

Tribhuvan University Faculty of Humanities and Social Sciences

Artificial Intelligence

A LAB REPORT

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Department of Computer Application

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Write a python program for the implementation of vacuum cleaner.

```
import random
class VacuumCleaner:
  def __init__(self, rows, cols):
    self.rows = rows
    self.cols = cols
    self.environment = [[random.choice(['Clean', 'Dirty']) for _ in range(cols)] for _ in range(rows)]
  def initial environment( self ):
    print("Current Environment:")
    print("========
    for row in self.environment:
      print( " | ".join( row ) )
    print("=
    print()
  def clean_environment( self ):
    print( "Cleaned Environment:" )
    print("======"")
    for row in self.get_clean_environment():
       print( " | ".join( row ) )
    print("======"")
    print()
  def get_clean_environment( self ) :
    row_count = 0
    clean_environment = []
    for row in self.environment:
      column\_count = 0
      clean_environment.append([])
       for column in row:
         if( column == 'Dirty' ) :
           column = 'Clean'
           ( clean_environment[ row_count ] ).append( column )
         else:
           ( clean_environment[ row_count ] ).append( column )
         column\_count = column\_count + 1
       row_count = row_count + 1
    return clean_environment
# Initialize and run the vacuum cleaner simulation
if __name__ == "__main__":
  rows = 3
  cols = 3
  vacuum_cleaner = VacuumCleaner(rows, cols)
  vacuum_cleaner.initial_environment()
  vacuum_cleaner.clean_environment()
```

Current Environment:		
Dirty Dirty Clean	Dirty	
Cleaned Environment:		
Clean Clean Clean	Clean	
	Dirty Dirty Clean Environ Clean Clean	

Write a python program for the implementation of depth first search (DFS).

```
class Graph:
  def __init__(self):
     self.graph = {} # Adjacency list representation
  def add_edge(self, vertex, neighbor):
     """Add an edge to the graph."""
     if vertex not in self.graph:
       self.graph[vertex] = []
     self.graph[vertex].append(neighbor)
  def dfs(self, start, visited=None):
     """Recursive implementation of Depth First Search."""
     if visited is None:
       visited = set()
     visited.add(start)
     print(start, end=" ") # Print the current vertex
     for neighbor in self.graph.get(start, []):
       if neighbor not in visited:
          self.dfs(neighbor, visited)
  def dfs_iterative(self, start):
     """Iterative implementation of Depth First Search."""
     visited = set()
     stack = [start]
     while stack:
       vertex = stack.pop()
       if vertex not in visited:
          print(vertex, end=" ")
          visited.add(vertex)
          # Add neighbors to the stack (in reverse order for correct traversal)
          stack.extend(reversed(self.graph.get(vertex, [])))
# Example usage
if __name__ == "__main__":
  g = Graph()
  g.add\_edge(0, 1)
  g.add_edge(0, 2)
  g.add edge(1, 3)
  g.add\_edge(1, 4)
  g.add_edge(2, 5)
  g.add\_edge(2, 6)
  print("DFS (Recursive):")
  g.dfs(0)
  print("\nDFS (Iterative):")
  g.dfs_iterative(0)
```

```
DFS (Recursive):
0 1 3 4 2 5 6
DFS (Iterative):
0 1 3 4 2 5 6
PS D:\College\7th Semester\Artificial Intelligence\Lab Codes>
```

Question -3

Write a python program for the implementation of breadth first search (BFS).

Code:

```
from collections import deque
class Graph:
  def __init__(self):
     self.graph = {} # Adjacency list representation
  def add edge(self, vertex, neighbor):
     """Add an edge to the graph."""
    if vertex not in self.graph:
       self.graph[vertex] = []
     self.graph[vertex].append(neighbor)
  def bfs(self, start):
     """Implementation of Breadth-First Search."""
     visited = set() # To track visited nodes
     queue = deque([start]) # Queue for BFS (FIFO)
     print("BFS Traversal:", end=" ")
     while queue:
       vertex = queue.popleft() # Dequeue the front of the queue
       if vertex not in visited:
          print(vertex, end=" ") # Process the current node
          visited.add(vertex)
         # Enqueue all unvisited neighbors
          for neighbor in self.graph.get(vertex, []):
            if neighbor not in visited:
              queue.append(neighbor)
# Example usage
if __name__ == "__main__":
  g = Graph()
  g.add edge(0, 1)
  g.add\_edge(0, 2)
  g.add\_edge(1, 3)
  g.add_edge(1, 4)
  g.add\_edge(2, 5)
  g.add\_edge(2, 6)
  print( "Graph: ", g.graph )
 g.bfs(0)
```

```
Graph: {0: [1, 2], 1: [3, 4], 2: [5, 6]}
BFS Traversal: 0 1 2 3 4 5 6
PS D:\College\7th Semester\Artificial Intelligence\Lab Codes>
```

