itemsets L(2): 47
Size of set of large itemsets L(3): 39
Size of set of large itemsets L(4): 6
Best rules found:
1. outlook=overcast 4 ==> play=yes 4 conf:(1)
2. temperature=cool 4 ==> humidity=normal 4 conf:(1)
3. humidity=normal windy=FALSE 4 ==> play=yes 4 conf:(1)
4. outlook=sunny play=no 3 ==> humidity=high 3 conf:(1)
5. outlook=sunny humidity=high 3 ==> play=no 3 conf:(1)
6. outlook=rainy play=yes 3 ==> windy=FALSE 3
7. outlook=rainy windy=FALSE 3 ==> play=yes 3
8. temperature=cool play=yes 3 ==> humidity=normal 3 conf:(1)
9. outlook=sunny temperature=hot 2 ==> humidity=high 2 conf:(1)
10. temperature=hot play=no 2 ==> outlook=sunny 2 conf:(1)
```

Weka retrouve les règles d'association les plus pertinentes.

Question 1.2.

Apriori - confidence	Apriori - lift
Minimum support: 0.15 (2 instances) Minimum metric <confidence>: 0.9 Number of cycles performed: 17</confidence>	Minimum support: 0.3 (4 instances) Minimum metric <lift>: 1.1 Number of cycles performed: 14</lift>
Generated sets of large itemsets:	Generated sets of large itemsets:
Size of set of large itemsets L(1): 12	Size of set of large itemsets L(1): 12
Size of set of large itemsets L(2): 47	Size of set of large itemsets L(2): 9
Size of set of large itemsets L(3): 39	Size of set of large itemsets L(3): 1
Size of set of large itemsets L(4): 6	
Apriori - leverage	Apriori - conviction
Minimum support: 0.3 (4 instances) Minimum metric <leverage>: 0.1 Number of cycles performed: 14</leverage>	Minimum support: 0.25 (3 instances) Minimum metric <conviction>: 1.1 Number of cycles performed: 15</conviction>
Generated sets of large itemsets:	Generated sets of large itemsets:
Size of set of large itemsets L(1): 12	Size of set of large itemsets L(1): 12
Size of set of large itemsets L(2): 9	Size of set of large itemsets L(2): 26
Size of set of large itemsets L(3): 1	Size of set of large itemsets L(3): 4

Ces 4 mesures permettent d'affiner les règles d'association selon le critère choisi.

Question 1.3.

Pour la règle suivante : 1. outlook=overcast 4 ==> play=yes 4 nous obtenons les valeurs

- Confidence = $\frac{4}{4}$ = 1
- Lift = $\frac{4/4}{9/14} = \frac{14}{9}$
- Leverage = $\frac{4}{14} (\frac{4}{9}) \times (\frac{9}{14}) = 0.10$

Question 1.4.

Il s'agit du ratio entre la fréquence de non-apparition de la partie droite de la règle et la fréquence de non-apparition de la partie droite de la règle avec la partie gauche de la règle. Autrement dit, on mesure la dépendance pour les contre-exemples.

Questions 1.5. à 1.9.

Cf. document joint.

Lorsque l'on applique l'algorithme Apriori, on obtient les données suivantes :

```
=== Run information ===
            weka.associations.Apriori -N 10 -T 0 -C 0.9 -D 0.05 -U 1.0 -M 0.1 -S -1.0 -c -1
Scheme:
Relation:
           adult1-weka.filters.unsupervised.attribute.Discretize-B3-M-420.0-Rfirst-last
Instances:
Attributes: 15
        age
        workclass
        fnlwgt
         education
         education-num
        marital-status
         occupation
        relationship
        race
         capital-gain
         capital-loss
        hours-per-week
        native-country
         gain
=== Associator model (full training set) ===
Apriori
Minimum support: 0.75 (187 instances)
Minimum metric <confidence>: 0.9
Number of cycles performed: 5
Generated sets of large itemsets:
Size of set of large itemsets L(1): 6
Size of set of large itemsets L(2): 7
Size of set of large itemsets L(3): 1
Best rules found:
1. gain= <=50K 191 ==> capital-gain='(-inf-11365]' 190 conf:(0.99)
2. hours-per-week='(27.333333-53.666667]' 199 ==> capital-gain='(-inf-11365]' 195 conf:(0.98)
3. capital-loss='(-inf-735.333333]' 232 ==> capital-gain='(-inf-11365]' 226 conf:(0.97)
4. native-country= United-States 217 ==> capital-gain='(-inf-11365]' 211 conf:(0.97)
5. race= White 202 ==> capital-gain='(-inf-11365]' 196 conf:(0.97)
6. capital-loss='(-inf-735.333333]' native-country= United-States 202 ==> capital-gain='(-inf-11365]' 196 conf:
(0.97)
7. native-country= United-States 217 ==> capital-loss='(-inf-735.333333]' 202 conf:(0.93)
8. capital-gain='(-inf-11365]' native-country= United-States 211 ==> capital-loss='(-inf-735.333333]' 196 conf:
(0.93)
9. capital-gain='(-inf-11365]' 244 ==> capital-loss='(-inf-735.333333]' 226 conf:(0.93)
10. race= White 202 ==> capital-loss='(-inf-735.333333]' 187 conf:(0.93)
```

En activant l'option « car », on obtient les règles pour les contre-exemples.