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Simple AI Technique implementation

AIM

To implement Simple AI techniques - Water Jug Problem.

PYTHON CODE:

```
x = 0
y = 0
m = 4
n = 3
print("Initial state = (0,0)")
print("Capacities = (4,3)")
print("Goal state = (2,y)")
while x = 2:
 r = int(input("Enter rule"))
 if(r == 1):
  x = m
 elif(r == 2):
  y = n
 elif(r == 3):
  \mathbf{x} = \mathbf{0}
 elif(r == 4):
  y = 0
 elif(r == 5):
  t = n - y
  y = n
  x = t
 elif(r == 6):
  t = m - x
```

x = m

```
y = t
elif(r == 7):
y += x
x = 0
elif(r == 8):
x += y
y = 0
print (x, y)
```

Initial state = (0,0)

Capacities = (4,3)

Goal state = (2,y)

Enter rule2

03

Enter rule8

3 0

Enter rule2

3 3

Enter rule6

4 2

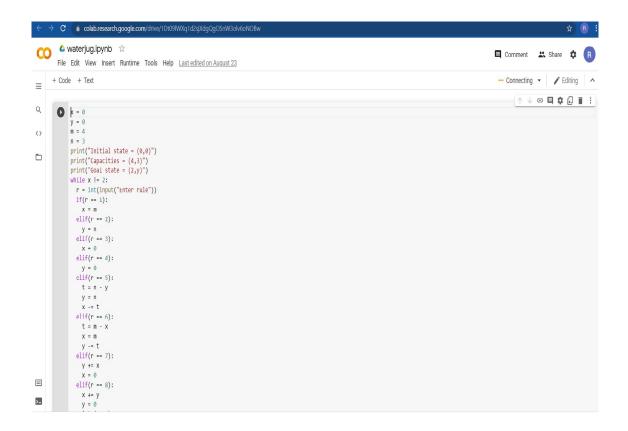
Enter rule3

02

Enter rule8

20

OUTPUT SCREENSHOT:



RESULT: The above python code has been successfully executed.

Implementation of Tic-Tac-Toe Game

AIM

To Implement Tic Tac Toe Game

```
def tic tac toe():
           board = [1, 2, 3, 4, 5, 6, 7, 8, 9]
           end = False
           win commbinations = ((0, 1, 2), (3, 4, 5), (6, 7, 8), (0, 3, 6), (1, 4, 7), (2, 5, 8), (0, 4, 8), (2, 4, 4, 5), (3, 4, 5), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 6), (4, 5, 
6))
           def draw():
                      print(board[0], board[1], board[2])
                      print(board[3], board[4], board[5])
                      print(board[6], board[7], board[8])
                      print()
           def p1():
                      n = choose number()
                      if board[n] == "X" or board[n] == "O":
                                  print("\nYou can't go there. Try again")
                                 p1()
                      else:
                                     board[n] = "X"
                                          def p2():
                      n = choose_number()
                      if board[n] == "X" or board[n] == "O":
                                  print("\nYou can't go there. Try again")
                                 p2()
                      else:
```

```
board[n] = "O"
def choose number():
  while True:
    while True:
       a = input()
       try:
         a = int(a)
         a -= 1
         if a in range(0, 9):
            return a
         else:
            print("\nThat's not on the board. Try again")
            continue
       except ValueError:
         print("\nThat's not a number. Try again")
         continue
def check board():
  count = 0
  for a in win_commbinations:
    if board[a[0]] == board[a[1]] == board[a[2]] == "X":
       print("Player 1 Wins!\n")
       print("Congratulations!\n")
       return True
    if board[a[0]] == board[a[1]] == board[a[2]] == "O":
       print("Player 2 Wins!\n")
       print("Congratulations!\n")
       return True
  for a in range(9):
     if board[a] == "X" or board[a] == "O":
       count += 1
    if count == 9:
       print("The game ends in a Tie\n")
       return True
```

```
while not end:
    draw()
    end = check_board()
    if end == True:
       break
    print("Player 1 choose where to place a cross")
    p1()
    print()
    draw()
    end = check_board()
    if end == True:
       break
    print("Player 2 choose where to place a nought")
    p2()
    print()
  if input("Play again (y/n)\n") == "y":
    print()
    tic_tac_toe()
tic_tac_toe()
OUTPUT:
123
4 5 6
789
Player 1 choose where to place a cross
12X
4 5 6
789
```

O 2 X
4 5 6
7 8 9
Player 1 choose where to place a cross
•
OXX
4 5 6
7 8 9
Player 2 choose where to place a nought
Thayer 2 choose where to place a hought
OXX
4 5 O
789
Player 1 choose where to place a cross
OXX
4 5 O
7 8 X
Player 2 choose where to place a nought
OXX
O X X 4 O O
400
400

Player 2 choose where to place a nought

OXX

XOO

78X

Player 2 choose where to place a nought

OXX

XOO

O 8 X

Player 1 choose where to place a cross

OXX

XOO

OXX

The game ends in a Tie

```
M Inbox (4,756) - rajkumad@srmist X 🔥 Priority - Google Drive X 🗯 Tictactoe.ipynb - Colaboratory X 🛨
△ Tictactoe.ipynb ☆
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↑ ↓ ⊕ 目 ‡ : :
       def tic_tac_toe():
    board = [1, 2, 3, 4, 5, 6, 7, 8, 9]
    end = False
    win_combinations = ((0, 1, 2), (3, 4, 5), (6, 7, 8), (0, 3, 6), (1, 4, 7), (2, 5, 8), (0, 4, 8), (2, 4, 6))
Q
                     def draw():
    print(board[0], board[1], board[2])
    print(board[0], board[4], board[5])
    print(board[0], board[7], board[0])
    print()
def p1():
    n = choose_number()
    if board[n] == "X" or board[n] == "0":
        print("\nYou can't go there. Try again")
        p1()
    else:
                                 board[n] - "x"
                    def p2():
    n choose number()
    if board[n] == "X" or board[n] == "0":
        print('\MYOU can't go there. Try again")
        p2()
    else:
        board[n] = "0"
                     def choose_number():
    while True:
    while True:
                                                                                                                                                                                                    Type here to search
                                                                   O 🙀 🔯 🦸
```