Write a python program for the implementation of uniform cost search.

```
import heapq
class Graph:
  def __init__(self):
    self.graph = {} # Adjacency list: {node: [(neighbor, cost), ...]}
  def add_edge(self, vertex, neighbor, cost):
    """Add an edge with a cost to the graph."""
    if vertex not in self.graph:
       self.graph[vertex] = []
    self.graph[vertex].append((neighbor, cost))
    # For undirected graphs, add the reverse edge
    if neighbor not in self.graph:
       self.graph[neighbor] = []
    self.graph[neighbor].append((vertex, cost))
  def uniform_cost_search(self, start, goal):
    """Uniform Cost Search (UCS) implementation."""
    # Priority queue to hold nodes and their cumulative costs
    priority_queue = [(0, start)] # (cumulative_cost, vertex)
    visited = set()
    while priority_queue:
       cost, current = heapq.heappop(priority_queue) # Pop the node with the lowest cost
       if current in visited:
         continue
       visited.add(current)
       print(f"Visited Node: {current}, Cost: {cost}")
       # Goal test
       if current == goal:
         print(f"Goal Node {goal} reached with Cost: {cost}")
         return
       # Expand neighbors
       for neighbor, edge_cost in self.graph.get(current, []):
         if neighbor not in visited:
            heapq.heappush(priority_queue, (cost + edge_cost, neighbor))
    print("Goal not reachable!")
    return
# Example usage
if __name__ == "__main__":
  g = Graph()
  g.add_edge("A", "B", 1)
  g.add_edge("A", "C", 4)
  g.add_edge("B", "C", 2)
  g.add_edge("B", "D", 6)
  g.add_edge("C", "D", 3)
  g.add_edge("C", "E", 5)
  g.add_edge("D", "E", 1)
  print("Uniform Cost Search from A to E:")
  g.uniform_cost_search("A", "E")
```

```
Uniform Cost Search from A to E:
Visited Node: A, Cost: 0
Visited Node: B, Cost: 1
Visited Node: C, Cost: 3
Visited Node: D, Cost: 6
Visited Node: E, Cost: 7
Goal Node E reached with Cost: 7
PS D:\College\7th Semester\Artificial Intelligence\Lab Codes>
```

Write a python program for the implementation of greedy best first search.

```
import heapq
class Graph:
  def __init__(self):
     self.graph = {} # Adjacency list: {node: [(neighbor, cost), ...]}
     self.heuristics = {} # Heuristic values: {node: heuristic value}
  def add edge(self, vertex, neighbor, cost):
     """Add an edge with a cost to the graph."""
     if vertex not in self.graph:
       self.graph[vertex] = []
     self.graph[vertex].append((neighbor, cost))
  def set_heuristic(self, node, value):
     """Set heuristic value for a node."""
     self.heuristics[node] = value
  def greedy_best_first_search(self, start, goal):
     """Greedy Best-First Search implementation."""
     priority_queue = [(self.heuristics[start], start)] # (heuristic_value, vertex)
     visited = set()
     print("Path Traversed:", end=" ")
     while priority queue:
       h_value, current = heapq.heappop(priority_queue) # Node with smallest heuristic value
       if current in visited:
          continue
       print(current, end=" -> ")
       visited.add(current)
       # Goal test
       if current == goal:
          print("Goal Reached!")
          return
       # Explore neighbors
       for neighbor, _ in self.graph.get(current, []):
          if neighbor not in visited:
             heapq.heappush(priority_queue, (self.heuristics[neighbor], neighbor))
     print("Goal not reachable!")
     return
```

```
# Example usage
if __name__ == "__main__":
  g = Graph()
  # Adding edges to the graph
  g.add_edge("A", "B", 1)
  g.add_edge("A", "C", 4)
  g.add_edge("B", "D", 6)
g.add_edge("B", "E", 2)
g.add_edge("C", "F", 5)
  g.add_edge("E", "G", 1)
  g.add_edge("F", "G", 2)
  # Setting heuristic values (lower = better)
  g.set_heuristic("A", 7)
  g.set_heuristic("B", 4)
  g.set_heuristic("C", 6)
  g.set_heuristic("D", 8)
  g.set_heuristic("E", 2)
  g.set_heuristic("F", 3)
  g.set_heuristic("G", 0)
  print("Greedy Best-First Search from A to G:")
  g.greedy_best_first_search("A", "G")
```

```
Greedy Best-First Search from A to G:
Path Traversed: A -> B -> E -> G -> Goal Reached!
PS D:\College\7th Semester\Artificial Intelligence\Lab Codes>
```

