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LDRP INSTITUTE OF TECHNOLOGY AND
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GANDHINAGAR



**Department of
Information Technology**

Subject: Soft Computing (CT605A-N)

Laboratory Manual

Prepared By

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19BEIT30033

6IT

**LDRP INSTITUTE OF TECHNOLOGY AND
RESEARCH
GANDHINAGAR**

**DEPARTMENT OF COMPUTER ENGINEERING
&
INFORMATION TECHNOLOGY**



CERTIFICATE

Mr. _____ of 6th IT Enrollment
No. _____ Exam No, _____ has
satisfactorily completed his term work in ***Soft Computing***
(CT605A-N) for the term ending in **Dec-May-2022**.

Date: _____

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HOD-IT

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1	Create a perceptron with appropriate no. of inputs and outputs. Train it using fixed increment learning algorithm until no change in weights is required. Output the final weights.		21-01-21	
2	Create a simple ADALINE network with appropriate no. of input and output nodes. Train it using delta learning rule until no change in weights is required. Output the final weights.			
3	Train the autocorrelator by given patterns: A1=(-1,1,-1,1), A2=(1,1,1,-1), A3=(-1, -1, - 1, 1). Test it using patterns: Ax=(-1,1,-1,1), Ay=(1,1,1,1), Az=(-1,-1,-1,-1).			
4	Train the hetrocorrelator using multiple training encoding strategy for given patterns: A1=(000111001) B1=(010000111), A2=(111001110) B2=(100000001), A3=(110110101) B3(101001010). Test it using pattern A2.			
5	Implement Union, Intersection, Complement and Difference operations on fuzzy sets. Also create fuzzy relation by Cartesian product of any two fuzzy sets and perform maxmin composition on any two fuzzy relations.			
6	Solve Greg Viot's fuzzy cruise controller using MATLAB Fuzzy logic toolbox.			
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PRACTICAL-1

AIM: Create a perceptron with appropriate no. of inputs and outputs. Train it using fixed increment learning algorithm until no change in weights is required. Output the final weights.

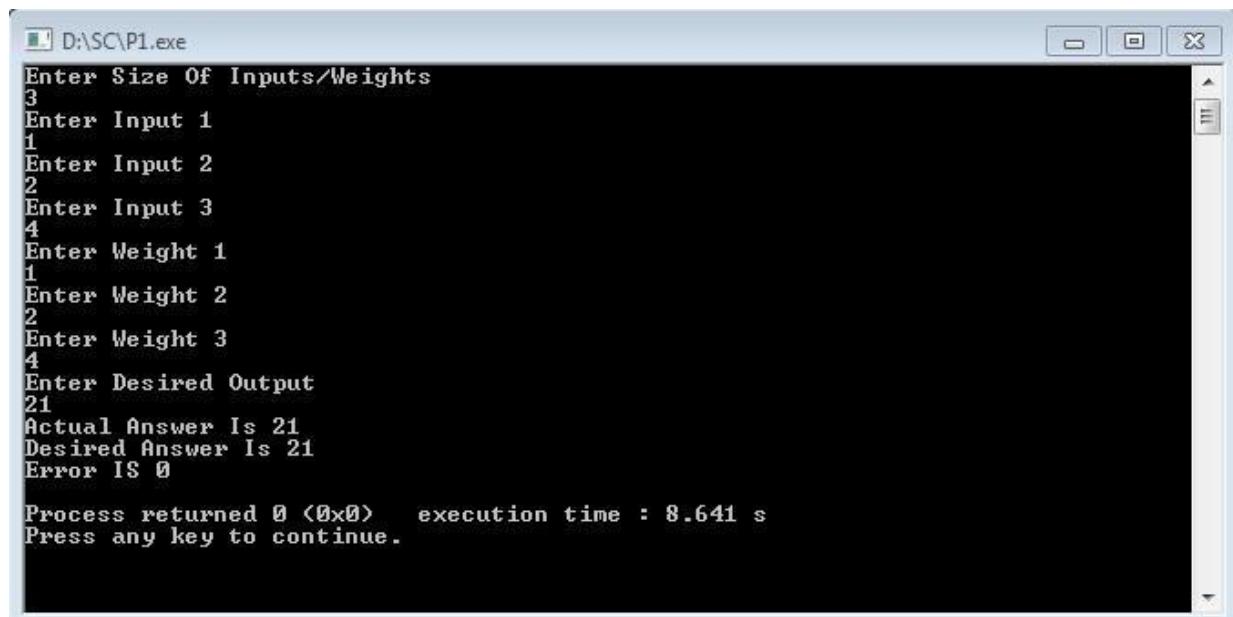
CODE:

1. C++

```
#include<iostream>
using namespace std;
int main()
{
    int n,i,j;
    cout<<"Enter Size Of Inputs/Weights "<<endl;
    cin>>n;
    int a[n],w[n],ans=0,desired,t,e,token=2;
    for (i=0;i<n;i++)
    {
        cout<<"Enter Input "<<i+1<<" "<<endl;
        cin>>a[i];
    }
    for (i=0;i<n;i++)
    {
        cout<<"Enter Weight "<<i+1<<" "<<endl;
        cin>>w[i];
    }
    cout<<"Enter Desired Output "<<endl;;
    cin>>desired;
    while(token!=1 && e!=0)
    {
        ans=0;
        for (i=0;i<n;i++)
        {
            ans=ans+a[i]*w[i];
        }
        cout<<"Actual Answer Is "<<ans<<endl;
        cout<<"Desired Answer Is "<<desired<<endl;
        e=desired-ans;
        cout<<"Error IS "<<e<<endl;
        if(e==0)
        {
            break;
        }
        cout<<"Enter 1 To Terminate"<<endl;;
        cin>>token;
        if(e<0)
        {
            for (i=0;i<n;i++)
            {
                w[i]=w[i]-1;
            }
        }
        else
        {
```

```
for (i=0;i<n;i++)
{
    w[i]=w[i]+1;
}
}
return 0;
}
```

OUTPUT:



```
D:\SC\P1.exe
Enter Size Of Inputs/Weights
3
Enter Input 1
1
Enter Input 2
2
Enter Input 3
4
Enter Weight 1
1
Enter Weight 2
2
Enter Weight 3
4
Enter Desired Output
21
Actual Answer Is 21
Desired Answer Is 21
Error IS 0

Process returned 0 (0x0)  execution time : 8.641 s
Press any key to continue.
```

PRACTICAL-2

AIM: Create a simple ADALINE network with appropriate no. of input and output nodes. Train it using delta learning rule until no change in weights is required. Output the final weights

CODE:

```

import numpy as np
inputs = np.array([-1, -1], [-1, 1], [1, -1], [1, 1])
outputs = np.array([-1, 1, 1, 1])

print(inputs, outputs)
weight=[1,1]bias=0 errors=[]
learning_rate=0.2epochs=10

for i in range(epochs): sum_squared_error = 0
for j in range(inputs.shape[0]):
    actual = outputs[j]
    x1 = inputs[j][0]
    x2 = inputs[j][1]
    unit = (x1 * weight[0]) + (x2 * weight[1]) + bias
    error = actual - unit
    #print("error = ", error)
    sum_squared_error += error * error
    weight[0] += learning_rate * error * x1
    weight[1] += learning_rate * error * x2
    bias += learning_rate * error

#print("sum of squared error = ", sum_squared_error/4, "\n\n")
print(" Weights are : ", weight)
print(" Bias is : ", bias)

print(" Sum Squared Error is : ", sum_squared_error/4)

#AND Gate

outputs = np.array([-1, -1, -1, 1])

weight=[1,1]bias=0 errors=[]
learning_rate=0.2epochs=10

for i in range(epochs): sum_squared_error = 0
for j in range(inputs.shape[0]):
    actual = outputs[j]
    x1 = inputs[j][0]
    x2 = inputs[j][1]
    unit = (x1 * weight[0]) + (x2 * weight[1]) + bias
    error = actual - unit
    #print("error = ", error)
    sum_squared_error += error * error
    weight[0] += learning_rate * error * x1
    weight[1] += learning_rate * error * x2
    bias += learning_rate * error

#print("sum of squared error = ", sum_squared_error/4, "\n\n")
print(" Weights are : ", weight)
print(" Bias is : ", bias)

print(" Sum Squared Error is : ", sum_squared_error/4)

