

INDEX

POAI

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```

def neumann_path(came_from, current):
    path = [current]
    while current in came_from:
        current = came_from[current]
        path.append(current)
    return path[1:-1]

```

```

def heuristic(a, b):
    return abs(a[0] - b[0]) + abs(a[1] - b[1])

```

```

def distance(a, b):
    return ((a[0] - b[0])**2 + (a[1] - b[1])**2)**0.5

```

```

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)

print(path)

OUTPUT

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

~~RESULT: The the program for A* Search has been
executed successfully.~~

EXN0:1

8 - QUEENS PROBLEM

AIM:-

To fm

To implement an 8-queens problem using python.

PROGRAM:

```

def share_diagonal(x0, y0, x1, y1):
    dx = abs(x0 - x1)
    dy = abs(y0 - y1)
    return dy == dx

```

def col_clashes(bs, c):

for i in range(c):

if share_diagonal(i, bs[i], c, bs[c]):

return True

return False

def has_clashes(the_board):

for col in range(1, len(the_board)):

if col_clashes(the_board - (col,)):

return True

return False

def main():

```

import random

n = int(input("Enter the number of queens: "))

rng = random.Random()
bd = list(range(n))

num_found = 0
tries = 0
result = []

while num_found < 1:
    rng.shuffle(bd)
    tries += 1

    if not has_clash_and_bd_not_in_result():
        print("found solution by in {} tries".format(tries))
        result.append(list(bd))
        num_found += 1

print(result)

main()

```

Output :-

Enter the number of queen: 9

Found solution [1, 6, 4, 7, 0, 3, 5, 2] in 94 tries.

Found Solution [3, 0, 4, 7, 6, 2, 6, 1] in 175 tries.

Found solution [2, 0, 6, 4, 7, 1, 3, 5] in 196 tries.

Found solution [5, 0, 4, 1, 7, 2, 6, 3] in 439 tries.

found solution [6, 2, 7, 1, 4, 0, 5, 3] in 156 tries.

[1, 6, 4, 7, 0, 3, 5, 2], [3, 0, 4, 7, 5, 2, 6, 1], [2, 0, 6, 4, 7, 1, 3, 5], [5, 0, 4, 1, 7, 2, 6, 3], [6, 2, 7, 0, 4, 0]

RESULT:-

thus the 8 queens problem was implemented successfully using backtracking algorithm.

EXNO:2

DEPTH FIRST SEARCH

16/08/2022

AIM:-

To implement a depth-first search problem using python.

PROGRAM:-

```

def dfs(graph, start, visited=None):
    if visited is None:
        visited = set()
    visited.add(start)
    print(f"start, end = {start,}")
    for neighbour in graph.get(start, {}):
        if neighbour not in visited:
            dfs(graph, neighbour, visited)

num_nodes = int(input("Enter the number of nodes:"))
graph = {}

for i in range(num_nodes):
    node = input("Enter node" + str(i+1) + ":")
    strip()

    neighbors = input("Enter neighbors of " + node +
                      " (comma-separated):").strip().split(',')

```

neighbors = [n.strip() for n in neighbors]

graph[node] = neighbors

print("Graph:", graph)

start_node = input("Enter the starting node:")
strip()

print("Starting node:", start_node)

dfs(graph, start_node)

OUTPUT:-

Enter the number of nodes: 5

Enter node 1: A

Enter neighbors of A (comma-separated): B,C,D

Enter node 2: B

Enter neighbors of B (comma-separated): C,D

Enter node 3: C

Enter neighbors of C (comma-separated): D

Enter node 4: D

Enter neighbors of D (comma-separated):

Enter node 5: E

Enter neighbors of E (comma-separated): F,G

Graph: { 'A': {'B', 'C', 'D'}, 'B': {'C', 'D'},

'C': {'D'}, 'D': {}, 'E': {'F', 'G'}} }

Enter the starting node A

starting node: A

A,B,C,D

RESULT:

~~SEARCH~~
RESULT:

Thus a ~~SEARCHING~~ searching technique using Depth-First Search algorithm was implemented successfully.

Exno.4

MINIMAX ALGORITHM

30/8/24

AIM:

To implement the minimax algorithm for a two-player game, with maximizer maximizing the score and minimizer minimizing it through DFS evaluation.

PROGRAM:

```
import math
```

```
def minimax (depth, node, index, is_maximiser, scores, height):
```

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

```
        return scores [node - index]
```

```
    if is_maximiser:
```

```
        return max (minimax (depth + 1, node - index * 2, False, scores, height))
```

else:

```
    return min (minimax (depth + 1, node - index * 2, True, scores, height))
```

```
minimax (depth + 1, node - index * 2 + 1, True, scores, height)
```

Scores = list (map (int, input ("Enter the scores separated by spaces: ") . split ())))

$t_{\text{tree}} - \text{height} = \text{calculate_tree_height}(\text{len}(\text{scores}))$

$\text{optimal_score} = \text{minimax}(0, 0, \text{time}, \text{scores}, t_{\text{tree}}$
 $- \text{height})$

print ("The optimal score is: " + optimal_score)

Output:

Enter the scores separated by spaces:-

1 4 2 6 3 5 0 7

The optimal score is: 4

Result:

Thus, the minimax algorithm successfully determines the players by evaluating the game-tree and selecting the best possible scores for maximizer and minimizer.