Write a python program for the implementation of A* search.

```
import heapq
class Graph:
  def __init__(self):
    self.graph = {} # Adjacency list: {node: [(neighbor, cost), ...]}
    self.heuristics = {} # Heuristic values: {node: heuristic_value}
  def add_edge(self, vertex, neighbor, cost):
     """Add an edge with a cost to the graph."""
    if vertex not in self.graph:
       self.graph[vertex] = []
    self.graph[vertex].append((neighbor, cost))
  def set_heuristic(self, node, value):
     """Set heuristic value for a node."""
    self.heuristics[node] = value
  def a_star_search(self, start, goal):
     """A* Search implementation."""
    # Priority queue to hold nodes and their f(n) = g(n) + h(n)
    open_list = [(0 + self.heuristics[start], 0, start)] # (f(n), g(n), node)
    g_{costs} = \{start: 0\} \# Dictionary to track g(n) values
    came_from = {} # To reconstruct the path
    visited = set()
    print("Path Traversed:", end=" ")
     while open_list:
       \underline{\ }, g, current = heapq.heappop(open_list) # Get the node with lowest f(n)
       if current in visited:
         continue
       print(current, end=" -> ")
       visited.add(current)
       # Goal test
       if current == goal:
         print("Goal Reached!")
          self.reconstruct_path(came_from, goal)
          return
       # Explore neighbors
       for neighbor, cost in self.graph.get(current, []):
          tentative g = g + cost
          if neighbor not in g_costs or tentative_g < g_costs[neighbor]:
            g_costs[neighbor] = tentative_g
            f = tentative_g + self.heuristics.get(neighbor, float('inf'))
            heapq.heappush(open_list, (f, tentative_g, neighbor))
            came_from[neighbor] = current
     print("Goal not reachable!")
    return
```

```
def reconstruct_path(self, came_from, goal):
     """Reconstruct and print the path from start to goal."""
     path = []
     current = goal
     while current in came_from:
        path.append(current)
        current = came_from[current]
     path.append(current)
     path.reverse()
     print(" -> ".join(path))
# Example usage
if __name__ == "__main__":
  g = Graph()
  # Adding edges (graph structure)
  # Adding edges (graph st
g.add_edge("A", "B", 1)
g.add_edge("A", "C", 4)
g.add_edge("B", "D", 6)
g.add_edge("B", "E", 2)
g.add_edge("C", "F", 5)
g.add_edge("E", "G", 1)
g.add_edge("F", "G", 2)
  # Setting heuristic values (lower = better)
  g.set_heuristic("A", 7)
  g.set_heuristic("B", 4)
  g.set_heuristic("C", 6)
  g.set_heuristic("D", 8)
  g.set_heuristic("E", 2)
  g.set_heuristic("F", 3)
  g.set_heuristic("G", 0)
  print("A* Search from A to G:")
  g.a_star_search("A", "G")
Output:
      A* Search from A to G:
      Path Traversed: A -> B -> E -> G -> Goal Reached!
      A \rightarrow B \rightarrow E \rightarrow G
```

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Question-7

Write a program to implement logic programming using prolog.

Code:

```
dog(rover).
dog(felix).
dog(benny).
animal(A):-dog(A).
```

```
1 ?- consult("question-1.pl").
true.
2 ?- animal(A).
A = rover;
A = felix;
A = benny.
3 ?- dog(rover).
true.
4 ?- dog(cat)
.
false.
```

Question - 8

Write a prolog program to represent few basic facts and perform queries (Elephant is an animal. Elephant is bigger than horse) etc.

Code:

```
animal(elephant).
animal(horse).
animal(dog).
animal(cat).

bigger_than(elephant, horse).
bigger_than(horse, dog).
bigger_than(dog, cat).

is_bigger(X, Y) :- bigger_than(X, Y).
is_bigger(X, Y) :- bigger_than(X, Z), is_bigger(Z, Y).
```

```
1 ?- consult("question-2.pl").
true.
2 ?- animal(elephant).
true.
3 ?- is_bigger(elephant, dog).
true .
4 ?- is_bigger(dog, elephant).
false.
5 ?- bigger_than(A, B).
A = elephant,
B = horse;
A = horse,
B = dog;
A = dog,
B = cat.
```

Question - 9

Write a program in prolog to implement simple arithmetic.

Code:

```
add(X, Y, Result) :- Result is X + Y.
subtract(X, Y, Result) :- Result is X - Y.
multiply(X, Y, Result) :- Result is X * Y.
divide(X, Y, Result) :- Y \= 0, Result is X / Y.
```

```
1 ?- consult("question-3.pl").
true.
2 ?- add(10, 1, 11).
true.
3 ?- add(10, 2, 20).
false.
4 ?- subtract(1,1, 0).
true.
5 ?- subtract(1,1, 2).
false.
6 ?- multiply(1, 1, 1).
true.
7 ?- multiply(1, 1, 2).
false.
8 ?- divide(1, 1, 1).
9 ?- divide(1, 1, 10).
false.
```

Write a prolog program to convert the sentences into FOPL and to execute queries

Code:

```
loves(likesh, lumanti).
loves(lumanti, likesh).
man(likesh).
woman(lumanti).
likes(om, yomari).
likes(puza, samebaji).
parent(om, sangita).
parent(puza, sangita).
parent(anish, om).

father(X, Y) :- parent(X, Y), man(X).
mother(X, Y) :- parent(X, Y), woman(X).
lovers(X, Y) :- loves(X, Y), loves(Y, X).
```

```
1 ?- consult("question-4.pl").
true.
2 ?- lovers(likesh, lumanti).
true.
3 ?- mother(Mother, sangita).
false.
4 ?- parent(A, B).
A = om,
B = sangita;
A = puza,
B = sangita ;
A = anish,
B = om.
5 ?- likes(A, B).
A = om,
B = yomari;
A = puza,
B = samebaji.
```

Write a prolog program to create a knowledge base of sentences and to execute queries.

```
male(rajesh).
male(sunil).
male(prakash).
male(ramesh).
female(anita).
female(sita).
female(gita).
female(laxmi).
parent(rajesh, sita).
parent(rajesh, sunil).
parent(gita, sita).
parent(gita, sunil).
parent(sunil, prakash).
parent(laxmi, prakash).
% Rules
father(X, Y) := parent(X, Y), male(X).
mother(X, Y) :- parent(X, Y), female(X).
sibling(X, Y) :- parent(Z, X), parent(Z, Y), X = Y.
grandparent(X, Y) :- parent(X, Z), parent(Z, Y).
Output:
1 ?- consult("question-5.pl").
                                 4 ?- parent(A, B).
                                                                 5 ?- father(rajesh, sita).
true.
                                                                 true.
                                 A = rajesh,
                                 B = sita;
2 ?- male(A).
                                                                 6 ?- father(rajesh, gita).
                                 A = rajesh,
A = rajesh;
                                                                 false.
                                 B = sunil;
A = sunil;
                                                                 11 ?- sibling(X, Y).
                                 A = gita,
A = prakash;
A = ramesh.
                                 B = sita ;
                                                                 X = sita
                                 A = gita,
                                                                 Y = sunil:
3 ?- female(A).
                                 B = sunil;
                                                                 X = sunil,
A = anita;
                                 A = sunil,
A = sita;
                                                                 Y = sita;
                                 B = prakash;
A = gita;
                                                                 X = sita,
A = laxmi.
                                 A = laxmi,
                                                                 Y = sunil;
                                 B = prakash.
7 ?- mother(gita, prakash).
                                                                 X = sunil,
                                10 ?- sibling(gita, sunil).
                                                                 Y = sita;
                                false.
8 ?- mother(gita, sunil).
                                                                 false.
```

Write a python program to implement AND, OR gate using perceptron algorithm.

```
import numpy as np
class Perceptron:
  def __init__(self, input_size, learning_rate=0.1, epochs=1000):
     self.learning_rate = learning_rate
     self.epochs = epochs
     self.weights = np.zeros(input size + 1) # Including bias term
  # Step 1: Define the activation function (Step function)
  def activation(self, x):
     return 1 if x \ge 0 else 0
  # Step 2: Train the Perceptron
  def train(self, X, y):
     for _ in range(self.epochs):
       for i in range(len(X)):
          # Input with bias term
          inputs = np.append(X[i], 1)
          prediction = self.activation(np.dot(inputs, self.weights))
          # Update weights using the perceptron learning rule
          self.weights += self.learning_rate * (y[i] - prediction) * inputs
  # Step 3: Make predictions
  def predict(self, X):
    predictions = []
     for i in range(len(X)):
       inputs = np.append(X[i], 1)
       prediction = self.activation(np.dot(inputs, self.weights))
       predictions.append(prediction)
    return predictions
# Step 4: Define input and output for AND and OR gates
# AND Gate
X_{and} = np.array([[0, 0], [0, 1], [1, 0], [1, 1]]) # Input
y_and = np.array([0, 0, 0, 1]) # Output
# OR Gate
X_{or} = np.array([[0, 0], [0, 1], [1, 0], [1, 1]]) # Input
y_or = np.array([0, 1, 1, 1]) # Output
# Step 5: Create perceptron models for AND and OR gates
perceptron and = Perceptron(input size=2)
perceptron_or = Perceptron(input_size=2)
# Train the models
perceptron_and.train(X_and, y_and)
perceptron_or.train(X_or, y_or)
```

```
# Step 6: Test the models
print("Testing AND Gate:")
predictions_and = perceptron_and.predict(X_and)
print(f"Predictions: {predictions_and}")
print(f"Expected: {y_and}\n")
print("Testing OR Gate:")
predictions\_or = perceptron\_or.predict(X\_or)
print(f"Predictions: {predictions_or}")
print(f"Expected: {y_or}")
Output:
   Testing AND Gate:
   Predictions: [0, 0, 0, 1]
   Expected: [0 0 0 1]
   Testing OR Gate:
   Predictions: [0, 1, 1, 1]
   Expected: [0 1 1 1]
   PS D:\College\7th Semester\Artificial Intelligence\Lab Codes>
```

