```
3. Write a program to implement Water jug program using AI.
Source Code:
from collections import deque
def water jug problem(jug1 capacity, jug2 capacity, target):
    queue = deque([(0, 0, 0)]) # (jug1, jug2, steps)
    visited = set((0, 0))
    while queue:
        jug1, jug2, steps = queue.popleft()
        if jug1 == target or jug2 == target:
            return steps
        if (jug1 capacity, jug2) not in visited:
            queue.append((jug1 capacity, jug2, steps + 1))
            visited.add((jug1 capacity, jug2))
        if (jug1, jug2 capacity) not in visited:
            queue.append((jug1, jug2 capacity, steps + 1))
            visited.add((jug1, jug2 capacity))
        if (0, jug2) not in visited:
            queue.append((0, jug2, steps + 1))
            visited.add((0, jug2))
        if (jug1, 0) not in visited:
            queue.append((jug1, 0, steps + 1))
            visited.add((jug1, 0))
        pour amount = min(jug1, jug2 capacity - jug2)
        if (jug1 - pour amount, jug2 + pour amount) not in
visited:
steps + 1)) queue.append((jug1 - pour amount, jug2 + pour amount,
            visited.add((jug1 - pour amount, jug2 + pour amount))
        pour amount = min(jug2, jug1_capacity - jug1)
        if (\overline{j}ug1 + pour amount, jug2 - pour amount) not in
visited:
steps + 1)) queue.append((jug1 + pour amount, jug2 - pour amount,
            visited.add((jug1 + pour amount, jug2 - pour amount))
return -1
jug1 capacity = 3
jug2 capacity = 5
target = 4
steps = water jug problem(jug1 capacity, jug2 capacity, target)
if steps !=-\overline{1}:
    print("Solution found in", steps, "steps.")
    print("Jug 1:", jug1_capacity) print("Jug
    2:", jug2 capacity) print("Target:", target)
else:
    print("No solution found.")
OUTPUT Solution found in 6
steps. Jug 1: 3 Jug 2: 5
Target: 4
```

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7. Build an Artificial Neural Network by implementing the Back propagation Algorithm and test the same using appropriate data sets Source Code: import numpy as np X=np.array(([2, 9], [1, 5], [3, 6]), dtype=float) y =np.array(([92], [86], [89]), dtype=float) X=X/np.amax(X,axis=0) def sigmoid (x):
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```
return (1/(1 + np.exp(-x)))
def derivatives sigmoid(x):
    return x * (1-x)
epoch=7000
lr=0.1
inputlayer neurons = 2
hiddenlayer neurons = 3
output neurons = 1
wh=np.random.uniform(size=(inputlayer neurons, hiddenlayer neurons)
bh=np.random.uniform(size=(1, hiddenlayer neurons))
wout=np.random.uniform(size=(hiddenlayer neurons,output neurons))
bout=np.random.uniform(size=(1,output neurons))
for i in range (epoch):
hinp1=np.dot(X,wh)
hinp=hinp1 + bh
hlayer act =sigmoid(hinp)
 outinp1=np.dot(hlayer act, wout)
 outinp= outinp1+ bout
 output = sigmoid(outinp)
 EO=y-output
 outgrad = derivatives sigmoid(output)
 d output = E0* outgra\overline{d}
 EH=d output.dot(wout.T)
 hiddengrad = derivatives sigmoid(hlayer act)
 d hiddenlayer = EH * hid\overline{d}engrad
 wout += hlayer act.T.dot(d output) *lr
 bout += np.sum(d output, axis=0,keepdims=True) *lr
 wh +=X.T.dot(d hiddenlayer) *lr
                  \n"
print("Input:
                                str(X))
print("Actual Output: \n" + str(y))
print("Predicted Output: \n" ,output)
OUTPUT Input: [[0.66666667 1.
 [0.33333333 0.55555556]
 [1.
            0.66666667]]
Actual Output:
[[92.]
[86.]
[89.]]
Predicted Output:
 [[0.999999]]
 [0.99999749]
 [0.99999891]]
```