Implementation of Travelling Sales man problem

AIM

To Implement Travelling Salesman Problem

```
from sys import maxsize
from itertools import permutations
V = 4
# implementation of traveling Salesman Problem
def travellingSalesmanProblem(graph, s):
  # store all vertex apart from source vertex
  vertex = []
  for i in range(V):
    if i != s:
       vertex.append(i)
  # store minimum weight Hamiltonian Cycle
  min path = maxsize
  next permutation=permutations(vertex)
  for i in next permutation:
    # store current Path weight(cost)
    current pathweight = 0
    # compute current path weight
    k = s
     for j in i:
       current pathweight += graph[k][j]
       k = j
     current pathweight += graph[k][s]
     # update minimum
```

```
min_path = min(min_path, current_pathweight)
return min_path

# Driver Code

if __name__ == "__main__":

# matrix representation of graph
graph = [[0, 10, 15, 20], [10, 0, 35, 25],

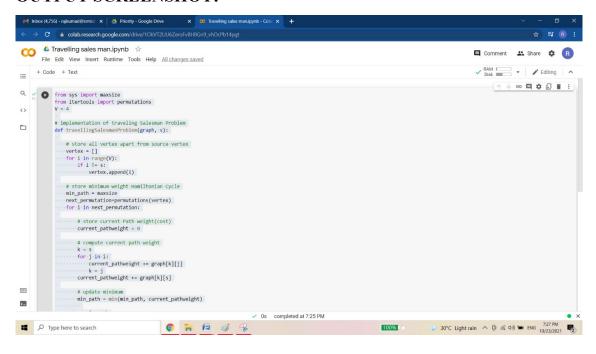
        [15, 35, 0, 30], [20, 25, 30, 0]]

s = 0

print(travellingSalesmanProblem(graph, s))
```

80

OUTPUT SCREENSHOT:



RESULT: The above python code has been successfully executed.

Knowledge implementation

AIM

To implement knowledge to intelligent agents

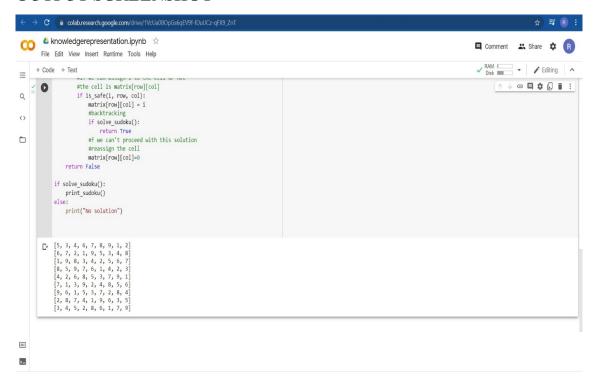
```
size = 9
#empty cells have value zero
matrix = [
  [5,3,0,0,7,0,0,0,0],
  [6,0,0,1,9,5,0,0,0],
  [0,9,8,0,0,0,0,6,0],
  [8,0,0,0,6,0,0,0,3],
  [4,0,0,8,0,3,0,0,1],
  [7,0,0,0,2,0,0,0,6],
  [0,6,0,0,0,0,2,8,0],
  [0,0,0,4,1,9,0,0,5],
  [0,0,0,0,8,0,0,7,9]
#print sudoku
def print_sudoku():
  for i in matrix:
     print (i)
#assign cells and check
def number_unassigned(row, col):
  num unassign = 0
  for i in range(0,size):
     for j in range (0,size):
       #cell is unassigned
       if matrix[i][j] == 0:
```

```
row = i
          col = j
          num\_unassign = 1
          a = [row, col, num_unassign]
          return a
  a = [-1, -1, num unassign]
  return a
#check validity of number
def is_safe(n, r, c):
  #checking in row
  for i in range(0,size):
     #there is a cell with same value
     if matrix[r][i] == n:
       return False
  #checking in column
  for i in range(0,size):
     #there is a cell with same value
     if matrix[i][c] == n:
       return False
  row start = (r//3)*3
  col start = (c//3)*3;
  #checking submatrix
  for i in range(row_start,row_start+3):
     for j in range(col_start,col_start+3):
       if matrix[i][j]==n:
          return False
  return True
#check validity of number
def solve_sudoku():
  row = 0
  col = 0
```

```
#if all cells are assigned then the sudoku is already solved
  #pass by reference because number_unassigned will change the values of row and col
  a = number unassigned(row, col)
  if a[2] == 0:
    return True
  row = a[0]
  col = a[1]
  #number between 1 to 9
  for i in range(1,10):
    #if we can assign i to the cell or not
    #the cell is matrix[row][col]
    if is safe(i, row, col):
       matrix[row][col] = i
       #backtracking
       if solve_sudoku():
          return True
       #f we can't proceed with this solution
       #reassign the cell
       matrix[row][col]=0
  return False
if solve_sudoku():
  print sudoku()
else:
  print("No solution")
OUTPUT
[5, 3, 4, 6, 7, 8, 9, 1, 2]
[6, 7, 2, 1, 9, 5, 3, 4, 8]
[1, 9, 8, 3, 4, 2, 5, 6, 7]
[8, 5, 9, 7, 6, 1, 4, 2, 3]
[4, 2, 6, 8, 5, 3, 7, 9, 1]
[7, 1, 3, 9, 2, 4, 8, 5, 6]
[9, 6, 1, 5, 3, 7, 2, 8, 4]
```

[2, 8, 7, 4, 1, 9, 6, 3, 5]

[3, 4, 5, 2, 8, 6, 1, 7, 9]



RESULT: The above python code has been successfully executed.

Implementation of FOPL and Rules

AIM:

To implement First order predicate logic and its rules.

