

## Experiment - 1a

Aim: Write a program for transformation numeric data using Z-score normalization.

Description:

Normalization: where the data values are scaled within the specific range the techniques are:

- 1) Min-Max normalization
- 2) Z-score normalization
- 3) Decimal scaling

Min-Max Normalization:

It performs a linear transformation, on the original data. It maps a value 'v' to v' as the value.

$$v' = \frac{v - \min_A}{\max_A - \min_A} (\text{new\_max}_A - \text{new\_min}_A) + \text{new\_min}_A$$

Z-score Normalization: The values of attributes of X' are normalized based on the mean and standard deviation of attribute X'.

$$v' = \frac{v - \bar{A}}{\sigma_A}$$

Decimal Scaling: It normalizes the attributes by moving the decimal point of values.

$$v' = \frac{v}{10^j}$$



Program:

```
import statistics as st
import math
array = []
temp = []
new_max = 1.0
new_min = 0.0
array = list(map(int, input("enter value into array").split()))
val = int(input("\n enter the value"))
print ("In primitive Normalization")
      z-score
new_val = (val - st.mean(array)) / st.stdev(array)
print (new_val)
```

O/P: enter value into array -123 256 -4548 -369 745  
enter the value 47  
z-score normalization  
-0.122.



## Experiment - 1b

Aim: write a program for transformation numeric data using Z-score normalization.

Description:

Normalization: Where the data values are scaled within the specific range the techniques are:

- 1) Min-Max normalization.
- 2) Z-score normalization
- 3) Decimal scaling.

Min Max - Normalization:

It performs a linear transformation on the original data it maps a value  $v$  to  $v'$  as the value.

$$v' = \frac{v - \min_A}{\max_A - \min_A} (\text{new\_max}_A - \text{new\_min}_A) + \text{new\_min}_A$$

Z-score Normalization: The values of attributes of  $x'$  are normalized based on the mean and standard deviation of attribute  $x'$

$$v' = \frac{v - \bar{A}}{\sigma_A}$$

Decimal Scaling :- It normalized the attributes by moving the decimal point of values

$$v' = \frac{v}{10^j}$$



Program:

```

import statistics as st
import math

array = []
temp = []
new-max = 1.0
new-min = 0.0
array = list(map(int, input("enter value into array").split()))
val = int(input("\n enter the value"))
print("\n Min Max Normalization")
new-val = ((val - min(array)) / (max(array) - min(array))) *
            (new-max - new-min) + new-min

print(new-val)
for i in array:
    temp.append(len(str(i)))
j = max(temp)
print("\n Decimal scaling")
new-val = val * (math.pow(10, j))

print(new-val)

```

O/P: ~~python~~ (7/7/20)

enter value into array -123 256 -4548 -369 745

enter the value 47

Min max Normalization

0.373



## Experiment-2a.

Aim: write a program for smoothing numeric data with data - bin sizes 2, 3, 5.

Description: Smoothing refers for removing noise data from data set. This can be done in following methods

1) Binning

- Binning by Mean
- Binning by Median
- Binning by most probable value
- Equal frequency binning.

2) Clustering

3) Combined human and Computer Inspection.

4) Regression

Binning by Mean:

Step 1: Take the values to bin

Step 2: Take the sizes of each bin

Step 3: Distribute the values into each bin according to its Capacity.

Step 4: Replace each of the value in the bin with the 'mean' of the values of that bin.

