

ANACONDA (SOFTWARE) INSTALLATION PROCEDURE

Guide to install Python with Anaconda

Step 1: Download Anaconda

Step 2: Download the Python 3 version for windows

Step 3: Double-click on the executable file

Step 4: Click Next-

Step 5: Click on "I agree to the terms and conditions" button.

Step 6: Select "Who You Want To Give Anaconda To"

Step 7: Select the installation location

Step 8: Select the environment variables

Step 9: Click Next and then "Finish"

Step 10: See if Python is installed.

OUTPUT:

[A', F', I', D', S']

Implement A* Search Algorithm.

PROGRAM :

```

def astarAlgo(start_node, stop_node):
    open_set = set([start_node])
    closed_set = set()
    g = {} # store distance from starting node.
    parents = {} # parents contains an adjacency map of all nodes

    #distance of starting node from itself is zero
    g[start_node] = 0
    # Start_node is root-node i.e. it has no parent nodes
    # so start_node is set to its own parent node
    parents[start_node] = start_node

    while len(open_set) > 0:
        n = None
        for v in open_set:
            if n == None or g[v] + heuristic(v) < g[n] + heuristic(n):
                n = v

        if n == stop_node or graph_nodes[n] == None:
            pass
        else:
            for (m, weight) in get_neighbours(n):
                #nodes 'm' not in first and last set are added to first.

```

n is set to its parent.

if m not in open-set and m not in closed-set:

open-set.add(m)

parents[m] = n

$g[m] = g[n] + \text{weight}$

#for each node m compare its distance from start i.e. $g(m)$

from start through n node

else:

if $g[m] > g[n] + \text{weight}$:

#update $g[m]$

$g[m] = g[n] + \text{weight}$

change parent of m to n

parents[m] = n

#if m in closed-set, remove and add to open

if m in closed-set:

closed-set.remove(m)

open-set.add(m)

if $r_i = \text{None}$:

print("Path does not exist!");

return None

if the current node is the stop-node

then we begin reconstructing the path from it to the start-node

if $n == \text{stop_node}$:

path = []

```

while parents[n] != n:
    path.append(n)
    n = parents[n]
path.append(start_node)
path.reverse()

print('path found: {}', format(path))
return path

```

remove n from the open-list and add it to the closed-list.
because all of his neighbours were inspected.

```

open_set.remove(n)
closed_set.add(n)

```

```

print("path does not exist!")
return None

```

define function to return neighbour & its distance
from the passed node

```

def get_neighbours(v):
    if v in graph_nodes[v]:
        return graph_nodes[v]
    else:
        return None

```

-# for simplicity we'll consider heuristic distances given
and this function returns heuristic distance for all nodes.

```

def heuristic(n):
    H-dist = {
        'A': 10,
        'B': 8,
        'C': 5,
        'D': 7,
        'E': 3,
        'F': 6,
        'G': 5,
        'H': 3,
        'I': 1,
        'J': 0
    }
    return H-dist[n]
}

```

#Describe your graph here

Graph nodes = 2

```

{'A': [(('B', 6), ('F', 3))],
 'B': [(('C', 3), ('D', 2))],
 'C': [(('D', 1), ('E', 5))],
 'D': [(('C', 1), ('E', 8))],
 'E': [(('I', 5), ('J', 5))],
 'F': [(('G', 1), ('H', 7))],
 'G': [(('I', 3))],
 'H': [(('I', 2))],
 'I': [(('E', 5), ('J', 3))],
 'J': None
}

```

astarAlg0 ('A'), ('J')

Implementation of A0* Search Algorithm

PROGRAM

```

class Graph:
    def __init__(self, graph, heuristicNodelist, startNode):
        #Instantiate graph object with graph topology, heuristic values, start node
        self.graph = graph
        self.H = heuristicNodelist
        self.start = startNode
        self.parent = {}
        self.status = {}
        self.solutionGraph = {}

    def applyAOstar(self): #starts a recursive A0* algorithm
        self.a0Star(self.start, False)

    def getNeighbors(self, v): #get the Neighbors of a given node
        return self.graph.get(v, {})

    def getStatus(self, v): #return the status of a given node
        self.status[v] = val

    def getHeuristicNodeValue(self, n):
        return self.H.get(n, 0) #always return heuristic value of given node

    def printSolution(self):
        print("FOR GRAPH SOLUTION, TRAVERSE THE GRAPH FROM START NODE: ", self.start)

