

Experiment NO.1

Dimension Product

```
create table product(  
    product_id int primary key,  
    product_name varchar(50),  
    product_category varchar(50),  
    unit_price int  
)  
  
insert into product values(1010, 'LG', 'TV',2000)  
insert into product values(1011, 'Sony', 'TV',35000)  
insert into product values(1012, 'Redmi', 'Mobile',20000)  
insert into product values(1013, 'Samsung', 'TV',50000)  
insert into product values(1014, 'Apple', 'Mobile',75000)  
insert into product values(1015, 'Nokia', 'Mobile',22000)  
select * from product
```

Dimension Employee

```
create table employee(  
    employee_id int primary key,  
    employee_name varchar(50),  
    title varchar(50),  
    department varchar(50),  
    region varchar(50)  
)  
  
insert into employee values(191041, 'rohit','rd','sales','kharghar')  
insert into employee values(191042, 'mohit','rd','sales','kharghar')  
insert into employee values(191043, 'shubham','rd','sales','kharghar')  
insert into employee values(191044, 'dipak','rd','sales','kharghar')  
insert into employee values(191045, 'aditi','rd','sales','kharghar')  
insert into employee values(191046, 'kaneshka','rd','sales','kharghar')  
select * from employee
```

Dimension Customer

```
create table customer(  
    customer_id int primary key,  
    customer_name varchar(50),  
    customer_address varchar(50),  
    city varchar(10),  
    zip int  
)  
  
insert into customer values(191056, 'rohit','kharghar','navimumbai',410210)  
insert into customer values(191057, 'mohit','kharghar','navimumbai',410210)  
insert into customer values(191058, 'shubham','kharghar','navimumbai',410210)  
insert into customer values(191059, 'kanishka','kharghar','navimumbai',410210)
```

```

insert into customer values(191060, 'dipak','kharghar','navimumbai',410210)
insert into customer values(191061, 'aditi','kharghar','navimumbai',410210)
insert into customer values(191062, 'rohan','kharghar','navimumbai',410210)
select * from customer

```

Dimension time

```

create table time(
    order_id int primary key,
    order_date int,
    year int,
    quater int,
    month int,
)
insert into time values(202101,25,2021,1,9)
insert into time values(202102,26,2021,1,10)
insert into time values(202103,24,2021,1,4)
insert into time values(202104,27,2021,1,7)
insert into time values(202105,29,2021,1,8)
insert into time values(202106,30,2021,1,10)
select * from time

```

Dimension Sales

```

create table sales(
    product_id int FOREIGN KEY REFERENCES product(product_id),
    order_id int FOREIGN KEY REFERENCES time(order_id),
    customer_id int FOREIGN KEY REFERENCES customer(customer_id),
    employee_id int FOREIGN KEY REFERENCES employee(employee_id),
    total int,
    quantity int,
    discount int
)
insert into sales values(1010,202101,191056,191041,20000,10,2000)
insert into sales values(1012,202102,191057,191042,20000,1,2000)
insert into sales values(1013,202103,191058,191043,50000,1,5000)
insert into sales values(1014,202104,191059,191044,75000,1,7500)
insert into sales values(1011,202105,191060,191045,35000,1,3500)
insert into sales values(1015,202106,191061,191046,22000,1,2200)
select * from sales

```

-- All

```

Select p.product_name,p.product_category,p.unit_price,t.order_id,c.customer_name,
e.employee_name,s.quantity,s.total,s.discount
from sales s
inner join product p on s.product_id=p.product_id

```

```

inner join time t on s.order_id=t.order_id
inner join employee e on s.employee_id=e.employee_id
inner join customer c on s.customer_id=c.customer_id

```

Experiment No.2

-- SLICE

```

Select p.product_name,p.product_category,p.unit_price,t.order_id,c.customer_name,
e.employee_name,s.quantity,s.total,s.discount
from sales s
inner join product p on s.product_id=p.product_id
inner join time t on s.order_id=t.order_id
inner join employee e on s.employee_id=e.employee_id
inner join customer c on s.customer_id=c.customer_id
where p.product_name = 'LG'

```

-- DICE

```

Select p.product_name,p.product_category,p.unit_price,t.order_id,c.customer_name,
e.employee_name,s.quantity,s.total,s.discount
from sales s
inner join product p on s.product_id=p.product_id
inner join time t on s.order_id=t.order_id
inner join employee e on s.employee_id=e.employee_id
inner join customer c on s.customer_id=c.customer_id
where p.product_name = 'LG' and quantity = 10

```

-- ROLLUP

```

Select product_name, SUM(s.total)
from sales s
inner join product p on s.product_id=p.product_id
inner join time t on s.order_id=t.order_id
inner join employee e on s.employee_id=e.employee_id
inner join customer c on s.customer_id=c.customer_id
where p.product_name = 'LG'
group by p.product_name

```

-- Drill Down

```

Select s.quantity, SUM(s.total) sum
from sales s
inner join product p on s.product_id=p.product_id
inner join time t on s.order_id=t.order_id
inner join employee e on s.employee_id=e.employee_id
inner join customer c on s.customer_id=c.customer_id
where p.product_name = 'LG'
group by s.quantity

