# DEPARTMENT OF NETWORKING AND COMMUNICATIONLAB MANUAL

### **ACADEMIC YEAR: 2021-22 EVEN SEMESTER**

Programme (UG/PG) : UG

Semester : VI

Course Code : 18CSC305J

Course Title : Artificial Intelligence Lab

Section : H2

Year : Third

### Submitted by

RAVI MYTRESH(RA1911003011019)

In partial fulfillment of the requirements for the degree of

### **BACHELOR OF TECHNOLOGY**



# FACULTY OF ENGINEERING AND TECHNOLOGYSRM UNIVERSITY

(Under section 3 of UGC Act, 1956) SRM Nagar, Kattankulathur- 603203 Kancheepuram District

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- MOVE UP through international alliances and collaborative initiatives to achieve global excellence.
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To Nurture as a globally recognizable department in imparting students high-quality education and providing high confidence, unique knowledge, and research experience in the field of networking, cyber security, forensics, information technology, cognitive computing, and the internet of things.

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- To provide world-class IT professionals with appropriate industry and research-based curriculum
- To train the students in such a way that leads to entrepreneurship and develops societal need-based industries
- To nourish the students as socially responsible professionals by providing them training in personality development, ethics and leadership program.

<b>Registration Number</b>	

# **LABORATORY RECORD**

Course Code:							
Name of the Course:							
Programme:							
It is certified that this	is a Bonafe	de record o	of the	work	carried	out	by
	_ ofclas	s during the ye	ear 2021	-2022			
Faculty In-Charge		HoD					
Internal Examiner							
Date of the Examination		_					

# LIST OF EXPERIMENTS & SCHEDULE

COL	URSE	COL	E.

# **COURSE TITLE:**

Exp No.	Name of the Experiment	Page No	Date	Signature
1				
2				
3				
4				
5				
6				
7				
8				
9				
10				
11				

Faculty In charge HOD

# HARDWARE AND SOFTWARE REQUIREMENTS

# HARDWARE REQUIREMENTS

- Thin Client
- o Broadband Internet Connection

# SOFTWARE REQUIREMENT

- AWS Login (Cloud9 Service)
- Google Colab

**Prepared By** 

Dr. S. Prabakeran

(Assistant Professor, Department of Networking and Communications)

# MARK SPLIT UP

1	Lab 1: Implementation of toy problems	
2	Lab 2: Developing agent programs for real world problems	CLAP1:5 marks
3	Lab 3: Implementation of constraint satisfaction problems	3 Exp= 2.5 marks Viva= 2.5 marks
4	Lab4: Implementation and Analysis of DFS and BFS for same application	
5	Lab 5: Developing Best first search and A* Algorithm for real world problems	CLAP2:7.5 marks
6	Lab 6: Implementation of uncertain methods for an application (Fuzzy logic/ Dempster Shafer Theory)	4 Exp= 4 marks Hackerrank= 3.5 marks (3 medium- 3 marks + 1 difficult – 0.5
7	Lab 7: Implementation of unification and resolution for real world problems.	mark)
8	Lab 8: Implementation of learning algorithms for an application	CLAP3:7.5 marks
9	Lab 9:Implementation of NLP programs	3 Exp= 5 marks
1 0	Lab 10: Applying deep learning methods to solve an application	Hackerrank= 2.5 marks (2 difficult + 1 Advanced level)
	Course Project:  Problem statement/objective with technical depth: 2 marks  Execution and Github upload: 2 marks  Purpose of the problem statement (societal benefit): 1 mark  Team Members: upto 4 max (expected 20-25 projects per faculty member in a section)	CLAP4: 5 marks

**Experiment No:-1 Date:-04-01-2022** 

Date: 4-01-2022

#### TOY PROBLEM

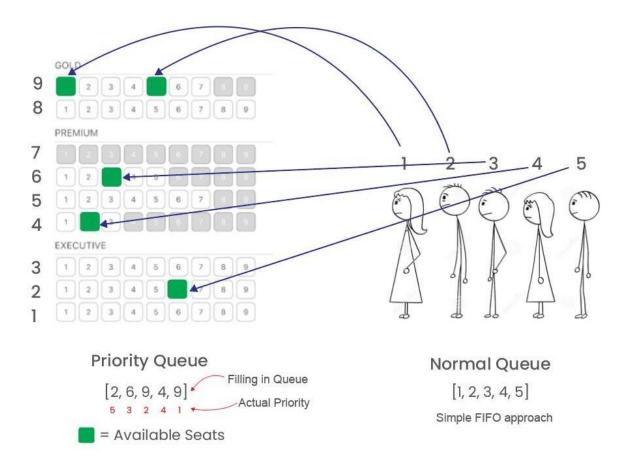
**Problem Statement :** Given an integer N and an array of seats[] where N is the number of people standing in a line to buy a movie ticket and seat[i] is the number of empty seats in the ith row of the movie theater. The task is to find the maximum amount a theater owner can make by selling movie tickets to N people. Price of a ticket is equal to the maximum number of empty seats among all the rows.

#### Algorithm:

- 1. Initialize queue q insert all seats array elements to the queue.
- 2. Tickets sold and the amount generated to be set to 0.
- 3. If tickets sold < N (People in the queue) and q top > 0
- 4. Then remove top element from queue and update total amount
- 5. Repeat step 3 and 4 until tickets sold = number of people in the queue.

**Optimization technique:** This problem can be solved by using a priority queue that will store the count of empty seats for every row and the maximum among them will be available at the top.

- 1. Create an empty priority\_queue q and traverse the seats[] array and insert all elements into the priority\_queue.
- 2. Initialize two integer variable ticketSold = 0 and ans = 0 that will store the number of tickets sold and the total collection of the amount so far.
- 3. Now check while ticketSold < N and q.top() > 0 then remove the top element from the priority\_queue and update ans by adding top element of the priority queue. Also store this top value in a variable temp and insert temp 1 back to the priority\_queue.
- 4. Repeat these steps until all the people have been sold the tickets and print the final result.



# Tool: jupyter notebook

#### **Programming code:**

```
def maxAmount(M, N, seats):
  q = []
           for i in
range(M):
    q.append(seats[i])
ticketSold = 0
                ans =
0
  q.sort(reverse = True)
                         while
(ticketSold < N and q[0] > 0):
     ans = ans + q[0]
temp = q[0]
                 q =
q[1:]
    q.append(temp - 1)
    q.sort(reverse = True)
ticketSold += 1
                  return
ans
```

### Output screen shots:

```
Enter number of rows available : 4
2
3
5
3
[2, 3, 5, 3]
Enter the number of People standing in the queue : 4
Maximum Profit generated = 15
```

**Result :** Successfully found out the maximum amount the theater owner can make by selling movie tickets to N people for a movie.

Experiment No:-1B Date:- -01-2022

# TIC TAC TOE PROBLEM

#### Problem Statement:

Two players, named 'player1' and 'player2', play a tic-tac-toe game on a grid of size '3 x 3'. Given an array 'moves' of size 'n', where each element of the array is a tuple of the form (row, column) representing a position on the grid. Players place their characters alternatively in the sequence of positions given in 'moves'. Consider that 'player1' makes the first move. Your task is to return the winner of the game, i.e., the winning player's name. If there is no winner and some positions remain unmarked, return 'uncertain'. Otherwise, the game ends in a draw, i.e., when all positions are marked without any winner, return 'draw'.

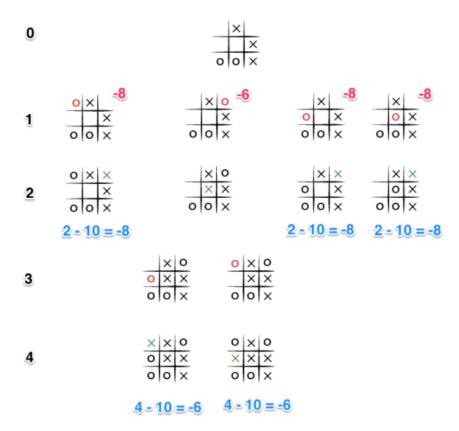
#### Algorithm:

☐ The game is to be played between two people (in this program between HUMAN and
COMPUTER).
☐ One of the player chooses 'O' and the other 'X' to mark their respective cells.
$\Box$ The game starts with one of the players and the game ends when one of the players has one
whole row/ column/ diagonal filled with his/her respective character ('O' or 'X').
$\Box$ If no one wins, then the game is said to be draw.

**Optimization technique:** The key is to use Minimax algorithm .A back and forth between the two players, where the player whose "turn it is" desires to pick the move with the maximum score. In turn, the scores for each of the available moves are determined by the opposing player deciding which of its available moves has the minimum score. And the scores for the opposing players moves are again determined by the turn-taking player trying to maximize its score and so on all the way down the move tree to an end state.

A description for the algorithm, assuming X is the "turn taking player,"

- If the game is over, return the score from X's perspective.
- Otherwise get a list of new game states for every possible move
- Create a scores list
- For each of these states add the minimax result of that state to the scores list
- If it's X's turn, return the maximum score from the scores list
- If it's O's turn, return the minimum score from the scores list



