

I N D E X

POAD OBSERVATION

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S.No.	Date	Title	Page No.	Teacher's Sign/ Remarks
1.		N-Queens Problem		✓
2.		Depth First Search		✓
3.		Water Jug problem using DFS		✓
4.		A* Search		✓
5.		Minimax Algorithm		✓
6.		Prolog		✓
7.		Introduction to prolog		✓
8.		Unification & Resolution		✓
9.		Fuzzy logic - Image processing		✓
10.		Implementing ANN for an application using Python		✓
11.		Implementing ANN for an appl using python regression		✓
12.		Decision tree classification		✓
13.		Implementation of decision tree classification		✓
14.		Implementation of clustering techniques (k Means)		✓
				Concl ✓

Exp No.: 1

N-Queens Problem

Date:

: (n, no. of board) for 8 job
: (1, 0) option of 8 job

AIM:

To solve the N-Queen Problem where the goal is to place n-queens on a $n \times n$ chessboard such that no two queens attack each other.

Algorithm:

- 1) Create a $n \times n$ chessboard with all cells set to 0.
- 2) Create a $n \times n$ chessboard with all cells set to 0, representing no queens placed.
- 3) Ensure no queen is in the same row, upper diagonal, or lower diagonal for a given position.
- 4) Try placing a queen in each row of the current column if it is safe using safe() function.
- 5) Move to the next column if placing a queen works, else backtrack by removing queen.
- 6) If queen one is placed in all columns return success.
- 7) Display the board.
- 8) If no solution exists print "solution does not exist".

Program:

```
def isSafe(board, row, coln, n):
```

for (i in range (col)):

if board[row][i] == 1 or not end[0] T

return false;

for i,j in zip(range(-1,-1), range(0,1,1)):
 range(0,1,1):

If board [i][j] == 1:

at the who will return false

for i, j in zip (orange, (row, col-1), start) (

range (Colpoptera) $\frac{1}{2}$ mm. long.

if $b_{i,j}$ and c_j are true then $i = j$ is true.

return false and no loops

Jefferson fore
for 1st time

def. Solve NQV till board is full.

new if $\text{col } z = n$:

refuse free breakfast ad

return tree word to who

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

if isSafe(board, i, col, n) :

board[i][col] = 1

web without "J" in it solve Nov 10 (board, col + n) 

If saline NQV fil (board, col + L n)

$\Sigma = \text{force}_\perp$

metacore tree

board[i][col] = 0
return false

def solveNQ(n):
 board[0:n] for i in range(n)]
 if solveNQUtil(board, 0, n) == False:
 point("solution does not exist")
 return false
 for q in board:

point(q)
return True

st + 100
n = int(input("enter n, value:"))
solveNQ(n)

OUTPUT:

Enter value: 5
[1, 0, 0, 0, 0]
[0, 1, 0, 0, 0]
[0, 0, 1, 0, 0]
[0, 0, 0, 1, 0]
[0, 0, 0, 0, 1]

marked [0, 0, 0, 0, 1] jump to last (1)
[0, 0, 1, 0, 0] which does not (1)
which is modified at start (1)

marked [0, 0, 0, 1, 0] jump to last (1)
[0, 1, 0, 0, 0] which does not (1)
which is modified at start (1)

marked [0, 1, 0, 0, 0] jump to last (1)
[0, 0, 1, 0, 0] which does not (1)
which is modified at start (1)

marked [0, 0, 1, 0, 0] jump to last (1)
[0, 0, 0, 1, 0] which does not (1)
which is modified at start (1)

marked [0, 0, 0, 1, 0] jump to last (1)
[0, 0, 0, 0, 1] which does not (1)
which is modified at start (1)

Exp 9

Depth First Search

o - [] Breadth

Search: not done

AIM:

[(n) steps in] (n) nodes in

Aim To implement depth first search

Draw a graph and explore all vertices

by visiting as far along each branch as possible before backtracking

Algorithm:

(1) Start

1) Start (with root node)

2) Initialize an empty stack and a list to

keep track of visited nodes

3) Push the starting node onto stack and

mark visited

4) While the stack is not empty, repeat

[0, 0, 0, 0, 1]

5) to step 7

[0, 1, 0, 0, 1]

6) Pop the top node from the stack

[0, 0, 0, 1, 0]

7) Print or process the popped node

[0, 0, 1, 0, 0]

8) For each adjacent unvisited neighbour of the popped node

[0, 0, 0, 1, 0]

9) Mark the neighbour as visited

[0, 0, 0, 1, 1]

- a) Push the unvisited neighbour onto the stack
- b) Repeat until all reachable node are visited
- c) Stop