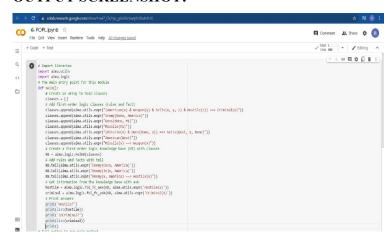
```
# Import libraries
import aima.utils
import aima.logic
# The main entry point for this module
def main():
  # Create an array to hold clauses
  clauses = []
  # Add first-order logic clauses (rules and fact)
  clauses.append(aima.utils.expr("(American(x) & Weapon(y) & Sells(x, y, z) & Hostile(z)) \Longrightarrow
Criminal(x)"))
  clauses.append(aima.utils.expr("Enemy(Nono, America)"))
  clauses.append(aima.utils.expr("Owns(Nono, M1)"))
  clauses.append(aima.utils.expr("Missile(M1)"))
  clauses.append(aima.utils.expr("(Missile(x) & Owns(Nono, x)) ==> Sells(West, x, Nono)"))
  clauses.append(aima.utils.expr("American(West)"))
  clauses.append(aima.utils.expr("Missile(x) \Longrightarrow Weapon(x)"))
  # Create a first-order logic knowledge base (KB) with clauses
  KB = aima.logic.FolKB(clauses)
  # Add rules and facts with tell
  KB.tell(aima.utils.expr('Enemy(Coco, America)'))
  KB.tell(aima.utils.expr('Enemy(Jojo, America)'))
  KB.tell(aima.utils.expr("Enemy(x, America) ==> Hostile(x)"))
  # Get information from the knowledge base with ask
  hostile = aima.logic.fol fc ask(KB, aima.utils.expr('Hostile(x)'))
  criminal = aima.logic.fol fc ask(KB, aima.utils.expr('Criminal(x)'))
  # Print answers
  print('Hostile?')
```

```
print(list(hostile))
print('\nCriminal?')
print(list(criminal))
print()
# Tell python to run main method
if __name__ == "__main__": main()

OUTPUT:
Hostile?
[{x: Coco}, {x: Nono}, {x: Jojo}]
Criminal?
```

OUTPUT SCREENSHOT:

 $[{x: West}]$



RESULT: The above python code has been successfully executed.

Implementation of Ontology and FOL

AIM

To implement ontology and First order logic

PROCEDURE:

```
from setuptools import setup, find packages # Always prefer setuptools over distutils
from codecs import open # To use a consistent encoding
from os import path
import os
# trick to manage package versions in one place only
# http://stackoverflow.com/questions/458550/standard-way-to-embed-version-into-python-
package
import re
VERSIONFILE="ontospy/VERSION.py"
verstrline = open(VERSIONFILE, "rt").read()
VSRE = r'' \land version = ['']([^'']*)[''']"
mo = re.search(VSRE, verstrline, re.M)
if mo:
  VERSIONSTRING = mo.group(1)
  raise RuntimeError("Unable to find version string in %s." % (VERSIONFILE,))
# Get the long description from the README file
with open(path.join(HERE, 'README.md'), encoding='utf-8') as f:
  long description = f.read()
# Parse requirements.txt file so to have one single source of truth
REQUIREMENTS DATA = []
with open(path.join(HERE, 'requirements.txt'), encoding='utf-8') as f:
  for l in f.readlines():
    if not l.startswith("#"):
       if (">=" in l):
         REQUIREMENTS DATA.append([1.split(">=")[0]])
```

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```
elif ("=" in 1):
         REQUIREMENTS DATA.append([l.split("=")[0]])
def get package folders(top folder, root path):
  Utility to generate dynamically the list of folders needed by the package data setting
    package data={
       'ontospy': ['viz/static/*.*', 'viz/templates/*.*', 'viz/templates/shared/*.*',
'viz/templates/splitter/*.*', 'viz/templates/markdown/*.*'],
    },
  *****
  dirs = []
  out = []
  for root, dirs, files in os.walk(top folder):
    for dir in dirs:
       dirs.append(os.path.join(root, dir))
  for d in dirs:
    d = os.path.join(d, "*.*")
    out.append( d.replace(root path+"/", ""))
  return out
PROJECT ROOT = os.path.join(HERE, "ontospy") # should be top level always
DATA STATIC FILES = os.path.join(PROJECT ROOT, "ontodocs", "media", "static")
DATA TEMPLATE FILES = os.path.join(PROJECT ROOT, "ontodocs", "media", "templates")
# dynamically generate list of data folders
PACKAGE DATA FOLDERS = get package folders(
  DATA STATIC FILES, PROJECT ROOT) + get package folders(
    DATA TEMPLATE FILES, PROJECT ROOT)
if True:
  print(PACKAGE DATA FOLDERS)
setup(
  name='ontospy',
  version=VERSIONSTRING,
  description=
```

```
'Query, inspect and visualize knowledge models encoded as RDF&OWL ontologies.',
long description=long description,
long description content type='text/markdown',
url='https://github.com/lambdamusic/ontospy',
author='Michele Pasin',
author email='michele.pasin@gmail.com',
license='MIT',
# See https://pypi.python.org/pypi?%3Aaction=list classifiers
classifiers=[
  # How mature is this project? Common values are
  # 3 - Alpha
  # 4 - Beta
  # 5 - Production/Stable
  'Development Status :: 5 - Production/Stable',
  # Indicate who your project is intended for
  'Intended Audience :: Science/Research',
  'Intended Audience :: Developers',
  'Topic :: Scientific/Engineering :: Artificial Intelligence',
  # Pick your license as you wish (should match "license" above)
  'License :: OSI Approved :: MIT License',
  # Specify the Python versions you support here. In particular, ensure
  # that you indicate whether you support Python 2, Python 3 or both.
  'Programming Language :: Python :: 3',
  'Programming Language :: Python :: 3.2',
  'Programming Language :: Python :: 3.3',
  'Programming Language :: Python :: 3.4',
  'Programming Language :: Python :: 3.5',
  'Programming Language :: Python :: 3.6',
  'Programming Language :: Python :: 3.7',
],
# requirements files see:
# http://python-packaging-user-guide.readthedocs.org/en/latest/requirements/
# NOTE: packages are installed in reverse order
```

```
install requires=REQUIREMENTS DATA,
# List additional groups of dependencies here (e.g. development dependencies).
# You can install these using the following syntax, for example:
# $ pip install -e .[dev,test]
extras require={
  'SHELL': ['readline'],
  'FULL': ['Django>=1.10.3', 'Pygments==2.1.3'],
},
package data={
  'ontospy': PACKAGE DATA FOLDERS
  },
entry points={
  'console scripts': [
    # 'ontospy-sketch=ontospy.extras.sketch:main',
     'ontospy=ontospy.cli:main_cli',
     'quicktest ontospy=ontospy.tests.quick:main'
  ],
},)
```

Concept Learning task

AIM:

To implement Concept learning task in machine learning.