Exhibit

INTRODUCTION TO PROLOG

25/10/27

INTRODUCTION TO PROLOG

AIM:

To learn PROLOG terminologies and write basic programs.

TERMINOLOGIES:

1. Atomic Terms:

Atomic terms are usually strings made up of lower- and uppercase letters, digits, and the underscore, starting with a lowercase letter.

Ex:

dog

ab_c_321

2. Variables:

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

Ex:

dog

apple - 420

3. Compound Terms:

Compound terms are made up of a PROLOG atom and a number of arguments PROLOG terms, i.e atoms, numbers, variables, or other compound terms)

Ex:

ps - bigger (elephant, x)

fig (x, -) ?

4. Facts:

A fact is a predicate followed by a dot.

Ex:

bigger - animal (whale)

ps - ps - beautiful

5. Rules:

A rule consists of a head and a

body

Ex:

ps. smaller (x, y) :- ps bigger (y, x)

qint (Aunt, child) :- sister (Aunt, parent)
parent, child)

SOURCE CODE

KB1:

woman (mia)

woman (lody)

woman (yolanda)

plays AirGuitar (lody)

party

query 1: ? - woman

query 2: ? plays

query 3: ? party

query 4: ? concert

KB2:

happy

listens 2 music

listens 2 music - happy

plays AirGuitar - listen 2 music

plays AirGuitar - listens 2 music

KB3:

likes (dan, sally)

likes (sally, dan)

likes (john, britney)

married (x, y) :- like (x, y) likes (y, x)

friends(X,Y) :- like(X,Y) like(Y,X)

KB1:

food(burger)

food(sandwich)

food(pizza)

lunch(sandwich)

dinner(pizza)

meal(X) :- food(X)

KB2:

owns(jack, car(bmw))

owns(john, car(s chevy))

owns(olivia, car(civic))

owns(sam, car(s chevy))

Sedan car(bmw)

truck car(s chevy)

~~RESOLV.~~

ExNo: 7

PROLOG FAMILY TREE

AIM:-

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

SOURCE CODE:-

KNOWLEDGE BASE:

1st FACS: ! *)

male(peter)

male(john)

male(chris)

female(betty)

female(sony)

female(lisa)

~~parent of(chris, peter)~~

~~parent of(chris, betty)~~

~~parent of(helen, betty)~~

~~parent of(kevin, lisa)~~

~~parent of(sony, john)~~

~~parent of(sony, helo)~~

thus we have written basic programs to learn Prolog terminologies.

RULES:-

son. parent

son. grand parent

Father (x,y) :- male (y) parent of (x,y)

Mother (x,y) :- female (y) parent of (x,y)

Grand mother (x,y) :- female (y) parent of (x,y)

Brother (x,y) :- male (y), Father (x,z)

Father (y,w), z=w.

Sister (x,y) :- female (y), Father (x,z),

Father (y,w), z=w

Ex No.: 8

IMPLEMENTING ARTIFICIAL NEURAL NETWORKS FOR AN APPLICATION USING PYTHON - REGRESSION.

AIM:-

To implement artificial neural networks for an application in regression using Python.

PROGRAM:-

```
import numpy as np
from keras.models import Sequential
from sklearn.datasets import make_regression
x,y = make_regression(n_samples=1000)
x_train, x_test, y_train, y_test = train-test-split
(x,y, test state=42)
x_train = scalar.fit_transform(x_train)
x_test = scalar.transform(x_test)
model = Sequential()
model.compile(optimizer="adam", loss='mean-squared-error')
```

RESULT:-

~~Thus,~~ Thus, we have developed a family tree program using PROLOG with all possible facts, rule, and queries.

```
model.fit(x_train, y_train, epochs=5,  
          batch_size=32, verbose=1)
```

```
loss = model.evaluate(x_test, y_test)
```

```
print(f"Model Loss: {loss}")
```

```
y_pred = model.predict(x_test)
```

```
print("Predictions for the first 5 test  
samples", y_pred[:5])
```

Output :-

EPOCH 1/5

25/25 ————— 2s m/s/step - loss 15025.5816

EPOCH 2/5

25/25 ————— 0s ms/step - loss 14661.4593

EPOCH 3/5

25/25 ————— 0s ms/step - loss 12759.1582

EPOCH 4/5

25/25 ————— 0s ms/step - loss 9095.9023

EPOCH 5/5

7/7 ————— 0s ms/step

model loss: 9905.6953125

7/7

Predictions for the first 5 test samples

{8-1e+06316}

{9,110759}

{-11,91479}

{-11,280347}

{-1,4355274})

Output :-

EPOCH 1/5

25/25 ————— 2s m/s/step - loss 15025.5816

EPOCH 2/5

25/25 ————— 0s ms/step - loss 14661.4593

EPOCH 3/5

25/25 ————— 0s ms/step - loss 12759.1582

EPOCH 4/5

25/25 ————— 0s ms/step - loss 9095.9023

EPOCH 5/5

7/7 ————— 0s ms/step

model loss: 9905.6953125

7/7

Predictions for the first 5 test samples

{8-1e+06316}

Result:-

~~Thus,~~ we have successfully implemented
artificial neural networks for an application in
regression using python.

Ex. NO 9

27/09/23

IMPLEMENTATION OF DECISION TREE

CLASSIFICATION TECHNIQUES

AIM:-

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

EXPLANATION:

- import tree from sklearn.
- call the function decision tree from tree
- Assign value for x and y
- Call the function predict for given random values for each given feature.
- display the output.

PROGRAM:-

```
from sklearn import tree
x = [{160, 50, 37}, {160, 60, 39}, {170, 70, 39}, {165, 55, 36}]
y = [0, 0, 1, 1]
```

clf = tree.DecisionTreeClassifier()

clf = clf.fit(x, y)

prediction = clf.predict([175, 75, 41])

print("predicted gender(0 = Female, 1 = Male):", prediction[0])

OUTPUT:

predicted gender(0 = Female, 1 = Male) : 1

RESULT:

Thus, we have successfully implemented a decision tree classification techniques for gender classification.

Exno:9

04/10/29

IMPLEMENTATION OF CLUSTERING TECHNIQUE: MEANS

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 k-means().
- Perform scatter operation and display the output.

PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.datasets import make_blobs
plt.title('Generated data points')
```

plt.show()

K-means = K-means(n_clusters=4, random_state=42)

labels = k-means.labels

plt.scatter(x[0:3], x[2], c=labels, cmap='viridis', s=30)

plt.title('K-means clustering')

plt.legend()

plt.show()

print(end="n")

print("Centroids of the clusters")

print(centroids, end="n")

print(inertia = {})

for k in range(1,11):

k-means = K-means(n_clusters=k)

k-means.fit(x)

plt.plot(range(1,11), inertia, marker="o")

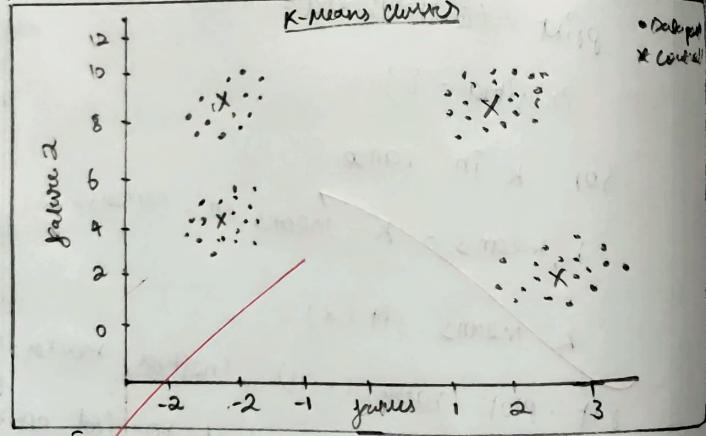
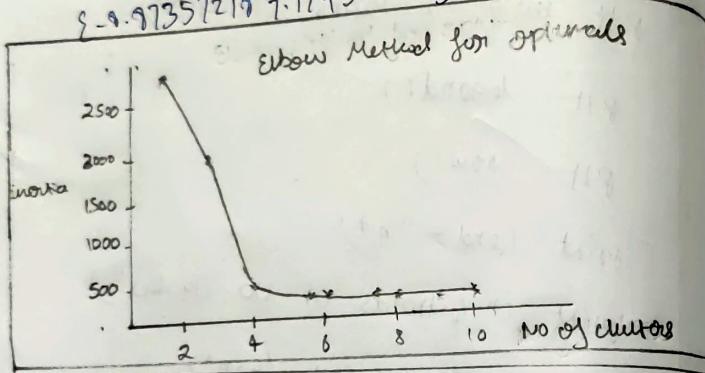
plt.title("Elbow method to find optimal k")

plt.ylabel("Inertia")

plt.show()

OUTPUT:

Centroids of the clusters:

$$\begin{cases} \{-2.70991136, 8.97143336\} \\ \{-6.9323205, -6.9304548\} \\ \{4.7192049, 2.049179676\} \\ \{-9.97357219, 7.174593427\} \end{cases}$$


RESULT:

Thus we have successfully implemented

a K-means clustering technique using
Python language.