Program:

```
def dfs(graph, start):
    stack = [start]
    visited = set()
```

while stack:

node = stack.pop()

If node not in visited:

Print(node, end=" ")

visited.add(node)

for neighbour in graph[node]:

If neighbor not in visited:

stack.append(neighbour)

```
graph = {
    'A': ['B', 'C'],
    'B': ['D', 'E'],
    'C': ['F'],
    'D': [],
    'E': ['F'],
    'F': []}
```

point C. DFS Traversal starting from node 'A':
dfs(graph, 'A') whatever the time deepest (0)

Output:

DFS traversal starting from node 'A':

ACFED

: (node, group), ths
[node] -> next

(0) for -> bitisn

: next value

: (group, next) : bca

: bcbca is last value for

(0) : bca is last value

(bca) bca, bcbca

: bcbca is group in modified list.

: bcbca is last modified list
(modified) bcbca, bca

✓

✓ [S, G], S -> group

✓ [S, G], S

Experiment 3

Water Jug Problem using DFS

Aim:

To solve the water jug problem using DFS

Algorithm:

- 1) Create a stack to store the states of the jugs
- 2) Initialise the stack with the initial state
- 3) While the stack is not empty, do the following
 - Pop a state from the stack
 - If the stack represents the desired quantity, stop and return the solution.
 - Generate all possible next states from the current state
 - Push the next state onto the stack
- 4) If the stack becomes empty and no solution is found, the problem is unsolvable.

Program:

```
def solveWaterJugProblem(capacity_jug1, capacity_jug2, desired_quantity):
```

stack = []

visited = set()

parent = {}

start-state = (0, 0)

stack.append(start-state)

visited.add(start-state)

parent[start-state] = None

while stack:

current-state = stack.pop()

If current-state[0] == desired-quantity

or current-state[1] == desired-quantity:

path = []

while current-state is not None:

path.append(current-state)

current-state = parent[current-state]

return path

next-states = generateNextStates(current-state,

Capacity Jug 1, Capacity Jug 2)

for state in next-states:

If state not in visited:

visited.add(state)

parent[state] = current-state

stack.append(state)

return "No solution found"

def generate_Next_States(state, capacity_jug1, capacity_jug2):

```
    next_states[i].append((capacity_jug1, state[i]))
```

```
next_states.append((state[0], capacity[jug2]))
```

next states. append (co, state[i]))

next states. append ([states[0], 0]))

flow_amount = min(state[0], capacity -
state[i]))

next.states.append((state[0] - poor_amount,
state[1] + poor_amount))

$\text{pour-amount} = \min(\text{state}[i], \text{capacity} - \text{state}[i])$

next_states.append((state[0] + pour_amount,
state[1] - pour_amount)))

restom next

rest are
solution = saline water bag (4, 2, 2)

point("solution": "solution")

OUTPUT:

~~solution~~: $\{(0, 0), (0, 3), (3, 0), (3, 3), (4, 2)\}$

Experiment 4

A* Search

AIM:

To find the shortest path from a start node to a goal node using A* search.

Algorithm:

1. Create open and closed sets; start with the initial node
2. Add the start node to the open set with an initial cost of 0
3. Remove the node with lowest f-value (cost + heuristic), from open set.
4. If the current node is goal node reconstruct the path
5. For each neighbour calculate $f = g + h$ if values
6. If the neighbour is not in open set or lower cost path is found - update costs
7. Add the neighbour to the open set if it is not already in the closed set

g. Repeat until the open set is empty or the goal is found.

Program:

```
import heapq

def a_star([start, goal], h, neighbors):
    open_set = []
    heapq.heappush(open_set, (0 + h(start), start))
    came_from = {}
    g_score = {start: 0}
    f_score = {start: h(start)}
    while open_set:
        current_f = heapq.heappop(open_set)[1]
        if current_f == goal:
            path = []
            current = current_f
            while current in came_from:
                path.append(current)
                current = came_from[current]
            path.append(goal)
            return path[::-1]
        current_g = g_score[current]
        for neighbor in neighbors(current):
            tentative_g = current_g + h(neighbor)
            if neighbor not in g_score or tentative_g < g_score[neighbor]:
                g_score[neighbor] = tentative_g
                f_score[neighbor] = tentative_g + h(neighbor)
                heapq.heappush(open_set, (f_score[neighbor], neighbor))
                came_from[neighbor] = current
```

for neighbor in neighbors[current]:
 if neighbor not in open-set:
 tentative-g = g-score[current] + 1
 if neighbor not in g-score or
 tentative-g < g-score[neighbor]:
 came-from[neighbor] = current
 g-score[neighbor] = tentative-g
 f-score[neighbor] = tentative-g + h(neighbor)
 if higher not in open-set
 heap.push((f-score[neighbor], came-from, neighbor))
 return Node

```

def heuristic(node):
    goal-position = (5, 5)
    return abs(node[0] - goal-position[0]) + abs(node[1] - goal-position[1])
  