```

OUTPUT:

```
[[[-1 -1]
 [-1  1]
 [ 1 -1]
 [ 1  1]] [-1  1  1  1]
Weights are : [0.42857118589015875, 0.3571404406252653]
Bias is : 0.500006995435774
Sum Squared Error is : 2.0408275145400836
```

AND GATE

```
Weights are : [0.5714319767623846, 0.6428611651212666]
Bias is : -0.5000053366683359
Sum Squared Error is : 0.5102118094738022
```

PRACTICAL-3

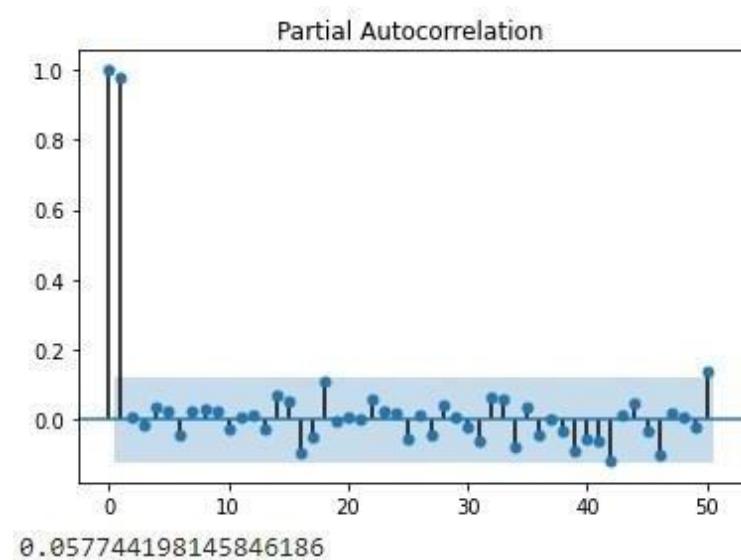
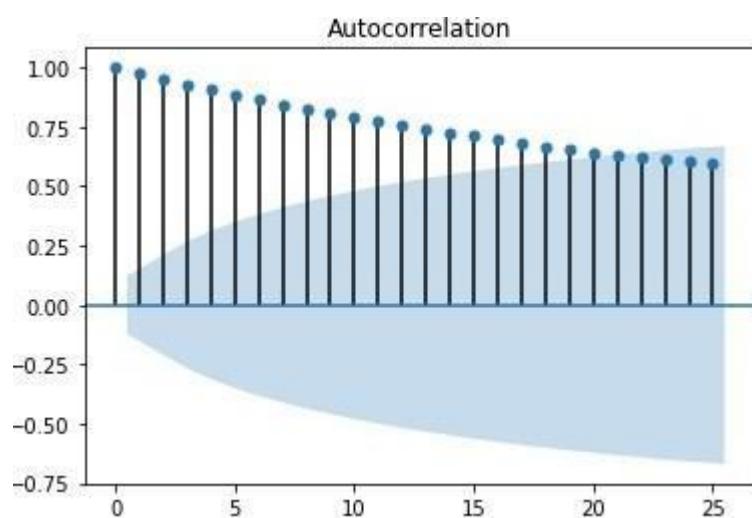
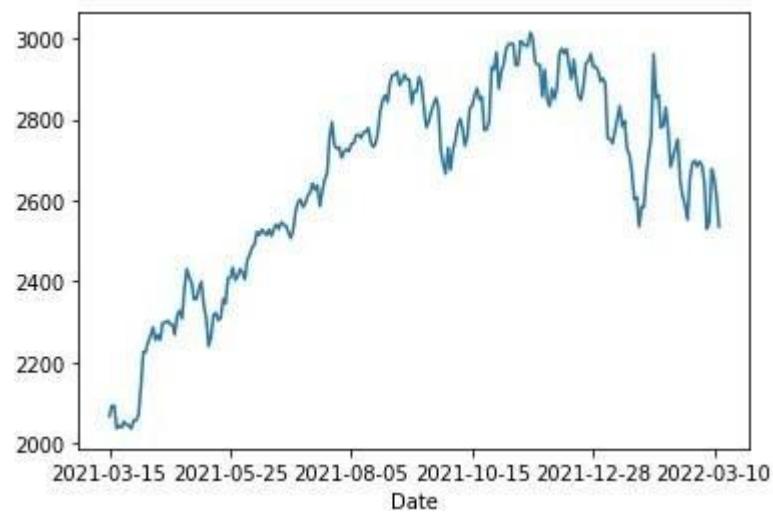
AIM: Train the autocorrelator by given patterns: A1=(-1,1,-1,1), A2=(1,1,1,-1), A3=(-1, -1, -1, 1). Test it using patterns: Ax=(-1,1,-1,1), Ay=(1,1,1,1), Az=(-1,-1,-1,-1).

CODE:

1_Python

```
import pandas as pd
import matplotlib.pyplot as plt
import statsmodels.api as sm
from statsmodels.stats.stattools import durbin_watson
from statsmodels.regression.linear_model import OLS
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
goog_stock_Data = pd.read_csv('GOOG.csv', header=0, index_col=0)
goog_stock_Data['Adj Close'].plot()
plt.show()
plot_acf(goog_stock_Data['Adj Close'], alpha =0.05)
plt.show()
plot_pacf(goog_stock_Data['Adj Close'], alpha =0.05, lags=50)
plt.show()
df = pd.DataFrame(goog_stock_Data,columns=['Date','Adj Close'])
X = np.arange(len(df[['Adj Close']]))

Y = np.asarray(df[['Adj Close']])
X = sm.add_constant(X)
ols_res = OLS(Y,X).fit()
durbin_watson(ols_res.resid)
```

OUTPUT:

PRACTICAL-4

AIM : Train the hetrocorrelator using multiple training encoding strategy for given patterns:
 A1=(000111001) B1=(010000111), A2=(111001110) B2=(100000001), A3=(110110101)
 B3(101001010). Test it using pattern A2.

CODE:

1_C

```
#include<stdio.h>
#include<conio.h>
int x1[1][9],x2[1][9],x3[1][9],p[1][9],pp[1][9],T[9][9]={0},s=0;
int y1[1][9],y2[1][9],y3[1][9],Hp[1][9],temp;
float q[3];
int i,j;
void printline();
x1[1][9]={-1,-1,-1,1,1,1,-1,1,1};
x2[1][9]={1,1,1,-1,-1,1,1,1,-1};
x3[1][9]={1,1,-1,1,1,-1,1,-1,1};
y1[1][9]={-1,1,-1,-1,-1,1,1,1,1};
y2[1][9]={1,-1,-1,-1,-1,-1,-1,1,1};
y3[1][9]={1,-1,1,-1,-1,1,-1,1,-1};
void main()
{
printf("\n\n=====HETEROCORRELATORS=====|\n|\n");
printf("Enter Scaling Vector:\n");
for(i=0;i<3;i++)
scanf("%f",&q[i]);
for(i=0;i<=8;i++)
{
  for(j=0;j<=8;j++)
  {
    temp=q[0]*x1[0][i];
    T[i][j]=T[i][j]+temp*y1[0][j];
  }
}
for(i=0;i<=8;i++)
{
  for(j=0;j<=8;j++)
  {
    temp=q[1]*x2[0][i];
    T[i][j]=T[i][j]+temp*y2[0][j];
  }
}
for(i=0;i<=8;i++)
{
  for(j=0;j<=8;j++)
  {
    temp=q[2]*x3[0][i];
    T[i][j]=T[i][j]+temp*y3[0][j];
  }
}
```

```

    }
printf("<-----Matrix ----->\n\n");
for(i=0;i<=8;i++)
{
for(j=0;j<=8;j++)
{
if(T[i][j]>=0)
printf(" %d ",T[i][j]);
else
printf("%d ",T[i][j]);
}
printf("\n");
}
printline();
printf("Enter any pattern to match:\n");
for(j=0;j<9;j++)
{
scanf("%d",&p[0][j]);
}
for(i=0;i<=8;i++)
{
for(j=0;j<=8;j++)
{
// printf("%d\n",T[0][j]);
s=s+T[j][i]*p[0][j];
}
pp[0][i]=s;
if(s>0)
Hp[0][i]=1;
if(s==0)
Hp[0][i]=p[0][i];
if(s<0)
Hp[0][i]=-1;
s=0;
}
printline();
printf("<-----alpha * M ----->\n\n");
for(j=0;j<9;j++)
printf("%d ",pp[0][j]);
printf("\n");
printline();
printf("\nSo,Recognized Pattern is:\n\n");
for(j=0;j<9;j++)
{
printf("%d ",Hp[0][j]);
}
printf("\n");
getch();
}
void printline()
{
int p;
for(p=0;p<25;p++)
printf("--");
printf("\n");
}

```

OUTPUT:

```
D:\Rohan\practice.exe
=====
=====HETEROCORRELATORS=====
Enter Scaling Vector:
2
2
1.5
<-----Matrix:----->
5 -5  1 -1 -1  1 -5 -3 -1
5 -5  1 -1 -1  1 -5 -3 -1
3 -3 -1  1  1 -1 -3 -5  1
-3  3  1 -1 -1  1  3  5 -1
-3  3  1 -1 -1  1  3  5 -1
-1  1 -5 -3 -3 -5  1 -1  5
5 -5  1 -1 -1  1 -5 -3 -1
3 -3 -1  1  1 -1 -3 -5  1
-3  3  1 -1 -1  1  3  5 -1
<-----alpha * M----->
29 -29 -7 -1 -1 -7 -29 -35 7
So, Recognized Pattern is:
1 -1 -1 -1 -1 -1 -1 -1 1
```

PRACTICAL-5

AIM: Implement Union, Intersection, Complement and Difference operations on fuzzy sets. Also create fuzzy relation by Cartesian product of any two fuzzy sets and perform maxmin composition on any two fuzzy relations

CODE:

```
a=dict() b=dict() c=dict()
a={'a':0.2,'b':0.3,'c':0.6,'d':0.6,}
b={'a':0.9,'b':0.9,'c':0.4,'d':0.5,}
print('The Fuzzy Set A ',a)
print('The Fuzzy Set B ',b)
for a_key,b_key in zip(a,b):
    a_value=a[a_key]
    b_value=b[b_key]
    if a_value>b_value:
        c[a_key]=a_value
    else:
        c[b_key]=b_value
print('Fuzzy Set Union ',c)

for a_key,b_key in zip(a,b):
    a_value=a[a_key]
    b_value=b[b_key]
    if a_value<b_value:
        c[a_key]=a_value
    else:
        c[b_key]=b_value
print('Fuzzy Set Intersction ',c)

for a_key in a:
    c[a_key]=1-a[a_key]
print('Fuzzy Set Complement OF A ',c)

for b_key in b:
    c[b_key]=1-b[b_key]
print('Fuzzy Set Complement OF B ',c)

for a_key,b_key in zip(a,b):
    a_value=a[a_key]
    b_value=b[b_key]
```

```
b_value=1-b_value if  
a_value<b_value:  
    c[a_key]=a_value  
else:  
    c[b_key]=b_value print('Fuzzy  
Set Difference ',c)
```

OUTPUT:

```
The Fuzzy Set A {'a': 0.2, 'b': 0.3, 'c': 0.6, 'd': 0.6}  
The Fuzzy Set B {'a': 0.9, 'b': 0.9, 'c': 0.4, 'd': 0.5}  
Fuzzy Set Union {'a': 0.9, 'b': 0.9, 'c': 0.6, 'd': 0.6}  
Fuzzy Set Intersction {'a': 0.2, 'b': 0.3, 'c': 0.4, 'd': 0.5}  
Fuzzy Set Complement OF A {'a': 0.8, 'b': 0.7, 'c': 0.4, 'd': 0.4}  
Fuzzy Set Complement OF B {'a': 0.0999999999999998, 'b': 0.0999999999999998, 'c': 0.6, 'd': 0.5}  
Fuzzy Set Difference {'a': 0.0999999999999998, 'b': 0.0999999999999998, 'c': 0.6, 'd': 0.5}
```

PRACTICAL-6

AIM:Solve Greg Viot's fuzzy cruise controller using MATLAB Fuzzy logic toolbox

CODE:

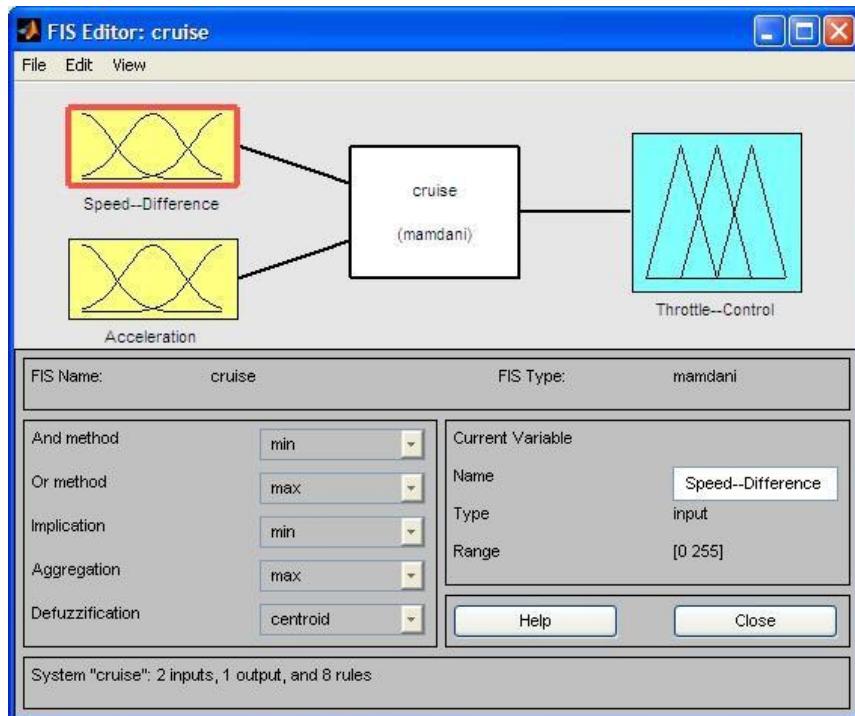
```
[System]
Name='Controller'
Type='mamdani'
Version=2.0
NumInputs=2
NumOutputs=1
NumRules=3
AndMethod='min'
OrMethod='max'
ImpMethod='min'
AggMethod='max'
DefuzzMethod='centroid'

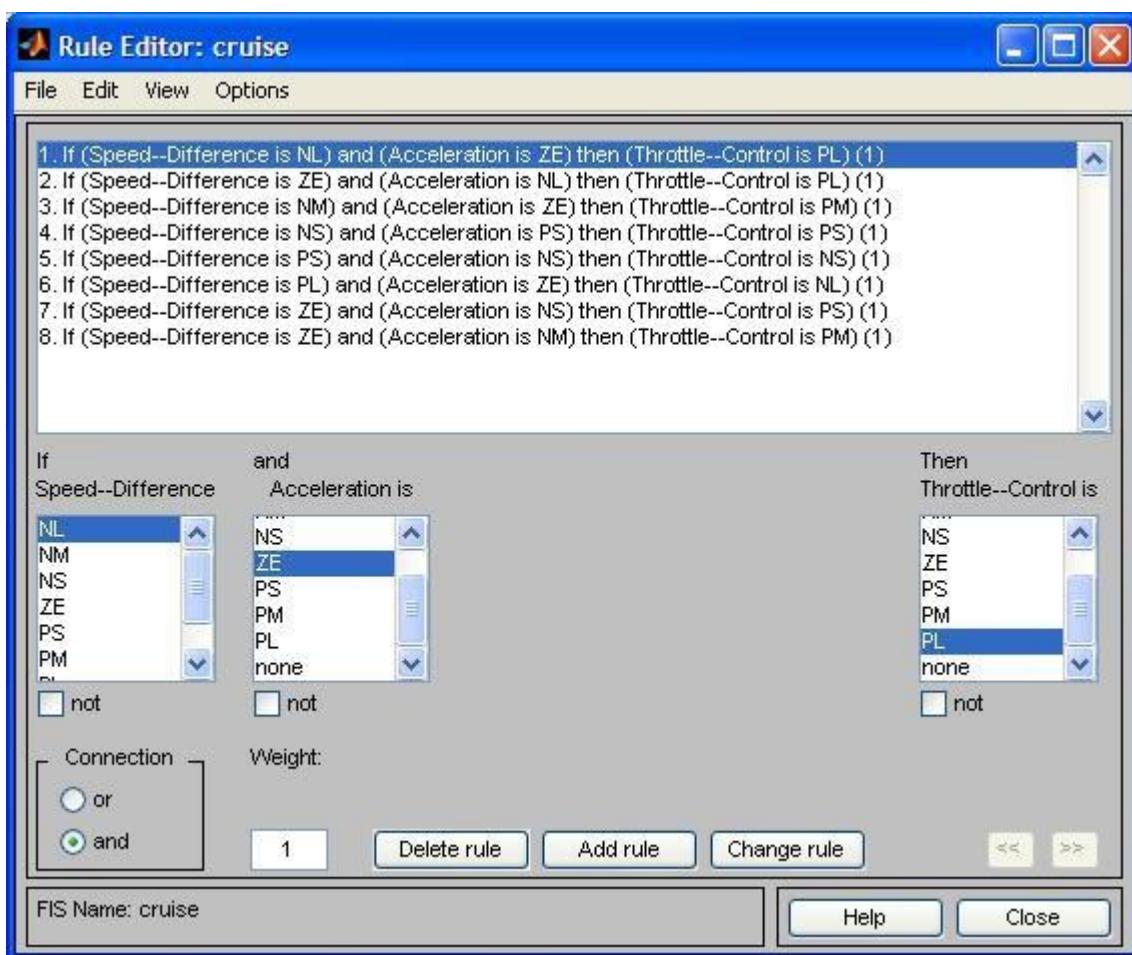
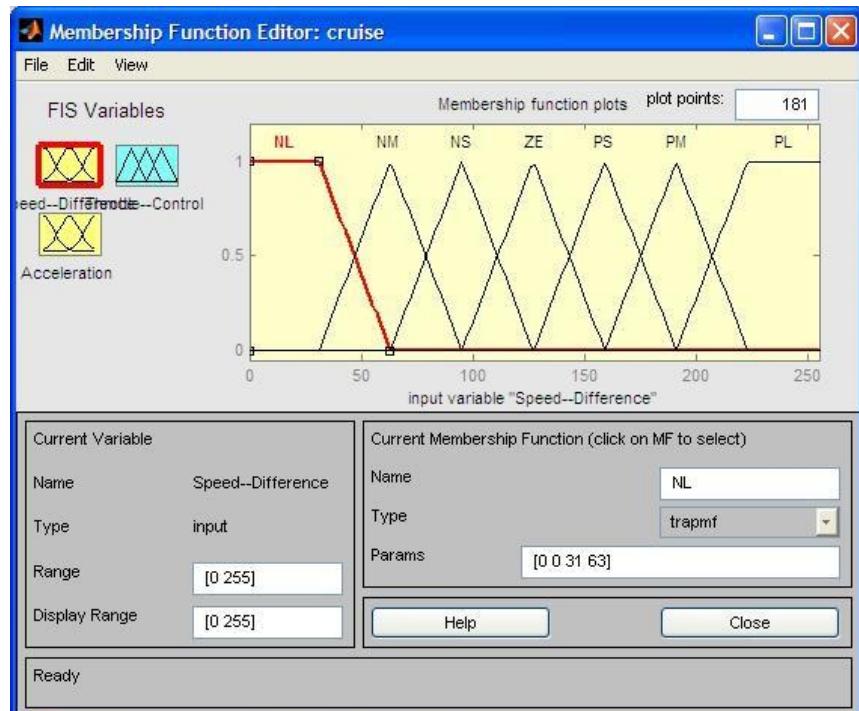
[Input1]
Name='DT'
Range=[-5 5]
NumMFs=3
MF1='PS':'trapmf',[-4 1.5 3.5 5]
MF2='PL':'trapmf',[0 0.8 5 5]
MF3='ZE':'trimf',[-5 0 5]