```
import pandas as pd
import numpy as np
#to read the data in the csv file
data = pd.read csv("data.csv")
print(data,"n")
#making an array of all the attributes
d = np.array(data)[:,:-1]
print("n The attributes are: ",d)
#segragating the target that has positive and negative examples
target = np.array(data)[:,-1]
print("n The target is: ",target)
#training function to implement find-s algorithm
def train(c,t):
  for i, val in enumerate(t):
     if val == "Yes":
       specific hypothesis = c[i].copy()
       break
  for i, val in enumerate(c):
     if t[i] == "Yes":
       for x in range(len(specific hypothesis)):
          if val[x] != specific hypothesis[x]:
            specific hypothesis[x] = '?'
          else:
            pass
  return specific hypothesis
#obtaining the final hypothesis
```

Time Weather Temperature Company Humidity Wind Goes

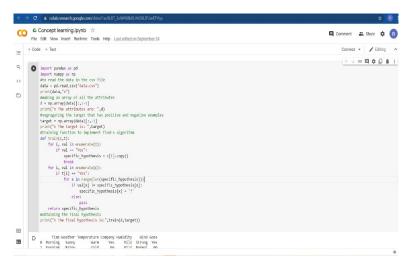
- 0 Morning Sunny Warm Yes Mild Strong Yes
- 1 Evening Rainy Cold No Mild Normal No
- 2 Morning Sunny Moderate Yes Normal Normal Yes
- 3 Evening Sunny Cold Yes High Strong Yes n
- n The attributes are: [['Morning' 'Sunny' 'Warm' 'Yes' 'Mild' 'Strong']

['Evening' 'Rainy' 'Cold' 'No' 'Mild' 'Normal']

['Morning' 'Sunny' 'Moderate' 'Yes' 'Normal' 'Normal']

['Evening' 'Sunny' 'Cold' 'Yes' 'High' 'Strong']]

- n The target is: ['Yes' 'No' 'Yes' 'Yes']
- n The final hypothesis is: ['?' 'Sunny' '?' 'Yes' '?' '?']



RESULT: The above python code has been successfully executed.

Design a Learning System

AIM

To design a learning system in machine learning

```
import sys
print('Python: {}'.format(sys.version))
import scipy
print('scipy: {}'.format(scipy. version ))
import numpy
print('numpy: {}'.format(numpy. version ))
import matplotlib
print('matplotlib: {}'.format(matplotlib. version ))
import pandas pd
print('pandas: {}'.format(pandas. version ))
import sklearn
print('sklearn: {}'.format(sklearn. version ))
url = "https://raw.githubusercontent.com/jbrownlee/Datasets/master/iris.csv"
names = ['sepal-length', 'sepal-width', 'petal-length', 'petal-width', 'class']
dataset = pd.read csv(url, names=names)
print(dataset.shape)
print(dataset.head(20))
print(dataset.describe())
print(dataset.groupby('class').size())
dataset.hist()
pyplot.show()
scatter matrix(dataset)
pyplot.show()
from pandas import read csv
from pandas.plotting import scatter matrix
from matplotlib import pyplot
url = "https://raw.githubusercontent.com/jbrownlee/Datasets/master/iris.csv"
```

```
names = ['sepal-length', 'sepal-width', 'petal-length', 'petal-width', 'class']
dataset = read csv(url, names=names)
dataset.plot(kind='box', subplots=True, layout=(2,2), sharex=False, sharey=False)
pyplot.show()
dataset.hist()
pyplot.show()
scatter matrix(dataset)
pyplot.show()
OUTPUT:
Python: 3.7.12 (default, Sep 10 2021, 00:21:48)
[GCC 7.5.0]
scipy: 1.4.1
numpy: 1.19.5
matplotlib: 3.2.2
pandas: 1.1.5
sklearn: 0.22.2.post1
(150, 5)
  sepal-length sepal-width petal-length petal-width
                                                          class
0
        5.1
                  3.5
                            1.4
                                     0.2 Iris-setosa
1
        4.9
                  3.0
                            1.4
                                     0.2 Iris-setosa
2
                  3.2
        4.7
                            1.3
                                     0.2 Iris-setosa
3
        4.6
                  3.1
                            1.5
                                     0.2 Iris-setosa
4
        5.0
                  3.6
                            1.4
                                     0.2 Iris-setosa
5
        5.4
                  3.9
                            1.7
                                     0.4 Iris-setosa
6
        4.6
                  3.4
                            1.4
                                     0.3 Iris-setosa
7
        5.0
                  3.4
                            1.5
                                     0.2 Iris-setosa
8
        4.4
                  2.9
                            1.4
                                     0.2 Iris-setosa
9
        4.9
                  3.1
                            1.5
                                     0.1 Iris-setosa
10
         5.4
                  3.7
                            1.5
                                      0.2 Iris-setosa
11
         4.8
                            1.6
                                      0.2 Iris-setosa
                  3.4
12
         4.8
                  3.0
                            1.4
                                      0.1 Iris-setosa
```

13

14

15

4.3

5.8

5.7

3.0

4.0

4.4

1.1

1.2

1.5

0.1 Iris-setosa

0.2 Iris-setosa

0.4 Iris-setosa

16	5.4	3.9	1.3	0.4 Iris-setosa
17	5.1	3.5	1.4	0.3 Iris-setosa
18	5.7	3.8	1.7	0.3 Iris-setosa
19	5.1	3.8	1.5	0.3 Iris-setosa

sepal-length sepal-width petal-length petal-width

count	150.000000	150.000000	150.0000	000 150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000
-1				

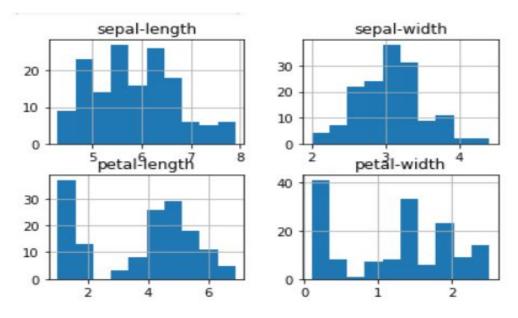
class

Iris-setosa 50

Iris-versicolor 50

Iris-virginica 50

dtype: int64



RESULT: The above python code has been successfully executed.

Implementation of candidate elimination algorithm

AIM

To implement candidate elimination algorithm