```

Experiment No.3

BINNING Method-

```
import pandas as pd
import numpy as np
import math
unsort_data = [34, 8, 9, 15, 26, 24, 21, 25, 21, 28, 29, 4]
data = np.sort(unsort_data)
print(data)
bin1 = np.zeros((1, 4))
for i in range(0, 3, 4):
    k = int(1/4)
    mean = (data[i] + data[i+1]+data[i+2] + data[i+3])/4
    for j in range(4):
        bin1[k, j] = mean
print("\n---Mean Bin-1:---\n", bin1)

bin2 = np.zeros((1, 4))
for i in range(0, 3, 4):
    k = int(1/4)
    mean = (data[i+4]+data[i+5]+data[i+6] + data[i+7])/4
    for j in range(4):
        bin2[k, j] = mean
print("\n---Mean Bin-2:---\n", bin2)

bin3 = np.zeros((1, 4))
for i in range(0, 3, 4):
    k = int(1/4)
    mean = (data[i+8]+data[i+9]+data[i+10] + data[i+11])/4
    for j in range(4):
        bin3[k, j] = mean
print("\n---Mean Bin-3:---\n", bin3)

for i in range(0, 3, 4):
    k = int(i/4)
    for j in range(4):
        bin1[k, j] = data[i+2]
print("\n---Median Bin-1:---\n", bin1)

for i in range(0, 3, 4):
    k = int(1/4)
    for j in range(4):
        bin2[k, j] = data[i+6]
print("\n---Median Bin-2:---\n", bin2)

for i in range(0, 3, 4):
    k = int(1/4)
    for j in range(4):
        bin3[k, j] = data[i+10]
print("\n---Median Bin-3:---\n", bin3)

for i in range(0, 3, 4):
    k = int(1/4)
    for j in range(4):
```

```

        if (data[i+j]-data[i+1]) < (data[i+3]-data[i+j]):
            bin1[k, j] = data[i]
        else:
            bin1[k, j] = data[i+3]
print("\n----Boundary Bin-1----\n", bin1)

for i in range(0, 3, 4):
    k = int(i/4)
    for j in range(4):
        if (data[i+j]-data[i+1]) < (data[i+3]-data[i+j]):
            bin2[k, j] = data[i+4]
        else:
            bin2[k, j] = data[i+7]
print("\n----Boundary Bin-2----\n", bin2)

for i in range(0, 3, 4):
    k = int(1/4)
    for j in range(4):
        if (data[i+j]-data[i+1]) < (data[i+3]-data[i+j]):
            bin3[k, j] = data[i+8]
        else:
            bin3[k, j] = data[i+11]
print("\n----Boundary Bin-3:----\n", bin3)

```

OUTPUT:

[4 8 9 15 21 21 24 25 26 28 29 34]

----Mean Bin-1:----

[[9. 9. 9. 9.]]

----Mean Bin-2:----

[[22.75 22.75 22.75 22.75]]

----Mean Bin-3:----

[[29.25 29.25 29.25 29.25]]

----Median Bin-1:----

[[9. 9. 9. 9.]]

-----Median Bin-2-----

[[24. 24. 24. 24.]]

----Median Bin-3-----

[[29. 29. 29. 29.]]

----Boundary Bin-1----

[[4. 4. 4. 15.]]

----Boundary Bin-2-----

[[21. 21. 21. 25.]]

----Boundary Bin-3:----

[[26. 26. 26. 29.]]

Experiment No.4

```
# Naive Bayes Algorithm
import pandas as pd
import csv
def pci(data):
    class_count = [0, 0]
    for i in range(len(data)):
        if data.iloc[i, -1] == 'Yes':
            class_count[0] += 1
        else:
            class_count[1] += 1
    return class_count[0], class_count[1]
def pcix(data, x, c1, c2):
    p1, p2 = 1, 1
    count_yes = [0 for i in range(len(data.columns) - 1)]
    count_no = [0 for i in range(len(data.columns) - 1)]
    for i in range(len(data)):
        for j in range(len(data.columns) - 1):
            if data.iloc[i, j] == x[j]:
                if data.iloc[i, -1] == 'Yes':
                    count_yes[j] += 1
            else:
                count_no[j] += 1
    for i in range(len(count_yes)):
        p1 = p1 * count_yes[i] / c1
        p2 = p2 * count_no[i] / c2
    return p1, p2
data = pd.read_csv('NB.csv')
data = data.iloc[:, 1:]
X = input('Enter tuple to classify: ')
X = X.split(',')
c1, c2 = pci(data)
p1, p2 = pcix(data, X, c1, c2)
if (p1 * c1 / len(data)) > (p2 * c2 / len(data)):
    print('Buys')
else:
    print('No Buy')
```

OUTPUT:

Enter tuple to classify: Youth, Medium, Yes, Excellent

No buy

NB.csv file for naïve bayes

```
id,age,income,student,credit_rating,buys_computer
1,youth,low,no,fair,yes
2,youth,high,yes,excellent,no
3,youth,high,no,fair,yes
4,middle,medium,yes,fair,yes
5,senior,low,yes,fair,yes
6,senior,low,yes,excellent,yes
7,middle,high,yes,excellent,yes
8,youth,medium,no,fair,no
9,middle,low,no,excellent,yes
10,senior,medium,yes,fair,yes
```

