

1) Problems in Search

a. Problem Statement for A* : ABC has to reach to Mumbai from Bangalore. As there are multiple paths to reach Mumbai help ABC to reach the destination using the shortest path by applying A* Algorithm

b. Problem Statement for uniform cost search : We have the Map of Romania. In this map, the distance between various places in Romania is given. If we have to reach from one place to another place there exist several paths. Write a Python Program to find the shortest distance between any two places using a uniform cost search.

Ans.

#PQueue() functions

class PQueue():

def __init__(self):

self.dict = { }

self.keys = []

self.sorted = False

#push fuction is used to push the keys into the stack with the given values. The push library is used

def push(self, k, v):

self.dict[k] = v

self.sorted = False

#sort fuction is used to sort the keys with the given values. The sort library is used

def _sort(self):

self.keys = sorted(self.dict, key=self.dict.get, reverse=True)

self.sorted = True

#pop fuction is used to pop the keys from the stack with the given values after sorting

def pop(self):

try:

if not self.sorted:

self._sort()

key = self.keys.pop()

value = self.dict[key]

self.dict.pop(key)

return key, value

except:

return None

Heuristics function is used in uniform cost search and finds the most promissing path.

#It takes the current state of the agent as its input and produces the estimation of how close agent is from the goal.

def heuristics(path):

h = { }

with open(path, 'r') as file:

for line in file:

k, v = line.split(", ")

h[k] = int(v)

print(h)

return h

def path_costs(path):

c = { }

with open(path, 'r') as file:

```

for line in file:
    line = line.split(", ")
    v = int(line.pop())
    e1 = line.pop()
    e2 = line.pop()
    if e1 not in c:
        c[e1] = {}
    if e2 not in c:
        c[e2] = {}
    c[e1][e2] = c[e2][e1] = v
    print(c)
return c

def a_star(start, goal, h, g):
    frontier = PQueue()
    # pushing path and cost to pqueue
    frontier.push(start, h[start])
    while True:
        # popping path with least cost
        path, cost = frontier.pop()
        print(path+ " " +str(cost))
        # splitting out end node in path
        end = path.split("->")[-1]
        # removing heuristic value of end node from cost
        cost -= h[end]
        if goal == end:
            break
        for node, weight in g[end].items():
            # adding edge weight(cost) and node heuristic to total cost
            new_cost = cost + weight + h[node]
            new_path = path + "->" + node
            # adding new path and cost to pqueue
            frontier.push(new_path, new_cost)

```

```

a_star('Arad', 'Bucharest', heuristics('./heuristics.txt'), path_costs('./paths.txt'))

```

```

Arad 366
Arad->Sibiu 393
Arad->Sibiu->Rimnicu Vilcea 413
Arad->Sibiu->Fagaras 415
Arad->Sibiu->Rimnicu Vilcea->Pitesti 417
Arad->Sibiu->Rimnicu Vilcea->Pitesti->Bucharest 418

```

2) Problem Statement for uniform cost search : For the Romania map, the distance between various places are given. If we have to reach from one place to another place there exist several paths. Write a Python Program to find the shortest distance between any two places using a uniform cost search.

```

class PQueue():
    def __init__(self):
        self.dict = {}
        self.keys = []
        self.sorted = False

```

```

def push(self, k, v):
    self.dict[k] = v
    self.sorted = False

def _sort(self):
    self.keys = sorted(self.dict, key=self.dict.get, reverse=True)
    self.sorted = True

def pop(self):
    try:
        if not self.sorted:
            self._sort()
        key = self.keys.pop()
        value = self.dict[key]
        self.dict.pop(key)
        return key, value
    except:
        return None

```

```

def path_costs(path):
    c = {}
    with open(path, 'r') as file:
        for line in file:
            line = line.split(" ")
            v = int(line.pop())
            e1 = line.pop()
            e2 = line.pop()
            if e1 not in c:
                c[e1] = {}
            if e2 not in c:
                c[e2] = {}
            c[e1][e2] = c[e2][e1] = v
    return c