```
import numpy as np
import pandas as pd
data = pd.read csv('enjoysport.csv')
concepts = np.array(data.iloc[:,0:-1])
print("\nInstances are:\n",concepts)
target = np.array(data.iloc[:,-1])
print("\nTarget Values are: ",target)
def learn(concepts, target):
  specific h = concepts[0].copy()
  print("\nInitialization of specific h and genearal h")
  print("\nSpecific Boundary: ", specific h)
  general h = [["?" for i in range(len(specific h))] for i in range(len(specific h))]
  print("\nGeneric Boundary: ",general h)
  for i, h in enumerate(concepts):
     print("\nInstance", i+1 , "is ", h)
     if target[i] == "yes":
       print("Instance is Positive ")
       for x in range(len(specific h)):
          if h[x]!= specific h[x]:
            specific h[x] = '?'
            general h[x][x] = "?"
     if target[i] == "no":
       print("Instance is Negative ")
       for x in range(len(specific h)):
          if h[x]!= specific h[x]:
            general_h[x][x] = \text{specific } h[x]
          else:
```

```
general h[x][x] = '?'
    print("Specific Bundary after ", i+1, "Instance is ", specific h)
    print("Generic Boundary after ", i+1, "Instance is ", general h)
    print("\n")
  indices = [i for i, val in enumerate(general h) if val == ['?', '?', '?', '?', '?', '?']]
  for i in indices:
    general h.remove(['?', '?', '?', '?', '?', '?'])
  return specific h, general h
s final, g final = learn(concepts, target)
print("Final Specific h: ", s final, sep="\n")
print("Final General h: ", g final, sep="\n")
OUTPUT:
Instances are:
[['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'high' 'strong' 'warm' 'same']
['rainy' 'cold' 'high' 'strong' 'warm' 'change']
['sunny' 'warm' 'high' 'strong' 'cool' 'change']]
Target Values are: ['yes' 'yes' 'no' 'yes']
Initialization of specific h and genearal h
Specific Boundary: ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
!?!, !?!], ['?!, !?!, !?!, !?!, !?!], ['?!, !?!, !?!, !?!, !?!, !?!]
Instance 1 is ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
Instance is Positive
Specific Bundary after 1 Instance is ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
'?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?']]
Instance 2 is ['sunny' 'warm' 'high' 'strong' 'warm' 'same']
Instance is Positive
Specific Bundary after 2 Instance is ['sunny' 'warm' '?' 'strong' 'warm' 'same']
!?!], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?']
Instance 3 is ['rainy' 'cold' 'high' 'strong' 'warm' 'change']
Instance is Negative
```

Specific Bundary after 3 Instance is ['sunny' 'warm' '?' 'strong' 'warm' 'same']

Generic Boundary after 3 Instance is [['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?']

'?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', 'same']]

Instance 4 is ['sunny' 'warm' 'high' 'strong' 'cool' 'change']

Instance is Positive

Specific Bundary after 4 Instance is ['sunny' 'warm' '?' 'strong' '?' '?']

Generic Boundary after 4 Instance is [['sunny', '?', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?',

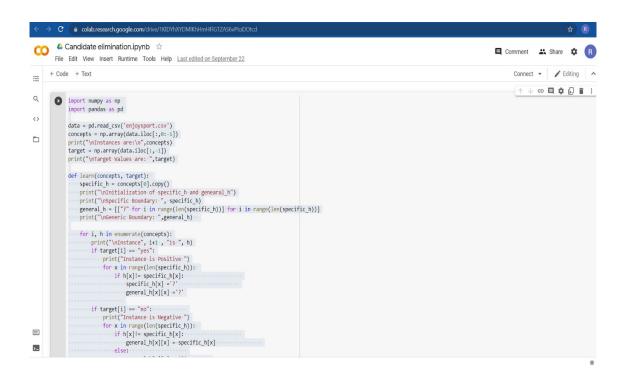
!?!, !?!, !?!, !?!], ['?!, !?!, !?!, !?!, !?!], ['?!, !?!, !?!, !?!, !?!, !?!], ['?!, !?!, !?!, !?!, !?!]

Final Specific h:

['sunny' 'warm' '?' 'strong' '?' '?']

Final General h:

[['sunny', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]



RESULT: The above python code has been successfully executed.

Decision tree implementation

AIM:

To implement Decision tree algorithm

```
import pandas as pd
from sklearn.tree import DecisionTreeClassifier
from sklearn.model selection import train test split
col names = ['pregnant', 'glucose', 'bp', 'skin', 'insulin', 'bmi', 'pedigree', 'age', 'label']
pima = pd.read csv("pima-indians-diabetes.csv", header = None, names = col names)
pima.head()
feature cols = ['pregnant', 'insulin', 'bmi', 'age', 'glucose', 'bp', 'pedigree']
X = pima[feature cols] # Features
y = pima.label # Target variable
X train, X test, y train, y test = train test split(X, y, test size = 0.3, random state = 1)
clf = DecisionTreeClassifier()
clf = clf.fit(X train,y train)
y pred = clf.predict(X test)
from sklearn.metrics import classification report, confusion matrix, accuracy score
result = confusion matrix(y test, y pred)
print("Confusion Matrix:")
print(result)
result1 = classification report(y test, y pred)
print("Classification Report:",)
print (result1)
result2 = accuracy score(y test,y pred)
print("Accuracy:",result2)
from sklearn.tree import export graphviz
from sklearn.externals.six import StringIO
from IPython.display import Image
import pydotplus
```

