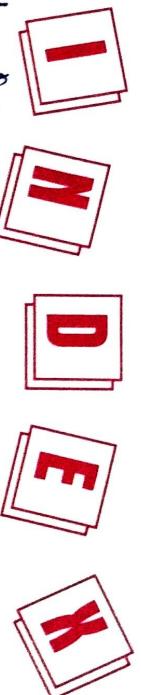


220701023



NAME : A. Ashwin STD : III year SEC. : A ROLL NO. : 20

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~~Completed~~

## N queen problem.

Aim:

To implement Nqueen problem

Program:

```
N = int(input("Enter the number of queens:"))
board = [[0]*N for _ in range(N)]
def is_safe(board, row, col):
    for i in range(len(board)):
        if board[i][col] == 1:
            return False
    for i, j in zip(range(row, -1, -1), range(col, -1, -1)):
        if board[i][j] == 1:
            return False
    return True
def solve_nqueens(board, col):
    if col >= N:
        return True
    for i in range(N):
        if solve_nqueens(board, col+1):
            board[col][i] = 1
            if solve_nqueens(board, col+1):
                return True
            board[col][i] = 0
    return False
if solve_nqueens(board, 0):
    print("Solution found:")
    for row in board
        print(row)
```

```
print(''.join(['o' if x == 1 else '*' for x in row]))
```

else:

Psinc ("No solution exists")

out put: can come up - as per the following

Enters the numbers of queens : 8

卷之三

卷之三

卷之三

卷之三

卷之三

卷之三

~~Pesult: (1911) 1911-1912~~

~~...the~~ N Green program has been  
successfully implemented.

## Depth first search.

Aim:  
To implement Depth First Search

Program:

def dfs - recursive (graph, start, visited = None):  
 if visited is None:  
 visited = set()

visited.add(start)

print(start)

for neighbor in graph[start]:

if neighbor not in visited:

dfs - recursive (graph, neighbor, visited)

graph = {

'A': ['B', 'C', 'D'],

'B': ['A', 'D', 'E'],

'C': ['A', 'F'],

'D': ['B', 'E'],

'E': ['B', 'F'],

'F': ['C', 'E']

print ("DFS recursive:")

dfs - recursive (graph, 'A')

def dfs - iterative (graph, start)

visited = set()

stack = [start]

while stack

vertex = stack.pop()

if vertex not in visited:

print (vertex)

visited.add (vertex)

stack.extend (neighbors for neighbor in

graph[vertex] if neighbor not in visited)

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

Print "DFS Iterative (graph, 'A')"

dfs\_iterative(graph, 'A')

Output: A B C D E H F

DFS Iterative:

A  
C  
F  
E  
B  
D



Result:

thus the depth first search has been implemented successfully.

Implementation of Depth First Search  
(using stack) using Python

## Algorithm

Alm. 1

## To Implement A<sup>t</sup> algorithm.

**Program:** *Programs will be developed by the*

from heap. import map.pop, heap.push

`def = init - (self, position, parent_name):`

self. Position = position

self. present = paixni

Self. 510

- 29 - (self, others) :

```
return self.position == others.position  
det_it_(self, others);
```

return cut. + 2 estates. +

`start = node = nodes[0];`

goal-node Node(goal)

open-list = set()

```
while open-list:
```

~~current - node - neapprop (open - list)~~

~~closed-list-add( $current-node$ , position)~~

Path = [ ]

卷之三

```
path.append(cusint_node.position)
current_node = current_node.parent
```

return path  $\tau : \vdash \Box$   
which becomes  $\vdash \Box (\Diamond \perp) \wedge \Diamond p$

for  $n$  in neighbours

    neighbour-position-current-node.position[0][ $i$ ]

        if  $o[i] = \text{neighbor\_position}[0]$  &  $\text{neighbor\_position}[1][i] < \text{len}(\text{grid}[0])$  and  $o[i] < \text{neighbor\_position}[1][i] < \text{len}(\text{grid}[0])$  and  $\text{grid}[\text{neighbor\_position}[0][i]][\text{neighbor\_position}[1][i]] == 0$ :

            neighbor-node = node + (neighbor-position, current-position) - node

            if neighbor-node == ~~neighbor-position~~ closed-list  
                continue

            neighbor-node.g = current-node.g + 1

            neighbor-node.h = abs(neighbor-node.position[0] - goal-node.position[0]) + abs(neighbor-

                node.position[1]).goal-node.position[1]

            neighbor-node.f = neighbor-node.g + neighbor-node.h

            if allneighbor-node != open-node for open-node in  
                neappush(open-list, neighbor-node)

        return None

grid = [

    [0, 0, 0, 0]

    [0, 1, 0, 1, 0]

    [0, 0, 0, 1, 0]

    [0, 1, 1, 1, 0]

    [0, 0, 0, 0, 0]

start = (0, 0)

goal = (4, 4)

path = a-star(start, goal, grid)

print ("Path found: ", path)

output:

path found = [(1, 0), (1, 1), (2, 0), (3, 0), (4, 0), (4, 1),  
(4, 2), (4, 3), (4, 4)]

on [0, 7+  
on [1, 7+  
d o l e n g  
A [neighbo  
r] = 20 :

current  
node  
used list

osited[0] -

neighbor -  
e - position  
bor-node[i]

n-node in  
n-list)



Result:

thus the A\* Algorithm has been  
implemented and executed successfully.

## A\* algorithm

aim:

To implement A\* algorithm

Program:

```
class graph:  
    def __init__(self, graph, heuristic):  
        self.graph = graph  
        self.heuristic = heuristic  
        self.solution = {}  
  
    def no_star(self, node):  
        print(f"Expanding: {node}")  
  
        if node not in self.graph or not self.graph[node]:  
            return  
  
        children = self.graph[node]  
        best = path = None  
        min_cost = float('inf')  
  
        for group in children:  
            cost = sum([self.heuristic[child] for child in group])  
            if cost < min_cost:  
                min_cost = cost  
                best = path = group  
  
        self.solution[node] = best  
  
    def print_lf("Best path for node  $i = \{best\_path\}$  with cost  $\{min\_cost\}$   
for child in best_path:  
    self.no_star(child)  
  
    def get_solution(self):  
        return self.solution  
  
graph = {  
    'n': ['C', 'B', 'D'], 'E': ['A', 'G']  
    'B': ['F', 'E']  
}
```

$A' = [C, D]$

$B' = [C, D]$

$C' = [D]$

$D' = [C]$

3

heuristic = 8

$A' = 0, B' = 1, C' = 2, D' = 4, E' = 1, F' = 3, G' = 5$

g

graph - obj = graph (graph, heuristic)

graph - obj & n - start('A')

solution = graph - obj . get - solution ()

print ("solution", solution)

output :

graph  
[node]:

Expanding : A

Best path for A = ['B', 'C'] with Best 3

Expanding : B

Best path for B = ['C', 'E'] with Best 1

Expanding : C, E, B, D, F, G by a path

Expanding = C

Best path = [C, G] with Best 1

Expanding : G

Solution : EA = [E, B, C], B' = [C] & C, G']  
3

path)  
min. cost)

result "No solution found" ->  
thus not implemented

successfully implemented - can't do

(graph - obj . get - solution)

graph - obj . get - solution

## Implementation of decision tree classification

Aim:

To implement a decision tree classification technique for gender classification technique using python.

Program:

```
import pandas as pd  
from sklearn import tree
```

Explanation:

- + Import the tree module from sklearn
- + Call the function from tree module
- + Assign value for X & Y
- + Call the function on basis of given random values can give feature + Display the output

```
program: # importing required modules  
import pandas as pd  
from sklearn.tree import DecisionTreeClassifier  
data = {  
    'Height': [170, 165, 155, 172, 185, 167, 180,  
               164, 177],  
    'Gender': ['female', 'female', 'male', 'male',  
              'female', 'male', 'female', 'male',  
              'female']  
}  
  
df = pd.DataFrame(data)  
X = df[['Height', 'Weight']]  
y = df['Gender']  
  
classifer = DecisionTreeClassifier()  
classifier = dt(X, y)  
  
height = float(input("Enter height"))  
weight = float(input("Enter weight"))
```

random values = np.random.rand(1000)

random values = np.random.rand(1000)

random values = np.random.rand(1000)

print("predicted gender for weight 75kg is male  
in and weight 60kg is female  
predicted gender for 75kg is male  
predicted gender for 60kg is female")

Random numbers generated from 0 to 1000  
predicted gender for weight 75kg is male  
predicted gender for weight 60kg is female  
predicted gender for 75kg is male  
predicted gender for 60kg is female

Output of the program is as follows  
~~predicted gender for weight 75kg is male~~

Thus the python program to implement a  
decision tree classification went well.  
~~gender classification is executed successfully.~~

(Program written by me in python)

(Program written by me in python)

night

## Implementation of ANN for an application using python - regression.

PL

Aim:

To implement artificial neural networks for an application in regression using

python

Program:

```
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from keras.layers import Input
from keras.optimizers import Adam
import matplotlib.pyplot as plt
# generating synthetic dataset for demo
x = np.random.rand(1000, 3)
y = 3*x[:, 0] + 2*x[:, 1] + 1.5 * np.sin(x[:, 2]) * np.random.normal(0, 0.1, 1000)
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)
x_train = StandardScaler().fit_transform(x_train)
x_test = StandardScaler().transform(x_test)
model = Sequential()
model.add(Dense(1, input_dim=x_train.shape[1], activation='relu'))
model.add(Dense(1, activation='relu'))
model.add(Dense(1, activation='relu'))
model.add(Dense(1, activation='relu'))
model.add(Dense(1, activation='relu'))
model.compile(optimizer=Adam(), loss='mean_squared_error', metrics=['accuracy'])
pred = model.predict(x_test)
plt.figure(figsize=(12, 6))
plt.plot(history.history['loss'], label='Train Loss')
plt.title('Training Loss')  
in r
```

plot (history), history (no loss), random plot (history), history (no loss), valid action loss

pre-train (training or validation loss)

pit. X 1400 ("open")

11) प्राचीन विद्या

Pit snow

Output:

Predicted: 0.06, Actual: 2500

卷之三十一

卷之三

卷之三

卷之三

L'ESPRESSO - 2 LUGLIO 1978

Re: (8) [REDACTED] [REDACTED]

~~100 80 100~~

1. *Leucosia* *leucostoma* *leucostoma* *leucostoma*

1922-1923

result.

~~This is the Python file.~~

~~an additional number~~ will be executed successfully.

~~in~~ regression

## Implementation of k-means clustering technique.

Output

Implementation of k-means clustering

### Aim:

To implement a k-means clustering technique using python language.

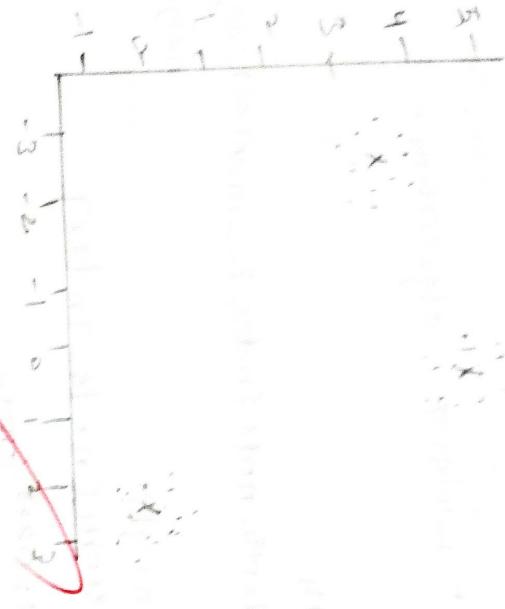
### Explanation:

- Import k-means from sklearn.cluster
- Assign x and y
- call the function kmeans()
- perform cluster operation and display the output

### Program:

```
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
x = [[1, 2], [1, 4], [1, 0], [4, 2], [4, 4]]
kmeans = KMeans(n_clusters=2)
kmeans.fit(x)
y = kmeans.labels_
centers = kmeans.cluster_centers_
for i, b in enumerate(y):
    plt.scatter(x[i][0], x[i][1], color='blue' if b == 0 else 'green')
plt.scatter(centers[:, 0], centers[:, 1], s=100, color='red', marker='x', zorder=100)
plt.xlabel('Feature 1')
plt.ylabel('Feature 2')
plt.title('K-means clustering')
plt.show()
```

output.



day  
blue

pos. result:

run the python program to implement  
k-means clustering technique is executed  
successfully.

## Minimax

Aim:

To implement minimax algorithm.

Program.

```
import math
def minimax(depth, node_index, is_maximizer, scores,
            height):
```

```
    if depth == height:
```

```
        return scores[node_index]
```

```
    if is_maximizer:
```

```
        return max(minimax(depth + 1, node_index,
                            False, scores, height),
                    minimax(depth + 1,
                            node_index - 1,
                            True, scores, height))
```

```
    else:
        return min(minimax(depth + 1, node_index,
                            True, scores, height),
                    minimax(depth + 1,
                            node_index - 1,
                            False, scores, height))
```

```
return max(minimax(0, 0, True, scores, height),
           minimax(0, 0, False, scores, height))
```

```
def calculate_tree_height(num_leaves)
```

```
    return math.ceil(math.log2(num_leaves))
```

```
scores = [3, 5, 6, 9, 1, 1, 2, 0, -1]
```

```
tree_height = calculate_tree_height(len(scores))
```

```
optimal_scores = minimax(0, 0, True, scores, tree_height)
```

```
print "The optimal scores are", optimal_scores
```

Optimal score

for i in range(tree\_height):

Output:  
The optimal score is = 5.