Program: import statistic as st

```
print ("Enter the values to bin")
```

```
a = list(map(int, input().split())) - bins = []
```

```
print ("In enter number of bins")  
n = int(input())
```

```
for i in range (n):
```



```

print("\n Enter Bin "+i+" size ")
size = int(input())
bins.append(k:k+size)
k += size
for i in bins:
    bin_mean = st.mean(i)
    for j in i:
        i[j] = bin_mean
    print("\n Print Bin "+i+" mean is")
    print(bin_mean)
    print("\n after binning elements of bin "+i+" are")
    for j in i:
        print(i[j], end=" ")

```

O/p:-

Enter the values to bin

3 2 2 4 3 3 1 0 0

Enter number of bins 3

Enter bin 0 size 2

Enter bin 1 size 3

Enter bin 2 size 5

Bin : 1

3 2

Bin : 2

2 4 3



### Experiment - 3

**Aim:** To implement the concept of data discretization

**Description:** Data discretization converts a large number of data value into similar one. so that data evaluation and data management becomes very closely.

It can be used to divide the range of continuous attribute into intervals. Numerous continuous attribute values are replaced by small interval labels. This leads to a concise, easy to use, knowledge level representation of mining results.

If the process starts by first finding once or a few points to split the entire attribute range and then repeats this recursively on resulting intervals then it is called top-down discretization or splitting.

If the process starts by considered all of the value continuous values as potential split-points, removes some by merging neighbourhood values to form intervals then it is called bottom-up discretization or merging.

Program:

```
import java.util.*;  
  
public class Discretization  
{  
    public static void main (String [] args)  
    {  
        int n, count=0, count1=0, count2=0, count3=0,  
            count4=0;
```



```
Float a[] = new Float[100];
String b[] = new String[100];
Scanner s = new Scanner(System.in);
System.out.println("enter n:");
n = s.nextInt();
System.out.println("enter elements:");
for (int i = 0; i < n; i++)
{
    a[i] = s.nextFloat();
}
for (int i = 0; i < n; i++)
{
    if (a[i] >= 70) {
        b[i] = "D";
        count1++;
    }
    else if (a[i] >= 60) {
        b[i] = "C";
        count1++;
    }
    else if (a[i] >= 50) {
        b[i] = "SC";
        count2++;
    }
    else if (a[i] >= 40) {
        b[i] = "tc";
        count3++;
    }
    else {
        b[i] = "f";
        count4++;
    }
}
```



```

    }
    System.out.println ( );
    System.out.println ("Marks and their status :");
    for (int i=0; i<n; i++)
    {
        System.out.println ("Marks and their status:");
        for (int i=0; i<n; i++)
        {
            System.out.println (a[i] + " " + b[i]);
        }
        System.out.println ("No. of student under distinction
            class are: " + count);
        System.out.println ("No. of student under first class
            are: " + count1);
        System.out.println ("No. of students under second class
            are: " + count2);
        System.out.println ("No. of students under third class
            are: " + count3);
        System.out.println ("No. of students under fail
            are: " + count4);
    }
}

```

O/p:

Enter n: 10

Enter elements:

90.0 85.1 76.4 27.8 38.5 67.8 88.7 49.8 71.9 30.8

Marks and their status

90.0	D
85.1	D
76.1	D



29.8	f
38.5	f
67.8	fc
88.9	D
49.8	tc
71.9	D
30.8	f

No. of Students under distinction class are : 5

No. of Students under first class are : 1

No. of Students under second class are : 0

No. of Student under third class are : 1

No. of Student under fail are : 3.



### Experiment-5

Aim: To create an arff file with all types of attributes and explore the steps using weka for Data load.

Procedure: How to create arff file:

1. first we define the name of a relation
2. list all the attributes with their type specified.
3. Enter values for all attributes of dataset. If missing values are there, specify it as '?'
4. Save the file as filename.arff.

Student.arff

@ relation student

@ attribute regdno numeric

@ attribute name string

@ attribute branch {cse, it, ece}

@ data

1, Anjana, cse

2, Puja, It

3, Devi, cse

4, Geetha, ece

Steps to load arff file :-

1. Open explorer in weka interface.
2. preprocessor is the default tab of the explorer.
3. select open file and search for required arff file.



4. Load the dataset

5. Screenshot the graph.

The attached screenshot represents the relation of the StudentId which is maximum, minimum and standard deviation.