```
P absent and friday = 0.03
P friday = 0.20
P absent given friday = P absent and friday / P friday
print(f'Probability that a student is absent given that today is
Friday: {P absent given friday:.2f}')
print("\nDetailed Calculation:")
print(f"P(Absent and Friday) = {P absent and friday:.2f}")
print(f"P(Friday) = {P friday:.2f}")
print(f"P(Absent | Friday) = P(Absent and Friday) / P(Friday)")
print(f"P(Absent | Friday) = {P absent and friday:.2f} /
{P friday:.2f}")
print(f"P(Absent | Friday) = {P absent given friday:.2f}")
print("\nVerification:")
expected result = 0.03 / 0.20
print(f"Verification of result: {expected result:.2f}")
assert P_absent_given_friday == expected result, "The calculated
result does not match the expected result."
print("The calculation is correct and verified.")
OUTPUT Will Be
Detailed Calculation:
P(Absent and Friday) = 0.03
P(Friday) = 0.20
P(Absent | Friday) = P(Absent and Friday) / P(Friday)
P(Absent | Friday) = 0.03 / 0.20
P(Absent | Friday) = 0.15
Verification:
Verification of result: 0.15
The calculation is correct and verified.
```

```
a* algoritham
import heapq
def a star search(grid, start, goal):
    open list = []
    heapq.heappush(open list, (0, start))
    came from = {start: None}
    cost so far = {start: 0}
    while open_list:
         , current = heapq.heappop(open list)
        if current == goal:
            break
        for dx, dy in [(1, 0), (-1, 0), (0, 1), (0, -1)]:
        next = (current[0] + dx, current[1] + dy)
        if 0 \le next[0] \le len(grid) and 0 \le next[1] \le next[1]
len(grid[0]) and grid[next[0]][next[1]] != 1:
                new cost = cost so far[current] + 1
                if next not in cost so far or new cost <
cost so far[next]:
                    cost so far[next] = new cost
                    priority = new cost + abs(next[0] - goal[0]) +
abs(next[1] - goal[1])
                    heapq.heappush(open list, (priority, next))
                    came from[next] = current
    return came from, cost so far
grid = [
    [0, 0, 1, 0, 0],
    [0, 0, 1, 0, 0],
    [0, 0, 0, 0, 1],
    [0, 1, 1, 0, 0],
    [0, 0, 0, 0, 0]
]
start = (0, 0)
goal = (4, 4)
came_from, cost_so_far = a_star_search(grid, start, goal)
current = goal
path = []
while current:
    path.append(current)
    current = came_from[current]
path.reverse()
print(path)
OUTPUT
[(0, 0), (0, 1), (1, 1), (2, 1), (2, 2), (2, 3), (3, 3), (3, 4),
(4, 4)
```

```
import csv
# Function to read training data from a CSV file
def read training data(filename):
    data = []
    with open(filename, 'r') as file:
        reader = csv.reader(file)
        for row in reader:
            data.append(row)
    return data
# Function to implement the FIND-S algorithm
def find s(data):
    # Initialize the most specific hypothesis to all positive attributes
    hypothesis = ['yes'] * (len(data[0]) - 1) # Excluding the class attribute
    # Iterate over each training example
    for example in data[1:]:
        # If the example is positive
        if example[-1] == 'yes':
            # Update the hypothesis to be more general
            for i in range(len(hypothesis)):
                if example[i] != hypothesis[i]:
                    hypothesis[i] = '?'
        # If the example is negative
        else:
            # Update the hypothesis to be more specific
            for i in range(len(hypothesis)):
                if hypothesis[i] == '?' and example[i] != 'no':
                    hypothesis[i] = 'no'
    return hypothesis
# Read training data from a CSV file
data = read training data('training data.csv')
# Print the training data
print("Training Data:")
for row in data:
    print(row)
# Apply the FIND-S algorithm
hypothesis = find s(data)
# Print the most specific hypothesis
print("\nMost Specific Hypothesis:", hypothesis)
```