10

12x2

۱۰۷

18

earns)

un (secy)

34

10

1

~~performed~~ the decision tree program has been executed successfully.

## Introduction to prolog

Aim:

To learn previous technology and write basic programs.

Terminologies.

1) Atomic terms

They are usually strings made up of tokens and uppercase letters, digits and the underscore, starting with a lowercase letter.

Eg. dog. ab-c-321

2) Variables

They are strings of letters, digits and the underscores, starting with a capital letter or an underscore,

Eg: dog, apple-420

3) compound terms.

Compound terms are made up of a prolog atom and a number of arguments enclosed in parenthesis and separated by commas

Eg: is - bigger (elephant , x) . figure( - ), T)

4) fact

A fact is a prediction

dot

followed by a

Eg: bigger - animal (monkey) . like - a - beautiful

5) rules

Rules consists of a head and body

Eg: is - smaller (Y, X) :- bigger (Y, X) , count (Aunt, child)

sister (Aunt , Parent) , parent (Parent, Child)

QUESTION

PB 1

worm

wo

room

play

pasture

grass

quarrel

quarrel

over

over

output

? - res

? - true

? - plus

? - plus

? - true

? - plus

? - true

sousse (odd):

KB 1

woman (mia).

woman (jody).

woman (yolanda).

playsnir (quittery)

pasty.

unsty 1 = ? - woman (mia)

unsty 2 = ? - playsAlguitar (mia)

unsty 3 = ? - party

unsty 4 = ? - concert.

output:

? - woman (mia)

true

? - playsAlguitar (mia)

false

? - connect

error = unk now procedure. const 10

KB 2

play

and

listens2music (mia)

listen2music (yolanda) = - happy (yolanda)

playsAlguitar (mia) = listen2music (mia)

playsAlguitar (yolanda) listens2music (yolanda)

output:

? - playsAlguitar (mia)

true.

? - playsAlguitar (yolanda)

false

KB 3:

lives dan, katty!

lives (sally, den)

lives (john, britney)

masked (kerry) = - lives (kerry), lives (kry)

friends (x,y) = - lives (x,y), lives (y,x)

child).

child).

output

? - likes(dan,x)

x = sally

? - married(dan,sally)

true

? - married(john,brittney)

false.

HB4

food(burgers)

food(sandwich)

food(pizza)

lunch(sandwich)

dinner(pizza)

meal(x) = food(x)

output:

? - food(pizza)

true

? - meal(x), lunch(x)

x = sandwich

HB5.

owns(jack, car(bmw)).

owns(john, car(ccrury)).

own(jane, car(ccrury))

sedan(car(bmw))

truck(car(ccrury)).

output:

? - owns(john,x)

x = car(ccrury)

? - own(john,...)

true.

~~Result.~~

thus the basic Prolog  
we can encounter successfull programs may

5

gr

gr

## Prolog - family tree

Aim : To develop a family tree program using Prolog with all possible facts, rules and queries.

Source - code :

knowledge - base :

/\* Facts \*/

male (pat).

male (john).

male (chris).

male (leavin).

female (batty).

female (jenny).

female (lucy).

parent of (lucy, leavin).

parent of (lucy, batty).

parent of (leavin, pat).

parent of (leavin, chris).

parent of (leavin, john).

parent of (jenny, john).

parent of (jenny, lucy).

/ \ RULES = /\* 1

\* son, parent \*

\* son, grandparent \*

\* father (x, y) :- male (x), parent of (x, y).

\* mother (x, y) :- female (y), parent of (x, y).

\* grand father (x, y) :- male (x), parent of (x, z), parent of (z, y).

\* grand mother (x, y) :- female (x), parent of (x, z), parent of (z, y).

parent of (z, y)

\* brother (x, y) :- male (x), father (x, z), father (y, w), z = w.

\* sister (x, y) :- female (y), father (x, z), father (y, w), z = w.

z = w.

output

wade (petes)

true

james (elvis - petes)

true

farm (elvis, wade)

false

mothers (elvis, k)

X = belly

brother (elvis, jason)

false.

Result.

~~Their apology for family tree program  
has been executed successfully.~~