```

OUTPUT:

Heuristic values : { 'A': 1, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 13 }

Solution graph = {}

Processing Node : A

Heuristic Values : { 'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 13 }

Solution Graph = {}

Processing Node : B

Heuristic value : { 'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 13 }

Solution graph = {}

Processing Node : A

Heuristic Values : { 'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 13 }

Solution graph = {}

Processing Node : G

Heuristic Value : { 'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 7, 'J': 13 }

Solution graph = {}

Processing Node : B

Heuristic Values : { 'A': 10, 'B': 8, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 7, 'J': 13 }

Solution graph = {}

Processing Node : A

Heuristic values : { 'A': 12, 'B': 8, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 0, 'J': 13 }

Solution graph : { 'I' : [] }

Processing Node : G

```

print( " - - - - - ")
print( self. solutionGraph )
print( " - - - - - ")

def computeMinimumCost( childNodeList self, v ):
    # computes the minimum cost of child nodes of a given node v
    minimumCost = 0
    costToChildNodeListDict = {}
    costToChildNodeListDict[minimumCost] = []
    flag = True
    for nodeInfoTupleList in self. getNeighbors( v ):
        # iterate over the set of child nodes
        cost = 0
        nodeList = []
        for c, weight in nodeInfoTupleList:
            cost = cost + self. getHeuristicNodeValue( c ) + weight
            nodeList.append( c )
        if flag == True: # initialize Minimum Cost with the cost of
                        # first set of child Nodes.
            minimumCost = cost
            costToChildNodeListDict[minimumCost] = nodeList
            # set the minimum cost child Node/s.
            flag = False
        else:          # checking the Minimum cost Node/s with current
                      # minimumCost > cost:
            if minimumCost > cost:
                minimumCost = cost
                costToChildNodeListDict[minimumCost] = nodeList
                # set the minimum cost child node/s.

```

Heuristic values: $\{A: 6, B: 2, C: 1, D: 10, E: 2, F: 1, G: 1,$
 $H: 7, I: 0, J: 0\}$

Solution graph = $\{J: [], G: [I], B: [G], J: [], C: [J]\}$

Processing Node: A

For Graph solution, Traverse the graph from the start Node: A

$\{J: [], G: [I], B: [G], J: [], C: [J], A: [B, C]\}$

{} - Initial state
{} - Environment state
{} - Environment state after
and post

(Corresponding the next step of action 1)
shown below (for soft move startill

[] - initial

change of state on action 1 ref

new env, initial state after (pre + 1) 200 = 200

Environment state

From this we can find minimum cost to go to J
initial state pre with

initial state (minimum cost)

initial state [] + min cost 200 = 200

Environment state after (pre + 1)

Expt. No. 2

return minimumCost, costToChildNodeListDict[minimumCost]
 #return the Minimum cost and Minimum cost child node/s.

def aStar(self, v, backtracking):

#A* algorithm for a start Node & backtracking status flag

print ("HEURISTIC VALUES:", self.H)

print ("SOLUTION GRAPH:", self.solutionGraph)

print ("PROCESSING NODE:", v)

print ("-----")

if self.getStatus(v) >= 0:

#if status node v >= 0, compute Minimum Cost nodes of v
 minimumCost, childNodeList = self.computeMinimumCost(childNode(v))
 self.setHeuristicNodeValue(v, minimumCost)
 self.setStatus(v, len(childNodeList))
 solved = True #check the Minimum Cost nodes of v are resolved

for childNode in childNodeList:

self.parent[childNode] = v

if self.getStatus(childNode) != -1:

solved = solved & False

if solved == True:

self.setStatus(v, -1)

self.solutionGraph[v] = childNodeList

if v != self.start:

self.aStar(self.parent[v], True)

```

if backtracking == False :
    for childNode in childNodeList :
        self.setStatus(childNode, 0)
        self.adStar(childNode, False)
    
```

$h = \{ 'A': 1, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7,$
 $'I': 7, 'J': 1 \}$

graph h = { 'A': [[('B', 1)], [('C', 1)], [('D', 1)]],
 $'B': [[('G', 1)], [('H', 1)]]$
 $'C': [[('J', 1)]]$
 $'D': [[('E', 1)], [('F', 1)]]$
 $'G': [[('F', 1)]]$

3

$g1 = \text{Graph}(\text{graph}(h, 'A'))$

$g1.\text{apply ADStar}()$

$g1.\text{printSolution}()$

For a given set of training data examples stored in a .csv file, implement and demonstrate the candidate elimination algorithm to output a description of the set of all hypothesis consistent with the training example.