```

```

def neighbours(node):
    x, y = node
    return [(x+1, y), (x-1, y), (x, y+1),
            (x, y-1)]
start = (0, 0)
goal = (5, 5)
path = a_star(start, goal, heuristic,
              neighbours)
print(path)

```

OUTPUT:

```

[(0, 0) (1, 0) (2, 0) (3, 0) (4, 0) (5, 0)
 (5, 1) (5, 2) (5, 3) (5, 4) (5, 5)]

```

~~Result,~~
~~thus the program is successfully created~~
~~and the output is working~~

Exp 10

(0000c = matlib.xmat) refixed 39.1M + libsvm

from AIM

Implementing Artificial Neural Networks

for an application using python classification

(libsvm.p, libsvm.py) www2.libsvm.org

CODE:

import numpy as np

from sklearn.datasets import make_circles

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

from sklearn.datasets import make_circles

from sklearn.model_selection import train_test_split

from sklearn.neural_network import MLPClassifier

X_train, y_train = make_circles(n_samples=100,
noise=0.05)

X_test, y_test = make_circles(n_samples=300,
noise=0.05)

sns.scatterplot(x=X_train[:, 0],

y=X_train[:, 1] # save line
hue=y_train)

plt.title("Train Data")

plt.show()

Model = MLPClassifier(max_iter=2000)

Model.fit(X_train, y_train)

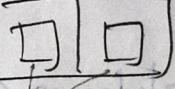
f.score(R2 score for training data =

{model.score(X_train, y_train)}

f.score(R2 score for test data =

{model.score(X_test, y_test)})

y_pred = model.predict(X_test)

fig, ax = plt.subplots(1, 2) 

sns.scatterplot(x=X_test[:, 0], y=X_test[:, 1])

test

hue = y_pred,

ax = ax[0] first plot same

ax[0].title.set_text("Predicted Data")

sns.scatterplot(x=X_test[:, 0], y=X_test[:, 1])

hue = y_test,

ax = ax[1] second plot same

ax[1].title.set_text("Test Data")

plt.show()

11.9.8

Result:

The code was successfully excited & the output was verified with output of
-microsoft word

plt & subplot

advant & Note \rightarrow droop^o & droop^u model - most
droop^o not able to be modelled most
droop^u not able to be modelled most
droop test most

①

model - droop^o double - layer model most

$$0.001 = \text{droop } o \rightarrow \text{no droop - dan} = p \times$$

$$20.3 = 2870.1$$

$$(0.01 = \text{water } o)$$

$$(0.0001, (0.01, 0.001)) = \text{sqrt } p \times \text{sqrt } X$$

$$p \times X \rightarrow \text{droop test most}, \text{test } X, \text{most } X$$

$$0.0001 = \text{droop}$$

$$\text{most } o \rightarrow \text{droop}$$

$$(0.01 = \text{water most})$$

Exp 11

ADM:

Implementing Artificial Neural Networks
for an application using Python
regression.

CODE:

```
Import numpy as np
Import matplotlib.pyplot as plt
Import seaborn as sns
from sklearn.datasets import make_regression
from sklearn.model_selection import train_test_split
from sklearn.neural_network import MLPRegressor
```

$x, y = \text{make_regression}(n_samples=1000,$
 $\text{noise}=0.05,$
 $n_features=100)$

~~$x_shape, y_shape = ((1000, 100), (1000, 1))$~~

$x\text{-train}, x\text{-test}, \text{train_test_split}(x, y,$
 $y\text{-train}, y\text{-test} =$
 $\text{test_size}=0.2,$
 $\text{shuffle}=\text{True},$
 $\text{random_state}=42)$

model = MLPRegressor(max_iter=2000)

model.fit(X_train, y_train)

print(f"R² score for training Data = {model.score(X_train, y_train)}")

print(f"R² score for test data = {model.score(X_test, y_test)}")

model = MLPRegressor(max_iter=2000).fit(X_train, y_train)

model = MLPRegressor(max_iter=2000).fit(X_train, y_train)

- ① Import $\text{from sklearn import MLPRegressor}$
- ② Database (snf_x) $\text{X} = \text{snf_x}$
- ③ snf_y ($\text{y} = \text{snf_y}$) $\text{y} = \text{snf_y}$
- ④ Model train $\text{model} = \text{MLPRegressor}(\text{max_iter}=2000)$
- ⑤ print score $\text{print(model.score(X, y))}$

Result:

The program was successfully executed

and the output was verified

1.01

Exp 12

(class = scikit-learn) reference: M.Labav
DECISION TREE CLASSIFICATION (code p. 180 & X) Jit. Labav

Topic -

AIM

To implement decision tree classification using Python

CODE:

```

from sklearn import preprocessing, import StandardScaler
from sklearn.tree import DecisionTreeClassifier
from google.colab import drive
drive.mount('/content/gdrive') →

```

```

import numpy as np | from sklearn import model_selection
import pandas as pd | import train_test_split
import matplotlib.pyplot as plt

```

```

dataset = pd.read_csv('/content/gdrive/My Drive/
                      Social_Network_Ads.csv')

```

dataset

```

X = dataset.iloc[:, [2, 3]].values
y = dataset.iloc[:, -1].values

```

split

```

X_train, X_test, y_train, y_test

```

```

= train_test_split(X, y, test_size=0.25,

```

random_state=0)

User ID	Credit Score	Age	Estimated Salary	Purchased
1	2	3	4	0, 1

Sc → | from sklearn.metrics import
Sc Model = StandardScaler() confusion_matrix

X_train = model.fit_transform(X_train)

X_test = model.transform(X_test) cr

model = DecisionTreeClassifier(criterion='entropy',
random_state=0)

model.fit(X_train, y_train)

y_pred = model.predict(X_test)

Cm = confusion_matrix(y_test, y_pred)

$$\begin{bmatrix} 62 & 9 \\ 3 & 29 \end{bmatrix}$$

| from matplotlib.colors import ListedColormap

X_set, y_set, X_train, y_train

X1, X2 = np.meshgrid(

np.arange(start=X_set[:, 0].min() - 1,

stop=X_set[:, 0].max() + 1,

step=0.01),

np.arange(start=X_set[:, 1].min() - 1,

stop=X_set[:, 1].max() + 1,

step=0.01)

)

$\text{plt.contourf}(X_1, X_2, \text{model.predict(np.array([X_1.ravel(), X_2.ravel()]).T)})$
 (flat) $\rightarrow \text{mesh} = \text{mesh} - X$
 $\cdot \text{ravel}(\text{X1.shape});$
 $\alpha = 0.75$
 $\text{cmap} = \text{ListedColormap}([\text{red}, \text{green}])$
 $\text{plt.xlim}(X_1.\min(), X_1.\max())$
 $\text{plt.ylim}(X_2.\min(), X_2.\max())$
 $(\text{berg}_{\text{p}} - \text{act}_{\text{p}}) \times \text{color norm} = \text{m}$

① Import

$[s \text{ } e8]$

$[e8 \text{ } e7]$

② Dataset

$[[x_{\text{p}}, y_{\text{p}}]]$

③ Split \rightarrow $x_{\text{train}}, y_{\text{train}}, x_{\text{test}}, y_{\text{test}}$

④ Scaling

$\rightarrow \text{normalization}$

⑤ Train model & predict

Result:

The program was successfully
executed & the output was printed
 $1.00 = 928$

$\rightarrow \text{new}([1, 1] \text{ for } X = \text{data})$
 $\rightarrow \text{new}([1, 1] \text{ for } X = \text{act})$
 $(1.0.0 = 928)$

EXP 13: IMPLEMENTATION OF DECISION TREE
CLASSIFICATION TECHNIQUE

AIM
To implement a decision tree classification technique using Python.

CODE:

```
from sklearn import tree
```

```
model = tree.DecisionTreeClassifier()
```

X = [[181, 80, 91]]

[182, 90, 92]

[183, 100, 92]

[184, 200, 93]

(1) [185, 800, 94]

[186, 400, 95]

[187, 500, 96]

[188, 600, 97]

[189, 700, 98]

[190, 800, 99]

[191, 900, 100]

[192, 1000, 101]

Y = [male
male]

female

male

female

male

female

male

female

male

female

male

model = model.fit(X, Y)

$\text{prediction}_f = \text{model}.\text{predict}([[15, 80, 91]])$

$\text{prediction}_m = \text{model}.\text{predict}([[183, 100, 92]])$

$\text{print}(\text{prediction}_f)$

$\text{print}(\text{prediction}_m)$

Output

with logistic model.

['male'] Unfixed, untrained, not learned

['female'] =Y | [15, 80, 181] =X

class

[SP, OP, 181]

[SP, OP, 81]

[SP, 001, 181]

Result:

[SP, 003, 181]

The program was ~~freezes~~ fully

class

[SP, 001, 181]

class

[SP, 002, 181]

class

[SP, 003, 181]

class

[SP, 001, 001]

class

[SP, 002, 181]

class

[SP, 003, 181]

class

[SP, 001, 281]

(X) A. I. don't know