```
dot_data = StringIO()
export_graphviz(clf, out_file=dot_data, filled=True, rounded=True,
special_characters=True,feature_names = feature_cols,class_names=['0','1'])
graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
graph.write_png('Pima_diabetes_Tree.png')
Image(graph.create_png())
```

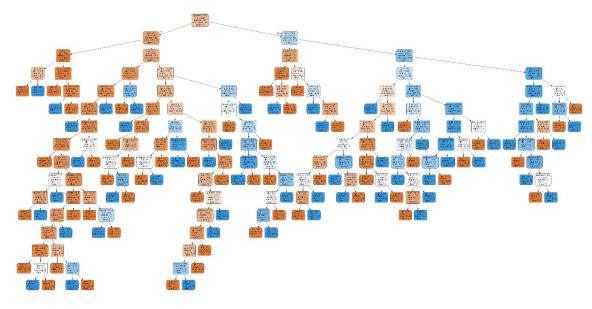
Confusion Matrix:

[[117 29]

[44 41]]

Classification Report:

precision recall f1-score support 0.73 0.80 0.76 146 1 0.59 0.480.53 85 0.68 231 accuracy macro avg 0.66 0.640.65 231 weighted avg 0.67 0.68 0.68 231



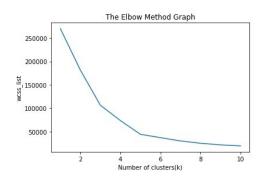
RESULT: The above python code has been successfully executed.

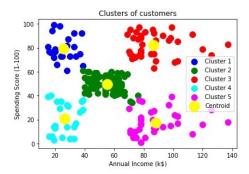
Implementation of K- Means algorithm

AIM:

To implement k-means algorithm

```
import numpy as nm
import matplotlib.pyplot as mtp
import pandas as pd
dataset = pd.read csv('Mall Customers.csv')
x = dataset.iloc[:, [3, 4]].values
from sklearn.cluster import KMeans
wess list=[] #Initializing the list for the values of WCSS
 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()
kmeans = KMeans(n clusters=5, init='k-means++', random state= 42)
y predict= kmeans.fit predict(x)
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()
```





RESULT: The above python code has been successfully executed.

Implementation of ID3 algorithm

AIM

To implement ID3 algorithm

```
from sklearn.model selection import train test split
from sklearn.tree import DecisionTreeClassifier
import numpy as np
import pandas as pd
dataset = pd.read csv("pima-indians-diabetes.csv")
dataset.head()
dataset.shape
features = dataset.drop(["Outcome"], axis=1)
X = np.array(features)
y = np.array(dataset["Outcome"])
X train, X val, y train, y val = train test split(X, y, random state=0, test size=0.20)
tree = DecisionTreeClassifier()
tree.fit(X train, y train)
tree.tree .max depth
validation prediction = tree.predict(X val)
training prediction = tree.predict(X train)
print('Accuracy training set: ', accuracy score(y true=y train, y pred=training prediction))
print('Accuracy validation set: ', accuracy score(y true=y val, y pred=validation prediction))
from sklearn.tree import export graphviz
feature names = features.columns
dot data = export graphviz(tree, out file=None,
              feature names=feature names,
               class names=True,
               filled=True, rounded=True,
               special characters=True)
graph = graphviz.Source(dot data)
graph
```

Accuracy training set: 0.7964169381107492
Accuracy validation set: 0.8116883116883117

RESULT: The above python code has been successfully executed.

Neural Network model implementation

AIM

To implement Neural Network.

```
from numpy import loadtxt
from keras.models import Sequential
from keras.layers import Dense
# load the dataset
dataset = loadtxt(', delimiter=',')
# split into input (X) and output (y) variables
X = dataset[:,0:8]
y = dataset[:,8]
# define the keras model
model = Sequential()
model.add(Dense(12, input dim=8, activation='relu'))
model.add(Dense(8, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
# compile the keras model
model.compile(loss='binary crossentropy', optimizer='adam', metrics=['accuracy'])
# fit the keras model on the dataset
model.fit(X, y, epochs=150, batch size=10)
# evaluate the keras model
_{-}, accuracy = model.evaluate(X, y)
print('Accuracy: %.2f' % (accuracy*100))
```

001101.	
Epoch 1/200	
77/77 [======	===] - 1s 1ms/step - loss: 9.3462 - accuracy: 0.5404
Epoch 2/200	
77/77 [===] - 0s 1ms/step - loss: 2.6633 - accuracy: 0.4896
Epoch 3/200	
77/77 [=================================	===] - 0s 1ms/step - loss: 1.4994 - accuracy: 0.5391
Epoch 4/200	
77/77 [=================================	===] - 0s 1ms/step - loss: 1.0731 - accuracy: 0.5938
Epoch 5/200	
77/77 [=================================	===] - 0s 1ms/step - loss: 0.8820 - accuracy: 0.6367
Epoch 6/200	
77/77 [=================================	===] - 0s 2ms/step - loss: 0.8197 - accuracy: 0.6289
Epoch 7/200	
77/77 [===] - 0s 1ms/step - loss: 0.7773 - accuracy: 0.6432
Epoch 8/200	
77/77 [===] - 0s 1ms/step - loss: 0.7406 - accuracy: 0.6523
Epoch 9/200	
77/77 [===] - 0s 1ms/step - loss: 0.7005 - accuracy: 0.6471
Epoch 10/200	
77/77 [===] - 0s 1ms/step - loss: 0.7090 - accuracy: 0.6549
Epoch 11/200	
77/77 [===] - 0s 1ms/step - loss: 0.6890 - accuracy: 0.6341
Epoch 12/200	
77/77 [===] - 0s 1ms/step - loss: 0.6738 - accuracy: 0.6615
Epoch 13/200	
77/77 [===] - 0s 1ms/step - loss: 0.6460 - accuracy: 0.6719
Epoch 14/200	
77/77 [===] - 0s 1ms/step - loss: 0.6386 - accuracy: 0.6680
Epoch 15/200	
77/77 [===] - 0s 1ms/step - loss: 0.6327 - accuracy: 0.6836
Epoch 16/200	
77/77 [===] - 0s 2ms/step - loss: 0.6208 - accuracy: 0.6680
Epoch 17/200	

