



#### **CHANDIGARH UNIVERSITY**

# UNIVERSITY INSTITUTE OF ENGINEERING DEPARTMENT OF COMPUTER SCIENCE & ENGINEERING



Submitted By: Adarsh Gupta	Submitted To: Er. Navneet Kaur
Subject Name	Data Mining Lab
Subject Code	20CSP-376
Branch	CSE
Semester	6



# UNIVERSITY INSTITUTE OF ENGINEERING

## **Department of Computer Science & Engineering**

Subject Name: Data Mining

**Subject Code:** 20CSP-376

**Submitted to:** Submitted by:

Faculty name: Er. Navneet Kaur Name: Adarsh Gupta

**UID:** 20BCS4884

Section: 607

**Group:** B

Ex. No	List of Experiments	Date	Conduc t (MM: 12)	Viva (MM: 10)	Record (MM: 8)	Total (MM: 30)	Remarks/Signature
1.1	Demonstration of preprocessing on .arff file using student data .arff.	15-02-2023					
1.2	To perform statistical analysis of data	22-02-2023					
1.3	Demonstration of associate rule mining using apriori algorithm on supermarket data.	22-02-2023					
1.4							
2.1							
2.3							
2.4							
3.1							
3.2							
3.3							

# **Experiment 1.1**

Student Name: Adarsh Gupta UID: 20BCS4884

Branch: CSE Section/Group: 607-B

Semester: 6<sup>TH</sup> Date of Performance: 15/02/2023

**Subject Name:** Data Mining Lab **Subject Code:** 20CSP-376

#### 1. Aim/Overview of the practical:

Demonstration of preprocessing on .arff file using student data .arff.

#### 2. Objective:

- 1. To learn how to create an .arff file.
- 2. To understand the process of file creation in R.
- 3. To learn the utilization of RWeka.

#### **3. Code:**

```
library(RWeka)
setwd("D:\\DataMining")
getwd()
rating <- 1:4
animal <- c('koala','hedgedog','sloth','panda')
country <-c('Australia','Italy','Peru','China')
avg_sleep_hours <- c(21,18,17,10)

super_sleepers <-data.frame(rating,animal,country,avg_sleep_hours,
stringsAsFactors = FALSE)
print(super_sleepers)</pre>
```

```
print(class(super_sleepers))
print(str(super_sleepers))
write.arff(super_sleepers, file="super_sleepers.arff")
```

#### 4. Output:

```
13:55 (Top Level) $
 Console Terminal ×
                 Background Jobs X
 R 4.2.2 · D:/DataMining/
> super_sleepers <-data.frame(rating,animal,country,avg_sleep_hours, stringsAsFactors = FALSE)</pre>
> print(super_sleepers)
          animal country avg_sleep_hours
  rating
           koala Australia
       1
       2 hedgedog
                    Italy
                                         18
3
           sloth
                                         17
           panda
       4
                      China
                                         10
> print(class(super_sleepers))
[1] "data.frame"
> print(str(super_sleepers))
 'data.frame': 4 obs. of 4 variables:
                : int 1234
 $ rating
                 : chr "koala" "hedgedog" "sloth" "panda"
 $ animal
 $ country : chr "Australia" "Italy" "Peru" "China"
 $ avg_sleep_hours: num 21 18 17 10
NULL
> write.arff(super_sleepers, file="super_sleepers.arff")
```

# **Experiment 1.2**

Student Name: Adarsh Gupta UID: 20BCS4884

Branch: CSE Section/Group: 607-B

**Semester:** 6<sup>TH</sup> **Date of Performance:** 22/02/2023

**Subject Name:** Data Mining Lab **Subject Code:** 20CSP-376

#### 1. Aim/Overview of the practical:

To perform statistical analysis of data

#### 2. Objective:

- To perform statistical analysis of data using RWeka.
- To learn how to calculate mean, median and standard deviation in R.
- To learn the utilization of RWeka.