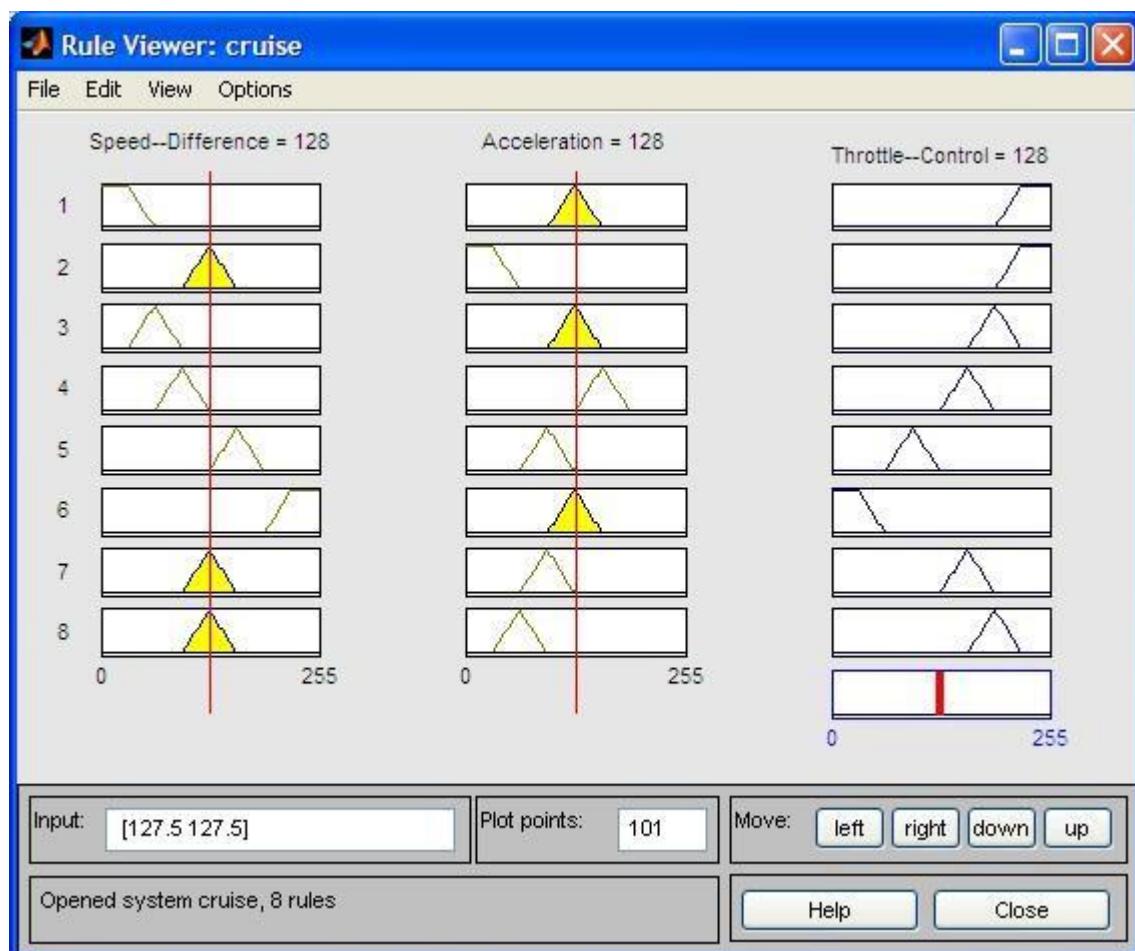
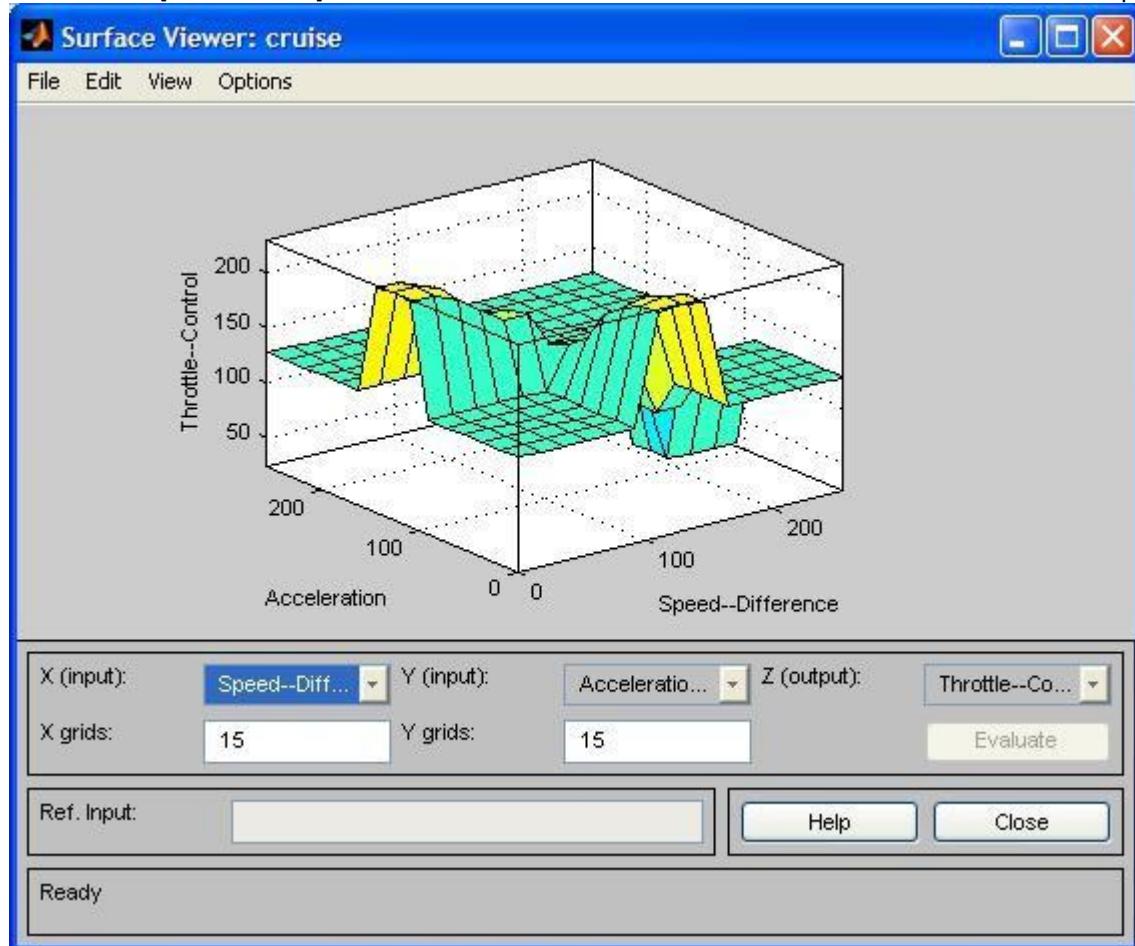
[Input2]
Name='dDT/dt'
Range=[-3 3]
NumMFs=3
MF1='PL':'trapmf',[0.6 1.5 3 3]
MF2='ZE':'trimf',[-3 0 3]
MF3='NS':'trimf',[-2.7 -0.5 1.9]

[Output1]
Name='Dial--Turn'
Range=[-1 1]
NumMFs=3
MF1='NL':'trapmf',[-1 -1 -0.5 -0.2]
MF2='NM':'trimf',[-1 -0.5 0]
MF3='ZE':'trimf',[-1 0 1]

[Rules]
3 1, 1 (1) : 1
2 2, 2 (1) : 1
1 3, 3 (1) : 1
```