Experiment No. 5

```
# importing libraries
import numpy as nm
import matplotlib.pyplot as mtp
import pandas as pd

# Importing the dataset
dataset = pd.read_csv('Mall_Customers_data.csv')

x = dataset.iloc[:, [3, 4]].values

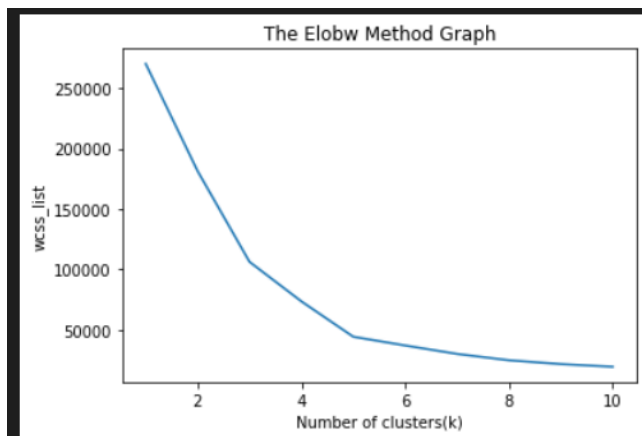
#finding optimal number of clusters using the elbow method
from sklearn.cluster import KMeans
wcss_list= [] #Initializing the list for the values of WCSS

#Using for loop for iterations from 1 to 10.
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state= 42)
    kmeans.fit(x)
    wcss_list.append(kmeans.inertia_)
mtp.plot(range(1, 11), wcss_list)
mtp.title('The Elbow Method Graph')
mtp.xlabel('Number of clusters(k)')
mtp.ylabel('wcss_list')
mtp.show()

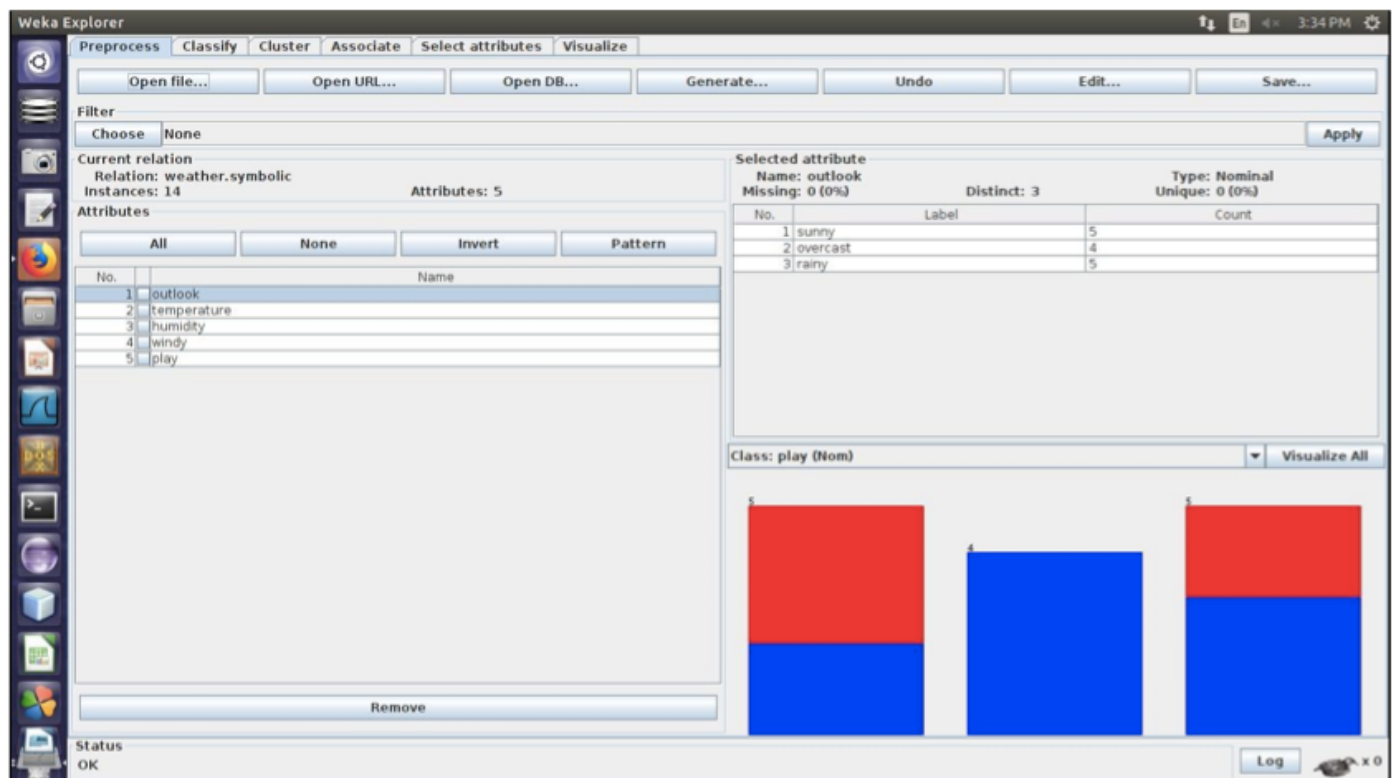
#training the K-means model on a dataset
kmeans = KMeans(n_clusters=5, init='k-means++', random_state= 42)
y_predict= kmeans.fit_predict(x)

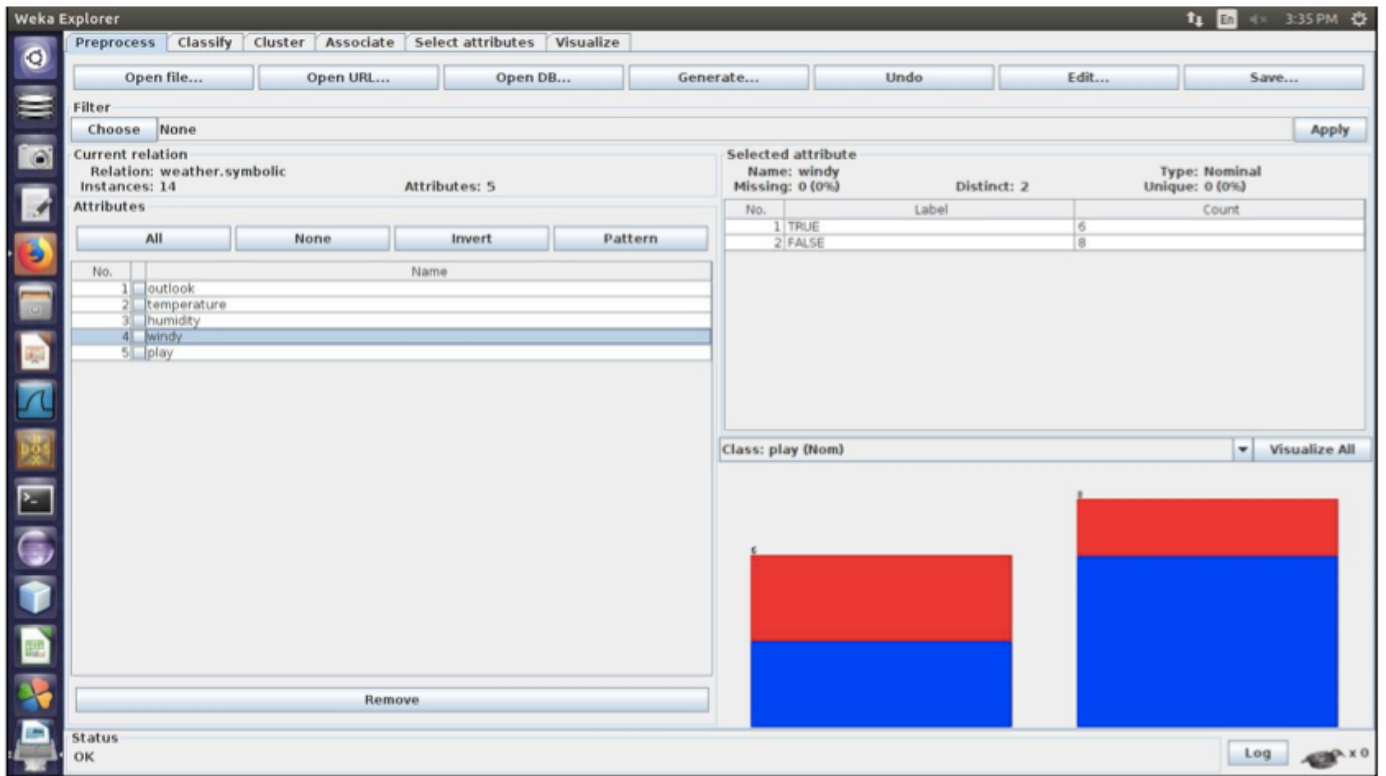
#visualizing the clusters
mtp.scatter(x[y_predict == 0, 0], x[y_predict == 0, 1], s = 100, c = 'blue', label = 'Cluster 1') #for first cluster
mtp.scatter(x[y_predict == 1, 0], x[y_predict == 1, 1], s = 100, c = 'green', label = 'Cluster 2') #for second cluster
mtp.scatter(x[y_predict== 2, 0], x[y_predict == 2, 1], s = 100, c = 'red', label = 'Cluster 3')
#for third cluster
mtp.scatter(x[y_predict == 3, 0], x[y_predict == 3, 1], s = 100, c = 'cyan', label = 'Cluster 4') #for fourth cluster
mtp.scatter(x[y_predict == 4, 0], x[y_predict == 4, 1], s = 100, c = 'magenta', label = 'Cluster 5') #for fifth cluster
mtp.scatter(kmeans.cluster_centers_[0], kmeans.cluster_centers_[1], s = 300, c = 'yellow', label = 'Centroid')
mtp.title('Clusters of customers')
mtp.xlabel('Annual Income (k$)')
mtp.ylabel('Spending Score (1-100)')
mtp.legend()
mtp.show()
```