Write a python program to illustrate working of backpropagation algorithm.

```
import numpy as np
# Sigmoid activation function
def sigmoid(x):
  return 1/(1 + np.exp(-x))
# Sigmoid derivative function
def sigmoid derivative(x):
  return x * (1 - x)
# Neural Network class
class NeuralNetwork:
  def __init__(self, input_size, hidden_size, output_size):
     # Initialize weights with random values
     self.input_size = input_size
     self.hidden_size = hidden_size
     self.output_size = output_size
     # Random weights and biases
     self.weights input hidden = np.random.rand(self.input size, self.hidden size)
     self.weights_hidden_output = np.random.rand(self.hidden_size, self.output_size)
     self.bias hidden = np.random.rand(self.hidden size)
     self.bias_output = np.random.rand(self.output_size)
  def forward(self, X):
     # Forward propagation through the network
     self.hidden_input = np.dot(X, self.weights_input_hidden) + self.bias_hidden
     self.hidden_output = sigmoid(self.hidden_input)
     self.final input = np.dot(self.hidden output, self.weights hidden output) + self.bias output
     self.final_output = sigmoid(self.final_input)
     return self.final output
  def backward(self, X, y, output):
     # Backpropagation
     output\_error = y - output
     output_delta = output_error * sigmoid_derivative(output)
    hidden_error = output_delta.dot(self.weights_hidden_output.T)
    hidden delta = hidden error * sigmoid derivative(self.hidden output)
     # Update weights and biases
     self.weights_hidden_output += self.hidden_output.T.dot(output_delta)
     self.weights_input_hidden += X.T.dot(hidden_delta)
     self.bias_output += np.sum(output_delta, axis=0)
     self.bias_hidden += np.sum(hidden_delta, axis=0)
```

```
def train(self, X, y, epochs):
    for epoch in range(epochs):
      # Perform forward pass
      output = self.forward(X)
      # Perform backward pass (backpropagation)
      self.backward(X, y, output)
      # Optionally, print error at intervals (for tracking learning)
      if epoch \% 1000 == 0:
       error = np.mean(np.square(y - output)) # Mean Squared Error
       print(f'Epoch {epoch}, Error: {error}')
  def predict(self, X):
    return self.forward(X)
# Define the XOR problem (training data)
X = \text{np.array}([[0, 0], [0, 1], [1, 0], [1, 1]]) # Input
y = np.array([[0], [1], [1], [0]]) # Output
# Create a neural network with 2 input neurons, 4 hidden neurons, and 1 output neuron
nn = NeuralNetwork(input size=2, hidden size=4, output size=1)
# Train the neural network
nn.train(X, y, epochs=10000)
# Test the neural network after training
print("\nTesting after training:")
print(nn.predict(X))
Output:
    Epoch 0, Error: 0.39541121161369497
    Epoch 1000, Error: 0.004090642495651932
    Epoch 2000, Error: 0.0009213306018762382
    Epoch 3000, Error: 0.0005042071744035696
    Epoch 4000, Error: 0.0003442509305999132
    Epoch 5000, Error: 0.0002603486273090009
    Epoch 6000, Error: 0.0002088730450990481
    Epoch 7000, Error: 0.00017414682216605005
    Epoch 8000, Error: 0.0001491747427014817
    Epoch 9000, Error: 0.00013037205228140204
   Testing after training:
    [[0.00626674]
     [0.98947872]
     [0.9887542 ]
     [0.01365362]]
   PS D:\College\7th Semester\Artificial Intelligence\Lab Codes>
```