OUTPUT:





PRACTICAL-7

AIM:Solve Air Conditioner Controller using MATLAB Fuzzy logic toolbox

CODE:

```
[System]
Name='cruise'
Type='mamdani'
Version=2.0
NumInputs=2
NumOutputs=1
NumRules=8
AndMethod='min'
OrMethod='max'
ImpMethod='min'
AggMethod='max'
DefuzzMethod='centroid'

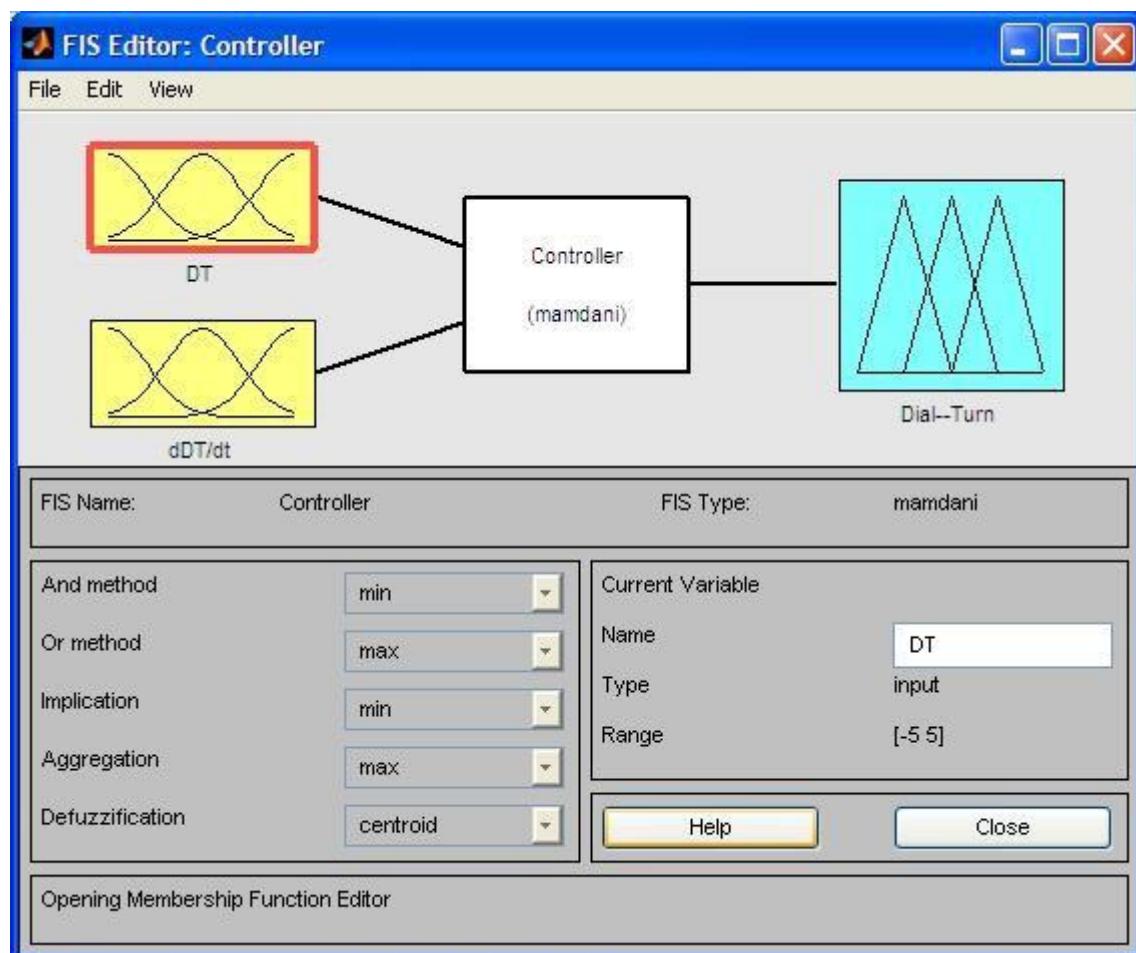
[Input1]
Name='Speed--Difference'
Range=[0 255]
NumMFs=7
MF1='NL':'trapmf',[0 0 31 63]
MF2='NM':'trimf',[31 63 95]
MF3='NS':'trimf',[63 95 127]
MF4='ZE':'trimf',[95 127 159]
MF5='PS':'trimf',[127 159 191]
MF6='PM':'trimf',[159 191 223]
MF7='PL':'trapmf',[191 223 255 255]

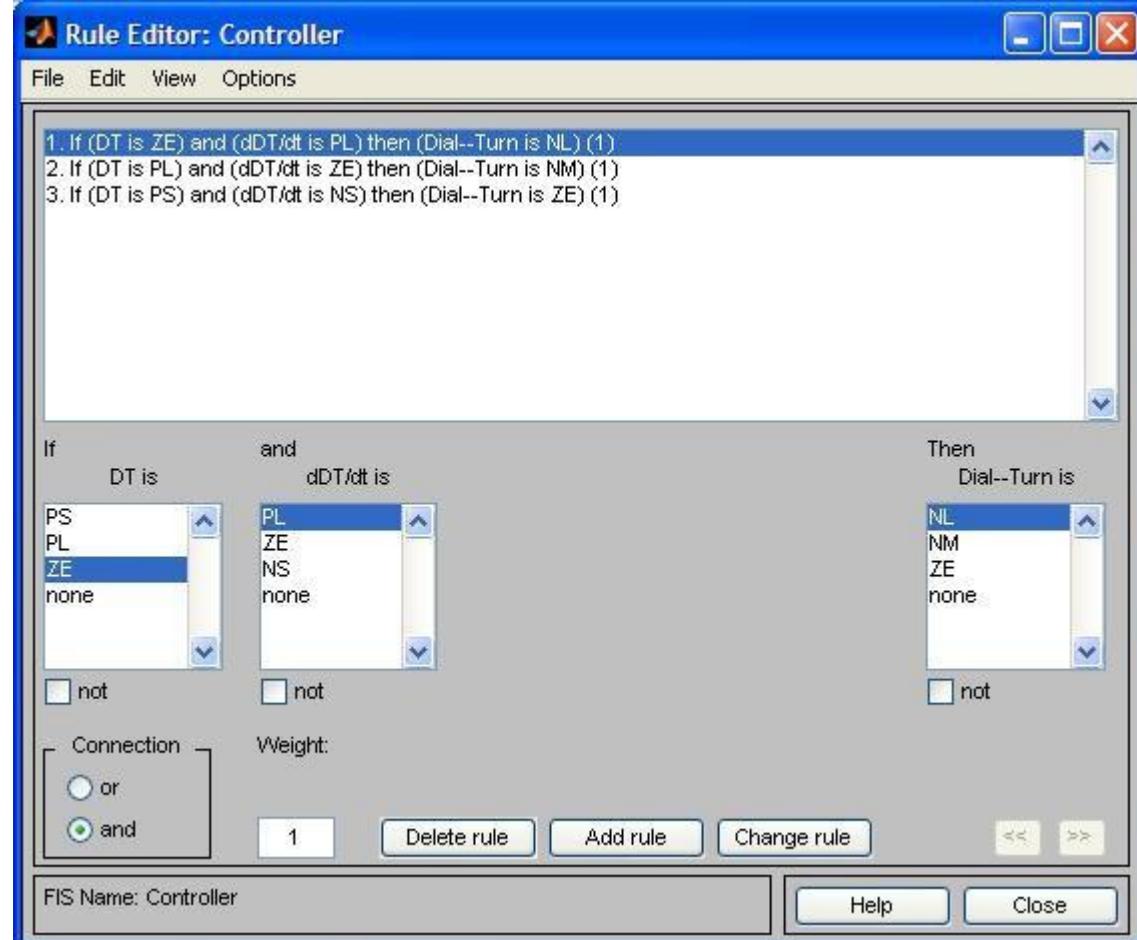