PROGRAM

```
import numpy as np
import pandas as pd
```

```
data = pd.read_csv("enjoysport.csv")
concepts = np.array(data.iloc[:, 0:-1])
print(concepts)
target = np.array(data.iloc[:, -1])
print(target)
```

```
def learn(concepts, target):
    specific_h = concepts[0].copy()
    print("Initialization of specific-h and general-h")
    print(specific_h)
    general_h = [[ "?" for i in range(len(specific_h)) ] for i in
                 range(len(specific_h))]
    print(general_h)
    for i, h in enumerate(concepts):
        print("For loop starts")
        if target[i] == "yes":
            print("if instance is positive")
            for x in range(len(specific_h)):
                if h[x] != specific_h[x]:
                    specific_h[x] = "?"
                    general_h[x][x] = "?"

```

```

if target[i] == "no":
    print("if instance is negative")
    for x in range(len(specific-h)):
        if h[x] != specific-h[x]:
            general-h[x][x] = specific-h[x]
        else:
            general-h[x][x] = '?'

```

```

print("Steps of candidate elimination algorithm", i+1)
print(specific-h)
print(general-h)
print("\n")
print("\n")

```

```

indices = (i for i, val in enumerate(general-h) if
           val == ['?', '?', '?', '?', '?', '?', '?', '?', '?'])
for i in indices:
    general-h.remove(['?', '?', '?', '?', '?', '?', '?', '?', '?'])
return specific-h, general-h

```

```

s-final, g-final = learn(concept, target)
print("Final Specific-h:", s-final, sep = "\n")
print("Final general-h:", g-final, sep = "\n")

```

PROGRAM

```
import pandas as pd  
import math  
import numpy as np  
  
data = pd.read_csv('3-dataset.csv')  
features = [feat for feat in data]  
features.remove("answer")
```

```
class Node:  
    def __init__(self):  
        self.children = []  
        self.value = ""  
        self.isLeaf = False  
        self.pred = ""
```

```
def entropy(examples):  
    pos = 0.0  
    neg = 0.0
```

```
for _, row in examples.iterrows():  
    if row["answer"] == "yes":  
        pos += 1  
    else:  
        neg += 1
```

```

if pos == 0.0 or neg == 0.0:
    return 0.0
else:

```

$$P = \text{pos} / (\text{pos} + \text{neg})$$

$$n = \text{neg} / (\text{pos} + \text{neg})$$

$$\text{return} - (P * \text{math.log}(P, 2)) - n * \text{math.log}(n, 2)$$

```
def info_gain(examples, attr):
```

$$\text{unique} = \text{np.unique}(\text{examples[attr]})$$

$$\text{gain} = \text{entropy}(\text{examples})$$

$$\text{for } u \text{ in unique:}$$

$$\text{subdata} = \text{examples}[\text{examples[attr]} == u]$$

$$\text{sub_e} = \text{entropy}(\text{subdata})$$

$$\text{gain} -= (\text{float}(\text{len}(\text{subdata})) / \text{float}(\text{len}(\text{examples}))) * \text{sub_e}$$

$$\text{return gain.}$$

```
def ID3(examples, attrs):
```

$$\text{root} = \text{Node}()$$

$$\text{max_gain} = 0$$

$$\text{max_feat} = "$$

$$\text{for feature in attrs:}$$

$$\text{gain} = \text{info_gain}(\text{examples}, \text{feature})$$

$$\text{if gain} > \text{max_gain}: \quad$$

$$\text{max_gain} = \text{gain}$$

$$\text{max_feat} = \text{feature.}$$

$$\text{root.value} = \text{max_feat}$$

$$\text{unique} = \text{np.unique}(\text{examples}[\text{max_feat}])$$

for c in uniq:

 subdata = examples[examples['max-feat'] == c]

 if entropy(subdata) == 0.0:

 newNode = Node()

 newNode.isLeaf = True

 newNode.value = c

 newNode.pred = np.unique(subdata["answer"])

 root.children.append(newNode)

 else:

 dummyNode = Node()

 dummyNode.value = None

 newAttrs = attrs.copy()

 newAttrs.remove("max-feat")

 child = ID3(subdata, newAttrs)

 dummyNode.children.append(child)

 root.children.append(dummyNode)

return root.

def printTree(root: Node, depth=0):

 for i in range(depth):

 print("\t", end=" ")

 print(root.value, end=" ")

 if root.isLeaf:

 print("→", root.pred)

 print()

 for child in root.children:

 printTree(child, depth+1)

root = ID3(data, features)

printTree(root)

Build an artificial Neural Network by implementing the Backpropagation algorithm and test the same using appropriate data links.

PROGRAM

```
import numpy as np
```

```
x = np.array([[8, 9], [1, 5], [3, 6]], dtype=float)
```

```
y = np.array([98, 86, 89], dtype=float)
```

```
x = x / np.amax(x, axis=0)
```

```
y = y / 100
```

```
def sigmoid(x):
```

```
    return 1 / (1 + np.exp(-x))
```

```
def derivatives_Sigmoid(x):
```

```
    return x * (1 - x)
```

```
epoch = 5000
```

```
lr = 0.1
```

```
inputlayer_neurons = 2
```

```
hiddenlayer_neurons = 3
```

```
output_neurons = 1
```

```
w_h = np.random.uniform(size=(inputlayer_neurons, hiddenlayer_neurons))
```

```
b_h = np.random.uniform(size=(1, hiddenlayer_neurons))
```

```
w_out = np.random.uniform(size=(hiddenlayer_neurons, output_neurons))
```

```
b_out = np.random.uniform(size=(1, output_neurons))
```

```
for i in range(epoch):
```

```
    hinpl = np.dot(x, wh)
```

```
    hinp = hinpl + bb
```

```
    hlayer_act = sigmoid(hinp)
```

```
    outinpl = np.dot(hlayer_act, wout)
```

```
    outinp = outinpl + bout
```

```
    output = sigmoid(outinp)
```

```
EO = y_output
```

```
outgrad = derivatives.sigmoid(y_output)
```

```
d_output = EO * outgrad
```

```
EH = d_output . dot(wout - r)
```

```
hiddengrad = derivatives.sigmoid(hlayer_act)
```

```
d_hiddenlayer = EH * hiddenlayer
```

```
wout_t = hlayer_act.T . dot(d_output) * lr
```

```
wh_t = x.T . dot(d_hiddenlayer) * lr
```

```
print("Input: \n" + str(x))
```

```
print("Actual Output: \n" + str(y))
```

```
print("Predicted Output: \n", output)
```

Write a program to implement the naïve Bayesian classifier for a sample training data set stored as a .csv file. Compute the accuracy of the classifier, considering few test data sets.

PROGRAM

```
import pandas as pd
```

```
PlayTennis = pd.read_csv("H.csv")
print("Given dataset :\n", PlayTennis, "\n")
```

```
from sklearn.preprocessing import LabelEncoder
Le = LabelEncoder()
```

```
PlayTennis['Outlook'] = Le.fit_transform(PlayTennis['Outlook'])
PlayTennis['temp'] = Le.fit_transform(PlayTennis['temp'])
PlayTennis['humidity'] = Le.fit_transform(PlayTennis['humidity'])
PlayTennis['wind'] = Le.fit_transform(PlayTennis['wind'])
PlayTennis['play'] = Le.fit_transform(PlayTennis['play'])
```

```
print("The encoded dataset is :\n", PlayTennis)
```

```
x = PlayTennis.drop(['play'], axis=1)
y = PlayTennis['play']
```

```
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import accuracy_score
```

```
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.20)
```

```
print ("\\n x-train:\\n", x-train)
print ("\\n y-train:\\n", y-train)
print ("\\n x-test:\\n", x-test)
print ("\\n y-test:\\n", y-test)
```

```
classifier = GaussianNB()
```

```
classifier.fit (x-train, y-train)
```

```
accuracy = accuracy_score (classifier.predict (x-test), y-test)
```

```
print ("\\n Accuracy is : ", accuracy).
```