OUTPUT:

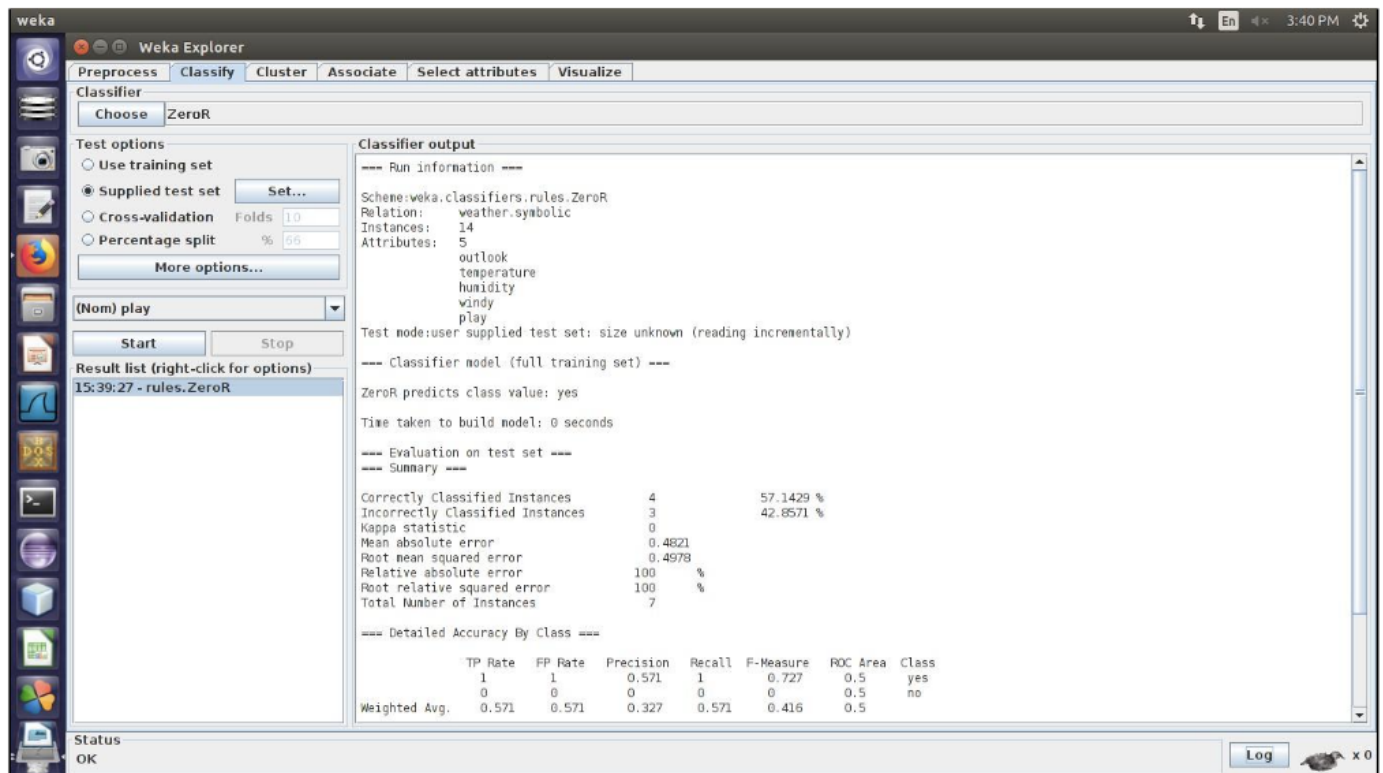


Experiment No.6:





Classify: Supplies test



Weka Explorer

Preprocess **Classify** Cluster Associate Select attributes Visualize

Classifier: Choose ZeroR

Test options:

- ☐ Use training set
- ☒ Supplied test set **Set...**
- ☐ Cross-validation Folds: 10
- ☐ Percentage split %: 66

More options...

(Nom) play

Start **Stop**

Result list (right-click for options):

- 15:39:27 - rules.ZeroR

Classifier output

--- Run information ---

Scheme: weka.classifiers.rules.ZeroR
 Relation: weather.symbolic
 Instances: 14
 Attributes: 5
 outlook
 temperature
 humidity
 windy
 play

Test mode: user supplied test set: size unknown (reading incrementally)

--- Classifier model (full training set) ---

ZeroR predicts class value: yes

Time taken to build model: 0 seconds

--- Evaluation on test set ---

--- Summary ---

Correctly Classified Instances	4	57.1429 %
Incorrectly Classified Instances	3	42.8571 %
Kappa statistic	0	
Mean absolute error	0.4621	
Root mean squared error	0.4978	
Relative absolute error	100 %	
Root relative squared error	100 %	
Total Number of Instances	7	

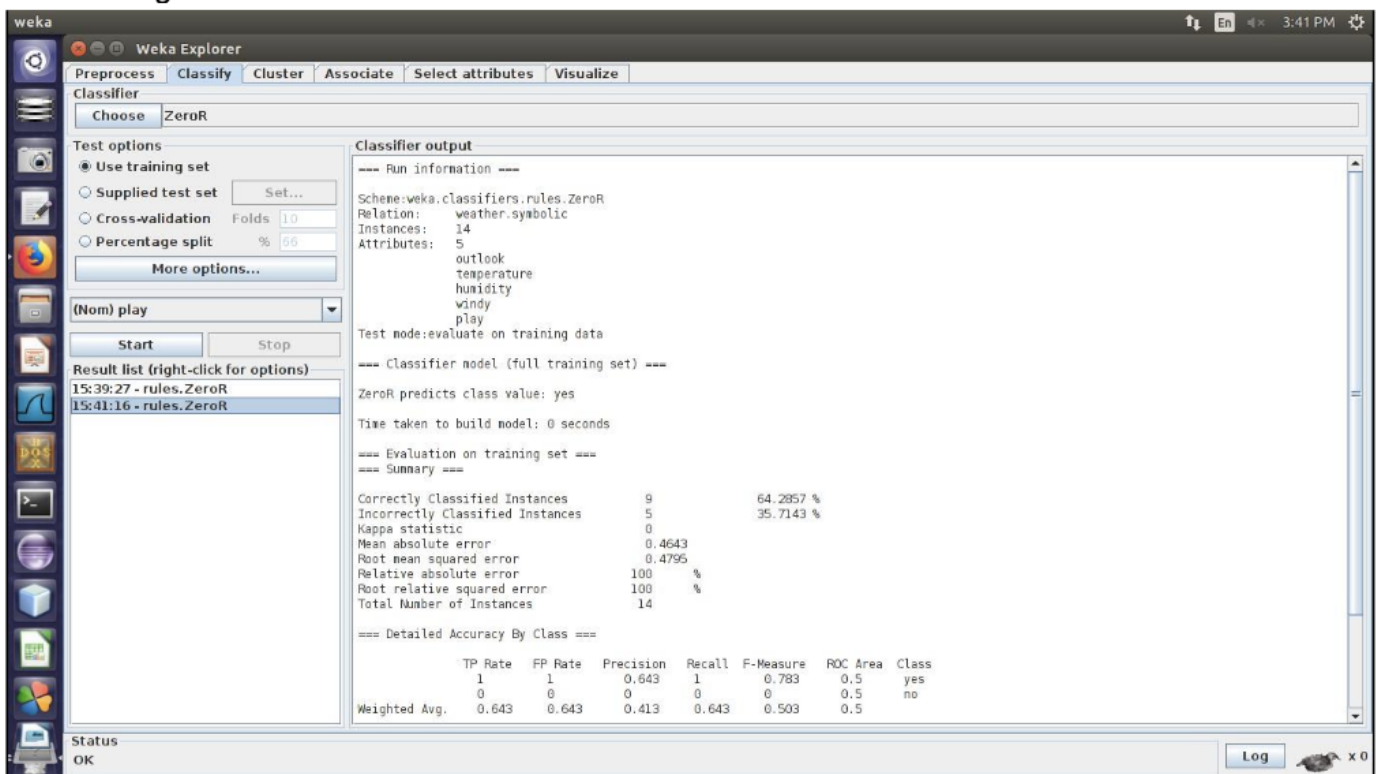
--- Detailed Accuracy By Class ---

	TP Rate	FP Rate	Precision	Recall	F-Measure	ROC Area	Class
1	1	1	0.571	1	0.727	0.5	yes
0	0	0	0	0	0	0.5	no
Weighted Avg.	0.571	0.571	0.327	0.571	0.416	0.5	

Status: OK

Log x 0

Use training set



Weka Explorer

Preprocess **Classify** Cluster Associate Select attributes Visualize

Classifier: Choose ZeroR

Test options:

- ☒ Use training set
- ☐ Supplied test set **Set...**
- ☐ Cross-validation Folds: 10
- ☐ Percentage split %: 66

More options...