#### 3. Code:

```
library("RWeka")

N = read.arff("D:\\DataMining")

print(N)

cat("\n\n\n")

print(head(N,2))
```

```
dim(N)
names(N)
N["animal"]
N["avg_sleep_hours"]
max(N["avg_sleep_hours"])
min(avg_sleep_hours)
sum(avg_sleep_hours)
mean(avg_sleep_hours)
median(sort(avg_sleep_hours))
sd(avg_sleep_hours)
summary(N)
```

#### 4. Output:

```
Background Jobs ×
        Terminal ×
R 4.2.2 · D:/DataMining/
> N["avg_s reep_nours"]
Error: object 'N' not found
> max(N["avg_sleep_hours"])
Error: object 'N' not found
> min(avg_sleep_hours)
[1] 10
> sum(avg_sleep_hours)
[1] 66
> mean(avg_sleep_hours)
[1] 16.5
> median(sort(avg_sleep_hours))
[1] 17.5
> sd(avg_sleep_hours)
[1] 4.654747
> summary(N)
```

# **Experiment 1.3**

Student Name: Adarsh Gupta UID: 20BCS4884

Branch: CSE Section/Group: 607-B

**Semester:** 6<sup>TH</sup> **Date of Performance:** 22/02/2023

**Subject Name:** Data Mining Lab **Subject Code:** 20CSP-376

#### 1. Aim/Overview of the practical:

Demonstration of associate rule mining using Apriori algorithm on supermarket data

## 2. Objective:

- a. To learn how to import.
- b. To learn how to perform associate rule mining using Apriori algorithm.
- c. To learn the utilization of arules, arulesviz, RcolorBrewer.

#### 3. Code:

library(arules)
library(arulesViz)
library(RColorBrewer)

data("Groceries")

rules <- apriori(Groceries, parameter = list(supp = 0.01, conf = 0.2))

#### 4. Output:

```
Console Terminal × Background Jobs ×
                                                                                                 -
R 4.2.2 · D:/DataMining/
> rules <- apriori(Groceries, parameter = list(supp = 0.01, conf = 0.2))</pre>
Apriori
Parameter specification:
 confidence minval smax arem aval original Support maxtime support minlen maxlen target ext
              0.1
                     1 none FALSE
                                              TRUE
                                                         5
                                                              0.01
                                                                        1
                                                                              10 rules TRUE
Algorithmic control:
 filter tree heap memopt load sort verbose
    0.1 TRUE TRUE FALSE TRUE
Absolute minimum support count: 98
set item appearances ...[0 item(s)] done [0.00s].
set transactions ...[169 item(s), 9835 transaction(s)] done [0.00s].
sorting and recoding items ... [88 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 done [0.00s].
writing ... [232 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].
> inspect(rules[1:10])
     1hs
                        rhs
                                           support
                                                      confidence coverage
                                                                                      count
                     => {whole milk}
                                           0.25551601 0.2555160 1.00000000 1.000000 2513
[1]
     {}
[2]
     {hard cheese}
                    => {whole milk}
                                           0.01006609 0.4107884 0.02450432 1.607682
                    => {other vegetables} 0.01037112 0.3709091 0.02796136 1.916916
[3]
     {butter milk}
                                                                                       102
[4]
     {butter milk}
                    => {whole milk}
                                           0.01159126 0.4145455
                                                                 0.02796136 1.622385
                                                                                       114
[5]
     {ham}
                     => {whole milk}
                                           0.01148958 0.4414062
                                                                 0.02602949 1.727509
                                                                                       113
     {sliced cheese} => {whole milk}
                                           0.01077783 0.4398340 0.02450432 1.721356
[6]
                                           0.01128622 0.4021739
[7]
     {oil}
                    => {whole milk}
                                                                 0.02806304 1.573968
     {onions}
                    => {other vegetables} 0.01423488 0.4590164
[8]
                                                                 0.03101169 2.372268
                                                                                       140
[9]
     {onions}
                    => {whole milk}
                                           0.01209964 0.3901639
                                                                 0.03101169 1.526965
                                                                                       119
                  => {yogurt}
[10] {berries}
                                           0.01057448 0.3180428 0.03324860 2.279848
> arules::itemFrequencyPlot(Groceries, topN = 20,
                            col = brewer.pal(8, 'Pastel2'),
                            main = 'Relative Item Frequency Plot',
                            type = "relative",
                            ylab = "Item Frequency (Relative)")
```



#### **Graph output:**