```
77/77 [=====
                   =======] - 0s 1ms/step - loss: 0.6492 - accuracy: 0.6667
Epoch 18/200
77/77 [======
                     ======] - 0s 1ms/step - loss: 0.6164 - accuracy: 0.6914
Epoch 19/200
77/77 [=====
                    =======] - 0s 1ms/step - loss: 0.6257 - accuracy: 0.6706
Epoch 20/200
77/77 [======
                ========] - 0s 1ms/step - loss: 0.6058 - accuracy: 0.6771
Epoch 21/200
77/77 [======] - 0s 1ms/step - loss: 0.5965 - accuracy: 0.6862
Epoch 22/200
77/77 [======
             Epoch 23/200
77/77 [======
              Epoch 24/200
77/77 [======
              Epoch 25/200
77/77 [======] - 0s 2ms/step - loss: 0.5788 - accuracy: 0.7070
.....
```

RESULT: The above python code has been successfully executed.

Implementation of Multi-layer neural network

AIM

To implement multi layer Neural Network.

```
from sklearn.base import BaseEstimator, ClassifierMixin, RegressorMixin
import random
class MultiLayerPerceptron(BaseEstimator, ClassifierMixin):
  def init (self, params=None):
    if (params == None):
       self.inputLayer = 4
                                        # Input Layer
       self.hiddenLayer = 5
                                        # Hidden Layer
       self.outputLayer = 3
                                        # Outpuy Layer
       self.learningRate = 0.005
                                          # Learning rate
       self.max epochs = 600
                                          # Epochs
       self.iasHiddenValue = -1
                                          # Bias HiddenLayer
       self.BiasOutputValue = -1
                                           # Bias OutputLayer
       self.activation = self.ativacao['sigmoid'] # Activation function
       self.deriv = self.derivada['sigmoid']
    else:
       self.inputLayer = params['InputLayer']
       self.hiddenLayer = params['HiddenLayer']
       self.OutputLayer = params['OutputLayer']
       self.learningRate = params['LearningRate']
       self.max epochs = params['Epocas']
       self.BiasHiddenValue = params['BiasHiddenValue']
       self.BiasOutputValue = params['BiasOutputValue']
       self.activation = self.ativacao[params['ActivationFunction']]
       self.deriv = self.derivada[params['ActivationFunction']]
```

^{&#}x27;Starting Bias and Weights'

```
self.WEIGHT hidden = self.starting weights(self.hiddenLayer, self.inputLayer)
    self.WEIGHT output = self.starting weights(self.OutputLayer, self.hiddenLayer)
    self.BIAS hidden = np.array([self.BiasHiddenValue for i in range(self.hiddenLayer)])
    self.BIAS output = np.array([self.BiasOutputValue for i in range(self.OutputLayer)])
    self.classes number = 3
  pass
   def starting weights(self, x, y):
    return [[2 * random.random() - 1 for i in range(x)] for j in range(y)]
  ativacao = {
     'sigmoid': (lambda x: 1/(1 + np.exp(-x))),
       'tanh': (lambda x: np.tanh(x)),
       'Relu': (lambda x: x*(x > 0)),
         }
  derivada = {
     'sigmoid': (lambda x: x*(1-x)),
       'tanh': (lambda x: 1-x**2),
       'Relu': (lambda x: 1 * (x>0))
        }
  def Backpropagation Algorithm(self, x):
    DELTA output = []
    'Stage 1 - Error: OutputLayer'
    ERROR output = self.output - self.OUTPUT L2
    DELTA output = ((-1)*(ERROR output) * self.deriv(self.OUTPUT L2))
arrayStore = []
    'Stage 2 - Update weights OutputLayer and HiddenLayer'
    for i in range(self.hiddenLayer):
       for j in range(self.OutputLayer):
         self.WEIGHT output[i][j] -= (self.learningRate * (DELTA output[j] *
self.OUTPUT L1[i]))
         self.BIAS output[j] -= (self.learningRate * DELTA output[j])
    'Stage 3 - Error: HiddenLayer'
    delta hidden = np.matmul(self.WEIGHT output, DELTA output)*
self.deriv(self.OUTPUT L1)
```

```
'Stage 4 - Update weights HiddenLayer and InputLayer(x)'
  for i in range(self.OutputLayer):
     for j in range(self.hiddenLayer):
       self.WEIGHT hidden[i][j] -= (self.learningRate * (delta hidden[j] * x[i]))
       self.BIAS hidden[j] -= (self.learningRate * delta hidden[j])
def show err graphic(self,v erro,v epoca):
  plt.figure(figsize=(9,4))
  plt.plot(v epoca, v erro, "m-",color="b", marker=11)
  plt.xlabel("Number of Epochs")
  plt.ylabel("Squared error (MSE) ");
  plt.title("Error Minimization")
  plt.show()
def predict(self, X, y):
  'Returns the predictions for every element of X'
  my predictions = []
  'Forward Propagation'
  forward = np.matmul(X,self.WEIGHT hidden) + self.BIAS hidden
  forward = np.matmul(forward, self.WEIGHT output) + self.BIAS output
  for i in forward:
     my predictions.append(max(enumerate(i), key=lambda x:x[1])[0])
  print(" Number of Sample | Class | Output | Hoped Output ")
  for i in range(len(my predictions)):
    if(my predictions[i] == 0):
       print("id:{} | Iris-Setosa | Output: {} ".format(i, my predictions[i], y[i]))
     elif(my predictions[i] == 1):
       print("id:{} | Iris-Versicolour | Output: {} ".format(i, my predictions[i], y[i]))
     elif(my predictions[i] == 2):
       print("id:{} | Iris-Iris-Virginica | Output: {} ".format(i, my predictions[i], y[i]))
  return my predictions
  pass
def fit(self, X, y):
  count epoch = 1
  total error = 0
```