(Nom) play

Start **Stop**

Result list (right-click for options):

- 15:39:27 - rules.ZeroR
- 15:41:16 - rules.ZeroR

Classifier output

--- Run information ---

Scheme: weka.classifiers.rules.ZeroR
 Relation: weather.symbolic
 Instances: 14
 Attributes: 5
 outlook
 temperature
 humidity
 windy
 play

Test mode: evaluate on training data

--- Classifier model (full training set) ---

ZeroR predicts class value: yes

Time taken to build model: 0 seconds

--- Evaluation on training set ---

--- Summary ---

Correctly Classified Instances	9	64.2857 %
Incorrectly Classified Instances	5	35.7143 %
Kappa statistic	0	
Mean absolute error	0.4643	
Root mean squared error	0.4795	
Relative absolute error	100 %	
Root relative squared error	100 %	
Total Number of Instances	14	

--- Detailed Accuracy By Class ---

	TP Rate	FP Rate	Precision	Recall	F-Measure	ROC Area	Class
1	1	1	0.643	1	0.783	0.5	yes
0	0	0	0	0	0	0.5	no
Weighted Avg.	0.643	0.643	0.413	0.643	0.503	0.5	

Status: OK

Log x 0

Cluster kmean: using numerical file

The screenshot shows the Weka Explorer interface with the 'Cluster' tab selected. The 'SimpleKMeans' algorithm is chosen with parameters: -N 2 -A "weka.core.EuclideanDistance" -R first-last -I 500 -S 10. The 'Cluster mode' section has 'Percentage split' selected at 69%, and 'Store clusters for visualization' is checked. The 'Cluster output' pane displays the following information:

Time taken to build model (full training data) : 0 seconds

=== Model and evaluation on test split ===

kMeans

Number of iterations: 2
Within cluster sum of squared errors: 7.9443946412391675
Missing values globally replaced with mean/mode

Cluster centroids:

Attribute	Full Data (9)	Cluster# 0 (3)	Cluster# 1 (6)
outlook	rainy	rainy	overcast
temperature	73.7778	72	74.6667
humidity	80.3333	83.6667	78.6667
windy	TRUE	TRUE	FALSE
play	yes	no	yes

Time taken to build model (percentage split) : 0 seconds

Clustered Instances

Cluster	Instances
0	1 (20%)
1	4 (80%)

Select Attributes

The screenshot shows the Weka Explorer interface with the 'Select attributes' tab selected. The 'CfsSubsetEval' evaluator is chosen with search method 'BestFirst -D 1 -N 5'. The 'Attribute selection output' pane displays the following information:

=== Run information ===

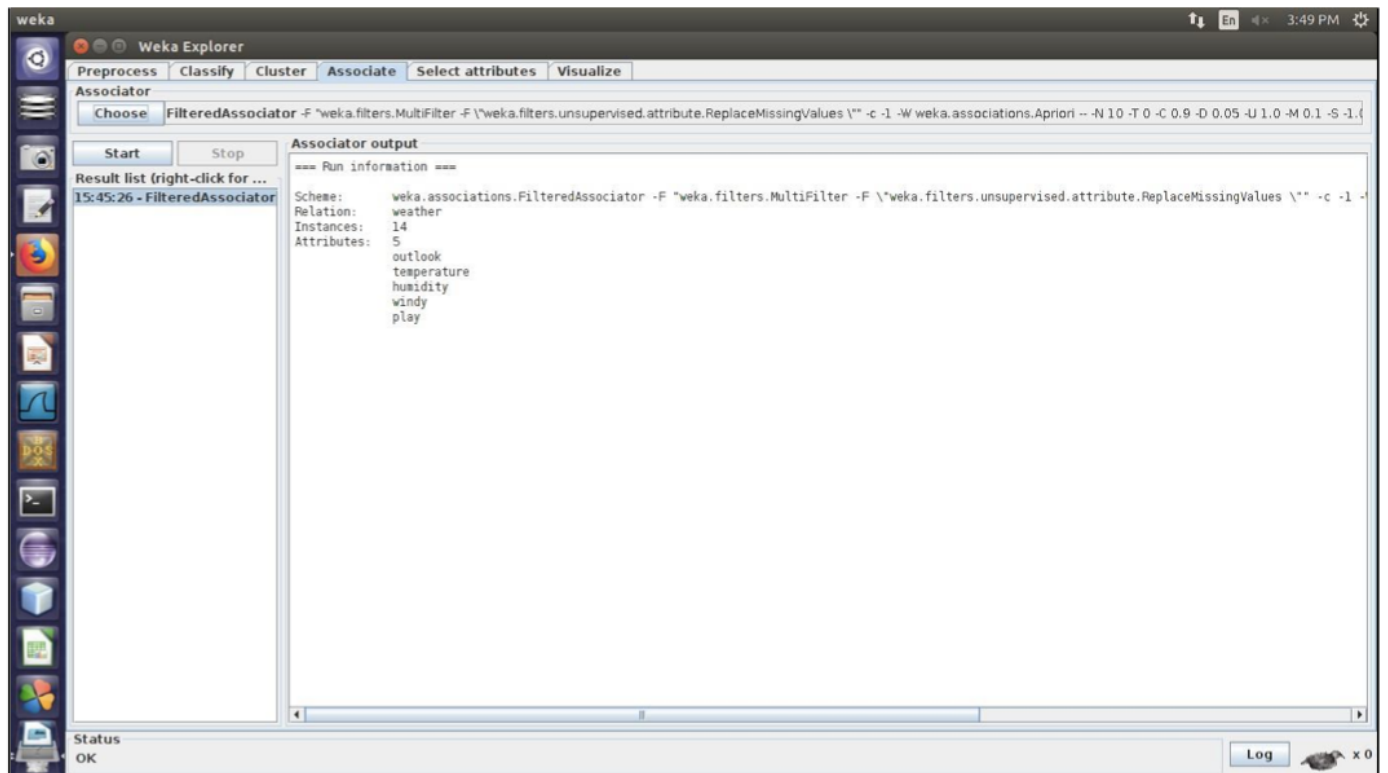
Evaluator: weka.attributeSelection.CfsSubsetEval
Search: weka.attributeSelection.BestFirst -D 1 -N 5
Relation: weather
Instances: 14
Attributes: 5
outlook
temperature
humidity
windy
play