**Tool:** VS Code and Python 3.9.0

#### **Programming code:**

```
import random
import copy as cp
# NOTE use cp.deepcopy() so the temp variable isn't linked with the other

class Cell:
    def __init__(self, position, location, max_val, min_val):
        self.position = position
        self.location = location # NOTE this is a list, [0] is row info and [1] is col info
        self.min_val = min_val
        self.max_val = max_val

def generate_cells(board):
    uboard = cp.deepcopy(board)
    for i in range(len(uboard)):
        for j in range(len(uboard[i])):
            if uboard[i][j] != 'X' and uboard[i][j] != 'O':
```

```
uboard[i][j] = Cell(uboard[i][j], [i,j], 0, 0)
           uboard[i][j].max_val = max_val(board, [i, j])
           uboard[i][j].min_val = min_val(board, [i, j])
           # NOTE convert uboard[i][j] into list of maxval and minval from objects
           uboard[i][j] = [uboard[i][j].position, uboard[i][j].max_val, uboard[i][j].min_val]
   return uboard
 def max_val(board, location): # NOTE only generates one max_val by a given location, not
entire board
   maxval = 0
   if board[location[0]][location[1]] != 'O' and board[location[0]][location[1]] != 'X':
      maxval += check_horizontal(board, location[0], 'max') # need row
      maxval += check_vertical(board, location[1], 'max') # need column
      # NOTE diagonal check is splitted into left and right diagonal for convienence
      maxval += left_diagonal(board, location[0], location[1], 'max')
      maxval += right diagonal(board, location[0], location[1], 'max')
   return maxval
def min_val(board, location):
   minval = 0
   if board[location[0]][location[1]] != 'O' and board[location[0]][location[1]] != 'X':
      minval -= check_horizontal(board, location[0], 'min')
      minval -= check vertical(board, location[1], 'min')
      minval -= left_diagonal(board, location[0], location[1], 'min')
      minval -= right diagonal(board, location[0], location[1], 'min')
   return minval
 def check_horizontal(board, row, u_type):
   opposed = 'X'
   sign = 'O'
   if u type == 'min':
      opposed = 'O'
      sign = 'X'
   \mathbf{v} = \mathbf{0}
   unfilled = 0
   for i in range(3): \# 3 == len(board)'s row
      if board[row][i] != opposed:
        unfilled += 1
        if board[row][i] == sign:
           v += 1
   if unfilled == 3:
      if v == 2:
        v = 10
      else:
```

```
v += 1
  elif unfilled < 3:
     v = 0
  return v
def check_vertical(board, col, u_type):
  opposed = 'X'
  sign = 'O'
  if u_type == 'min':
     opposed = 'O'
     sign = 'X'
  v = 0
  unfilled = 0
  for i in range(3): # 3 == len(board)'s column
    if board[i][col] != opposed:
       unfilled += 1
       if board[i][col] == sign:
          v += 1
  if unfilled == 3:
    if v == 2:
       v = 10
     else:
       v += 1
  elif unfilled < 3:
    v = 0
  return v
def left_diagonal(board, row, col, u_type): # NOTE top_left to bottom_right diagonal check
  opposed = 'X'
  sign = 'O'
  if u_type == 'min':
     opposed = 'O'
     sign = 'X'
  v = 0
  unfilled = 0
  if row == col:
     for i in range(3):
       if board[i][i] != opposed:
          unfilled += 1
          if board[i][i] == sign:
            v += 1
  if unfilled == 3:
    if v == 2:
       v = 10
```

```
else:
       v += 1
  elif unfilled < 3:
     v = 0
  return v
def right_diagonal(board, row, col, u_type):
  opposed = 'X'
  sign = 'O'
  if u_type == 'min':
     opposed = 'O'
     sign = 'X'
  v = 0
  unfilled = 0
  state = False
  for i in range(len(board)):
     if board[i][abs(i-2)] == board[row][col]:
       state = True
     if board[i][abs(i-2)] != opposed:
       unfilled += 1
       if board[i][abs(i-2)] == sign:
          v +=1
  if unfilled == 3 and state == True:
     if v == 2:
       v = 10
     else:
       v += 1
  elif unfilled < 3:
     v = 0
  return v
def dispUboard(uboard):
  print('\n')
  count = 0
  print("Utility Board:\n")
  for i in range(len(uboard)):
     for j in range(len(uboard[i])):
       count += 1
       if uboard[i][j] == 'O' or uboard[i][j] == 'X':
          print(' ',uboard[i][j],end='
       else:
          print(uboard[i][j],end=' ')
       if count\%3 == 0:
          print('\n')
```

```
def checkWin(board, sign):
  if checkHorizontal(board, sign) == True:
     return True
  if checkVertical(board, sign) == True:
     return True
  if checkDiagonal(board, sign) == True:
     return True
  return False
def checkTie(board):
  filled = 0
  for i in range(len(board)):
     for j in range(len(board[i])):
       if board[i][j] == 'O' or board[i][j] == 'X':
          filled += 1
  if filled == 9:
     return True
  return False
def checkDiagonal(board, sign):
  for i in range(len(board)):
     filled = 0
     if board[0][0] == sign:
       for j in range(len(board[i])):
          if board[i][i] == sign:
            filled += 1
     elif board[0][2] == sign:
       for j in range(len(board[i])):
          if board[0+j][2-j] == sign:
            filled += 1
  if filled == 3:
     return True
  return False
def checkHorizontal(board, sign): # NOTE BUGGY so fix it
  for i in range(len(board)):
     if board[i][0] == sign:
       filled = 0
       for j in range(len(board[i])):
          if board[i][j] == sign:
            filled += 1
       if filled == 3:
          return True
  return False
def checkVertical(board, sign):
  for i in range(len(board)):
     if board[0][i] == sign:
```

```
filled = 0
       for j in range(len(board[i])):
          if board[j][i] == sign:
            filled += 1
       if filled == 3:
         return True
  return False
def dispboard(board):
  print('\n')
  count = 0
  print('Tictactoe Board:\n')
  for i in range(len(board)):
     for j in range(len(board[i])):
       count += 1
       print(board[i][j],end=' ')
       if count\%3 == 0:
          print('\n')
def checkCompatible(board, move, sign):
  i = 2
  if move \leq 2:
     i = 0
  elif move \geq 3 and move \leq 5:
    i = 1
  loc = [i,(move-(i*3))]
  if board[loc[0]][loc[1]] == move:
     board[loc[0]][loc[1]] = sign
     return True
     print("Please select an empty spot and try again.")
     return False
def computerDecision(board):
  while (checkTie(board) == False) and (checkWin(board, 'X') == False):
     uboard = generate_cells(board)
     dispUboard(uboard)
     dispboard(board)
     # TODO run minimax algorithm here
     computer_decision = minimax_algorithm(uboard)
     computer_decision = int(computer_decision)
     if checkCompatible(board, computer_decision, 'O') == True:
       if checkTie(board) == True:
          dispboard(board)
```

```
play_again = input("\nThis is a tie game, to play again enter any key, otherwise enter 'q'
to quit.\nYour decision: ")
           if play_again == 'q':
             return
           else:
             board = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
             GameInitializer(board)
        elif checkWin(board, 'O') == True:
           dispboard(board)
           print("The computer won!")
           return
        else:
           playerDecision(board)
      else:
        computerDecision(board)
 def playerDecision(board):
   while (checkTie(board) == False) and (checkWin(board, 'O') == False):
      dispboard(board)
      player decision = input("\n(The player's turn) Enter the empty position you want to place
your 'X': ")
      player_decision = int(player_decision)
      if checkCompatible(board, player_decision, 'X') == True:
        if checkTie(board) == True:
           dispboard(board)
           play_again = input("\nThis is a tie game, if you want to play again enter 'p', to quit
enter any key.\nYour decision: ")
           if play_again == 'q':
             return
           else:
             board = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
             GameInitializer(board)
        elif checkWin(board, 'X') == True:
           dispboard(board)
           print("The player won!")
          return
        else:
           computerDecision(board)
      else:
        playerDecision(board)
 def GameInitializer(board):
   choice = input("\nDo you want to go first or the computer goes first?\nEnter 'c' for computer
first, or 'p' if you would like to go first\nYour Choice: ")
   if choice == 'c':
```

```
computerDecision(board)
   elif choice == 'p':
      playerDecision(board)
   else:
      print("\nPlease enter 'c' or 'p' and try again.")
      GameInitializer(board)
 def minimax_algorithm(ub): # should return a pos, such as 4, not index[1,1]
   optimal = 0
   options = \Pi
   redundant_optimal = [] # This adds the random feature for the computer decision.
   for i in range(len(ub)):
      for j in range(len(ub[i])):
        if ub[i][j] != 'X' and ub[i][j] != 'O':
           # NOTE uboard[i][j's 0 is position, 1 is maxval, 2 is minval
           if ub[i][i][1] >= 10:
             return ub[i][i][0]
           elif ub[i][j][2] <= -10:
             return ub[i][j][0]
           else:
             if abs(ub[i][i][1]) == abs(ub[i][i][2]):
                # NOTE if abs of max = abs of min, add 1 to their sum. Why? because we want to
win more more than limiting the enemy
                options.append([abs(ub[i][j][1]) + abs(ub[i][j][2])+1, ub[i][j][0]])
             else: # NOTE, [0] is the total val of abs(max + min). [1] is the index
                options.append([abs(ub[i][j][1]) + abs(ub[i][j][2]), ub[i][j][0]])
   optimal = max(options) # NOTE for redundant optimal, [0] is index, [1] is val
   for i in range(len(options)):
      if options[i][0] == optimal[0]:
        redundant optimal.append(options[i][1])
   redundant optimal.append(optimal[1])
   randnum = random.randint(0,len(redundant_optimal)-1)
   return redundant_optimal[randnum]
# NOTE play game here
init\_board = [[0, 1, 2],
         [3, 4, 5],
          [6, 7, 8]]
GameInitializer(init board)
```

#### **Output:-**

Do you want to go first or the computer goes first? Enter 'c' for computer first, or 'p' if you would like to go first Your Choice: P

Please enter 'c' or 'p' and try again.

Do you want to go first or the computer goes first? Enter 'c' for computer first, or 'p' if you would like to go first Your Choice: p

#### Tictactoe Board:

- 0 1 2
- 3 4 5
- 6 7 8

(The player's turn) Enter the empty position you want to place your 'X': 4

#### Utility Board:

#### Tictactoe Board:

- 0 1 2
- 3 X 5

6 7 8

Tictactoe Board:

- 0 1 2
- 3 X 5
- 6 7 8

(The player's turn) Enter the empty position you want to place your 'X': 2

Utility Board:

Tictactoe Board:

- 0 1 X
- 3 X 5
- 6 7 8

Tictactoe Board: 0 1 X 3 X 5 0 7 8 (The player's turn) Enter the empty position you want to place your 'X': 3 Utility Board: 0 [1, 0, -2] X X [5, 0, -12] X [7, 2, -2] [8, 2, -2] Tictactoe Board: 0 1 X X X 5 0 7 8 Tictactoe Board: 0 1 X  $\mathbf{x} \quad \mathbf{x} \quad \mathbf{o}$ 0 7 8 (The player's turn) Enter the empty position you want to place your 'X': 8 Utility Board: [1, 0, -2] X Χ X 0 0 [7, 0, -2] X Tictactoe Board:

```
Tictactoe Board:
0 1 X
X \quad X \quad O
0 0 X
(The player's turn) Enter the empty position you want to place your 'X': 1
Tictactoe Board:
o x x
X \quad X \quad O
0 0 X
  0 0 X
  (The player's turn) Enter the empty position you want to place your 'X': 1
  Tictactoe Board:
  0 X X
  X \quad X \quad O
  0 0 X
  This is a tie game, if you want to play again enter 'p', to quit enter any key.
  Your decision:
```

**Result :** The Tic Tac Toe problem was implemented successfully using minmax algorithm to evaluate the best moves with the highest score.

#### **GRAPH COLORING PROBLEM**

**Problem Statement: graph coloring** is a special case of graph labeling; it is an assignment of labels traditionally called "colors" to elements of a graph subject to certain constraints. The problem is, given m colors, find a way of coloring the vertices of a graph such that no two adjacent vertices are colored using same color. The other graph coloring problem is *Edge Coloring* (No vertex is incident to two edges of same color).

#### Algorithm:

- 1.Color first vertex with first color.
- 2. Do following for remaining V-1 vertices.
- a) Consider the currently picked vertex and color it with the lowest numbered color that has not been used on any previously colored vertices adjacent to it.
- b) If all previously used colors appear on vertices adjacent to v, assign a new color to it.

**Optimization technique:** The idea is to assign colors one by one to different vertices, starting from the vertex 0. Before assigning a color, check for safety by considering already assigned colors to the adjacent vertices i.e check if the adjacent vertices have the same color or not. If there is any color assignment that does not violate the conditions, mark the color assignment as part of the solution. If no assignment of color is possible then backtrack and return false.

#### Algorithm:

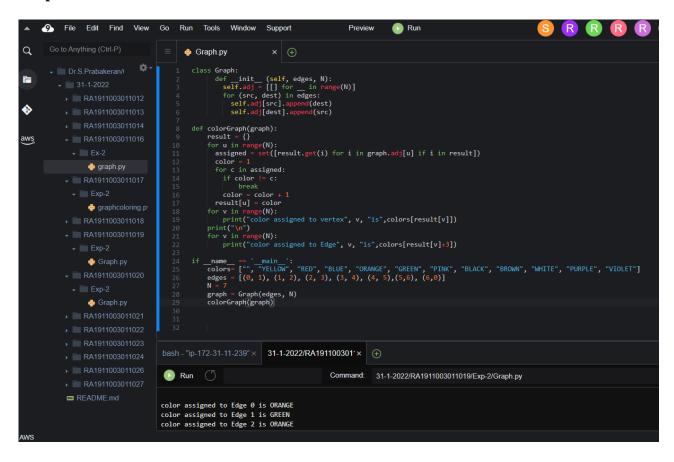
- 1. Create a recursive function that takes the graph, current index, number of vertices, and output color array.
- 2. If the current index is equal to the number of vertices. Print the color configuration in output array.
- 3. Assign a color to a vertex (1 to m).