```
n = len(X);
    epoch array = []
    error array = []
    W0 = []
    W1 = []
    while(count epoch <= self.max epochs):
       for idx,inputs in enumerate(X):
         self.output = np.zeros(self.classes number)
         'Stage 1 - (Forward Propagation)'
         self.OUTPUT L1 = self.activation((np.dot(inputs, self.WEIGHT hidden) +
self.BIAS hidden.T))
         self.OUTPUT L2 = self.activation((np.dot(self.OUTPUT L1, self.WEIGHT output) +
self.BIAS output.T))
         'Stage 2 - One-Hot-Encoding'
         if(y[idx] == 0):
            self.output = np.array([1,0,0]) #Class1 \{1,0,0\}
         elif(y[idx] == 1):
            self.output = np.array([0,1,0]) \#Class2 \{0,1,0\}
         elif(y[idx] == 2):
           self.output = np.array([0,0,1]) #Class3 \{0,0,1\}
         square error = 0
         for i in range(self.OutputLayer):
            erro = (self.output[i] - self.OUTPUT L2[i])**2
            square error = (square error + (0.05 * erro))
            total error = total error + square error
         'Backpropagation: Update Weights'
         self.Backpropagation Algorithm(inputs)
       total error = (total error / n)
       if((count epoch \% 50 == 0)or(count epoch == 1)):
         print("Epoch ", count epoch, "- Total Error: ",total error)
         error array.append(total error)
         epoch array.append(count epoch)
       W0.append(self.WEIGHT hidden)
       W1.append(self.WEIGHT output)
```

```
count_epoch += 1
self.show_err_graphic(error_array,epoch_array)
plt.plot(W0[0])
plt.title('Weight Hidden update during training')
plt.legend(['neuron1', 'neuron2', 'neuron3', 'neuron4', 'neuron5'])
plt.ylabel('Value Weight')
plt.show()
plt.plot(W1[0])
plt.title('Weight Output update during training')
plt.legend(['neuron1', 'neuron2', 'neuron3'])
plt.ylabel('Value Weight')
plt.show()
return self
```

Epoch 1 - Total Error: 0.08939914265311079

Epoch 50 - Total Error: 0.05376852115527734

Epoch 100 - Total Error: 0.03773105860576402

Epoch 150 - Total Error: 0.0311644504972821

Epoch 200 - Total Error: 0.028406909135682064

Epoch 250 - Total Error: 0.02669256194214157

Epoch 300 - Total Error: 0.02524003405939519

Epoch 350 - Total Error: 0.02384525367109564

Epoch 400 - Total Error: 0.022427464754471382

Epoch 450 - Total Error: 0.020948415886624147

Epoch 500 - Total Error: 0.01936513137210054

Epoch 550 - Total Error: 0.017597424493928912

Epoch 600 - Total Error: 0.01579217059564909

Epoch 650 - Total Error: 0.014194648016335552

Epoch 700 - Total Error: 0.01285558332929372

RESULT: The above python code has been successfully executed.

Applying Backpropagation and genetic algorithm

AIM

To implement genetic algorithm

```
# genetic algorithm search of the one max optimization problem
from numpy.random import rand
# objective function
def onemax(x):
    return -sum(x)
# tournament selection
def selection(pop, scores, k=3):
    # first random selection
    selection_ix = randint(len(pop))
    for ix in randint(0, len(pop), k-1):
        # check if better (e.g. perform a tournament)
        if scores[ix] < scores[selection_ix]:</pre>
```

```
selection ix = ix
       return pop[selection ix]
# crossover two parents to create two children
def crossover(p1, p2, r_cross):
       # children are copies of parents by default
       c1, c2 = p1.copy(), p2.copy()
       # check for recombination
       if rand() < r cross:
               # select crossover point that is not on the end of the string
       pt = randint(1, len(p1)-2)
               # perform crossover
               c1 = p1[:pt] + p2[pt:]
               c2 = p2[:pt] + p1[pt:]
       return [c1, c2]
# mutation operator
def mutation(bitstring, r mut):
       for i in range(len(bitstring)):
               # check for a mutation
               if rand() < r mut:
                       # flip the bit
                       bitstring[i] = 1 - bitstring[i]
# genetic algorithm
def genetic algorithm(objective, n bits, n iter, n pop, r cross, r mut):
       # initial population of random bitstring
       pop = [randint(0, 2, n bits).tolist() for in range(n pop)]
       # keep track of best solution
       best, best eval = 0, objective(pop[0])
       # enumerate generations
       for gen in range(n iter):
               # evaluate all candidates in the population
               scores = [objective(c) for c in pop]
               # check for new best solution
               for i in range(n pop):
                       if scores[i] < best eval:
```

```
best, best eval = pop[i], scores[i]
                               print(">\%d, new best f(\%s) = \%.3f" \% (gen, pop[i], scores[i]))
               # select parents
               selected = [selection(pop, scores) for _ in range(n_pop)]
               # create the next generation
               children = list()
               for i in range(0, n \text{ pop}, 2):
                       # get selected parents in pairs
                       p1, p2 = selected[i], selected[i+1]
                       # crossover and mutation
                       for c in crossover(p1, p2, r cross):
                               # mutation
                               mutation(c, r mut)
                               # store for next generation
                               children.append(c)
               # replace population
               pop = children
       return [best, best eval]
# define the total iterations
n iter = 100
# bits
n bits = 20
# define the population size
n pop = 100
# crossover rate
r cross = 0.9
# mutation rate
r mut = 1.0 / float(n bits)
# perform the genetic algorithm search
best, score = genetic algorithm(onemax, n bits, n iter, n pop, r cross, r mut)
print('Done!')
print('f(\%s) = \%f' \% (best, score))
OUTPUT:
>0, new best f([1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1]) = -13.000
```

RESULT: The above python code has been successfully executed.