Evaluation mode: 2-fold cross-validation

=== Attribute selection 2 fold cross-validation (stratified), seed: 1 ===

number of folds (%)	attribute
2(100 %)	1 outlook
0(0 %)	2 temperature
1(50 %)	3 humidity
1(50 %)	4 windy

Associate:



Experiment No.7:

```
import numpy as np
X = np.array([[8,4],[10,15],[16,12],[22,12],[30,30],[85,70],[74,80],[60,76],[65,50],[85,92]])
import matplotlib.pyplot as plt
labels = range(1,11)
plt.figure(figsize=(10,7))
plt.subplots_adjust(bottom=0.1)
plt.scatter(X[:,0],X[:,1],label='True Position')
for label,x,y in zip(labels,X[:,0],X[:,1]):
    plt.annotate(label,xy=(x,y),xytext=(-3,3),textcoords='offset points',ha='right',va='bottom')
plt.show()
```

Single Linkage

```
from scipy.cluster.hierarchy import dendrogram,linkage
from matplotlib import pyplot as plt
linked = linkage(X,'single')
labellist = range(1,11)
plt.figure(figsize=(10,7))
dendrogram(linked,orientation='top',labels=labellist,
distance_sort='descending',show_leaf_counts=True)
plt.show()
```

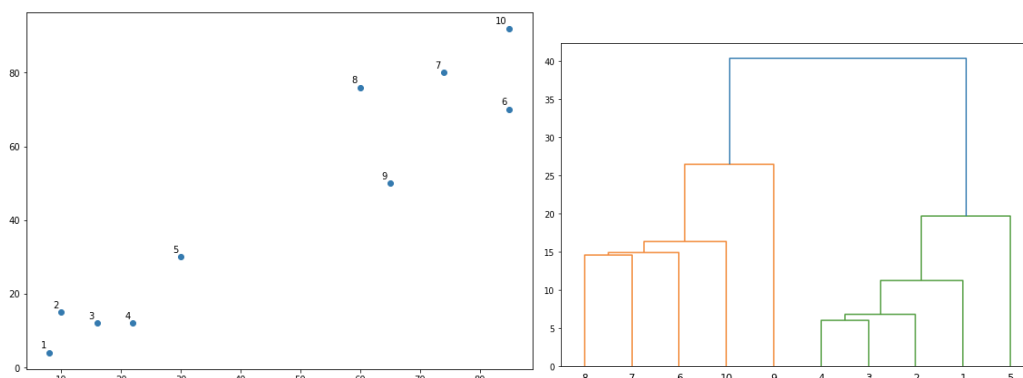
#Complete Linkage

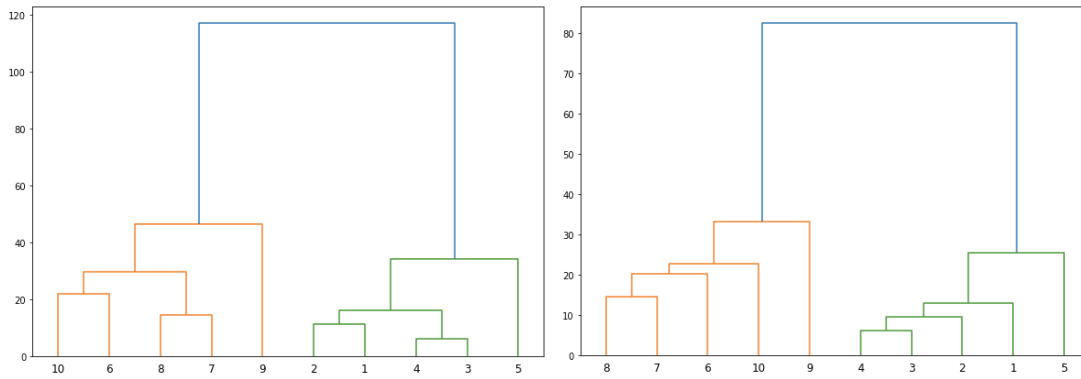
```
from scipy.cluster.hierarchy import dendrogram,linkage
from matplotlib import pyplot as plt
linked = linkage(X,'complete')
labellist = range(1,11)
plt.figure(figsize=(10,7))
dendrogram(linked,orientation='top',labels=labellist,
distance_sort='descending',show_leaf_counts=True)
plt.show()
```

#Average Linkage

```
from scipy.cluster.hierarchy import dendrogram,linkage
from matplotlib import pyplot as plt
linked = linkage(X,'average')
labellist = range(1,11)
plt.figure(figsize=(10,7))
dendrogram(linked,orientation='top',labels=labellist,
distance_sort='descending',show_leaf_counts=True)
plt.show()
```

OUTPUT:





Experiment No.8

```
data = [
    ['T100',['I1','I2','I5']],
    ['T200',['I2','I4']],
    ['T300',['I2','I3']],
    ['T400',['I1','I2','I4']],
    ['T500',['I1','I3']],
    ['T600',['I2','I3']],
    ['T700',['I1','I3']],
    ['T800',['I1','I2','I3','I5']],
    ['T900',['I1','I2','I3']]
]
```

```
init = []
for i in data:
    for q in i[1]:
        if(q not in init):
            init.append(q)
init = sorted(init)
print(init)
sp = 0.4
s = int(sp*len(init))
s
from collections import Counter
c = Counter()
for i in init:
    for d in data:
        if(i in d[1]):
            c[i]+=1
print("C1:")
for i in c:
    print(str([i])+"": "+str(c[i]))
print()
l = Counter()
for i in c:
    if(c[i] >= s):
        l[frozenset([i])]+=c[i]
print("L1:")
for i in l:
    print(str(list(i))+"": "+str(l[i]))
print()
pl = l
```

```

pos = 1
for count in range (2,1000):
    nc = set()
    temp = list(l)
    for i in range(0,len(temp)):
        for j in range(i+1,len(temp)):
            t = temp[i].union(temp[j])
            if(len(t) == count):
                nc.add(temp[i].union(temp[j]))
    nc = list(nc)
    c = Counter()
    for i in nc:
        c[i] = 0
        for q in data:
            temp = set(q[1])
            if(i.issubset(temp)):
                c[i]+=1
    print("C"+str(count)+":")
    for i in c:
        print(str(list(i))+": "+str(c[i]))
    print()
    l = Counter()
    for i in c:
        if(c[i] >= s):
            l[i]+=c[i]
    print("L"+str(count)+":")
    for i in l:
        print(str(list(i))+": "+str(l[i]))
    print()
    if(len(l) == 0):
        break
    pl = l
    pos = count
print("Result: ")
print("L"+str(pos)+":")
for i in pl:
    print(str(list(i))+": "+str(pl[i]))
print()