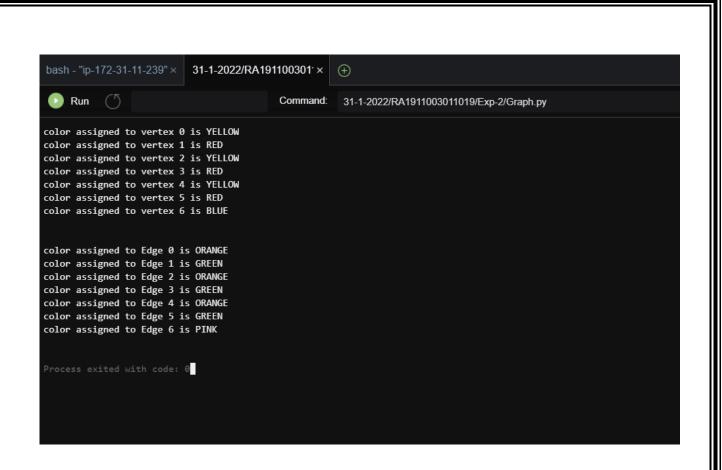
- 4. For every assigned color, check if the configuration is safe, (i.e. check if the adjacent vertices do not have the same color) recursively call the function with next index and number of vertices
- 5. If any recursive function returns true break the loop and return true.
- 6. If no recursive function returns true then return false.

Tool: Cloud9 ide and Python 3.9.0

```
Programming code:
class Graph:
   def _init_ (self, edges, N):
     self.adj = [[] for _ in range(N)]
     for (src, dest) in edges:
      self.adj[src].append(dest)
      self.adj[dest].append(src)
def colorGraph(graph):
  result = \{ \}
  for u in range(N):
   assigned = set([result.get(i) for i in graph.adj[u] if i in result])
   color = 1
   for c in assigned:
    if color != c:
       break
     color = color + 1
   result[u] = color
  for v in range(N):
     print("color assigned to vertex", v, "is",colors[result[v]])
  print("\n")
  for v in range(N):
     print("color assigned to Edge", v, "is",colors[result[v]+3])
if _name_ == '_main_':
  colors= ["", "YELLOW", "RED", "BLUE", "ORANGE", "GREEN", "PINK", "BLACK",
"BROWN", "WHITE", "PURPLE", "VIOLET"]
edges=[(0,1),(1,2),(2,3),(3,4),(4,5),(5,6),(6,0)]
  N = 7
  graph = Graph(edges, N)
  colorGraph(graph)
```

#### **Output screen shots:**





**Result:** A unique color was successfully assigned to each vertex and edge of the graph.

**Experiment No:-3 Date:-01-02-2022** 

#### **Implementation of constraint satisfaction problem (Cryptarithmetic Problem )**

A I	gorithm	•
$\Delta$	201141111	

☐ If a	t, create a list of all the characters that need assigning to pass to Solve I characters are assigned, return true if puzzle is solved, false otherwise erwise, consider the first unassigned character every possible choice among the digits not in use)
if recu	nat choice and then recursively try to assign the rest of the characters sion successful, return true essful, unmake assignment and try another digit

**Optimization technique:** The algorithm above actually has a lot in common with the permutations algorithm, it pretty much just creates all arrangements of the mapping from characters to digits and tries each until one works or all have been successfully tried. For a large puzzle, this could take a while. A smarter algorithm could take into account the structure of the puzzle and avoid going down dead-end paths. For example, if we assign the characters starting from the one's place and moving to the left, at each stage, we can verify the correctness of what we have so far before we continue

onwards. This definitely complicates the code but leads to a tremendous improvement in

• Start by examining the rightmost digit of the topmost row, with a carry of 0

efficiency, making it much more feasible to solve large puzzles.

- If we are beyond the leftmost digit of the puzzle, return true if no carry, false otherwise
- If we are currently trying to assign a char in one of the addends
  If char already assigned, just recur on the row beneath this one, adding value into the sum
  If not assigned, then

- for (every possible choice among the digits not in use)
   make that choice and then on row beneath this one, if successful, return true if !successful, unmake assignment and try another digit
- return false if no assignment worked to trigger backtracking
- Else if trying to assign a char in the sum
- If char assigned & matches correct, recur on next column to the left with carry, if success return true,
- If char assigned & doesn't match, return false
- If char unassigned & correct digit already used, return false
- If char unassigned & correct digit unused, assign it and recur on next column to left with carry, if success return true
- return false to trigger backtracking.

**Tool:** aws cloud9 and Python 3.9.0

#### **Programming code:**

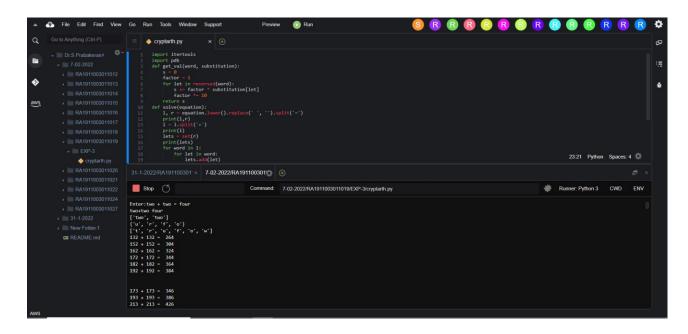
```
import itertools
import pdb

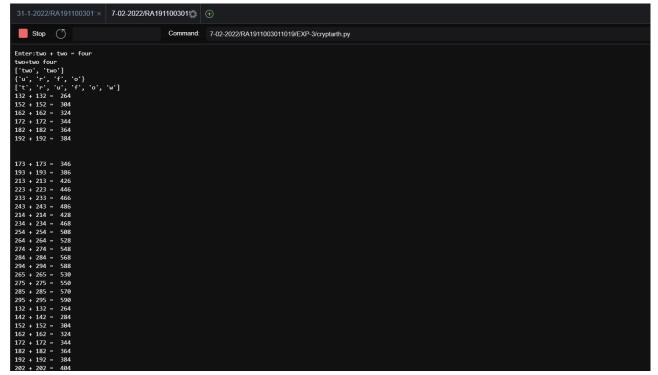
def get_val(word, substitution):
    s = 0
    factor = 1
    for let in reversed(word):
        s += factor * substitution[let]
        factor *= 10
    return s

def solve(equation):
    l, r = equation.lower().replace(' ', ").split('=')
    print(l,r)
```

```
1 = 1.split('+')
  print(l)
  lets = set(r)
  print(lets)
  for word in 1:
     for let in word:
        lets.add(let)
  lets = list(lets)
  print(lets)
  digits = range(20)
  for perm in itertools.permutations(digits, len(lets)):
     sol = dict(zip(lets, perm))
     if sum(get_val(word, sol) for word in l) == get_val(r, sol):
        print('+'.join(str(get\_val(word, sol)) \ for \ word \ in \ l) + " = ",get\_val(r, sol))
equation = input("Enter:")
solve(equation)
```

#### **Output screen shots:**





**Result :** Successfully solved the given constraint satisfaction problem.

#### **Implementation And Analysis Of DFS And BFS**

**Breadth-First Search :** Breadth-First Search (BFS) is an algorithm used for traversing graphs or trees. Traversing means visiting each node of the graph. Breadth-First Search is a recursive algorithm to search all the vertices of a graph or a tree. BFS in python can be implemented by using data structures like a dictionary and lists. Breadth-First Search in tree and graph is almost the same. The only difference is that the graph may contain cycles, so we may traverse to the same node again.

#### Algorithm:

- 1. Start by putting any one of the graph's vertices at the back of the queue.
- 2. Now take the front item of the queue and add it to the visited list.
- 3. Create a list of that vertex's adjacent nodes. Add those which are not within the visited list to the rear of the queue.
- 4. Keep continuing steps two and three till the queue is empty.

**The Depth-First Search:** The Depth-First Search is a recursive algorithm that uses the concept of backtracking. It involves thorough searches of all the nodes by going ahead if potential, else by backtracking. Here, the word backtrack means once you are moving forward and there are not any more nodes along the present path, you progress backward on an equivalent path to seek out nodes to traverse. All the nodes are progressing to be visited on the current path until all the unvisited nodes are traversed after which subsequent paths are going to be selected.

#### **Algorithm:**

- 1. We will start by putting any one of the graph's vertex on top of the stack.
- 2. After that take the top item of the stack and add it to the visited list of the vertex.
- 3. Next, create a list of that adjacent node of the vertex. Add the ones which aren't in the visited list of vertexes to the top of the stack.
- 4. Lastly, keep repeating steps 2 and 3 until the stack is empty.

**Tool:** Aws Cloud9 and Python 3.9.0

#### **Programming code:**

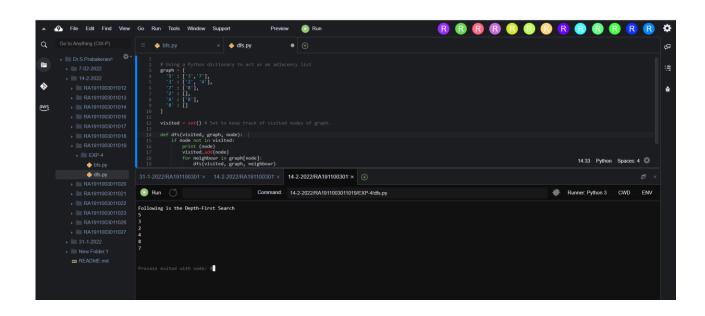
#### **BFS**:

```
graph = {
'5': ['3','7'],
'3' : ['2', '4'],
'7' : ['8'],
'2' : [],
'4' : ['8'],
'8' : []
visited = [] # List for visited nodes.
queue = [] #Initialize a queue
defbfs(visited, graph, node): #function for BFS
  visited.append(node)
  queue.append(node)
                      # Creating loop to visit each node
while queue:
    m = queue.pop(0)
print (m, end =" ")
for neighbour in graph[m]:
if neighbour notin visited:
        visited.append(neighbour)
        queue.append(neighbour)
# Driver Code
print("Following is the Breadth-First Search")
bfs(visited, graph, '5') # function calling
```

#### **DFS**:

```
# Using a Python dictionary to act as an adjacency list
graph = {
'5' : ['3','7'],
'3' : ['2', '4'],
'7' : ['8'],
'2' : [],
'4' : ['8'],
'8' : []
visited =set() # Set to keep track of visited nodes of graph.
defdfs(visited, graph, node): #function for dfs
if node notin visited:
print (node)
        visited.add(node)
for neighbour in graph[node]:
            dfs(visited, graph, neighbour)
# Driver Code
print("Following is the Depth-First Search")
dfs(visited, graph, '5')
```

#### **Output screen shots:**



**Result :** Successfully Implemented BFS and DFS.

**Experiment No:-5 Date:-15-02-2022** 

## **Best First Search (Informed Search)**

In BFS and DFS, when we are at a node, we can consider any of the adjacent as next node. So both BFS and DFS blindly explore paths without considering any cost function. The idea of Best First Search is to use an evaluation function to decide which adjacent is most promising and then explore. Best First Search falls under the category of Heuristic Search or Informed Search.

We use a priority queue to store costs of nodes. So the implementation is a variation of BFS, we just need to change Queue to Priority Queue.

#### Algorithm:

```
1) Create an empty PriorityQueue
PriorityQueue pq;
2) Insert "start" in pq.
pq.insert(start)
3) Until PriorityQueue is empty
u = PriorityQueue.DeleteMin
If u is the goal
Exit
Else
Foreach neighbor v of u
If v "Unvisited"
Mark v "Visited"
pq.insert(v)
Mark u "Examined"
End procedure
```

#### A\* Algorithm

A heuristic algorithm sacrifices optimality, with precision and accuracy for speed, to solve problems faster and more efficiently.

All graphs have different nodes or points which the algorithm has to take, to reach the final node. The paths between these nodes all have a numerical value, which is considered as the weight of the path. The total of all paths transverse gives you the cost of that route.

Initially, the Algorithm calculates the cost to all its immediate neighboring nodes,n, and chooses the one incurring the least cost. This process repeats until no new nodes can be 34

chosen and all paths have been traversed. Then, you should consider the best path among them. If f(n) represents the final cost, then it can be denoted as:

$$f(n) = g(n) + h(n)$$
, where:

g(n) = cost of traversing from one node to another. This will vary from node to node

h(n) = heuristic approximation of the node's value. This is not a real value but an approximation cost

#### **Algorithm**

- Make an open list containing starting node o If it reaches the destination node : o Make a closed empty list
  - o If it does not reach the destination node, then consider a node with the lowest f-score in the open list

We are finished

• Else:

Put the current node in the list and check its neighbors

- For each neighbor of the current node:
  - o If the neighbor has a lower g value than the current node and is in the closed list:

Replace neighbor with this new node as the neighbor's parent

• Else If (current g is lower and neighbor is in the open list):

Replace neighbor with the lower g value and change the neighbor's parent to the current node.

• Else If the neighbor is not in both lists:

Add it to the open list and set its g

**Tool:** VS Code and Python 3.9.0

#### **Programming code:**

self.parent = parent

#### A-star

```
# graph class class Graph:
  # init class
                def __init__(self, graph_dict=None,
directed=True):
     self.graph_dict = graph_dict or { }
     self.directed = directed
not directed:
       self.make_undirected()
  # create undirected graph by adding symmetric edges
make_undirected(self):
     for a in list(self.graph_dict.keys()):
                                                 for (b,
dist) in self.graph_dict[a].items():
self.graph_dict.setdefault(b, { })[a] = dist
  # add link from A and B of given distance, and also add the inverse link if the graph is undirected
def connect(self, A, B, distance=1):
     self.graph_dict.setdefault(A, { })[B] = distance
                                                          if
not self.directed:
       self.graph\_dict.setdefault(B, {})[A] = distance
  # get neighbors or a neighbor
get(self, a, b=None):
     links = self.graph_dict.setdefault(a, { })
if b is None:
                     return links
                                      else:
       return links.get(b)
  # return list of nodes in the graph
                                     def nodes(self):
s1 = set([k for k in self.graph_dict.keys()])
set([k2 for v in self.graph_dict.values() for k2, v2 in
                nodes = s1.union(s2)
v.items()])
     return list(nodes)
# node class class
Node:
  # init class
                def init (self, name:str,
parent:str):
                 self.name = name
```

self.g = 0 # distance to

```
start node
               self.h = 0 # distance to goal
node
    self.f = 0 # total cost
  # compare nodes
  _eq__(self, other):
    return self.name == other.name
  # sort nodes
                 def
 lt (self, other):
                        return
self.f < other.f
  # print node
                 def
__repr__(self):
    return ((\{0\},\{1\}))'.format(self.name, self.f))
# A* search def astar_search(graph, heuristics, start,
end):
  # lists for open nodes and closed nodes
  open = []
  closed = []
  # a start node and an goal node start_node =
                    goal_node = Node(end, None)
Node(start, None)
  # add start node
  open.append(start_node)
  # loop until the open list is empty
                                      while
len(open) > 0:
    open.sort()
                                    # sort open list to get the node with the lowest cost first
                                            # get node with the lowest cost
    current\_node = open.pop(0)
closed.append(current_node)
                                       # add current node to the closed list
     # check if we have reached the goal, return the path
                                                              if current node
== goal node:
                      path = []
                                       while current node != start node:
path.append(current_node.name + ': ' + str(current_node.g))
current node = current node.parent
       path.append(start_node.name + ': ' + str(start_node.g))
                                                                     return
path[::-1]
     neighbors = graph.get(current_node.name) # get neighbours
    # loop neighbors
                           for key, value in
neighbors.items():
       neighbor = Node(key, current node) # create neighbor node
                                                                              if(neighbor in
                  # check if the neighbor is in the closed list
closed):
                                                                       continue
       # calculate full path cost
```

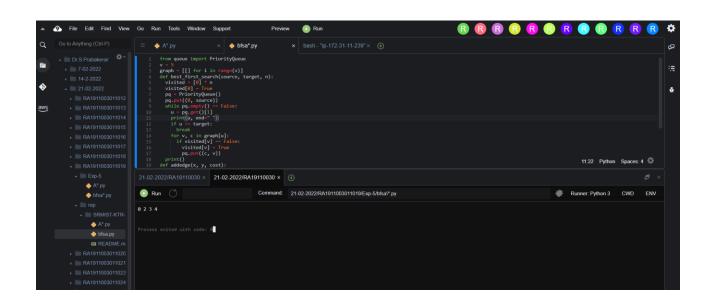
```
neighbor.g = current_node.g + graph.get(current_node.name, neighbor.name)
neighbor.h = heuristics.get(neighbor.name)
                                                   neighbor.f = neighbor.g + neighbor.h
       # check if neighbor is in open list and if it has a lower f value
if(add_to_open(open, neighbor) == True):
         # everything is green, add neighbor to open list
                                                                    open.append(neighbor)
  # return None, no path is found
None
# check if a neighbor should be added to open list def
add_to_open(open, neighbor): for node in open:
                                                        if
(neighbor == node and neighbor.f > node.f):
       return False return
True
# create a graph graph =
Graph() # create graph
connections (Actual
distance)
graph.connect('S', 'A', 2) graph.connect('S', 'G',
20) graph.connect('A', 'C', 7) graph.connect('C',
'G', 8) graph.connect('C', 'D', 9)
graph.connect('D', 'G', 10)
# make graph undirected, create symmetric connections graph.make_undirected()
# create heuristics (straight-line distance, air-travel distance)
heuristics = {} heuristics['A'] = 5 heuristics['C'] = 8 heuristics['G'] =
7 heuristics ['D'] = 6 heuristics ['S'] = 9 # run the search algorithm path
= astar_search(graph, heuristics, 'S', 'G') print("Path:", path)
```

# **Best First Search**

```
print(u, end=" ")  if u
== target:
    break    for v, c in graph[u]:
if visited[v] == False:
visited[v] = True
pq.put((c, v))    print() def
addedge(x, y, cost):
graph[x].append((y, cost))
graph[y].append((x, cost))
addedge(0, 1, 5) addedge(0, 2,
1) addedge(2, 3, 2) addedge(1,
4, 1) addedge(3, 4, 2) source =
0 target = 4
best_first_search(source, target, v)
```

## Output screen shots:

#### **BFSA:**



**Result :** A\* and Best first search algorithms were implemented successfully.

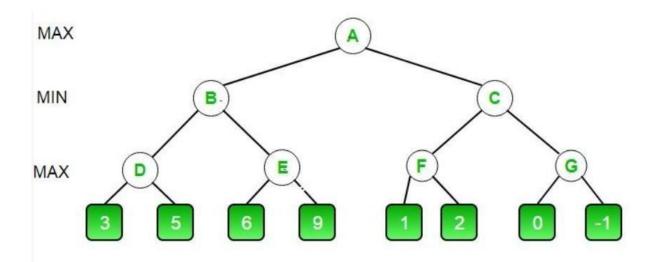
**Experiment No: 6** 

Date: 22-02-2022

### MINIMAX ALGORITHM

**AIM**: Developing a mini max algorithm for real world problems.

**PROBLEM**: Find the optimal value in the given tree of integer values in the most optimal way possible under the time complexity O(B^D).



#### **ALGORITHM MINIMAX APPROACH:**

- 1. Start traversing the given tree in top to bottom manner.
- 2. If node is a leaf node then return the value of the node.
- 3. If isMaximizingPlayer exist then bestVal = -INFINITY
- 4. For each child node, value = minimax(node, depth+1, false, alpha, beta)
- 5. bestVal = max(bestVal, value) and alpha = max(alpha, bestVal)
- 6. If beta <= alpha then stop traversing and return bestVal
- 7. Else, bestVal = +INFINITY
- 8. For each child node, value = minimax(node, depth+1, true, alpha, beta)
- 9. bestVal = min(bestVal, value) and beta = min(beta, bestVal)

10. if beta <= alpha the stop traversing and return bestVal

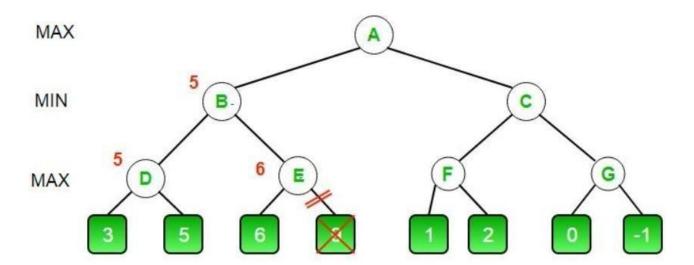
## **OPTIMIZATION TECHNIQUE:**

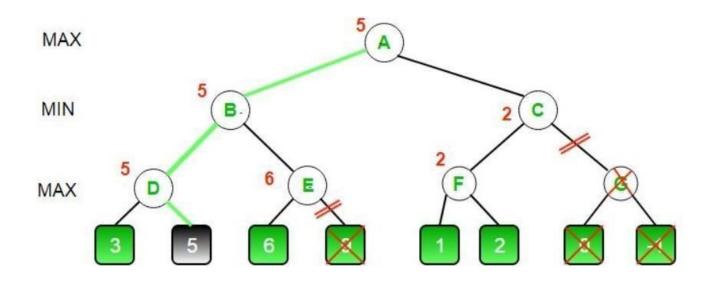
Alpha-Beta pruning is not actually a new algorithm, rather an optimization technique for minimax algorithms. It reduces the computation time by a huge factor. This allows us to search much faster and even go into deeper levels in the game tree. It cuts off branches in the game tree which need not be searched because there already exists a better move available. It is called Alpha-Beta pruning because it passes 2 extra parameters in the minimax function, namely alpha and beta.

Let's define the parameters alpha and beta.

**Alpha** is the best value that the **maximizer** currently can guarantee at that level or above.

Beta is the best value that the minimizer currently can guarantee at that level or above.





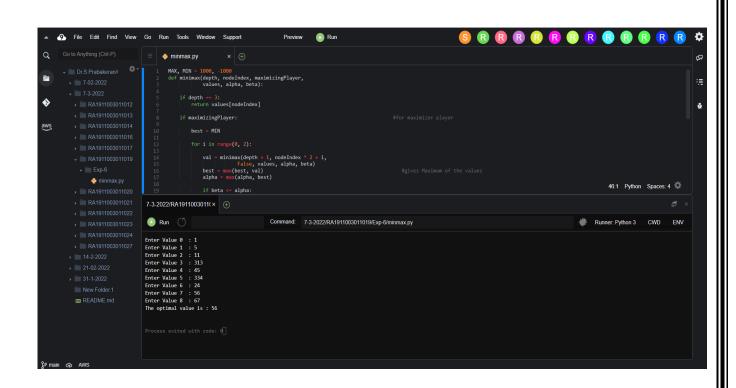
### **CODE (MINIMAX ALGORITHM):**

```
MAX, MIN = 1000, -1000 def
minimax(depth, nodeIndex,
maximizingPlayer,
values, alpha, beta):
       if depth == 3:
               return values[nodeIndex]
       if
maximizingPlay
er:
               best = MIN
               for i
in range(0, 2):
                       val = minimax(depth + 1, nodeIndex * 2 + i,
                                               False, values,
alpha, beta)
                               best = max(best, val)
                       alpha = max(alpha, best)
                       if
beta <= alpha:
                               break
               return best
```

else:

```
best =
MAX
                for i
in range(0, 2):
        val =
minimax(depth + 1,
nodeIndex * 2 + i,
                                                        True,
values, alpha, beta)
                                        best = min(best, val)
        beta = min(beta, best)
                                                if beta <= alpha:
                        break
                return best
if __name__ == "__main__":
  values
=[]
for i in
range(0,
8):
    x = int(input(f"Enter Value {i} : "))
values.append(x)
  print ("The optimal value is:", minimax(0, 0, True, values, MIN,
MAX))
```

**OUTPUT:** 



**RESULT:** The Optimal value of the given tree successfully found using Minimax Algorithm with Alpha Beta Pruning in time complexity O(B^D).

**Experiment No: 7** 

Date: 14-03-2022

#### IMPLEMENTATION OF UNCERTAIN METHODS OF AN APPLICATION

#### **Problem Statement:**

To implement Fuzzy logic using matplotlib in python and find the graph of temperature, humidity and speed in different conditions.

#### Algorithm:

- 1. Locate the input, output, and state variables of the plane under consideration.
- 2. Split the complete universe of discourse spanned by each variable into a number of fuzzy subsets, assigning each with a linguistic label. The subsets include all the elements in the universe.
- 3. Obtain the membership function for each fuzzy subset.
- 4. Assign the fuzzy relationships between the inputs or states of fuzzy subsets on one side and the output of fuzzy subsets on the other side, thereby forming the rule base.
- 5. Choose appropriate scaling factors for the input and output variables for normalizing the variables between [0, 1] and [-1, I] interval.
- 6. Carry out the fuzzification process.
- 7. Identify the output contributed from each rule using fuzzy approximate reasoning.
- 8. Combine the fuzzy outputs obtained from each rule.
- 9. Finally, apply defuzzification to form a crisp output.

#### **Optimization Technique:**

- 1. Decomposing the large-scale system into a collection of various subsystems.
- 2. Varying the plant dynamics slowly and linearizing the nonlinear plane dynamics about a set of operating points.
- 3. Organizing a set of state variables, control variables, or output features for the system under consideration.
- 4. Designing simple P, PD, PID controllers for the subsystems. Optimal controllers can also be designed.

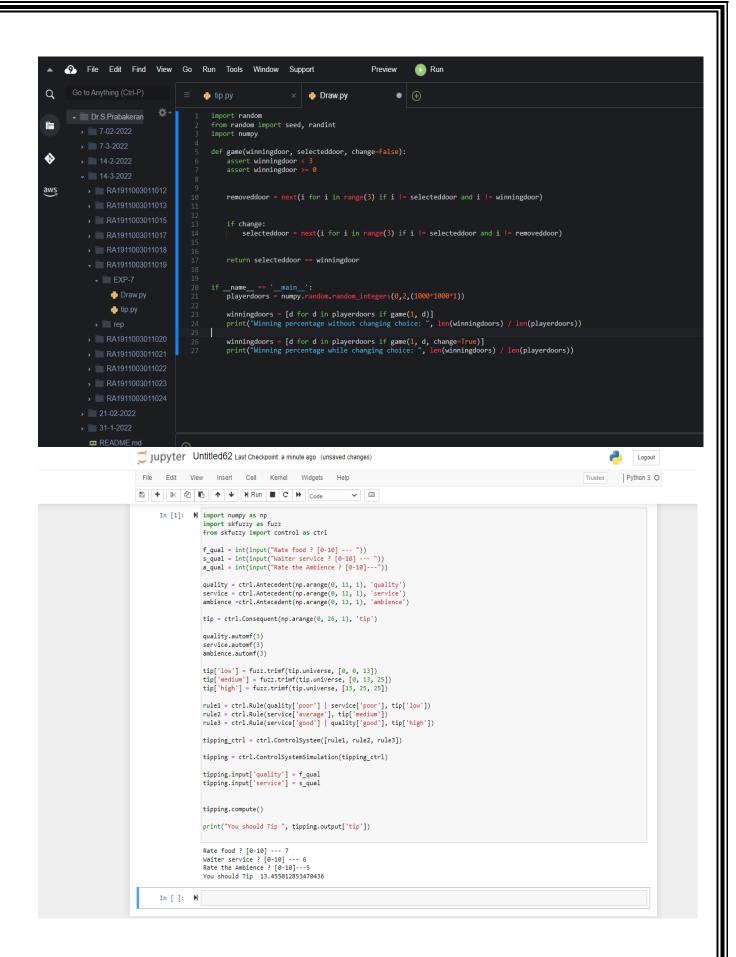
**Uncertainty In this problem :** Fuzzy Logic - Temperature, Humidity.

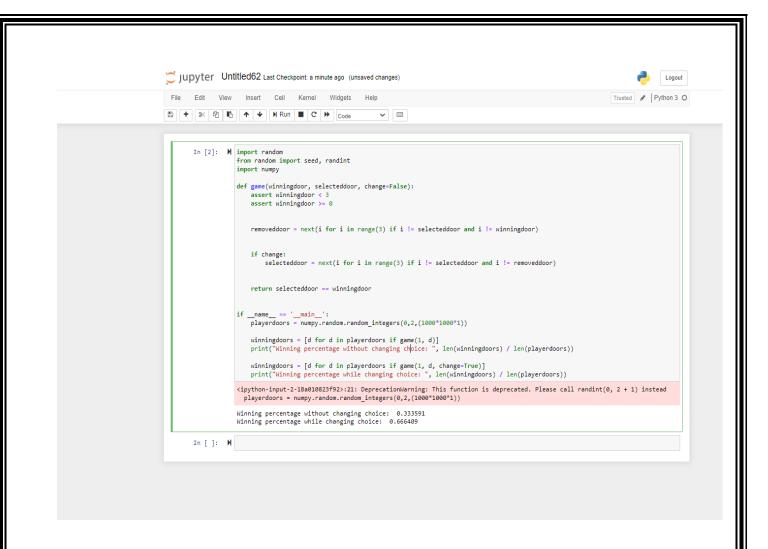
```
CODE: M = 9
def puzzle(a):
for i in range(M):
for j in range(M):
print(a[i][j],end =
    print() def
solve(grid, row, col,
num):
  for x in
range(9):
              if
grid[row][x] ==
num:
       return False
  for x in
              if
range(9):
grid[x][col] ==
num:
       return False
  startRow = row - row \% 3
                              startCol
= col - col % 3 for i in range(3):
for j in range(3):
                        if grid[i+
startRow][j + startCol] == num:
         return
False
        return
True
def Suduko(grid, row, col):
  if (row == M - 1 and col == M):
    return True if col ==
M:
        row += 1
                      col =
0 if grid[row][col] > 0:
return Suduko(grid, row, col
+1) for num in range(1,
M + 1, 1):
    if solve(grid, row, col, num):
       grid[row][col] =
num
Suduko(grid, row, col + 1):
         return
True
grid[row][col] = 0
return False
"0 means the cells where no value is assigned"
```

```
grid = [[2, 5, 0, 0, 3, 0, 9, 0, 1],
        [0, 1, 0, 0, 0, 4, 0, 0, 0],
        [4, 0, 7, 0, 0, 0, 2, 0, 8],
        [0, 0, 5, 2, 0, 0, 0, 0, 0],
        [0, 4, 0, 0, 0, 3, 0, 0, 0],
        [0, 0, 0, 3, 6, 0, 0, 7, 2],
        [0, 7, 0, 0, 0, 0, 0, 0, 3],
        [9, 0, 3, 0, 0, 0, 6, 0, 4]]

if
(Suduko(grid, 0, 0, 0)):
puzzle(grid)) else:
    print("Solution does not exist:(")
```

#### **OUTPUT:**





**Result:** We have successfully implemented fuzzy uncertainty problem using matplotlib and output is received

Experiment No:-8 Date:-08-03-2022

#### **UNIFICATION AND RESOLUTION**

**<u>AIM:</u>** Developing an algorithm for implementation of **UNIFICATION AND RESOLUTION** 

**PROBLEM STATEMENT:** Developing an optimized technique using an appropriate artificial intelligence algorithm to solve the Unification and Resolution.

### **ALGORITHM:**

- 1. function PL-RESOLUTION (KB, Q) returns true or false inputs: KB,
- 2. the knowledge base, group of sentences/facts in propositional logic
- 3. Q, the query, a sentence in propositional logic
- 4. clauses  $\rightarrow$  the set of clauses in the CNF representation of KB  $^{\land}$  Q new  $\rightarrow$  {}
- 5. loop do for each Ci, Cj in clauses do
- 6. resolvents  $\rightarrow$  PL-RESOLVE (Ci, Cj)
- 7. if resolvents contains the empty clause the return true
- 8. new  $\rightarrow$  new union resolvents
- 9. if new is a subset of clauses then return false
- 10. clauses  $\rightarrow$  clauses union true

#### **OPTIMIZATION TECHNIQUE:**

Resolution basically works by using the principle of proof by contradiction. To find the conclusion we should negate the conclusion. Then the resolution rule is applied to the resulting clauses. Each clause that contains complementary literals is resolved to produce a2. new clause, which can be added to the set of facts (if it is not already present). This process continues until one of the two things happen:•There are no new clauses that can be added. An application of the resolution rule derives the empty clauseAn empty clause shows that the negation of the conclusion is a complete contradiction,hence the negation of the conclusion is invalid or false or the assertion is completely valid or true.

- 1. Convert the given statements in Predicate/Propositional Logic
- 2. Convert these statements into Conjunctive Normal Form
- 3. Negate the Conclusion (Proof by Contradiction)
- 4. Resolve using a Resolution Tree (Unification)

#### **UNIFICATION CODE:**

```
def get_index_comma(string):
  index_list = list()
  par_count = 0
```

```
for i in range(len(string)):
     if string[i] == ',' and par_count == 0:
       index_list.append(i)
     elif string[i] == '(':
       par_count += 1
     elif string[i] == ')':
       par_count -= 1
  return index_list
def is_variable(expr):
  for i in expr:
     if i == '(' or i == ')':
       return False
  return True
def process_expression(expr):
  expr = expr.replace(' ', ")
  index = None
  for i in range(len(expr)):
     if expr[i] == '(':
       index = i
       break
  predicate_symbol = expr[:index]
  expr = expr.replace(predicate_symbol, ")
  expr = expr[1:len(expr) - 1]
  arg_list = list()
  indices = get_index_comma(expr)
  if len(indices) == 0:
     arg_list.append(expr)
  else:
     arg_list.append(expr[:indices[0]])
     for i, j in zip(indices, indices[1:]):
       arg list.append(expr[i + 1:j])
     arg_list.append(expr[indices[len(indices) - 1] + 1:])
  return predicate_symbol, arg_list
def get_arg_list(expr):
  _, arg_list = process_expression(expr)
  flag = True
  while flag:
     flag = False
```

```
for i in arg_list:
       if not is_variable(i):
          flag = True
          _, tmp = process_expression(i)
          for j in tmp:
            if i not in arg_list:
               arg_list.append(j)
          arg_list.remove(i)
  return arg_list
def check_occurs(var, expr):
  arg_list = get_arg_list(expr)
  if var in arg_list:
     return True
  return False
def unify(expr1, expr2):
  if is_variable(expr1) and is_variable(expr2):
     if expr1 == expr2:
       return 'Null'
     else:
       return False
  elif is_variable(expr1) and not is_variable(expr2):
     if check_occurs(expr1, expr2):
       return False
     else:
       tmp = str(expr2) + '/' + str(expr1)
       return tmp
  elif not is_variable(expr1) and is_variable(expr2):
     if check occurs(expr2, expr1):
       return False
     else:
       tmp = str(expr1) + '/' + str(expr2)
       return tmp
  else:
     predicate_symbol_1, arg_list_1 = process_expression(expr1)
     predicate_symbol_2, arg_list_2 = process_expression(expr2)
     # Step 2
     if predicate_symbol_1 != predicate_symbol_2:
       return False
     # Step 3
     elif len(arg_list_1) != len(arg_list_2):
```

```
return False
     else:
       # Step 4: Create substitution list
        sub_list = list()
       # Step 5:
       for i in range(len(arg_list_1)):
          tmp = unify(arg_list_1[i], arg_list_2[i])
          if not tmp:
             return False
          elif tmp == 'Null':
             pass
          else:
             if type(tmp) == list:
                for j in tmp:
                  sub_list.append(j)
             else:
                sub_list.append(tmp)
        # Step 6
        return sub_list
if __name__ == '__main__':
  f1 = 'Q(a, g(x, a), f(y))'
  f2 = 'Q(a, g(f(b), a), x)'
  # f1 = input('f1 : ')
  # f2 = input('f2 : ')
  result = unify(f1, f2)
  if not result:
     print('The process of Unification failed!')
  else:
     print('The process of Unification successful!')
     print(result)
```

### **RESOLUTION CODE:**

```
import copy
import time

class Parameter:
   variable_count = 1

   def __init__(self, name=None):
```

```
if name:
       self.type = "Constant"
       self.name = name
     else:
       self.type = "Variable"
       self.name = "v" + str(Parameter.variable_count)
       Parameter.variable_count += 1
  def isConstant(self):
     return self.type == "Constant"
  def unify(self, type_, name):
     self.type = type_
     self.name = name
  def __eq__(self, other):
     return self.name == other.name
  def __str__(self):
     return self.name
class Predicate:
  def __init__(self, name, params):
     self.name = name
     self.params = params
  def eq (self, other):
     return self.name == other.name and all(a == b for a, b in zip(self.params, other.params))
  def __str__(self):
    return self.name + "(" + ",".join(str(x) for x in self.params) + ")"
  def getNegatedPredicate(self):
     return Predicate(negatePredicate(self.name), self.params)
class Sentence:
  sentence\_count = 0
  def __init__(self, string):
     self.sentence_index = Sentence.sentence_count
     Sentence_count += 1
     self.predicates = []
     self.variable_map = { }
     local = \{ \}
     for predicate in string.split("|"):
       name = predicate[:predicate.find("(")]
```

```
params = []
       for param in predicate[predicate.find("(") + 1: predicate.find(")")].split(","):
         if param[0].islower():
            if param not in local: # Variable
              local[param] = Parameter()
               self.variable_map[local[param].name] = local[param]
            new_param = local[param]
          else:
            new_param = Parameter(param)
            self.variable_map[param] = new_param
          params.append(new_param)
       self.predicates.append(Predicate(name, params))
  def getPredicates(self):
     return [predicate.name for predicate in self.predicates]
  def findPredicates(self, name):
     return [predicate for predicate in self.predicates if predicate.name == name]
  def removePredicate(self, predicate):
     self.predicates.remove(predicate)
     for key, val in self.variable_map.items():
       if not val:
          self.variable_map.pop(key)
  def contains Variable (self):
     return any(not param.isConstant() for param in self.variable_map.values())
  def __eq__(self, other):
     if len(self.predicates) == 1 and self.predicates[0] == other:
       return True
     return False
  def __str__(self):
     return "".join([str(predicate) for predicate in self.predicates])
class KB:
  def __init__(self, inputSentences):
     self.inputSentences = [x.replace(" ", "") for x in inputSentences]
     self.sentences = []
     self.sentence_map = {}
  def prepareKB(self):
     self.convertSentencesToCNF()
     for sentence_string in self.inputSentences:
```

```
sentence = Sentence(sentence string)
    for predicate in sentence.getPredicates():
       self.sentence_map[predicate] = self.sentence_map.get(
         predicate, []) + [sentence]
def convertSentencesToCNF(self):
  for sentenceIdx in range(len(self.inputSentences)):
    # Do negation of the Premise and add them as literal
    if "=>" in self.inputSentences[sentenceIdx]:
       self.inputSentences[sentenceIdx] = negateAntecedent(
         self.inputSentences[sentenceIdx])
def askQueries(self, queryList):
  results = []
  for query in queryList:
     negatedQuery = Sentence(negatePredicate(query.replace(" ", "")))
     negatedPredicate = negatedQuery.predicates[0]
    prev sentence map = copy.deepcopy(self.sentence map)
    self.sentence_map[negatedPredicate.name] = self.sentence_map.get(
       negatedPredicate.name, []) + [negatedQuery]
     self.timeLimit = time.time() + 40
    try:
       result = self.resolve([negatedPredicate], [
                    False * (len(self.inputSentences) + 1))
    except:
       result = False
    self.sentence_map = prev_sentence_map
    if result:
       results.append("TRUE")
    else:
       results.append("FALSE")
  return results
def resolve(self, queryStack, visited, depth=0):
  if time.time() > self.timeLimit:
    raise Exception
  if queryStack:
    query = queryStack.pop(-1)
    negatedQuery = query.getNegatedPredicate()
     queryPredicateName = negatedQuery.name
    if queryPredicateName not in self.sentence map:
       return False
    else:
       queryPredicate = negatedQuery
```

```
if not visited[kb_sentence.sentence_index]:
              for kbPredicate in kb_sentence.findPredicates(queryPredicateName):
                canUnify, substitution = performUnification(
                   copy.deepcopy(queryPredicate), copy.deepcopy(kbPredicate))
                if canUnify:
                   newSentence = copy.deepcopy(kb_sentence)
                   newSentence.removePredicate(kbPredicate)
                   newQueryStack = copy.deepcopy(queryStack)
                   if substitution:
                     for old, new in substitution.items():
                        if old in newSentence.variable map:
                          parameter = newSentence.variable_map[old]
                          newSentence.variable_map.pop(old)
                          parameter.unify(
                             "Variable" if new[0].islower() else "Constant", new)
                          newSentence.variable_map[new] = parameter
                     for predicate in newQueryStack:
                        for index, param in enumerate(predicate.params):
                          if param.name in substitution:
                             new = substitution[param.name]
                             predicate.params[index].unify(
                               "Variable" if new[0].islower() else "Constant", new)
                   for predicate in newSentence.predicates:
                     newQueryStack.append(predicate)
                   new_visited = copy.deepcopy(visited)
                   if kb_sentence.containsVariable() and len(kb_sentence.predicates) > 1:
                     new visited[kb sentence.sentence index] = True
                   if self.resolve(newQueryStack, new visited, depth + 1):
                     return True
         return False
    return True
def performUnification(queryPredicate, kbPredicate):
  substitution = {}
  if queryPredicate == kbPredicate:
     return True, {}
     for query, kb in zip(queryPredicate.params, kbPredicate.params):
       if query == kb:
         continue
```

for kb sentence in self.sentence map[queryPredicateName]:

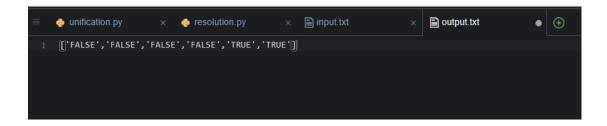
```
if kb.isConstant():
          if not query.isConstant():
            if query.name not in substitution:
               substitution[query.name] = kb.name
            elif substitution[query.name] != kb.name:
               return False, {}
            query.unify("Constant", kb.name)
          else:
            return False, {}
       else:
          if not query.isConstant():
            if kb.name not in substitution:
               substitution[kb.name] = query.name
            elif substitution[kb.name] != query.name:
               return False, {}
            kb.unify("Variable", query.name)
          else:
            if kb.name not in substitution:
               substitution[kb.name] = query.name
            elif substitution[kb.name] != query.name:
               return False, {}
  return True, substitution
def negatePredicate(predicate):
  return predicate[1:] if predicate[0] == "\sim" else "\sim" + predicate
def negateAntecedent(sentence):
  antecedent = sentence[:sentence.find("=>")]
  premise = []
  for predicate in antecedent.split("&"):
     premise.append(negatePredicate(predicate))
  premise.append(sentence[sentence.find("=>") + 2:])
  return "|".join(premise)
def getInput(filename):
  with open(filename, "r") as file:
     noOfQueries = int(file.readline().strip())
     inputQueries = [file.readline().strip() for _ in range(noOfQueries)]
     noOfSentences = int(file.readline().strip())
     inputSentences = [file.readline().strip()
                for _ in range(noOfSentences)]
     return inputQueries, inputSentences
```

```
def printOutput(filename, results):
    print(results)
    with open(filename, "w") as file:
        for line in results:
            file.write(line)
            file.write("\n")
        file.close()

if __name__ == '__main__':
    inputQueries_, inputSentences_ = getInput('input.txt')
    knowledgeBase = KB(inputSentences_)
    knowledgeBase.prepareKB()
    results_ = knowledgeBase.askQueries(inputQueries_)
    printOutput("output.txt", results_)
```

#### **UNIFICATION OUTPUT:**

## **RESOLUTION OUTPUT:**



## **RESULT:**

Developed Unification and Resolution Algorithm in Python for solving logical problems.

**Experiment No:-9 Date:-21-03-2022** 

## <u>Implementation of learning algorithms for an application</u>

#### Aim:

- a) Implementation of Linear Regression algorithm to predict students score using the given dataset.
- b) Implementation of Support Vector Classification algorithm to classify the cases of breast cancer

using the given dataset.

c) Implementation of K-means clustering algorithm to group the customers based on their demographic detail using the given dataset.

import pandas as pd import numpy as np import matplotlib.pyplot as plt from sklearn.model\_selection import train\_test\_split from sklearn.linear model import LinearRegression from sklearn import metrics %matplotlib inline Import required modules and packages dataset = pd.read\_csv('....\student\_scores.csv') dataset.head() Import data set Choose the right path for the dataset dataset.describe() Descriptive statistics of the attributes available in the dataset dataset.plot(x='Hours', y='Scores', style='o') plt.title('Hours vs Percentage') plt.xlabel('Hours Studied') plt.ylabel('Percentage Score')

```
plt.show()
Visualize the data.
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, 1].values
Identify the independent (X) and
dependent variables (y) in the data set
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=0)
print('X train shape: ', X_train.shape)
print('Y train shape: ', Y_train.shape)
print('X test shape: ', X_test.shape)
Splitting the given data in to training set
(80%) and testing set (20%)
Beginners Level
Lab 11 - Implementation of Learning Algorithms for an Application
18CSC305J - ARTIFICIAL INTELLIGENCE Page 4
print('Y test shape: ', Y_test.shape)
regressor = LinearRegression()
Model instantiation
regressor.fit(X_train, y_train) Model Training
print(regressor.intercept_)
print(regressor.coef )
Finding out the coefficient (a) and
intercept (b) value of linear model
(y=aX+b)
y_pred = regressor.predict(X_test)
df = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})
print(df)
Testing the model
print('Mean Absolute Error:',
metrics.mean absolute error (y test, y pred))
print('Mean Squared Error:',
metrics.mean_squared_error (y_test, y_pred))
print('Root Mean Squared Error:',
np.sqrt(metrics.mean squared error (y test, y pred)))
MAE, MSE, RMSE – Evaluation metrics
of Model
Discussion:
```

```
[ ] dataset = pd.read_csv('C:\\Users\DELL\Desktop\student_scores.csv')
    dataset.shape()
```

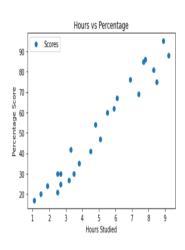
	Hours	Scores
0	2.5	21
1	5.1	47
2	3.2	27
3	8.5	75
4	3.5	30

| dataset.shape

(25, 2)

| dataset.describe()

	Hours	Scores
count	25.000000	25.000000
mean	5.012000	51.480000
std	2.525094	25.286887
min	1.100000	17.000000
25%	2.700000	30.000000
50%	4.800000	47.000000
75%	7.400000	75.000000
max	9.200000	95.000000



```
[ ] print (regressor.coef )
   [9.91065648]
 [ ] y pred = regressor.predict(X test)
     = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})
     Actual Predicted
      20 16.884145
        27 33.732261
        30 26.794801
       62 60,491033
[ ] print('Mean Absolute Error:', metrics.mean absolute_error(y_test, y_pred))
print('Mean Squared Error:', metrics.mean_squared_error(y_test, y_pred))
print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test, y_pred)))
   Mean Absolute Error: 4.183859899002975
Mean Squared Error: 21.5987693072174
Root Mean Squared Error: 4.6474476121003665
B)
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix,
classification report
Import required modules and packages
dataset = pd.read_csv('....\diabetes data.csv')
print(dataset.head())
Import data set
Choose the right path for the dataset
def diagnosis(x):
if x=='M':
return 1
if x=='B':
return 0
dataset['diagnosis'] = dataset['diagnosis'].apply(diagnosis)
print(dataset)
Data cleaning process. Converting
categorical value in to numerical value.
M = malignant, B = benign
print("Any missing sample in data set:",
dataset.isnull().values.any(), "\n")
Check for any missing values in the data
dataset = dataset.replace([np.inf, -np.inf], np.nan)
dataset= dataset.fillna(dataset.mean())
dataset
Replace the missing value with its mean
value of the respective attribute
dataset= dataset.drop(columns=["Unnamed: 32"]) drop this column because it's not
necessary (null)
                                                                                                                         66
```

```
Y = dataset['diagnosis']
X = dataset.drop(columns=['diagnosis'])
X_train, X_test, Y_train, Y_test = train_test_split(X, Y,
test_size=0.2, random_state=9)
print('X train shape: ', X_train.shape)
print('Y train shape: ', Y train.shape)
print('X test shape: ', X_test.shape)
print('Y test shape: ', Y_test.shape)
Splitting the given data in to training set
(80%) and testing set (20%)
svc_classifier= SVC(kernel='poly') Model instantiation. Apply SVM with
different kernels 'linear', 'poly', 'rbf',
'sigmoid' and verify the accuracy of the
model
svc classifier.fit(X train, Y train) Model Training
y_pred=svc_classifier.predict(X_test) Testing the model
print(confusion_matrix(Y_test,y_pred))
print(classification_report(Y_test,y_pred))
Evaluation metrics to measure the
performance of the model
```

```
import pandas as po
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix,classification_report
%matplotlib inline
dataset = pd.read csv('C:\\Users\DELL\Desktop\data.csv')
dataset.head()
         id diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean compactness_mean concavity_mean
                                                                                                                                                       texture_worst perimeter
                                                                                                                                     points mean
     842302
                                17.99
                                              10.38
                                                              122.80
                                                                         1001.0
                                                                                          0.11840
                                                                                                            0.27760
                                                                                                                              0.3001
                                                                                                                                          0.14710
                                                                                                                                                                 17.33
                               20.57
                                              17.77
                                                                         1326.0
                                                                                          0.08474
                                                                                                            0.07864
                                                                                                                              0.0869
                                                                                                                                                                 23.41
     842517
                                                              132.90
                                                                                                                                          0.07017
                                19.69
2 84300903
                                              21.25
                                                              130.00
                                                                         1203.0
                                                                                          0.10960
                                                                                                            0.15990
                                                                                                                              0.1974
                                                                                                                                          0.12790
                                                                                                                                                                 25.53
                                                                                                                              0.2414
3 84348301
                                11.42
                                              20.38
                                                               77.58
                                                                          386.1
                                                                                          0.14250
                                                                                                            0.28390
                                                                                                                                          0.10520
                                                                                                                                                                 26.50
                               20.29
                                                              135.10
                                                                         1297.0
                                                                                          0.10030
                                                                                                            0.13280
                                                                                                                              0.1980
4 84358402
                                                                                                                                          0.10430
5 rows × 33 columns
```

```
id diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean compactness_mean concavity_mean
                                                                                                                                                               ... texture worst perimete
                                                                                                                                               points mean
       842302
                                   17.99
                                                   10.38
                                                                   122.80
                                                                               1001.0
                                                                                                 0.11840
                                                                                                                     0.27760
                                                                                                                                      0.30010
                                                                                                                                                     0.14710
                                                                                                                                                                             17.33
       842517
                                                   17.77
                                                                   132.90
                                                                               1326.0
                                                                                                 0.08474
                                                                                                                     0.07864
                                                                                                                                      0.08690
                                                                                                                                                     0.07017
                                                                                                                                                                             23.41
                                                                               1203.0
                                                                                                 0.10960
                                                                                                                     0.15990
                                                                                                                                      0.19740
 2 84300903
                                   19.69
                                                   21.25
                                                                   130.00
                                                                                                                                                     0.12790
                                                                                                                                                                             25.53
                                                   20.38
                                                                                386.1
                                                                                                 0.14250
                                                                                                                     0.28390
                                                                                                                                      0.24140
                                                                                                                                                                             26.50
     84348301
                                   11.42
                                                                    77.58
                                                                                                                                                     0.10520
     84358402
                                   20.29
                                                   14.34
                                                                   135.10
                                                                               1297.0
                                                                                                 0.10030
                                                                                                                     0.13280
                                                                                                                                      0.19800
                                                                                                                                                     0.10430
                                                                                                                                                                             16.67
                                   21.56
                                                   22 39
                                                                   142.00
                                                                               1479 0
                                                                                                 0.11100
                                                                                                                     0 11590
                                                                                                                                      0.24390
                                                                                                                                                                             26.40
564
       926424
                                                                                                                                                     0.13890
       926682
                                   20.13
                                                   28.25
                                                                   131.20
                                                                               1261.0
                                                                                                 0.09780
                                                                                                                     0.10340
                                                                                                                                      0.14400
                                                                                                                                                     0.09791
                                                                                                                                                                             38.25
                                   16.60
                                                   28.08
                                                                                858.1
                                                                                                 0.08455
                                                                                                                     0.10230
                                                                                                                                      0.09251
                                                                                                                                                     0.05302
                                                                                                                                                                             34.12
       926954
                                                                   108.30
566
567
       927241
                                   20.60
                                                   29.33
                                                                   140.10
                                                                               1265.0
                                                                                                 0.11780
                                                                                                                     0.27700
                                                                                                                                      0.35140
                                                                                                                                                     0.15200
                                                                                                                                                                             39.42
                                                                                                 0.05263
                                                                                                                     0.04362
                                    7.76
                                                   24.54
                                                                    47.92
                                                                                181.0
                                                                                                                                      0.00000
                                                                                                                                                     0.00000
                                                                                                                                                                             30.37
568
        92751
569 rows × 33 columns
```

```
svc classifier= SVC(kernel='rbf')
  svc classifier
  SVC()
  svc classifier=svc classifier.fit(X train, Y train)
  y pred=svc classifier.predict(X test)
print(confusion matrix(Y_test,y_pred))
         0]
   [[74
        0]]
    [40
print(classification report(Y test, y pred))
                 precision
                               recall f1-score
                                                   support
              0
                      0.65
                                 1.00
                                           0.79
                                                        74
              1
                      0.00
                                 0.00
                                           0.00
                                                        40
                                           0.65
                                                       114
      accuracy
                      0.32
                                 0.50
                                           0.39
     macro avg
                                                       114
  weighted avg
                      0.42
                                 0.65
                                           0.51
                                                       114
```

(c) Implementation of K-means clustering algorithm to group the customers based on their demographic detail using the given dataset.

Problem: Client is owing a supermarket mall and through membership cards, client have some basic data 68

about your customers like Customer ID, age, gender, annual income and spending score. Help the client

to understand the customers like who are the target customers so that the sense can be given to marketing

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette score
%matplotlib inline
data=pd.read_csv('C:\\Users\DELL\Desktop\mall_customers.csv')
print(data.head())
  CustomerID Genre Age Annual Income (k$) Spending Score (1-100)
     1 Male 19 15
        2 Male 21
                                  15
       3 Female 20
                                 16
                                                        6
       4 Female 23
                                                        77
       5 Female 31
inVsout=data.iloc[:,[3,4]]
inVsout
```

18

83

0	15	39
1	15	81
2	16	6
3	16	77
4	17	40
195	120	79
196	126	28
197	126	74

137

137

Annual Income (k\$) Spending Score (1-100)

200 rows × 2 columns

198

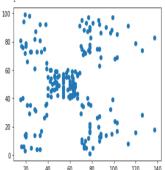
199

inVsout=data.iloc[:,[3,4]]

inVsout

plt.scatter(inVsout.iloc[:,0],inVsout.iloc[:,1])

<matplotlib.collections.PathCollection at 0x1fa26e7ca90>



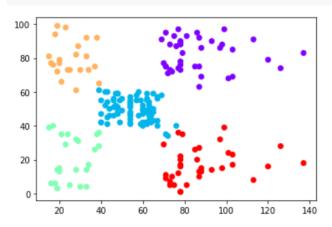
kmeans=KMeans(n\_clusters=5)

kmeans.fit(inVsout)

KMeans(n\_clusters=5)

plt.scatter(inVsout.iloc[:,0],inVsout.iloc[:,1], c=kmeans.labels\_, cmap='rainbow')
plt.show()

plt.scatter(inVsout.iloc[:,0],inVsout.iloc[:,1], c=kmeans.labels\_, cmap='rainbow')
plt.show()



	Annual	Income	(k\$)	Spending	Score	(1-100)
0			15			39
2			16			6
4			17			40
6			18			6
8			19			3
10			19			14
12			20			15
14			20			13
16			21			35
18			23			29
20			24			35
22			25			5
24			28			14
26			28			32
28			29			31
30			30			4
32			33			4
34			33			14

silhouette\_score(inVsout,kmeans.labels\_)

0.553931997444648

Experiment No:-10 Date:-10-04-2022

# To Implement NLP programs

Aim:-To Implement NLP programs

NLP stands for Natural Language Processing, which is a part of Computer Science, Human language, and Artificial Intelligence. It is the technology that is used by machines to understand, analyse, manipulate, and interpret human's languages. It helps developers to organize knowledge for performing tasks such as translation, automatic summarization, Named Entity Recognition (NER), speech recognition, relationship extraction, and topic segmentation.

```
code:-
```

```
!pip install -q wordcloud
import wordcloud
import nltk
nltk.download('stopwords')
nltk.download('wordnet')
nltk.download('punkt')
nltk.download('averaged_perceptron_tagger')
import pandas as pd
import matplotlib.pyplot as plt
import io
import unicodedata
import numpy as np
import re
```

```
import string
# Constants
# POS (Parts Of Speech) for: nouns, adjectives, verbs and adverbs
DI_POS_TYPES = {'NN':'n', 'JJ':'a', 'VB':'v', 'RB':'r'}
POS_TYPES = list(DI_POS_TYPES.keys())
# Constraints on tokens
MIN\_STR\_LEN = 3
RE_VALID = '[a-zA-Z]'
# Upload from google drive
from google.colab import files
uploaded = files.upload()
print("len(uploaded.keys():", len(uploaded.keys()))
for fn in uploaded.keys():
 print('User uploaded file "{name}" with length {length} bytes'.format(name=fn,
length=len(uploaded[fn])))
# Get list of quotes
df_quotes = pd.read_csv(io.StringIO(uploaded['quotes.txt'].decode('utf-8')), sep='\t')
# Display
print("df_quotes:")
print(df_quotes.head().to_string())
print(df_quotes.describe())
# Convert quotes to list
li_quotes = df_quotes['Quote'].tolist()
print()
print("len(li_quotes):", len(li_quotes)
# Get stopwords, stemmer and lemmatizer
```

```
stopwords = nltk.corpus.stopwords.words('english')
stemmer = nltk.stem.PorterStemmer()
lemmatizer = nltk.stem.WordNetLemmatizer()
# Remove accents function
def remove_accents(data):
  return ".join(x for x in unicodedata.normalize('NFKD', data) if x in
string.ascii_letters or x == " ")
# Process all quotes
li_tokens = []
li_token_lists = []
li_lem_strings = []
for i,text in enumerate(li_quotes):
  # Tokenize by sentence, then by lowercase word
  tokens = [word.lower() for sent in nltk.sent_tokenize(text) for word in
nltk.word_tokenize(sent)]
  # Process all tokens per quote
  li_tokens_quote = []
  li_tokens_quote_lem = []
  for token in tokens:
    # Remove accents
    t = remove_accents(token)
    # Remove punctuation
     t = str(t).translate(string.punctuation)
     li_tokens_quote.append(t)
    # Add token that represents "no lemmatization match"
```

```
li_tokens_quote_lem.append("-") # this token will be removed if a lemmatization
match is found below
    # Process each token
    if t not in stopwords:
       if re.search(RE_VALID, t):
         if len(t) >= MIN_STR_LEN:
           # Note that the POS (Part Of Speech) is necessary as input to the
lemmatizer
           # (otherwise it assumes the word is a noun)
           pos = nltk.pos_tag([t])[0][1][:2]
           pos2 = 'n' # set default to noun
           if pos in DI_POS_TYPES:
             pos2 = DI_POS_TYPES[pos]
            stem = stemmer.stem(t)
           lem = lemmatizer.lemmatize(t, pos=pos2) # lemmatize with the correct
POS
           if pos in POS_TYPES:
              li_tokens.append((t, stem, lem, pos))
              # Remove the "-" token and append the lemmatization match
              li_tokens_quote_lem = li_tokens_quote_lem[:-1]
              li_tokens_quote_lem.append(lem)
  # Build list of token lists from lemmatized tokens
  li_token_lists.append(li_tokens_quote)
  # Build list of strings from lemmatized tokens
  str_li_tokens_quote_lem = ''.join(li_tokens_quote_lem)
```

```
li_lem_strings.append(str_li_tokens_quote_lem)
# Build resulting dataframes from lists
df_token_lists = pd.DataFrame(li_token_lists)
print("df_token_lists.head(5):")
print(df_token_lists.head(5).to_string())
# Replace None with empty string
for c in df_token_lists:
  if str(df_token_lists[c].dtype) in ('object', 'string_', 'unicode_'):
    df_token_lists[c].fillna(value=", inplace=True)
df_lem_strings = pd.DataFrame(li_lem_strings, columns=['lem quote'])
print()
print("")
print("df_lem_strings.head():")
print(df_lem_strings.head().to_string())
# Add counts
print("Group by lemmatized words, add count and sort:")
df_all_words = pd.DataFrame(li_tokens, columns=['token', 'stem', 'lem', 'pos'])
df_all_words['counts'] = df_all_words.groupby(['lem'])['lem'].transform('count')
df_all_words = df_all_words.sort_values(by=['counts', 'lem'], ascending=[False,
True]).reset_index()
print("Get just the first row in each lemmatized group")
df_words = df_all_words.groupby('lem').first().sort_values(by='counts',
ascending=False).reset_index()
print("df_words.head(10):")
print(df_words.head(10))
```

```
df_words = df_words[['lem', 'pos', 'counts']].head(200)
for v in POS_TYPES:
  df_pos = df_words[df_words['pos'] == v]
  print()
  print("POS_TYPE:", v)
  print(df_pos.head(10).to_string())
li\_token\_lists\_flat = [y for x in li\_token\_lists for y in x] # flatten the list of token lists
to a single list
print("li_token_lists_flat[:10]:", li_token_lists_flat[:10])
di_freq = nltk.FreqDist(li_token_lists_flat)
del di_freq["]
li_freq_sorted = sorted(di_freq.items(), key=lambda x: x[1], reverse=True) # sorted
list
print(li_freq_sorted)
di_freq.plot(30, cumulative=False)
li_lem_words = df_all_words['lem'].tolist()
di_freq2 = nltk.FreqDist(li_lem_words)
li_freq_sorted2 = sorted(di_freq2.items(), key=lambda x: x[1], reverse=True) # sorted
list
print(li_freq_sorted2)
di_freq2.plot(30, cumulative=False)
```

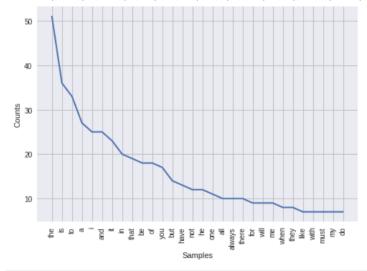
# Output:-

Group by lemmatized words, add count and sort: Get just the first row in each lemmatized group df\_words.head(10):

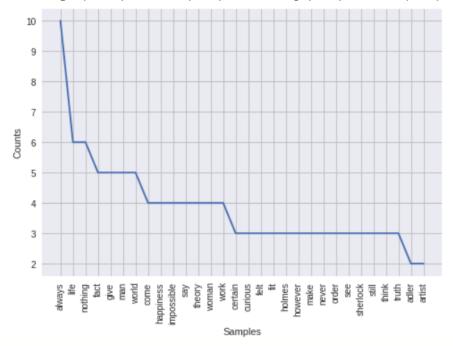
	lem	index	token	stem	pos	counts
0	always	50	always	alway	RB	10
1	nothing	116	nothing	noth	NN	6
2	life	54	life	life	NN	6
3	man	74	man	man	NN	5
4	give	39	gave	gave	VB	5
5	fact	106	fact	fact	NN	5
6	world	121	world	world	NN	5
7	happiness	119	happiness	happi	NN	4
8	work	297	work	work	NN	4
9	theory	101	theory	theori	NN	4

POS_TYPE: NN										
	lem	pos	cou	nts						
1	nothing	NN	6							
2	life	NN	6							
3	man	NN	5							
5	fact	NN	5							
6	world	NN	5							
7	happiness	NN	4							
8	work	NN	4							
9	theory	NN	4							
10	woman	NN	4							
17	holmes	NN	3							
POS	TYPE: JJ									
100		1em	pos	cour	nts					
11	impossi			0041	4					
15	cert			3						
18		ious			3					
34		nice			2					
43		JJ		2						
48		good			2					
61	improba	-			2					
62	-	best			2					
	philosophi				1					
81	possi				1					
0.1	Pood	-22-0			_					
POS_TYPE: VB										
	lem	pos	cou	nts						
4	give	VB		5						
12	say	VB		4						
13	come	VB		4						
22	see	VB	3							
23	make	VB		3						
2.6	t.hink	VB		.3						

```
li_token_lists_flat[:10]: ['i', 'like', 'living', '', 'i', 'have', 'sometimes', 'been', 'wildly', '']
[('the', 51), ('is', 36), ('to', 33), ('a', 27), ('i', 25), ('and', 25), ('it', 23), ('in', 20), ('that
```



[('always', 10), ('life', 6), ('nothing', 6), ('fact', 5), ('give', 5), ('man', 5),



Result:-Thus the NPL program was implemented

**Experiment No:-11 Date:-06-04-2022** 

## **Deep Learning**

**AIM:** Implementation of Deep Learning

### **CODE:**

```
import tensorflow as tf
from utils.DL utils import myCallback, build model,
compile_train_model, plot_loss_acc
from itertools import product
accuracy_desired = [0.85, 0.9, 0.95]
num_neurons = [16,32,64,128]
cases = list(product(accuracy_desired,num_neurons)) print("So, the cases we
are considering are as follows...\n")
for i,c in enumerate(cases):
 print("Accuracy target { }, number of neurons:
\{\}".format(c[0],c[1]))
 if (i+1)%4==0 and (i+1)!=len(cases):
     print("-"*50)
  for c in cases:
 # Create a mycallback class with the specific accuracy target
 callbacks = myCallback(c[0], print_msg=False)
 # Build a model with a specific number of neurons
 model = build_model(num_layers=1,architecture=[c[1]])
 # Compile and train the model passing on the callback
class, choose suitable batch size and a max epoch limit
 model = compile_train_model(model,
x_train,y_train,callbacks=callbacks,
                        batch_size=32,epochs=30)
 # Construct a suitable title string for displaying the results
properly
 title = "Loss and accuracy over the epochs for\naccuracythreshold \
  {} and number of neurons {}".format(c[0],c[1])
```

# Use the plotting utility function, pass on the accuracy target,# trained model, and the custom title string plot\_loss\_acc(model,target\_acc=c[0],title=title)

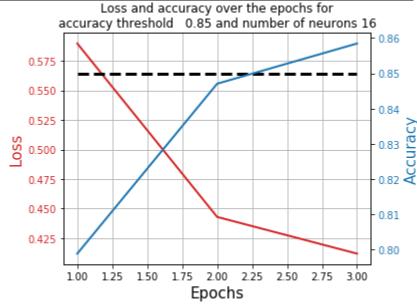
#### **OUTPUT:**

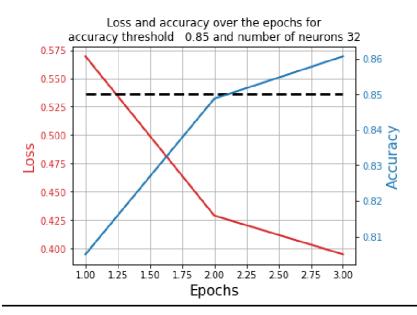
```
So, the cases we are considering are as follows...

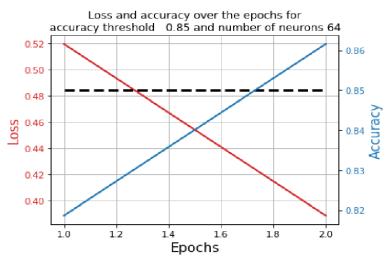
Accuracy target 0.85, number of neurons: 16
Accuracy target 0.85, number of neurons: 32
Accuracy target 0.85, number of neurons: 64
Accuracy target 0.85, number of neurons: 128

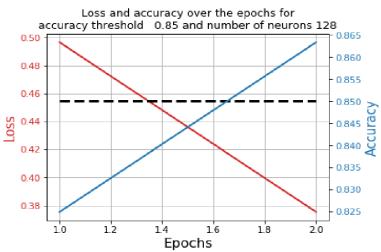
Accuracy target 0.9, number of neurons: 16
Accuracy target 0.9, number of neurons: 32
Accuracy target 0.9, number of neurons: 64
Accuracy target 0.9, number of neurons: 128

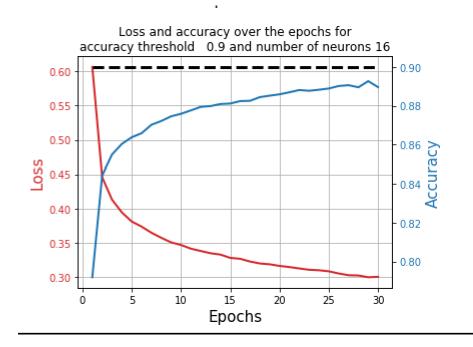
Accuracy target 0.95, number of neurons: 16
Accuracy target 0.95, number of neurons: 32
Accuracy target 0.95, number of neurons: 32
Accuracy target 0.95, number of neurons: 64
Accuracy target 0.95, number of neurons: 64
Accuracy target 0.95, number of neurons: 128
```











**RESULT:** Deep Learning Model Implemented