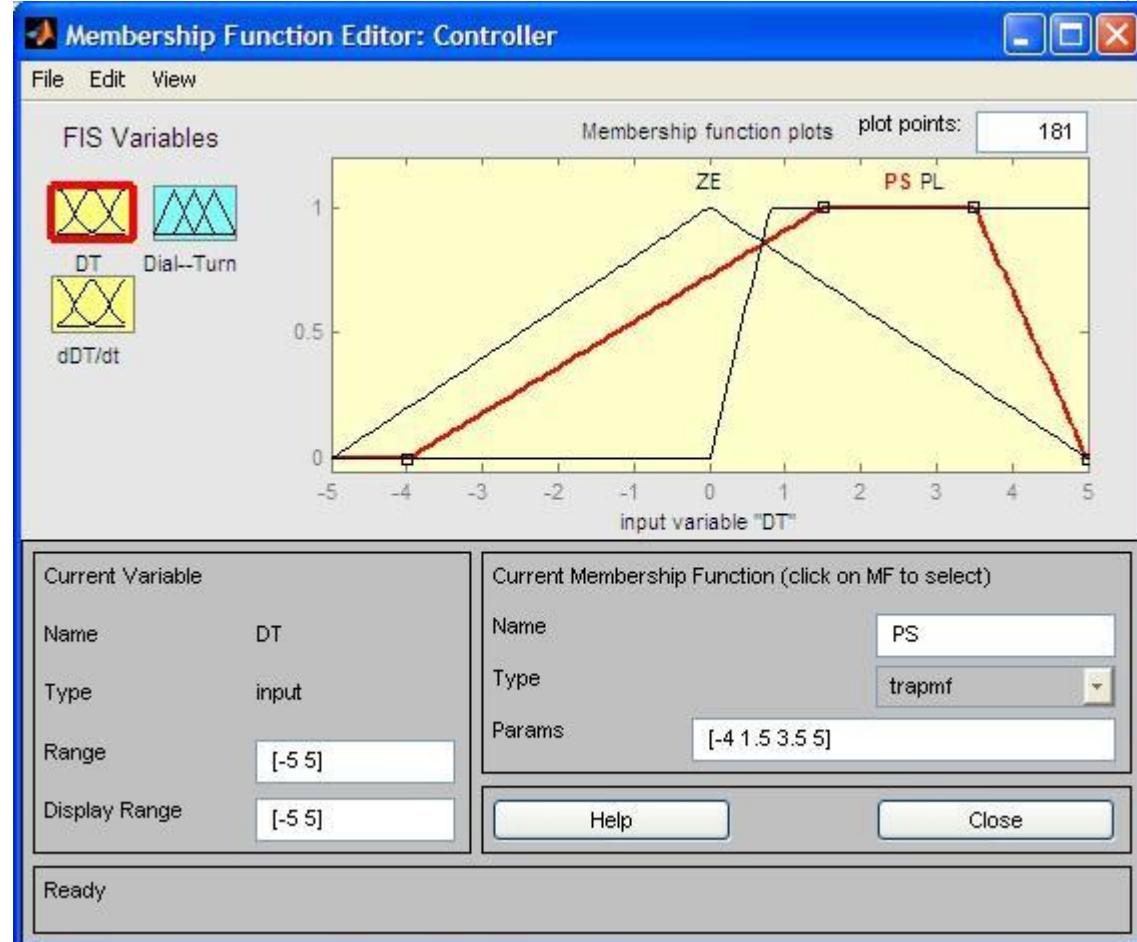
[Input2]
Name='Acceleration'
Range=[0 255]
NumMFs=7
MF1='NL':'trapmf',[0 0 31 63]
MF2='NM':'trimf',[31 63 95]
MF3='NS':'trimf',[63 95 127]
MF4='ZE':'trimf',[95 127 159]
MF5='PS':'trimf',[127 159 191]
MF6='PM':'trimf',[159 191 223]
MF7='PL':'trapmf',[191 223 255 255]

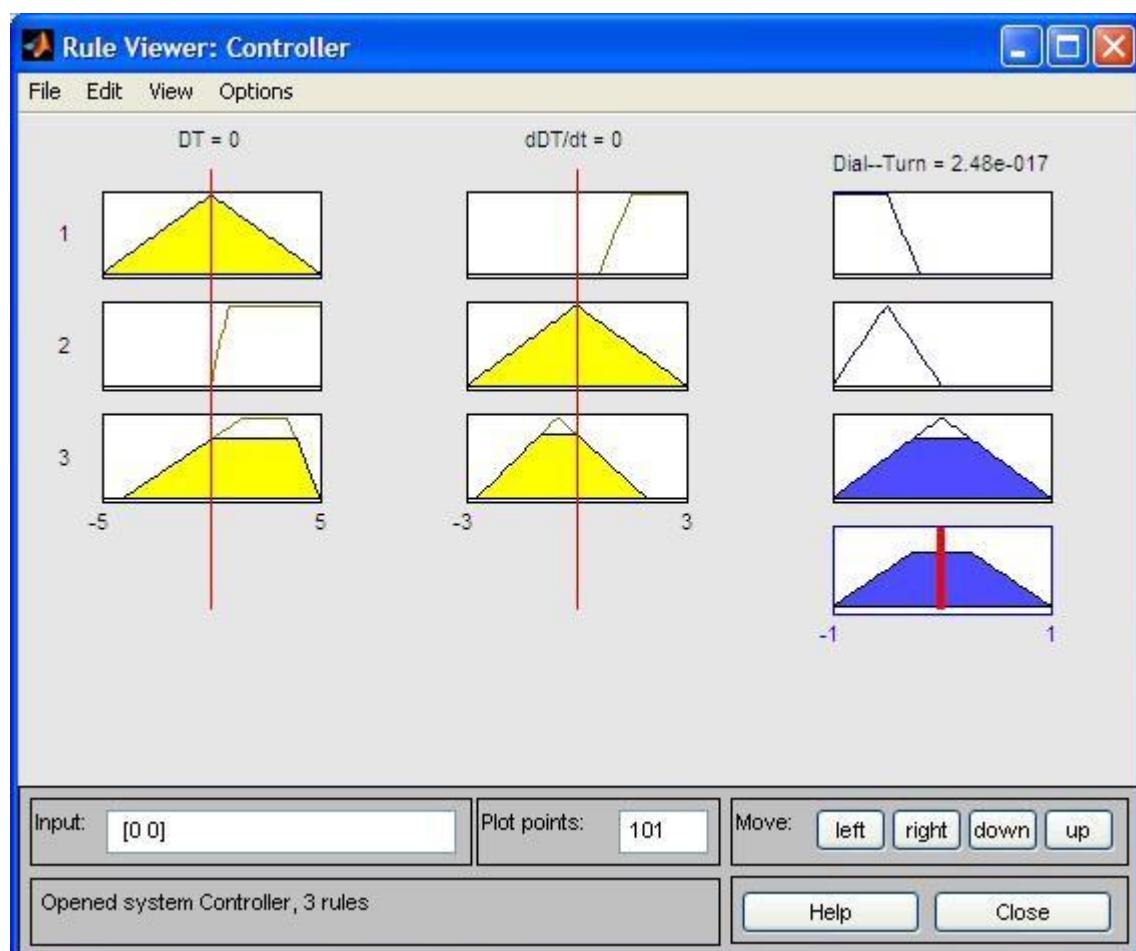
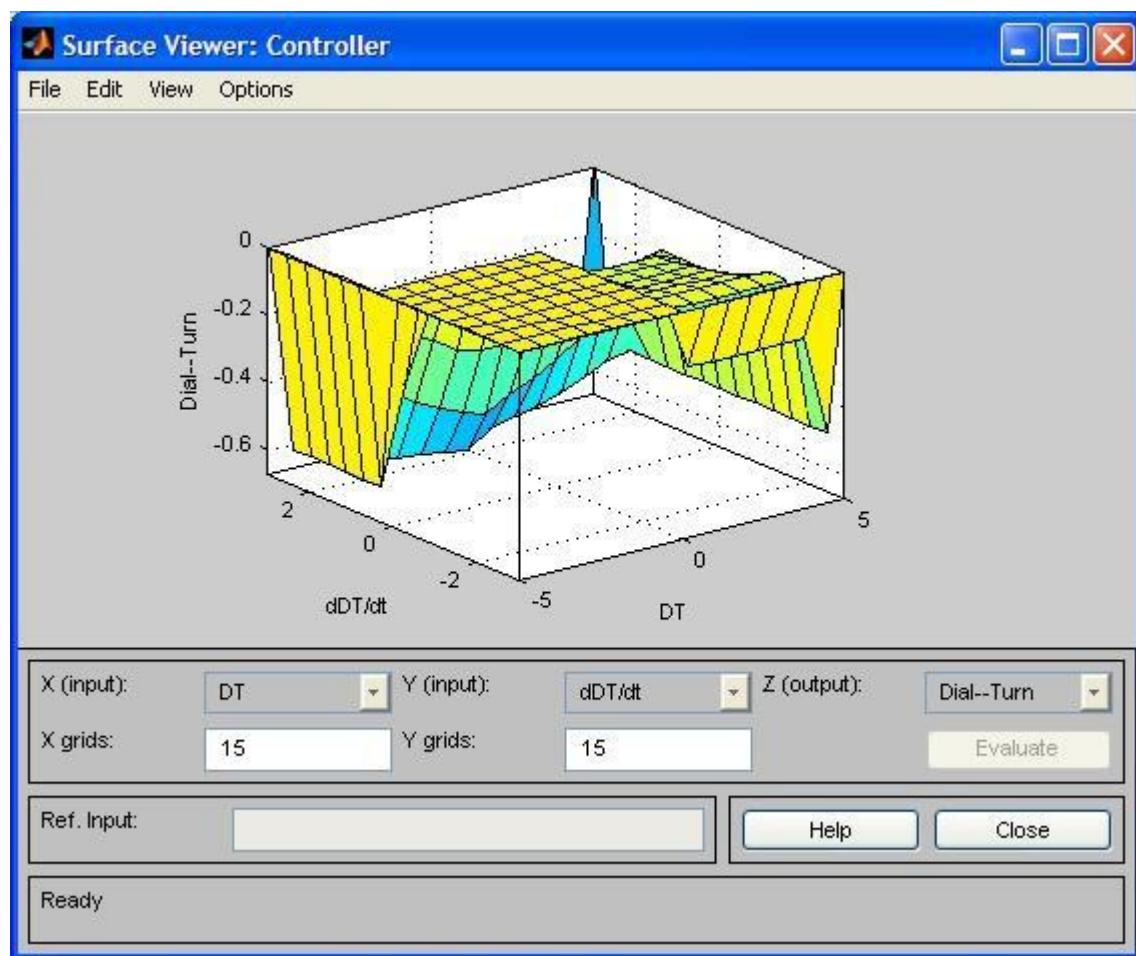
[Output1]
Name='Throttle--Control'
Range=[0 255]
NumMFs=7
MF1='NL':'trapmf',[0 0 31 63]
MF2='NM':'trimf',[31 63 95]
MF3='NS':'trimf',[63 95 127]
MF4='ZE':'trimf',[95 127 159]
MF5='PS':'trimf',[127 159 191]
MF6='PM':'trimf',[159 191 223]
MF7='PL':'trapmf',[191 223 255 255]
```