```

OUTPUT:

```
['I1', 'I2', 'I3', 'I4', 'I5']
```

C1:

```
['I1']: 6
```

```
['I2']: 7
```

```
['I3']: 6
```

```
['I4']: 2
```

```
['I5']: 2
```

L1:

```
['I1']: 6
```

```
['I2']: 7
```

```
['I3']: 6
```

```
['I4']: 2
```

```
['I5']: 2
```


C2:

```
['I1', 'I3']: 4
['I2', 'I4']: 2
['I4', 'I5']: 0
['I1', 'I4']: 1
['I3', 'I5']: 1
['I1', 'I5']: 2
['I1', 'I2']: 4
['I3', 'I4']: 0
['I2', 'I3']: 4
['I2', 'I5']: 2
```

L2:

```
['I1', 'I3']: 4
['I2', 'I4']: 2
['I1', 'I5']: 2
['I1', 'I2']: 4
['I2', 'I3']: 4
['I2', 'I5']: 2
```

C3:

```
['I2', 'I4', 'I5']: 0
['I1', 'I3', 'I2']: 2
['I1', 'I3', 'I5']: 1
['I2', 'I3', 'I4']: 0
['I1', 'I2', 'I5']: 2
['I2', 'I3', 'I5']: 1
['I2', 'I1', 'I4']: 1
```

L3:

```
['I1', 'I3', 'I2']: 2
['I1', 'I2', 'I5']: 2
```

C4:

```
['I5', 'I1', 'I3', 'I2']: 1
```

L4:

Result:

L3:

```
['I1', 'I3', 'I2']: 2
['I1', 'I2', 'I5']: 2
```

Experiment No.9

```
import networkx as nx
import numpy as np
from numpy import array
import matplotlib.pyplot as plt

with open('HITS.txt') as f:
    lines = f.readlines()

G = nx.DiGraph()
for line in lines:
    t = tuple(line.strip().split(','))
    G.add_edge(*t)

h, a = nx.hits(G, max_iter=100)
h = dict(sorted(h.items(), key=lambda x: x[0]))
a = dict(sorted(a.items(), key=lambda x: x[0]))

print(np.round(list(a.values()), 3))
print(np.round(list(h.values()), 3))

pr = nx.pagerank(G)
pr = dict(sorted(pr.items(), key=lambda x: x[0]))
print(np.round(list(pr.values()), 3))

sim = nx.simrank_similarity(G)
lol = [[sim[u][v] for v in sorted(sim[u])] for u in sorted(sim)]
sim_array = np.round(array(lol), 3)
print(sim_array)

nx.draw(G, with_labels=True, node_size=2000, edge_color='#eb4034', width=3, font_size=16,
font_weight=500, arrowsize=20, alpha=0.8)
plt.savefig("graph.png")
```

OUTPUT:

```
[0.088 0.187 0.369 0.128 0.059 0.11 0. 0.059]
[0.043 0.144 0.03 0.187 0.268 0.144 0.154 0.03 ]
[0.241 0.137 0.218 0.24 0.077 0.035 0.019 0.034]
[[1. 0.207 0.221 0.193 0.217 0.269 0. 0.171]
 [0.207 1. 0.355 0.369 0.302 0.553 0. 0.369]
 [0.221 0.355 1. 0.242 0.4 0.324 0. 0.427]
 [0.193 0.369 0.242 1. 0.229 0.548 0. 0.243]
 [0.217 0.302 0.4 0.229 1. 0.271 0. 0.498]
 [0.269 0.553 0.324 0.548 0.271 1. 0. 0.244]
 [0. 0. 0. 0. 0. 0. 1. 0. ]
 [0.171 0.369 0.427 0.243 0.498 0.244 0. 1. ]]
```

//HITS.txt file

```
1,4
2,3
2,5
3,1
```

	4,2
	4,3
	5,3
	5,2
	5,4
	5,6
	6,3
	6,8
	7,1
	7,3
	8,1

Experiment N0.10

```
import networkx as nx
import matplotlib.pyplot as plt
G = nx.DiGraph()
G.add_edges_from([('A', 'D'), ('B', 'C'), ('B', 'E'), ('C', 'A'),
('D', 'C'), ('E', 'D'), ('E', 'B'), ('E', 'F'),
('E', 'C'), ('F', 'C'), ('F', 'H'), ('G', 'A'), ('G', 'C'), ('H', 'A')])
plt.figure(figsize=(10, 10))
nx.draw_networkx(G, with_labels=True)
hubs, authorities = nx.hits(G, max_iter=50, normalized=True)
print("Hub Scores:", hubs)
print("Authority Scores:", authorities)
```

OUTPUT:

```
Hub Scores: {'A': 0.04642540403219994, 'D': 0.13366037526115382, 'B': 0.15763599442967322, 'C':
0.03738913224642654, 'E': 0.25881445984686646, 'F': 0.1576359944296732, 'H':
0.03738913224642654, 'G': 0.17104950750758036} Authority Scores: {'A': 0.10864044011724346, 'D':
0.13489685434358, 'B': 0.11437974073336447, 'C': 0.3883728003876181, 'E': 0.06966521184241478,
'F': 0.11437974073336447, 'H': 0.06966521184241475, 'G': 0.0}
```