```

```

def ucs(start, goal, g):
    frontier = PQueue()
    # pushing path and cost to pqueue
    frontier.push(start, 0)
    while True:
        # popping path with least cost
        path, cost = frontier.pop()
        print(path + " " + str(cost))
        # splitting out end node in path
        end = path.split("->")[-1]
        if goal == end:
            break
        for node, weight in g[end].items():
            # adding edge weight(cost) to total cost
            new_cost = cost + weight
            new_path = path + "->" + node
            # adding new path and cost to pqueue
            frontier.push(new_path, new_cost)

```

```
ucs('Arad', 'Bucharest', path_costs('./paths.txt'))
```

Heuristics.txt

Arad, 366
Bucharest, 0
Craiova, 160
Dobreta, 242
Eforie, 161
Fagaras, 176
Giurgiu, 77
Hirsova, 151
Lasi, 226
Lugoj, 244
Mehadia, 241
Neamt, 234
Oradea, 380
Pitesti, 100
Rimnicu Vilcea, 193
Sibiu, 253
Timisoara, 329
Urziceni, 80
Vaslui, 199
Zerind, 374

Paths.txt

Arad, Zerind, 75
Arad, Sibiu, 140
Arad, Timisoara, 118
Zerind, Oradea, 71
Oradea, Sibiu, 151
Timisoara, Lugoj, 111
Sibiu, Fagaras, 99
Sibiu, Rimnicu Vilcea, 80
Lugoj, Mehadia, 70
Fagaras, Bucharest, 211
Rimnicu Vilcea, Pitesti, 97
Rimnicu Vilcea, Craiova, 146
Mehadia, Dobreta, 75
Bucharest, Pitesti, 101
Bucharest, Urziceni, 85
Bucharest, Giurgiu, 90
Pitesti, Craiova, 138
Craiova, Dobreta, 120
Urziceni, Hirsova, 98
Urziceni, Vaslui, 142
Hirsova, Eforie, 86
Vaslui, Lasi, 92
Lasi, Neamt, 87

Output:-

```
PS C:\Users\tarun\AI LAB PROGRAMS> python uniformcostsearch.py
Arad 0
Arad->Zerind 75
Arad->Timisoara 118
Arad->Sibiu 140
Arad->Zerind->Oradea 146
Arad->Zerind->Arad 150
Arad->Zerind->Oradea->Zerind 217
Arad->Sibiu->Rimnicu Vilcea 220
Arad->Zerind->Arad->Zerind 225
Arad->Timisoara->Lugoj 229
Arad->Timisoara->Arad 236
Arad->Sibiu->Fagaras 239
Arad->Zerind->Arad->Timisoara 268
Arad->Sibiu->Arad 280
Arad->Zerind->Oradea->Zerind->Oradea 288
Arad->Zerind->Arad->Sibiu 290
Arad->Sibiu->Oradea 291
Arad->Zerind->Oradea->Zerind->Arad 292
Arad->Zerind->Arad->Zerind->Oradea 296
Arad->Zerind->Oradea->Sibiu 297
Arad->Timisoara->Lugoj->Mehadia 299
Arad->Zerind->Arad->Zerind->Arad 300
Arad->Sibiu->Rimnicu Vilcea->Sibiu 300
Arad->Timisoara->Arad->Zerind 311
Arad->Sibiu->Rimnicu Vilcea->Pitesti 317
Arad->Sibiu->Fagaras->Sibiu 338
Arad->Timisoara->Lugoj->Timisoara 340
Arad->Timisoara->Arad->Timisoara 354
Arad->Sibiu->Arad->Zerind 355
Arad->Zerind->Oradea->Zerind->Oradea->Zerind 359
Arad->Sibiu->Oradea->Zerind 362
Arad->Sibiu->Rimnicu Vilcea->Craiova 366
Arad->Zerind->Arad->Zerind->Oradea->Zerind 367
Arad->Zerind->Oradea->Zerind->Arad->Zerind 367
Arad->Timisoara->Lugoj->Mehadia->Lugoj 369
Arad->Zerind->Arad->Sibiu->Rimnicu Vilcea 370
Arad->Timisoara->Lugoj->Mehadia->Dobreta 374
Arad->Zerind->Arad->Zerind->Arad->Zerind 375
Arad->Timisoara->Arad->Sibiu 376
Arad->Zerind->Oradea->Sibiu->Rimnicu Vilcea 377
Arad->Zerind->Arad->Timisoara->Lugoj 379
Arad->Sibiu->Rimnicu Vilcea->Sibiu->Rimnicu Vilcea 380
Arad->Timisoara->Arad->Zerind->Oradea 382
Arad->Timisoara->Arad->Zerind->Arad 386
Arad->Zerind->Arad->Timisoara->Arad 386
Arad->Zerind->Arad->Sibiu->Fagaras 389
Arad->Zerind->Oradea->Sibiu->Fagaras 396
Arad->Sibiu->Arad->Timisoara 398
Arad->Sibiu->Rimnicu Vilcea->Sibiu->Fagaras 399
Arad->Zerind->Oradea->Zerind->Arad->Timisoara 410
Arad->Sibiu->Rimnicu Vilcea->Pitesti->Rimnicu Vilcea 414
Arad->Sibiu->Fagaras->Sibiu->Rimnicu Vilcea 418
Arad->Sibiu->Rimnicu Vilcea->Pitesti->Bucharest 418
```

3) Problem Statement for Depth Limited Search : Design and develop a program in Python to print all the nodes reachable from a given starting node in a graph by using the Depth Limited Search method. Repeat the experiment for different Graphs.

```
from collections import defaultdict

class Graph:
    def __init__(self, vertices):
        self.V = vertices
        self.graph = defaultdict(list)

    def addEdge(self, u, v):
        self.graph[u].append(v)

    def DLS(self, source, target, maxDepth):
        if source == target : return True

        if maxDepth <= 0 : return False
```

```

        # recursively traversing the graph while searching
        for i in self.graph[source]:
            if(self.DLS(i, target, maxDepth-1)):
                return True
        return False

g = Graph(9)# creating the graph
g.addEdge(0, 1)
g.addEdge(0, 2)
g.addEdge(1, 3)
g.addEdge(1, 4)
g.addEdge(2, 5)
g.addEdge(2, 6)
g.addEdge(3,7)
g.addEdge(3,8)

target = 3
maxDepth = 3
source = 0

if g.DLS(source, target, maxDepth) == True:
    print(f"Target {target} is reachable from source {source} within
max depth {maxDepth}")
else:
    print(f"Target {target} is NOT reachable from source {source}
within max depth {maxDepth}")

```

Heuristics.txt

Arad, 366
 Bucharest, 0
 Craiova, 160
 Dobreta, 242
 Eforie, 161
 Fagaras, 176
 Giurgiu, 77
 Hirsowa, 151
 Lasi, 226
 Lugoj, 244
 Mehadia, 241
 Neamt, 234
 Oradea, 380
 Pitesti, 100
 Rimnicu Vilcea, 193

Sibiu, 253
Timisoara, 329
Urziceni, 80
Vaslui, 199
Zerind, 374

Paths.txt

Arad, Zerind, 75
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Sibiu, Rimnicu Vilcea, 80
Lugoj, Mehadia, 70
Fagaras, Bucharest, 211
Rimnicu Vilcea, Pitesti, 97
Rimnicu Vilcea, Craiova, 146
Mehadia, Dobreta, 75
Bucharest, Pitesti, 101
Bucharest, Urziceni, 85
Bucharest, Giurgiu, 90
Pitesti, Craiova, 138
Craiova, Dobreta, 120
Urziceni, Hirsova, 98
Urziceni, Vaslui, 142
Hirsova, Eforie, 86
Vaslui, Lasi, 92
Lasi, Neamt, 87

Outupt:-

```

Arad 0
Arad->Zerind 75
Arad->Timisoara 118
Arad->Sibiu 140
Arad->Zerind->Oradea 146
Arad->Zerind->Arad 150
Arad->Zerind->Oradea->Zerind 217
Arad->Sibiu->Rimnicu Vilcea 220
Arad->Zerind->Arad->Zerind 225
Arad->Timisoara->Lugoj 229
Arad->Timisoara->Arad 236
Arad->Sibiu->Fagaras 239
Arad->Zerind->Arad->Timisoara 268
Arad->Sibiu->Arad 280
Arad->Zerind->Oradea->Zerind->Oradea 288
Arad->Zerind->Arad->Sibiu 290
Arad->Sibiu->Oradea 291
Arad->Zerind->Oradea->Zerind->Arad 292
Arad->Zerind->Arad->Zerind->Oradea 296
Arad->Zerind->Oradea->Sibiu 297
Arad->Timisoara->Lugoj->Mehadia 299
Arad->Zerind->Arad->Zerind->Arad 300
Arad->Sibiu->Rimnicu Vilcea->Sibiu 300
Arad->Timisoara->Arad->Zerind 311
Arad->Sibiu->Rimnicu Vilcea->Pitesti 317
Arad->Sibiu->Fagaras->Sibiu 338
Arad->Timisoara->Lugoj->Timisoara 340
Arad->Timisoara->Arad->Timisoara 354
Arad->Sibiu->Arad->Zerind 355
Arad->Zerind->Oradea->Zerind->Oradea->Zerind 359
Arad->Sibiu->Oradea->Zerind 362
Arad->Sibiu->Rimnicu Vilcea->Craiova 366
Arad->Zerind->Arad->Zerind->Oradea->Zerind 367
Arad->Zerind->Oradea->Zerind->Arad->Zerind 367
Arad->Timisoara->Lugoj->Mehadia->Lugoj 369
Arad->Zerind->Arad->Sibiu->Rimnicu Vilcea 370
Arad->Timisoara->Lugoj->Mehadia->Dobreta 374
Arad->Zerind->Arad->Zerind->Arad->Zerind 375
Arad->Timisoara->Arad->Sibiu 376
Arad->Zerind->Oradea->Sibiu->Rimnicu Vilcea 377
Arad->Zerind->Arad->Timisoara->Lugoj 379
Arad->Sibiu->Rimnicu Vilcea->Sibiu->Rimnicu Vilcea 380
Arad->Timisoara->Arad->Zerind->Oradea 382
Arad->Timisoara->Arad->Zerind->Arad 386
Arad->Zerind->Arad->Timisoara->Arad 386
Arad->Zerind->Oradea->Sibiu->Fagaras 396
Arad->Sibiu->Arad->Timisoara 398
Arad->Sibiu->Rimnicu Vilcea->Sibiu->Fagaras 399
Arad->Zerind->Oradea->Zerind->Arad->Timisoara 410
Arad->Sibiu->Rimnicu Vilcea->Pitesti->Rimnicu Vilcea 414
Arad->Sibiu->Fagaras->Sibiu->Rimnicu Vilcea 418
Arad->Sibiu->Rimnicu Vilcea->Pitesti->Bucharest 418

```

4) Write a program to implement a Minimax decision-making algorithm, typically used in a turn-based, two player games. The goal of the algorithm is to find the optimal next move.

Ans

```

import math
import random
#minimax class
def minimax (currentDepth, nodeIndex,
             maxTurn, score,
             tarDepth):

    # base case : tarDepth reached
    if (currentDepth == tarDepth):
        return score[nodeIndex]

    if (maxTurn):
        return max(minimax(currentDepth + 1, nodeIndex * 2,
                           False, score, tarDepth),
                   minimax(currentDepth + 1, nodeIndex * 2 + 1,
                           False, score, tarDepth))

    else:
        return min(minimax(currentDepth + 1, nodeIndex * 2,
                           True, score, tarDepth),
                   minimax(currentDepth + 1, nodeIndex * 2 + 1,
                           True, score, tarDepth))

```



```
True, score, tarDepth))
```

```
# Driver code
```

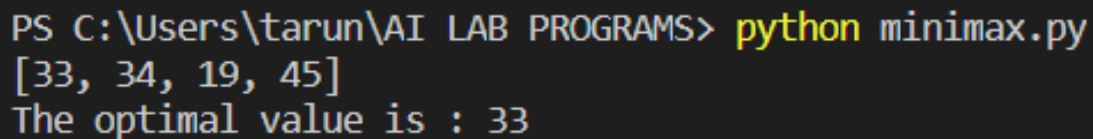
```
score = random.sample(range(1, 50), 4)
```

```
print(str(score))
```

```
treeDepth = math.log(len(score), 2)
```

```
print("The optimal value is : ", end = "")
```

```
print(minimax(0, 0, True, score, treeDepth))
```



```
PS C:\Users\tarun\AI LAB PROGRAMS> python minimax.py
[33, 34, 19, 45]
The optimal value is : 33
```

5) Write a program to implement Alpha Beta pruning in Python. The algorithm can be applied to any depth of tree by not only pruning the tree leaves but also the entire subtree. Order the nodes in the tree such that the best nodes are checked first from the shallowest node. Use domain knowledge while finding the best move. Ex: for Chess, try order: captures first, then threats, then forward moves, backward moves.

Ans

```
tree = [[[5, 1, 2], [8, -8, -9]], [[9, 4, 5], [-3, 4, 3]]]
```

```
root = 0
```

```
pruned = 0
```

```
def children(branch, depth, alpha, beta):
```

```
    global tree
```

```
    global root
```

```
    global pruned
```

```
    i = 0
```

```
    for child in branch:
```

```
        if type(child) is list:
```

```
            (nalpha, nbeta) = children(child, depth + 1, alpha, beta)
```

```
            if depth % 2 == 1:
```

```
                beta = nalpha if nalpha < beta else beta
```

```
            else:
```

```
                alpha = nbeta if nbeta > alpha else alpha
```

```
            branch[i] = alpha if depth % 2 == 0 else beta
```

```
            i += 1
```

```
        else:
```

```
            if depth % 2 == 0 and alpha < child:
```

```
                alpha = child
```

```
            if depth % 2 == 1 and beta > child:
```

```

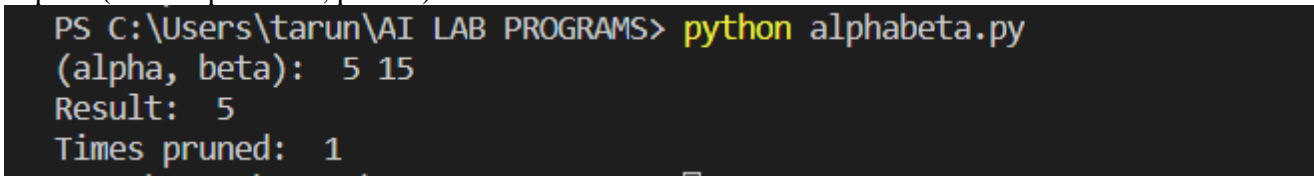
        beta = child
        if alpha >= beta:
            pruned += 1
            break
    if depth == root:
        tree = alpha if root == 0 else beta
    return (alpha, beta)

def alpha_beta(in_tree=tree, start=root, up=-15, low=15):
    global tree
    global pruned
    global root
    (alpha, beta) = children(tree, start, up, low)
    return (alpha, beta, tree, pruned)

if __name__ == "__main__":
    res=[]
    (alpha, beta, tree, pruned)=alpha_beta(None)

    print("(alpha, beta): ", alpha, beta)
    print("Result: ", tree)
    print("Times pruned: ", pruned)

```



```

PS C:\Users\tarun\AI LAB PROGRAMS> python alphabeta.py
(alpha, beta): 5 15
Result: 5
Times pruned: 1

```

6) Assume that you are organizing a party for N people and have been given a list L of people who, for social reasons, should not sit at the same table. Furthermore, assume that you have C tables (that are infinitely large). Write a function $\text{layout}(N, C, L)$ that can give a table placement (ie. a number from $0 \dots C - 1$) for each guest such that there will be no social mishaps.

For simplicity we assume that you have a unique number $0 \dots N - 1$ for each guest and that the list of restrictions is of the form $[(X, Y), \dots]$ denoting guests X, Y that are not allowed to sit together.

Answer with a dictionary mapping each guest into a table assignment, if there are no possible layouts of the guests you should answer False.

Ans

```

def backtrack(x, enemies, domain, assigned):
    if -1 not in assigned: # checking for unassigned people
        return x
    v = 999

    for i in range(len(domain)):
        if v > len(domain[i]) and assigned[i] != 1: # finding unassigned people

```

```

        v = i
order = []

for i in domain[v]:
    min = 1000
    for j in enemies[v]:
        temp = len(domain[j])
        if i in domain[j]:
            temp -= 1
        if temp < min:
            min = temp
    order.append((i, min))

order = sorted(order, key=lambda x:x[1], reverse=True)
ordered = [i[0] for i in order]

```

```

for i in ordered:
    new_d = [[j for j in i] for i in domain]
    for j in enemies[v]:
        if i == x[j]:
            continue
    x[v] = i
    assigned[v] = 1
    new_d[v] = [z for z in new_d[v] if z==i]
    temp = []
    for j in range(len(new_d)):
        if j!=v and j in enemies[v]:
            new_d[j] = [z for z in new_d[j] if z!=i]
    res = backtrack(x, enemies, new_d, assigned)
    if res!=0:
        return res
x[v] = ""
assigned[v] = -1
return 0

```

```

if __name__ == "__main__":
    people = int(input("Number of people = "))
    tables = int(input("Number of tables = "))
    edges = []
    rows = input("People who should not sit together = ").split()
    while(rows):
        edges.append((int(rows[0]),int(rows[1])))
        rows = input().split()

    x = ["" for i in range(people)]
    # filling out the enemies matrix
    enemies = [[] for i in range(people)]
    for i in edges:
        enemies[i[0]].append(i[1])
        enemies[i[1]].append(i[0])

```

```

for i in range(people):
    j = list(set(enemies[i])) # deduplicating the each row
    enemies[i] = j
assigned = [-1 for i in range(people)]
domain = [[x for x in range(tables)] for i in range(people)]

res = backtrack(x, enemies, domain, assigned)

if res == 0:
    print("Tables could not be assigned")
else:
    for i in range(len(res)):
        print(f"{i} : {res[i]}")

```

7 Implementation of Tic Tac Toe game here ,the player needs to take turns marking the spaces in a 3x3 grid with their own marks, if 3 consecutive marks (Horizontal, Vertical,Diagonal) are formed then the player who owns these moves get won. Noughts and Crosses or X's and O's abbreviations can be used to play.

Ans

```
import os
```

```

turn = 'X'
win = False
spaces = 9

```

```

def draw(board):
    for i in range(6, -1, -3):
        print(' ' + board[i] + '|' +
              board[i+1] + '|' + board[i+2])

```

```

def takeinput(board, spaces, turn):
    pos = -1
    print(turn + "'s turn:")

```

```

while pos == -1:
    try:
        print("Pick position 1-9:")
        pos = int(input())
        if(pos < 1 or pos > 9):
            pos = -1
        elif board[pos - 1] != ' ':
            pos = -1
    except:
        print("enter a valid position")
    spaces -= 1
    board[pos - 1] = turn
    if turn == 'X':
        turn = 'O'
    else:

```

```

    turn = 'X'
    return board, spaces, turn

def checkwin(board):
    # could probably make this better
    for i in range(0, 3):
        # rows
        r = i*3
        if board[r] != ' ':
            if board[r] == board[r+1] and board[r+1] == board[r+2]:
                return board[r]
        # columns
        if board[i] != ' ':
            if board[i] == board[i+3] and board[i] == board[i+6]:
                return board[i]
    # diagonals
    if board[0] != ' ':
        if (board[0] == board[4] and board[4] == board[8]):
            return board[0]
    if board[2] != ' ':
        if (board[2] == board[4] and board[4] == board[6]):
            return board[2]

    return 0

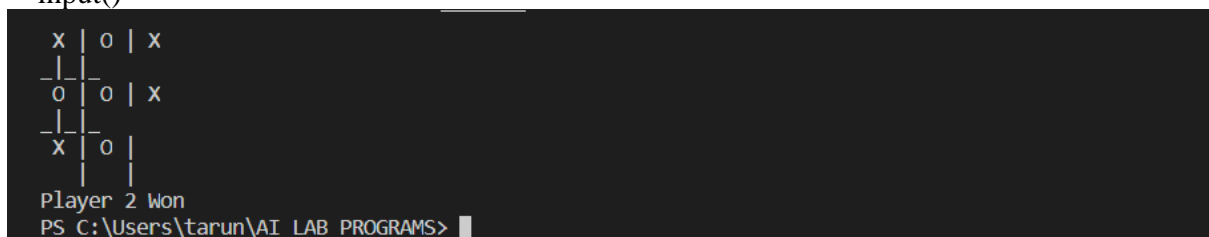
board = [' ']*9

while not win and spaces:
    draw(board)
    board, spaces, turn = takeinput(board, spaces, turn)
    win = checkwin(board)
    os.system('cls')

draw(board)

if not win and not spaces:
    print("draw")
elif win:
    print(f'{win} wins')
    input()

```



```

  x | o | x
  -|-
  o | o | x
  -|-
  x | o |
  -|-
Player 2 Won
PS C:\Users\tarun\AI LAB PROGRAMS>

```

9) Implement the Perceptron Learning single layer Algorithm by Initializing the weights and threshold. Execute the code and check, how many iterations are needed, until the network converge.

Ans

```

import numpy as np
theta = 1
epoch = 3

class Perceptron(object):
    def __init__(self, input_size, learning_rate=0.2):
        self.learning_rate = learning_rate
        self.weights = np.zeros(input_size + 1) # zero init for weights and bias

    def predict(self, x):
        return (np.dot(x, self.weights[1:]) + self.weights[0]) #  $X \cdot W + B$ 

    def train(self, x, y, weights):
        for inputs, label in zip(x, y):
            net_in = self.predict(inputs)
            if net_in > theta:
                y_out = 1
            elif net_in < -theta:
                y_out = -1
            else:
                y_out = 0
            if y_out != label: # updating the net on incorrect prediction
                self.weights[1:] += self.learning_rate * label * inputs #  $W = \alpha * Y * X$ 
                self.weights[0] += self.learning_rate * label #  $B = \alpha * Y$ 
            print(inputs, net_in, label, y_out, self.weights)

if __name__ == "__main__":
    x = []
    x.append(np.array([1, 1]))
    x.append(np.array([1, -1]))
    x.append(np.array([-1, 1]))
    x.append(np.array([-1, -1]))

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

    perceptron = Perceptron(2)

    for i in range(epoch):
        print("Epoch", i)
        print("X1 X2 ", " Net ", " T ", " Y ", " B Weights")
        weights = perceptron.weights
        print("Initial Weights", weights)
        perceptron.train(x, y, weights)

```