[Rules]

1 4, 7 (1) : 1
4 1, 7 (1) : 1
2 4, 6 (1) : 1
3 5, 5 (1) : 1
5 3, 3 (1) : 1
7 4, 1 (1) : 1
4 3, 5 (1) : 1
4 2, 6 (1) : 1

OUTPUT:





PRACTICAL-8

AIM: Implement TSP using GA.

CODE:

1.Python

```

import numpy as np, random, operator, pandas as pd
import matplotlib.pyplot as plt
def create_starting_population(size,Number_of_city):
    population = []
    for i in range(0,size):
        population.append(create_new_member(Number_of_city))
    return population
def pick_mate(N):
    i=random.randint(0,N)
    return i
def distance(i,j):
    return np.sqrt((i[0]-j[0])**2 + (i[1]-j[1])**2)
def score_population(population, CityList):
    scores = []
    for i in population:
        scores.append(fitness(i, CityList))
    return scores
def fitness(route,CityList):
    score=0
    for i in range(1,len(route)):
        k=int(route[i-1])
        l=int(route[i])
        score = score + distance(CityList[k],CityList[l])
    return score
def create_new_member(Number_of_city):
    pop=set(np.arange(Number_of_city,dtype=int))
    route=list(random.sample(pop,Number_of_city))
    return route
def crossover(a,b):
    child=[]
    childA=[]
    childB=[]
    geneA=int(random.random()* len(a))
    geneB=int(random.random()* len(a))
    start_gene=min(geneA,geneB)
    end_gene=max(geneA,geneB)
    for i in range(start_gene,end_gene):
        childA.append(a[i])
    
```

```

childB=[item for item in a if item not in childA]
child=childA+childB
return child
def mutate(route,probability): route=np.array(route)
for swaping_p in range(len(route)):
    if(random.random() < probability):
        swapedWith = np.random.randint(0,len(route))
        temp1=route[swaping_p]
        temp2=route[swapedWith]
        route[swapedWith]=temp1 route[swaping_p]=temp2
return route
def selection(popRanked, eliteSize):
    selectionResults=[]
    result=[]
    for i in popRanked:
        result.append(i[0])
    for i in range(0,eliteSize):
        selectionResults.append(result[i])
    return selectionResults
def rankRoutes(population,City_List):
    fitnessResults = {}
    for i in range(0,len(population)):
        fitnessResults[i] = fitness(population[i],City_List)
    return sorted(fitnessResults.items(), key = operator.itemgetter(1), reverse
= False)
def breedPopulation(mating_pool):
    children=[]
    for i in range(len(mating_pool)-1):
        children.append(crossover(mating_pool[i],mating_pool[i+1]))
    return children
def mutatePopulation(children,mutation_rate):
    new_generation=[]
    for i in children:
        mutated_child=mutate(i,mutation_rate)
        new_generation.append(mutated_child)
    return new_generation
def matingPool(population, selectionResults):
    matingpool = []
    for i in range(0, len(selectionResults)):
        index = selectionResults[i]
        matingpool.append(population[index])
    return matingpool
def next_generation(City_List,current_population,mutation_rate,elite_size):
    population_rank=rankRoutes(current_population,City_List)
    selection_result=selection(population_rank,elite_size)
    mating_pool=matingPool(current_population,selection_result)
    children=breedPopulation(mating_pool)
    next_generation=mutatePopulation(children,mutation_rate)
    return next_generation

```

```

def genetic_algorithm(City_List, size_population=1000, elite_size=75, mutation_Rate=0.05, generation=2000):

    pop = []
    progress = []
    Number_of_cities = len(City_List)
    population = create_starting_population(size_population, Number_of_cities)
    progress.append(rankRoutes(population, City_List)[0][1])
    print(f"initial route distance {progress[0]}")
    print(f"initial route {population[0]}")
    for i in range(0, generation):
        pop = next_generation(City_List, population, mutation_Rate, elite_size)
        progress.append(rankRoutes(pop, City_List)[0][1])
    rank_ = rankRoutes(pop, City_List)[0]
    print(f"Best Route :{pop[rank_[0]]} ")
    print(f"best route distance {rank_[1]}")
    plt.plot(progress)
    plt.ylabel('Distance')
    plt.xlabel('Generation')
    plt.show()
    return rank_, pop
cityList = []
for i in range(0, 25):
    x = int(random.random() * 200)
    y = int(random.random() * 200)
    cityList.append((x, y))

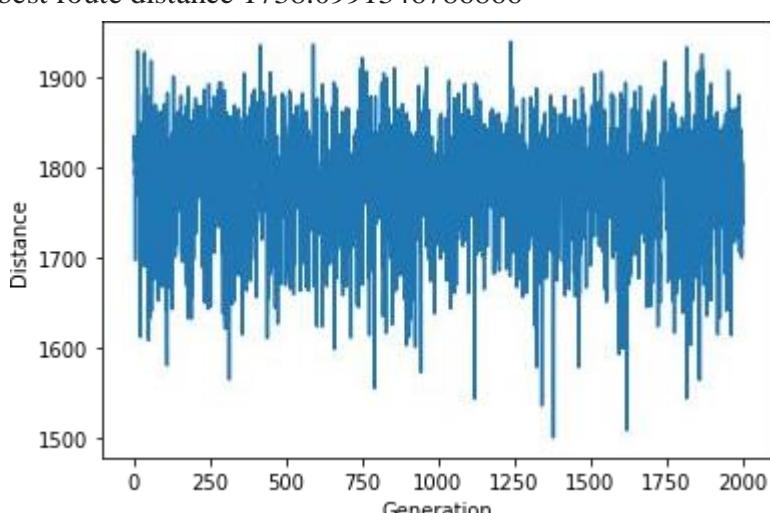
```

OUTPUT:

```

initial route distance 1808.5488313262974
initial route [14, 9, 20, 18, 5, 19, 10, 0, 7, 23, 12, 4, 3, 15, 8, 22, 13, 24, 16, 11, 1, 2, 21, 6, 17]
Best Route :[12 0 1 23 22 19 11 18 14 6 3 5 10 2 7 9 20 24 17 8 21 13 16 4
15]
best route distance 1738.0991346786866

```



```
x_axis=[]
y_axis=[]
for i in cityList:
    x_axis.append(i[0])
    y_axis.append(i[1])
plt.scatter(x_axis,y_axis)
plt.show()
```

OUTPUT:

