

University of Global Village (UGV), Barishal



Content of the Theory Course **University Student (UGV) Format**

Program: Bachelor of Science in Computer Science Engineering (CSE)

Course Code	CSE-406
Name of Course Title	Artificial Intelligence Lab
Course Type	Core Course
Level	7th Semester
Academic Session	Summer 2024
Name(s) of Academic Course teacher(s)	Md. Riadul Islam Assistant Professor Mobile: 01842611852 E-mail: riadnwu@gmail.com
Consultation Hours:	

Artificial Intelligence Lab Student (UGV) Format	
Course Code: CSE-406	Credits: 01
Exam Hours: 03	CIE Marks: 30
Course for 7th Semester, Bachelor of Science in Computer Science Engineering (CSE)	SEE Marks: 20

1. Course Objectives:

- **Master basic search algorithms:** Students will gain a comprehensive understanding of fundamental search algorithms such as depth-first search, breadth-first search, and A* search. They will learn to apply these algorithms to solve a variety of problems efficiently.
- **Implement effective knowledge representation and reasoning techniques:** This objective focuses on equipping students with techniques to represent knowledge effectively within computational systems. They will explore various formalisms such as logic-based representations, semantic networks, and probabilistic models. Additionally, students will learn how to reason with this knowledge to draw meaningful conclusions.
- **Develop and deploy supervised and unsupervised learning algorithms:** Students will delve into the theory and practical implementation of both supervised (e.g., linear regression, support vector machines) and unsupervised (e.g., clustering, dimensionality reduction) learning algorithms. They will understand the principles behind these methods and gain proficiency in applying them to analyze datasets and make predictions.
- **Design and train neural networks proficiently:** This objective involves mastering the principles of neural network architectures (e.g., feedforward, convolutional, recurrent networks) and learning algorithms (e.g., backpropagation). Students will gain hands-on experience in designing, training, and optimizing neural networks for various tasks such as image recognition, natural language processing, and time series prediction.
- **Apply advanced natural language processing techniques:** Students will explore advanced techniques in natural language processing (NLP) such as sentiment analysis, named entity recognition, and machine translation. They will learn how to preprocess text data effectively, apply state-of-the-art models like transformers, and evaluate NLP systems' performance.

- **Implement reinforcement learning algorithms:** This objective focuses on understanding the principles of reinforcement learning (RL) algorithms such as Q-learning, policy gradients, and deep RL approaches. Students will learn how to apply RL to build agents that can learn optimal behavior through interaction with an environment, applicable to tasks such as game playing and robotics.
- **Gain hands-on experience in AI and machine learning projects:** Throughout the course, students will work on practical projects that integrate concepts learned across different modules. These projects will provide them with real-world experience in applying AI and machine learning techniques to solve complex problems and address specific challenges.
- **Explore cutting-edge topics and real-world applications in AI:** The course will cover advanced topics such as generative adversarial networks (GANs), explainable AI, ethical considerations in AI, and emerging trends in the field. Students will gain insights into how AI is shaping industries and society, preparing them to tackle future challenges and opportunities in the AI domain.

4. Course Learning Outcome (CLO) at the end of the course, the students will be able to-

CLO1	Search Algorithms Description: Implement basic search algorithms like BFS, DFS, and A*. Specific Learning Objectives: • Understand the concepts of BFS and DFS. • Implement BFS, DFS, and A* search algorithms. • Apply these algorithms to solve search problems.
CLO2	Knowledge Representation and Reasoning Description: Develop logic-based knowledge representation and reasoning systems. Specific Learning Objectives: • Implement propositional and predicate logic. • Develop inference rules and apply them. • Use forward and backward chaining.
CLO3	Supervised Learning

	Description: Implement and evaluate supervised learning algorithms. Specific Learning Objectives: • Implement Linear and Logistic Regression. • Develop Decision Trees and SVM models. • Evaluate models using accuracy, precision, recall, and F1-score.
CLO4	Unsupervised Learning Description: Implement clustering and dimensionality reduction techniques. Specific Learning Objectives: • Implement K-means and Hierarchical clustering. • Perform PCA and t-SNE. • Visualize clustering and dimensionality reduction results.
CLO5	Neural Networks and Deep Learning

	<i>Description:</i> Develop and train neural networks. <i>Specific Learning Objectives:</i> • Build and train Feedforward, CNN, and RNN models. • Use TensorFlow or PyTorch for model development. • Evaluate and optimize neural networks.
CLO6	Natural Language Processing (NLP) <i>Description:</i> Apply NLP techniques to text data. <i>Specific Learning Objectives:</i> • Preprocess text data. • Implement Sentiment Analysis and Named Entity Recognition (NER). • Generate text using NLP models.
CLO7	Reinforcement Learning

	<i>Description:</i> Implement reinforcement learning algorithms. <i>Specific Learning Objectives:</i> • Develop reinforcement learning environments. • Implement Q-learning and DQN. • Apply Policy Gradient methods.
CLO8	AI and Machine Learning Project Development <i>Description:</i> Apply AI techniques to real-world projects. <i>Specific Learning Objectives:</i> • Define project scope and objectives. • Collect and preprocess data. • Implement and evaluate AI/ML algorithms.
CLO9	Advanced Topics and Applications

	<i>Description:</i> Explore and implement advanced AI topics. <i>Specific Learning Objectives:</i> • Implement GANs and Transfer Learning. • Explore AI in robotics and autonomous systems.
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5. Alignment to Program Learning Outcomes (PLOs)

PLOs/CLOs	CLO1: Basic Search Algorithms	CLO2: Knowledge Representation and Reasoning	CLO3: Supervised and Unsupervised Learning	CLO4: Neural Networks	CLO5: Natural Language Processing	CLO6: Reinforcement Learning
PLO1	3	3	3	3	3	3
PLO2	3	3	3	3	3	3
PLO3	3	3	3	3	3	3
PLO4	3	3	3	3	3	3
PLO5	3	3	3	3	3	3

6. Mapping of CLOs to Program Learning Outcomes (PLOs)

PLO1: Acquire a strong foundation in theoretical and practical aspects of computer science and engineering.

- CLO1: Basic Search Algorithms: Students will grasp foundational algorithms crucial for problem-solving in AI.
- CLO2: Knowledge Representation and Reasoning: Students will learn formalisms essential for representing and reasoning about knowledge in AI.
- CLO3: Supervised and Unsupervised Learning: Students will acquire practical knowledge in implementing machine learning algorithms, fundamental to AI applications.

- d. CLO4: Neural Networks: Students will develop skills in designing and training neural networks, a cornerstone of modern AI.
- e. CLO5: Natural Language Processing: Students will apply advanced techniques to process and understand human language, a critical aspect of AI.
- f. CLO6: Reinforcement Learning: Students will implement algorithms that enable learning through interaction, essential for AI systems.

PLO2: Apply knowledge of mathematics, science, and engineering principles to solve complex engineering problems.

- g. All CLOs contribute to applying theoretical knowledge to develop solutions in AI, from algorithms to practical applications like neural networks and natural language processing.

PLO3: Design and conduct experiments, as well as analyze and interpret data to provide valid conclusions.

- h. CLO3 (Supervised and Unsupervised Learning) involves designing experiments to analyze data and draw conclusions, essential for evaluating AI models.
- i. CLO6 (Reinforcement Learning) involves experimenting with AI agents in simulated environments to evaluate performance.

PLO4: Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings.

- j. Group projects and assignments throughout the lab manual promote teamwork and collaborative problem-solving, preparing students for multidisciplinary AI applications.

PLO5: Recognize the need for, and engage in, lifelong learning to continually update knowledge and skills in computer science and engineering.

- k. Exploration of advanced AI topics (CLO5 and CLO6) encourages students to stay updated with current AI trends and technologies, fostering a commitment to lifelong learning.

7. Topics to be covered/Content of the course

Week	Topics	Specific Outcomes	Teaching Learning Strategy(s)	Assessment Strategy(s)	Alignment to CLO
1.	Introduction to AI Tools and Environment (Part 1)	Set up and navigate Python and relevant libraries.	Hands-on practice, instructor-led setup	Feedback, observation	CLO1
2.	Acting Humanly: The Turing Test	Acting Humanly: The Turing Test	Practical exercises, group work	Q&A, practical assignments	CLO1
3.	Simple Rational Agent Chatbot in Python	Implement logic-based representation and inference rules.	Code implementation, peer review	Lab exercises, quizzes	CLO2
4.	BFS With MAZE	Implement Linear Regression, Logistic	Practical coding sessions, group work	Feedback, practical assignments	CLO3

		Regression, Decision Trees, SVM.			
5.	DFS With MAZE	Implement K-means, Hierarchical clustering, PCA, t-SNE.	Hands-on exercises, instructor guidance	Observation, quizzes	CLO4
6.	N Queen Using DFS	Build and train neural networks using TensorFlow/PyTorch.	Practical coding sessions, instructor guidance	Practical assignments, observation	CLO5
7.	8 Puzzle Using BFS	Text preprocessing, Sentiment Analysis, NER, text generation.	Lecture, multimedia, interactive sessions	Feedback, Q&A, quizzes	CLO6

8.	A* Search	Develop and train reinforcement learning models.	Hands-on project work, peer collaboration	Observation, feedback	CLO7
Mid Term Examination					
9.	Conditional probability		Define project scope, collect and preprocess data, and implement algorithms.	Group discussions, instructor feedback	Observation, project planning CL O8
10.	Genetic Algorithms To Solve The Traveling Salesman Problem (TSP)		Define project scope, collect and preprocess data, and implement algorithms.	Group discussions, instructor feedback	Observation, project planning CL O8

11.	ACO : Ant Colony Optimization for TSP	Implement GANs, Transfer Learning, explore AI in robotics.	Hands-on project work, peer collaboration	Observation, feedback	CL O9
12.	ANN Forward pass and Back propagation	Implement advanced search algorithms and optimization techniques.	Practical exercises, group work	Q&A, practical assignments	CL O1
13.	Fuzzy logic control in automated car system	Implement Ensemble Methods (Random Forest, Gradient Boosting).	Practical coding sessions, group work	Feedback, practical assignments	CL O3
14.	Neural Networks and Deep Learning (Extended)	Implement advanced neural network architectures (ResNet, LSTM).	Practical coding sessions, instructor guidance	Practical assignments, observation	CL O5
15.	AI and Machine Learning Project Development (Extended)	Implement and refine project scope, collect and preprocess data, advanced algorithms.	Group discussions, instructor feedback	Observation, project planning	CL O8

16.	Final Project Implementation	Implement and refine final AI/ML projects.	Hands-on project work, peer collaboration	Observation, feedback	All CL Os
Final Term Examination					

8. Teaching-Learning Strategy:

- **Lectures:** Brief interactive sessions to explain AI concepts.
- **Practical Sessions:** Extensive hands-on practice to apply AI principles.
- **Group Discussions:** Collaborative activities to enhance understanding and problem-solving skills.
- **Assignments:** Regular tasks to reinforce learning and assess understanding.
- **Quizzes and Tests:** Periodic assessments to evaluate progress and mastery of concepts.

9. Assessment Strategy:

- **Continuous In-course Evaluation (CIE):** 30 marks
 - Lab Participation: 10 marks

- Assignments: 10 marks
- Quizzes: 10 marks
- **Final Project Evaluation:** 20 marks
 - Project Implementation: 10 marks
 - Project Presentation: 5 marks
 - Project Report: 5 marks
- **Total Marks:** 50

10. **Instructional Materials and References:**

- **Textbooks:**
 - "Artificial Intelligence: A Modern Approach" by Stuart Russell and Peter Norvig.
 - "Deep Learning" by Ian Goodfellow, Yoshua Bengio, and Aaron Courville.
- **Reference Books:**
 - "Machine Learning: A Probabilistic Perspective" by Kevin P. Murphy.
- **Online Resources:**
 - AI Reference (<http://www.aireference.com/>)
 - Stack Overflow (<https://stackoverflow.com/>)
 - GeeksforGeeks (<https://www.geeksforgeeks.org/>)

WEEK-1

Set up and navigate Python and relevant libraries

1. Objective

- Learn how to install and configure Python.
 - Understand and navigate development environments (Anaconda, Jupyter Notebook, VS Code).
 - Explore Python syntax and environment interaction.
 - Install and use key Python libraries such as NumPy, Pandas, and Matplotlib.
-

2. Tools and Software

- **Python 3.10+**
 - **Anaconda / Miniconda** (recommended)
 - **Jupyter Notebook or VS Code**
 - Internet access for installing additional libraries
-

3. Lab Tasks

1. Python Installation and Setup

- Download and install Python or Anaconda.
- Verify installation using the command:
 - `python --version`
- Launch the environment (Jupyter Notebook or VS Code).

2. Understanding Python Environment

- Create and run a simple Python program (`print("Hello, Python!")`).
- Explore Python data types: integers, floats, strings, lists, tuples, and dictionaries.
- Practice basic operations and variable declarations.

3. Installing and Importing Libraries

- Use **pip** to install common libraries:
- `pip install numpy pandas matplotlib`
- Import and test each library in Python:
- `import numpy as np`
- `import pandas as pd`
- `import matplotlib.pyplot as plt`

4. Hands-on with Core Libraries

- Use **NumPy** to create and manipulate arrays.
- Use **Pandas** to create DataFrames and handle simple datasets.
- Use **Matplotlib** to plot data visually.

5. Practical Example

```

6. import numpy as np
7. import pandas as pd
8. import matplotlib.pyplot as plt
9.
10. # Create data
11. data = np.array([5, 10, 15, 20, 25])
12. df = pd.DataFrame(data, columns=['Values'])
13.
14. # Display and plot
15. print(df)
16. plt.plot(df['Values'], marker='o')
17. plt.title('Sample Data Visualization')
18. plt.xlabel('Index')
19. plt.ylabel('Values')
20. plt.show()

```

4. Evaluation Criteria

Criteria	Marks	Description
Setup Verification	20	Successful Python and library installation
Program Execution	30	Ability to run and interpret basic scripts
Library Usage	30	Correct use of NumPy, Pandas, and Matplotlib
Report Submission	20	Summary with code and screenshots

5. Expected Learning Outcomes

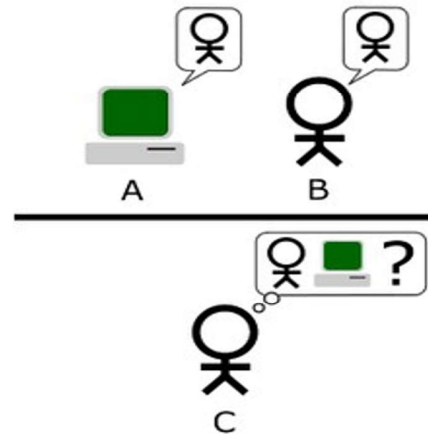
- Successfully install and configure Python.
- Operate in Jupyter Notebook or VS Code.
- Understand Python's basic syntax and environment.
- Use standard libraries for data handling and visualization.

WEEK-2

Acting Humanly: The Turing Test

Acting Humanly: The Turing Test

http://en.wikipedia.org/wiki/Turing_test



Alan Turing

1912-1954

- To be intelligent, a program should simply act like a human

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CSC 361 Artificial
Intelligence

```
# eliza_bot.py
import re
import random

reflections = {
    "i am": "you are",
    "i'm": "you are",
    "i": "you",
    "me": "you",
    "my": "your",
    "you are": "I am",
    "you": "I",
}

pairs = [
    (r'.*\bhello\b.*', ["Hello. How can I help you today?", "Hi – what's on your mind?"]),
    (r'.*\b(sad|unhappy|depressed|upset)\b.*', ["I'm sorry you feel {0}. Do you want to talk about"]),
    (r'.*\b(happy|great|good|fine)\b.*', ["That's good to hear! What made you {0}?"]),
    (r'.*\b(my|i am|i\'m) (.*)', ["Why do you say {1}?", "How long have you been {1}?"]),
    (r'.*\b(yes|no)\b.*', ["Why do you say {0}?", "Can you tell me more?"]),
    (r'.*\b(bye|goodbye|see you)\b.*', ["Goodbye. Take care!", "See you later!"]),
    (r'(.*)', ["Tell me more about {0}.", "Why do you think {0}?", "How does {0} make you feel?"])
]
```

```
def reflect(fragment):
    words = fragment.lower().split()
    for i,w in enumerate(words):
        if w in reflections:
            words[i] = reflections[w]
    return " ".join(words)

def respond(sentence):
    sentence = sentence.strip()
    for pattern, responses in pairs:
        m = re.match(pattern, sentence, re.IGNORECASE)
        if m:
            template = random.choice(responses)
            # insert captured groups if needed
            groups = [g for g in m.groups() if g is not None]
            if groups:
                # reflect first captured group
                return template.format(reflect(groups[-1]))
            else:
                return template.format(sentence)
    return "I see."
```

 Copy code

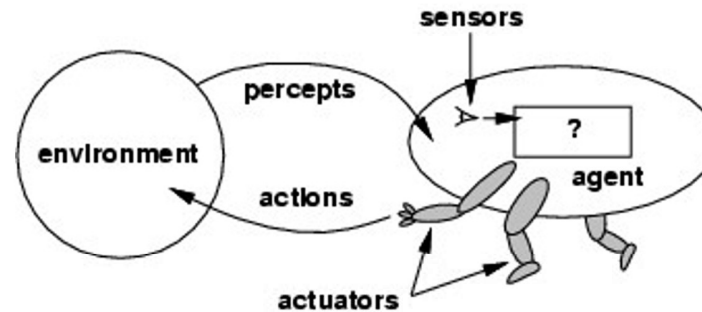


```
def chat():
    print("ELIZA-like bot. Type 'quit' to exit.")
    while True:
        user = input("You: ")
        if user.lower() in ['quit', 'exit', 'bye', 'goodbye']:
            print("Bot: Goodbye.")
            break
        print("Bot:", respond(user))

if __name__ == "__main__":
    chat()
```

WEEK-3

Simple Rational Agent Chatbot in Python



- The **agent function** maps from **percept** histories to actions:
- $[f: \mathcal{P}^* \rightarrow \mathcal{A}]$
- Percept history is also known as **percept sequence**.

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How it works:

1. **Perceive:** Takes user input.
2. **Think:** Checks its knowledge base for keywords.
3. **Act:** Responds with the best matching answer.
4. **Goal:** Give meaningful, rational replies to user questions.

```

class RationalAgent:
    def __init__(self):
        # Agent knowledge base
        self.knowledge = {
            "hello": "Hello! How can I assist you today?",
            "your name": "I am a Rational Agent designed to help you.",
            "what is ai": "AI stands for Artificial Intelligence – systems that can think and learn",
            "what is rational agent": "A rational agent acts to achieve the best outcome based on the information it has",
            "bye": "Goodbye! Have a great day."
        }

    def perceive(self, environment_input):
        """Perceive the environment (user input)."""
        return environment_input.lower()

    def think(self, perception):
        """Decide the best action based on knowledge."""
        for key in self.knowledge:
            if key in perception:
                return self.knowledge[key]
        return "I'm not sure about that. Could you please explain more?"

    def act(self, action):
        """Perform the chosen action (output message)."""
        print("Agent:", action)

```

```

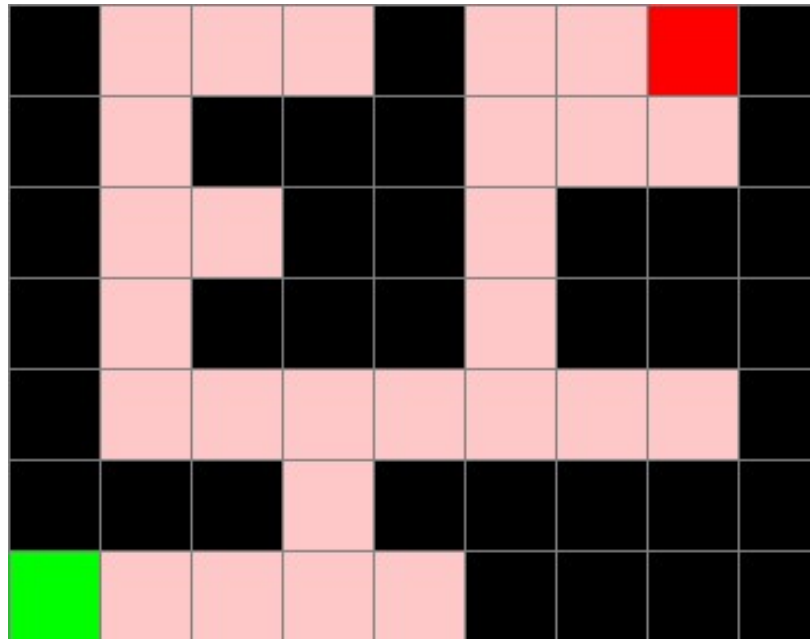
def run(self):
    """Main loop – perceive, think, act."""
    print("Rational Agent: Hello! Ask me something. (type 'bye' to exit)\n")
    while True:
        user_input = input("You: ")
        perception = self.perceive(user_input)
        response = self.think(perception)
        self.act(response)
        if "bye" in perception:
            break

# Run the agent
if __name__ == "__main__":
    agent = RationalAgent()
    agent.run()

```

WEEK-4

BFS With MAZE



```

maze=open('maze.txt')
maze= [list(i) for i in maze.readlines()]
n=len(maze)
m=len(maze[0])
G=[0,8]
S=[6,0]
queue=[]
index=S
queue.append(index)

```

```

while index !=G:
    #print(queue)
    queue.pop()
    i=index[0]
    j=index[1]
    if i-1>-1:
        if maze[i-1][j] == ' ':
            queue.append([i-1,j])
            maze[i - 1][j] = '.'
        elif maze[i-1][j] == 'G':
            break
    if i + 1 < n - 1:
        if maze[i+1][j] == ' ':
            queue.append([i+1,j])
            maze[i + 1][j] = '.'

```

```

        elif maze[i+1][j] == 'G':
            break

    if j + 1 < m-1:
        if maze[i][j+1] == ' ':
            queue.append([i,j+1])
            maze[i][j+1] = '.'
        elif maze[i][j+1] == 'G':
            break
    if j-1>-1:
        if maze[i][j-1] == ' ':
            queue.append([i,j-1])
            maze[i][j-1] = '.'
        elif maze[i][j-1] == 'G':
            break
    print(queue[0])

    index=queue[-1]
    for k in maze:
        for l in k:
            print(l, end='')
        print()

    from PIL import Image, ImageDraw

```

```

# Size of each cell
cell_size = 40
rows = len(maze)
cols = max(len(row) for row in maze)

# Create a new image
img = Image.new("RGB", (cols * cell_size, rows * cell_size),
color="white")
draw = ImageDraw.Draw(img)

# Define colors for each character
colors = {
    '#': (0, 0, 0),          # Wall - Black
    '.': (255, 200, 200),    # Path - Light Gray
    'S': (0, 255, 0),        # Start - Green
    'G': (255, 0, 0),        # Goal - Red
    ' ': (255, 255, 255)    # Empty - White
}

# Draw each cell
for i, row in enumerate(maze):
    for j, ch in enumerate(row):
        color = colors.get(ch, (255, 255, 255)) # Default to white
        x0 = j * cell_size

```

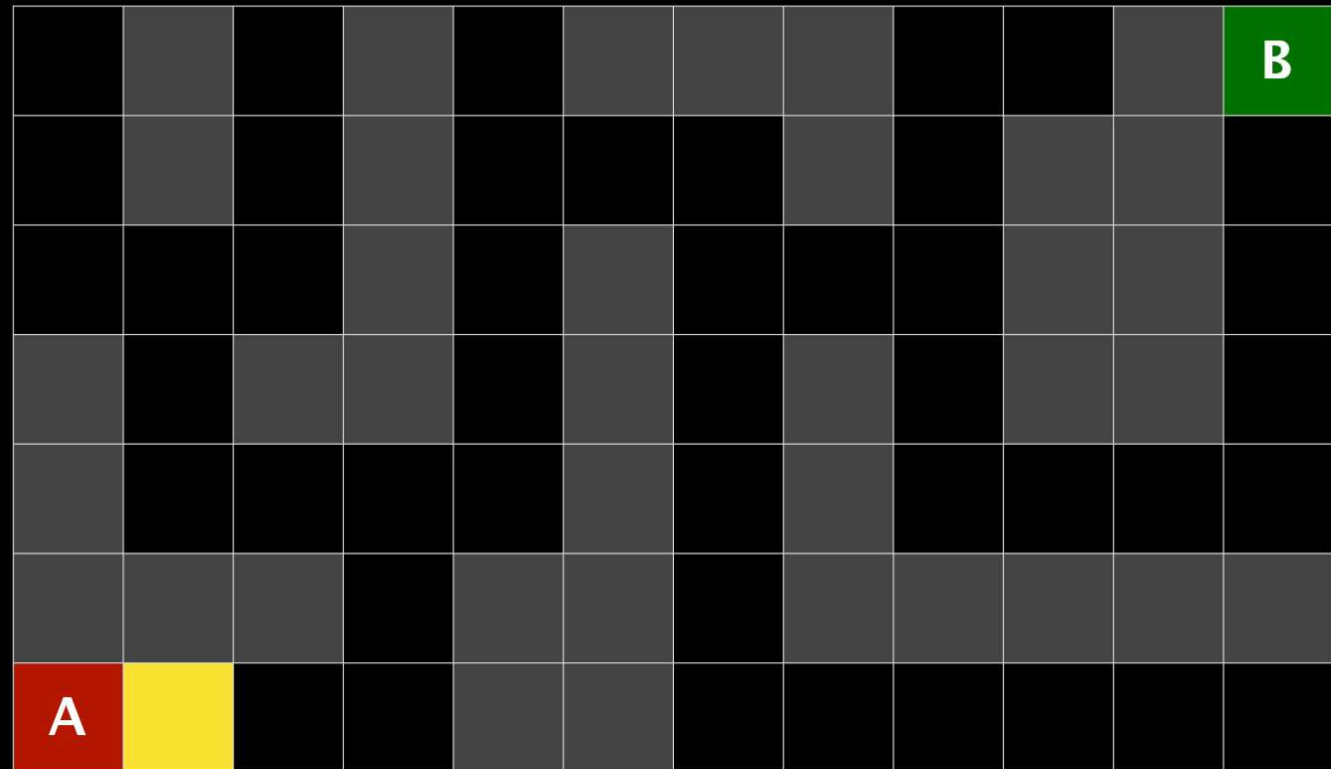
```
y0 = i * cell_size
x1 = x0 + cell_size
y1 = y0 + cell_size
draw.rectangle([x0, y0, x1, y1], fill=color, outline="gray")

# Save or show the image
img.show()          # Show in default viewer
# img.save("maze.png") # Or save to a file
```

WEEK-5

DFS With MAZE

Depth-First Search



```

maze=open('maze.txt')
maze= [list(i) for i in maze.readlines()]
n=len(maze)
m=len(maze[0])
G=[0,4]
S=[6,0]
stack=[]
index=S
stack.append(index)

while index !=G:
    #print(stack)
    stack.pop(-1)
    i=index[0]
    j=index[1]
    if i-1>-1:
        if maze[i-1][j] == ' ':
            stack.append([i-1,j])
            maze[i - 1][j] = '.'
        elif maze[i-1][j] == 'G':
            break
    if i + 1 < n - 1:
        if maze[i+1][j] == ' ':
            stack.append([i+1,j])
            maze[i + 1][j] = '.'

```

```

        elif maze[i+1][j] == 'G':
            break
    if j + 1 < m-1:
        if maze[i][j+1] == ' ':
            stack.append([i,j+1])
            maze[i][j+1] = '.'
        elif maze[i][j+1] == 'G':
            break
    if j-1>-1:
        if maze[i][j-1] == ' ':
            stack.append([i,j-1])
            maze[i][j-1] = '.'
        elif maze[i][j-1] == 'G':
            break

```

```

    index=stack[len(stack)-1]
    for k in maze:
        for l in k:
            print(l, end='')
        print()

```

```

from PIL import Image, ImageDraw

```

```

# Size of each cell
cell_size = 40
rows = len(maze)
cols = max(len(row) for row in maze)

```

```

# Create a new image
img = Image.new("RGB", (cols * cell_size, rows * cell_size), color="white")
draw = ImageDraw.Draw(img)

# Define colors for each character
colors = {
    '#': (0, 0, 0),          # Wall - Black
    '.': (255, 200, 200),    # Path - Light Gray
    'S': (0, 255, 0),        # Start - Green
    'G': (255, 0, 0),        # Goal - Red
    ' ': (255, 255, 255)    # Empty - White
}

# Draw each cell
for i, row in enumerate(maze):
    for j, ch in enumerate(row):
        color = colors.get(ch, (255, 255, 255)) # Default to white
        x0 = j * cell_size
        y0 = i * cell_size
        x1 = x0 + cell_size
        y1 = y0 + cell_size
        draw.rectangle([x0, y0, x1, y1], fill=color, outline="gray")

# Save or show the image
img.show() # Show in default viewer
# img.save("maze.png") # Or save to a file

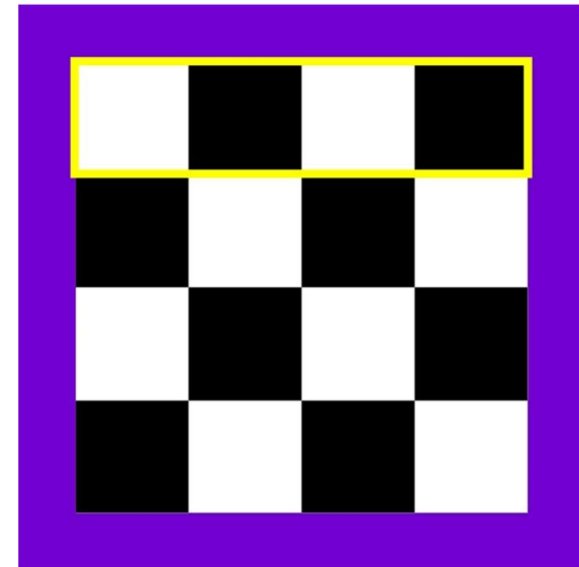
```

WEEK-6

N Queen Using DFS

N-Queens: Example

- Let's illustrate the backtracking algorithm with a mathematical example for $N = 4$. The chessboard as an $N \times N$ matrix.
- Now, we start placing queens using the backtracking algorithm:
 1. Set Q1 to 1st row of 1st column.
 2. Set Q2 to 2nd row of 3rd column.
 3. Q3 has no place to 3rd row.
 4. Backtrack to Q2 and set 2nd row of 4th column.
 5. Set Q3 to 3rd row of 2nd column.



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127

```

import numpy as np
from numpy import bool

def q_pos(index,type):
    x=index[0]
    y=index[1]
    if type==1:
        bord[x][y]=-1
    else:
        bord[x][y] = 0
    q_over=((1,0),(0,1),(-1,0),(0,-1),(-1,-1),(-1,1),(1,1),(1,-1))
    for a in q_over:
        i=a[0]+x
        j=a[1]+y
        while(i>=0 and j>=0 and i<n and j<n and bord[i][j]!=-1):
            if type==1:
                bord[i][j]= bord[i][j]+1
            else:
                bord[i][j] = bord[i][j] - 1
            i =i+a[0]
            j =j+a[1]

def n_quee(num_queen):
    if num_queen==n:
        return 1
    free_index=np.where(bord[num_queen]==0)
    free_index=free_index[0].tolist()
    #print(free_index)

```

```

while len(free_index) != 0:
    i = free_index[0]
    #print("Q", num_queen)
    free_index = np.delete(free_index, 0)
    q_pos((num_queen, i), 1)
    #print(board)
    flag = n_queen(num_queen + 1)
    if flag == 1:
        return 1
    q_pos((num_queen, i), 0)

import numpy as np
from PIL import Image, ImageDraw, ImageFont

def display(board):
    n = board.shape[0]
    cell_size = 80
    img_size = n * cell_size

    # Create blank image
    img = Image.new("RGB", (img_size, img_size), "white")
    draw = ImageDraw.Draw(img)

    # Load font
    try:
        font = ImageFont.truetype("arial.ttf", 30)
    except:
        font = ImageFont.load_default()

```

```

for i in range(n):
    for j in range(n):
        x0, y0 = j * cell_size, i * cell_size
        x1, y1 = x0 + cell_size, y0 + cell_size

        # Chessboard pattern
        fill = (240, 240, 240) if (i + j) % 2 == 0 else (180, 180, 180)
        draw.rectangle([x0, y0, x1, y1], fill=fill, outline="black")

        value = board[i, j]
        if value == -1: # queen
            draw.text((x0 + cell_size//3, y0 + cell_size//4), "Q", fill="red",
font=font)
            elif value > 0: # overlap number
                # Use textbbox for proper centering
                bbox = draw.textbbox((0, 0), str(value), font=font)
                w, h = bbox[2] - bbox[0], bbox[3] - bbox[1]
                draw.text((x0 + (cell_size - w)//2, y0 + (cell_size - h)//2),
                    str(value), fill="black", font=font)

img.show()

n=8
bord=np.zeros((n,n),dtype=int)
num_queen=0
n_quee(num_queen)
display(bord)

```

Q	3	3	3	2	4	3	3
2	3	3	3	Q	3	4	3
3	2	3	4	4	4	3	Q
2	3	3	4	4	Q	4	3
3	3	Q	3	4	4	4	4
4	4	3	4	3	4	Q	3
3	Q	4	3	4	3	3	3
3	3	4	Q	3	3	2	3

WEEK-7

8 Puzzle Using BFS

Solve this Eight Puzzle using BFS

2	5	1
8		3
4	7	6

Initial board

1	2	3
4	5	6
7	8	

Goal board

```

import copy
def find_zero(p_state):
    x=0
    while True:
        if 0 in p_state[x]:
            y=p_state[x].index(0)
            return x,y
        else:
            x=x+1

def position_check():
    while queue[0] != t_state:
        print(queue[0])
        print('@@@@@@@@@@@@@@@@')
        index=[[1,0],[-1,0],[0,1],[0,-1]]

        x, y =find_zero(queue[0])
        for a in index:
            i=a[0]+x
            j=a[1]+y
            if i>=0 and j>=0 and i<n and j<n:
                temp_state =copy.deepcopy(queue[0])
                temp_state[x][y] =temp_state[i][j]
                temp_state[i][j] =0
                print(temp_state)

            if temp_state in pop_list:
                a=1
            elif temp_state == t_state:

```

```

        print(temp_state)
        print("result Found")
        return 0
    else:
        queue.append(temp_state)
        pop_list.append(temp_state)
        queue.pop(0)
        print('-----')
print(queue[0])

```

```

p_state=[[1,2,3],[4,5,6],[7,8,0]]
t_state=[[1,3,0],[6,2,4],[7,5,8]]
print(t_state)
print('+++++')
n=len(p_state)
queue=[]
pop_list=[]
queue.append(p_state)
pop_list.append(p_state)
position_check()

```

WEEK-8

8 Puzzle Using BFS

A* search

search algorithm that expands node with lowest value of $g(n) + h(n)$

$g(n)$ = cost to reach node

$h(n)$ = estimated cost to goal

A* Search

	10	9	8	7	6	5	4	3	2	1	B
	11										1
	12		10	9	8	7	6	5	4		2
	13		11						5		3
	14	13	12		10	9	8	7	6		4
			13		11						5
A	16	15	14		12	11	10	9	8	7	6

```

import heapq

# Define the maze: 0 = free space, 1 = obstacle
maze = [
    [0, 1, 0, 0, 0, 0],
    [0, 1, 0, 1, 1, 0],
    [0, 0, 0, 1, 0, 0],
    [0, 1, 1, 1, 0, 1],
    [0, 0, 0, 0, 0, 0]
]

start = (0, 0) # Starting position
end = (4, 5)   # Goal position

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

# A* algorithm
def astar(maze, start, end):
    rows, cols = len(maze), len(maze[0])
    open_set = []
    heapq.heappush(open_set, (0 + heuristic(start, end), 0, start, [start]))
    visited = set()

```



```

while open_set:
    _, cost, current, path = heapq.heappop(open_set)

    if current in visited:
        continue
    visited.add(current)

    if current == end:
        return path

    # Explore neighbors (up, down, left, right)
    neighbors = [(0,1),(1,0),(0,-1),(-1,0)]
    for dx, dy in neighbors:
        nx, ny = current[0] + dx, current[1] + dy
        if 0 <= nx < rows and 0 <= ny < cols and maze[nx][ny] == 0:
            if (nx, ny) not in visited:
                new_cost = cost + 1
                heapq.heappush(open_set, (new_cost + heuristic((nx, ny), end), new_cost, (nx,

return None # No path found

```

```
# Run A*
path = astar(maze, start, end)

if path:
    print("Path found:")
    for step in path:
        print(step)
else:
    print("No path found.")
```

WEEK-9

Conditional Probability



Given clouds in the morning,
what's the probability of rain in the afternoon?

80% of rainy afternoons start with cloudy
mornings.

40% of days have cloudy mornings.

10% of days have rainy afternoons.



$C = \text{cloud}$	$C = \neg \text{cloud}$
0.4	0.6



$R = \text{rain}$	$R = \neg \text{rain}$
0.1	0.9



	$R = \text{rain}$	$R = \neg \text{rain}$
$C = \text{cloud}$	0.08	0.32
$C = \neg \text{cloud}$	0.02	0.58

$$P(C \mid \text{rain})$$

$$P(C \mid \text{rain}) = \frac{P(C, \text{rain})}{P(\text{rain})} = \alpha P(C, \text{rain})$$

$$= \alpha \langle 0.08, 0.02 \rangle = \langle 0.8, 0.2 \rangle$$

	$R = \text{rain}$	$R = \neg \text{rain}$
$C = \text{cloud}$	0.08	0.32
$C = \neg \text{cloud}$	0.02	0.58

Given data:

- $P(\text{Cloudy Morning} \mid \text{Rainy Afternoon}) = 0.8$

- $P(\text{Cloudy Morning}) = 0.4$
 - $P(\text{Rainy Afternoon}) = 0.1$
-

We need to find:

Usually, such problems ask for $P(\text{Rainy Afternoon} \mid \text{Cloudy Morning})$ —the probability that the afternoon is rainy given the morning is cloudy.

We can use Bayes' theorem:

$$P(R \mid C) = \frac{P(C \mid R) \times P(R)}{P(C)}$$

Substitute the values:

$$P(R \mid C) = \frac{0.8 \times 0.1}{0.4} = 0.2$$

So,

The probability of a rainy afternoon given a cloudy morning is 0.2 (or 20%).

python



```
# Given probabilities
P_cloudy_given_rainy = 0.8
P_cloudy = 0.4
P_rainy = 0.1

# Using Bayes' theorem
P_rainy_given_cloudy = (P_cloudy_given_rainy * P_rainy) / P_cloudy

print(f"Probability of rainy afternoon given cloudy morning: {P_rainy_given_cloudy:.2f}")
```

✓ **Output:**

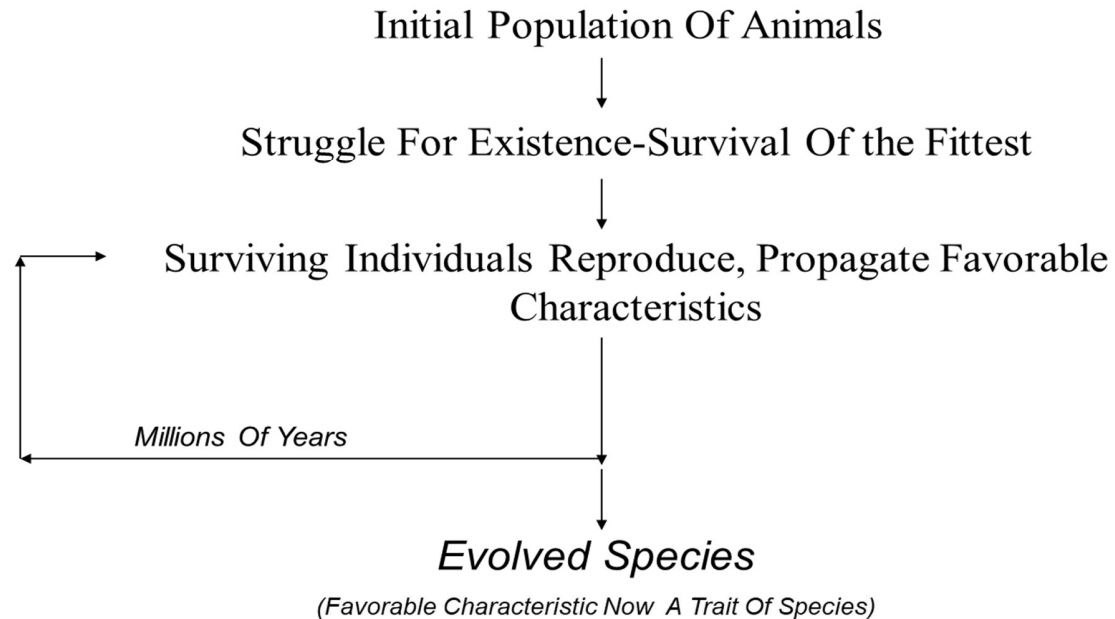
Probability of rainy afternoon given cloudy morning: 0.20

WEEK-10

Genetic Algorithms To Solve The Traveling Salesman Problem (TSP)



Evolution Through Natural Selection



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75



Simple Genetic Algorithm

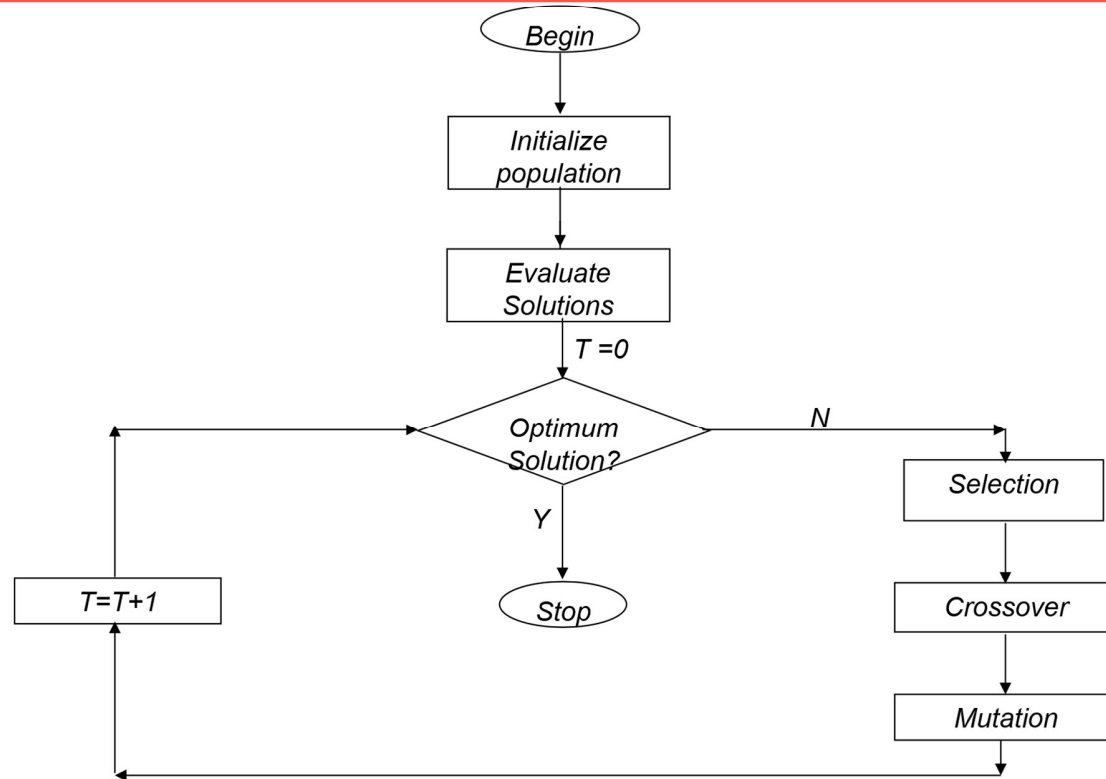
```
Simple_Genetic_Algorithm()  
{  
    Initialize the Population;  
    Calculate Fitness Function;  
  
    While(Fitness Value != Optimal Value)  
    {  
        Selection;//Natural Selection, Survival  
Of Fittest  
        Crossover;//Reproduction, Propagate  
favorable characteristics  
  
        Mutation;//Mutation  
        Calculate Fitness Function;  
    }  
}
```

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78



Working Mechanism Of GAs



77

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Problem Setup: 5 Cities

- Cities: A, B, C, D, E
- Distance Matrix:

–		A		B		C		D		E	
–		A		0		2		9		10	
–		B		2		0		6		4	
–		C		9		6		0		8	
–		D		10		4		8		0	
–		E		7		3		5		6	

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Step 1: Initial Population

- Population size = 4 (example):
 - Individual 1: $A \rightarrow B \rightarrow C \rightarrow D \rightarrow E \rightarrow A$
 - Individual 2: $A \rightarrow C \rightarrow E \rightarrow B \rightarrow D \rightarrow A$
 - Individual 3: $A \rightarrow D \rightarrow E \rightarrow C \rightarrow B \rightarrow A$
 - Individual 4: $A \rightarrow E \rightarrow B \rightarrow D \rightarrow C \rightarrow A$

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Step 2: Fitness Calculation

- Fitness = $1 / \text{total distance}$
- Example:
 - Individual 1: Total distance = 29, Fitness = $1 / 29 \approx 0.0345$
 - Individual 2: Total distance = 31, Fitness = $1 / 31 \approx 0.0322$
 - Individual 3: Total distance = 29, Fitness = $1 / 29 \approx 0.0345$
 - Individual 4: Total distance = 31, Fitness = $1 / 31 \approx 0.0322$
- Total fitness
 - = $(0.0345 + 0.0322 + 0.0345 + 0.0322)$
 - = 0.1334

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Step 3: Selection (Roulette Wheel)

- Total fitness = 0.1334
- Selection probabilities:
 - Individual 1: $(0.0345 \div 0.1334) = 0.2586$
 - Individual 2: $(0.0322 \div 0.1334) = 0.2412$
 - Individual 3: $(0.0345 \div 0.1334) = 0.2586$
 - Individual 4: $(0.0322 \div 0.1334) = 0.2412$
- Selection: Randomly pick parents based on fitness probabilities.

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```
# A B C D E  
# 0 1 2 3 4
```

```
import random
```

```

def population_gen(n):
    population=[]
    for i in range(n):
        l = [0, 1, 2, 3, 4]
        random.shuffle(l)
        population.append(l)
    return population
def fitness_cal():
    fitness=[0]*n
    for i in range(n):
        for j in range(len(l)-1):
            fitness[i]=fitness[i]+
g[population[i][j]][population[i][j+1]]
            fitness[i]=1/fitness[i]
    print(fitness)
    total_fitness = sum(fitness)
    print(total_fitness)
    for i in range(n):
        fitness[i]=fitness[i]/total_fitness
    return fitness

def selection(population,fitness):
    parent=random.choices(population,weights=fitness,k=2)
    print(parent)
    return parent

def corssover(point):
    child=parent[0][:point]+parent[1][point:]
    child=list(child)
    print(child)
    miss_val=[]
    rep_index=[]
    for i in range(len(l)):
        if child.count(i)==0:
            miss_val.append(i)
        elif child.count(i)>1:
            rep_index.append(child.index(i))
    print(miss_val)

```

```

    print(rep_index)
    for i in range(len(rep_index)):
        child[rep_index[i]]=miss_val[i]
    print(child)
    return child
def mutation(child):
    random_index= random.sample(1, 2)
    print(random_index[0])
    print(random_index[1])
    temp1=child[random_index[0]]
    temp2=child[random_index[1]]
    child[random_index[0]]=temp2
    child[random_index[1]]=temp1
    print(child)
    return child
g=[[0,3,4,5,6],
   [2,0,5,8,9],
   [4,5,0,3,1],
   [2,3,5,0,4],
   [7,2,3,4,0],
  ]
print(g)

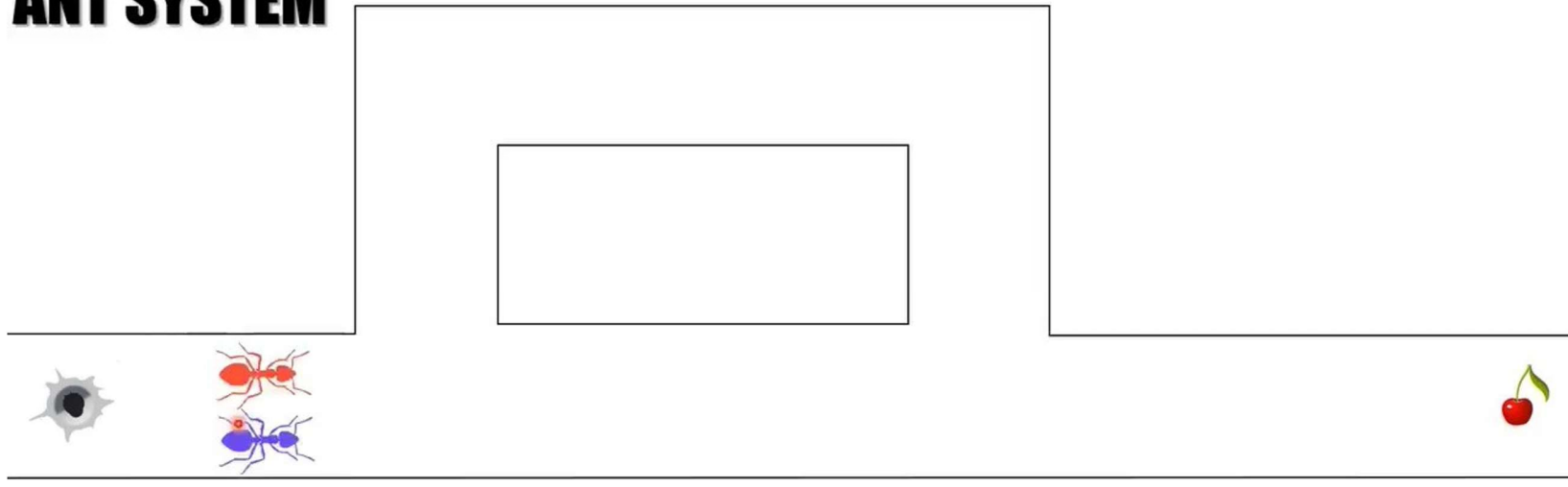
l=[0,1,2,3,4]
n=5
print(l)
population=population_gen(n)
print(population)
m=10
for i in range(m):
    fitness=fitness_cal()
    print(fitness)
    parent=selection(population,fitness)
    point=3
    child=corssover(point)
    child=mutation(child)
    population[0]=child

```

WEEK-10

ACO Algorithm for TSP

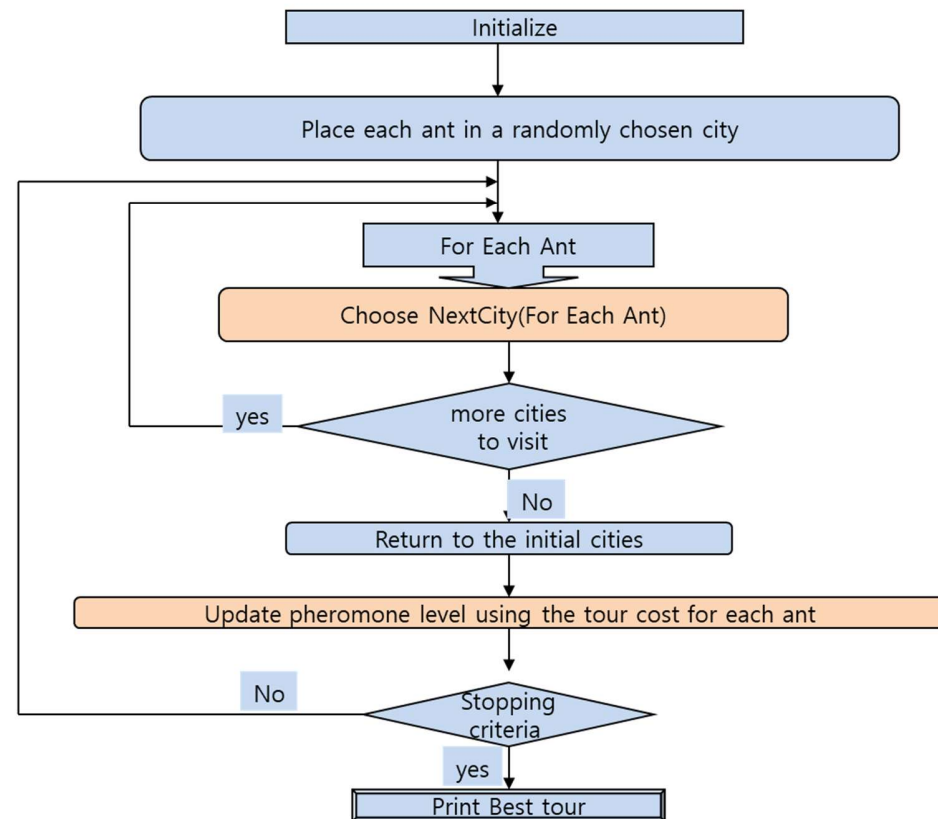
ANT SYSTEM



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145

ACO Algorithm for TSP



Loop

Randomly position m artificial ants on n cities

For city=1 to n

For ant=1 to m

{Each ant builds a solution by adding one city after the other}

Select probabilistically the next city according to
exploration and exploitation mechanism

Apply the **local trail updating** rule

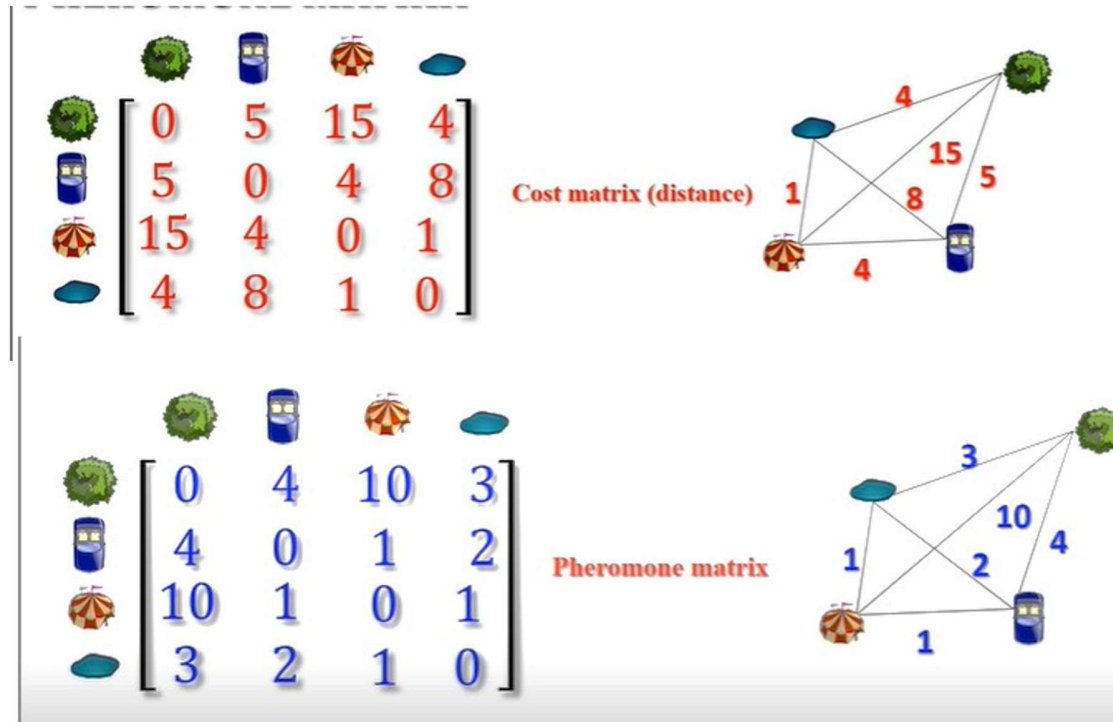
End for

End for

Apply the **global trail updating** rule using **the best ant**

Until End_condition

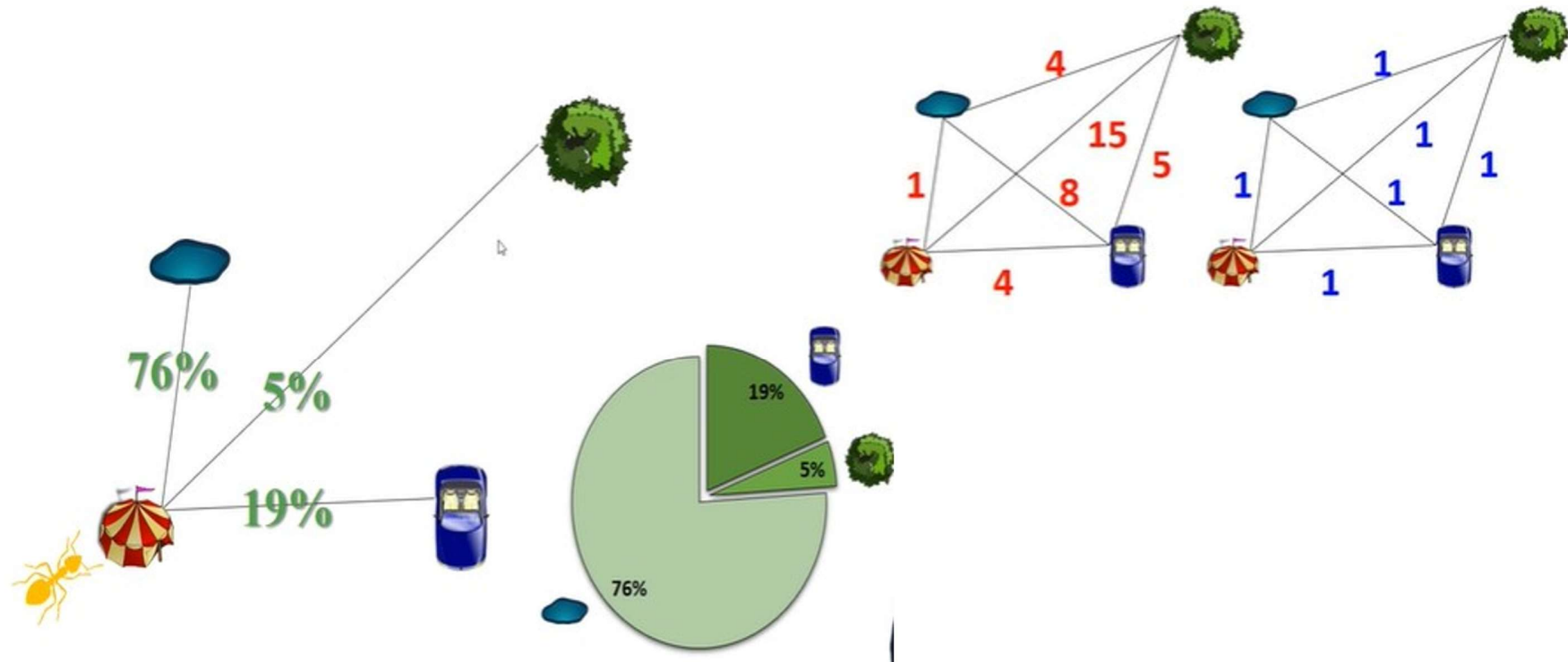
Ant Colony Optimization algorithm



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148

Ant Colony Optimization algorithm

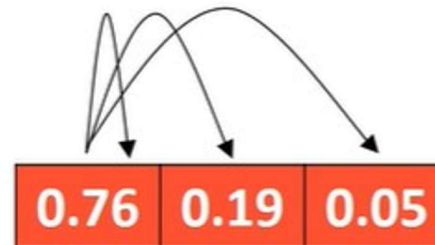


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153

Ant Colony Optimization algorithm

Probabilistic

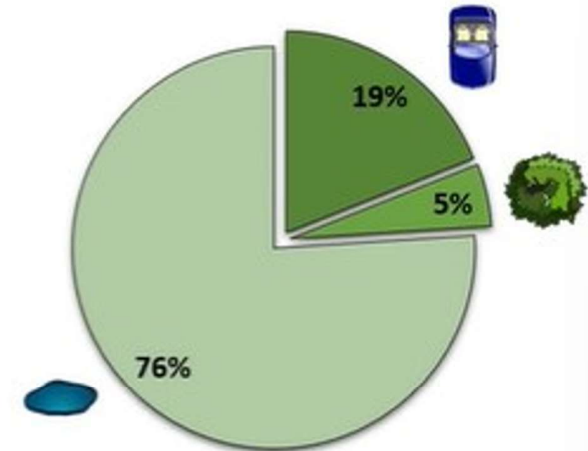


Cumulative sum

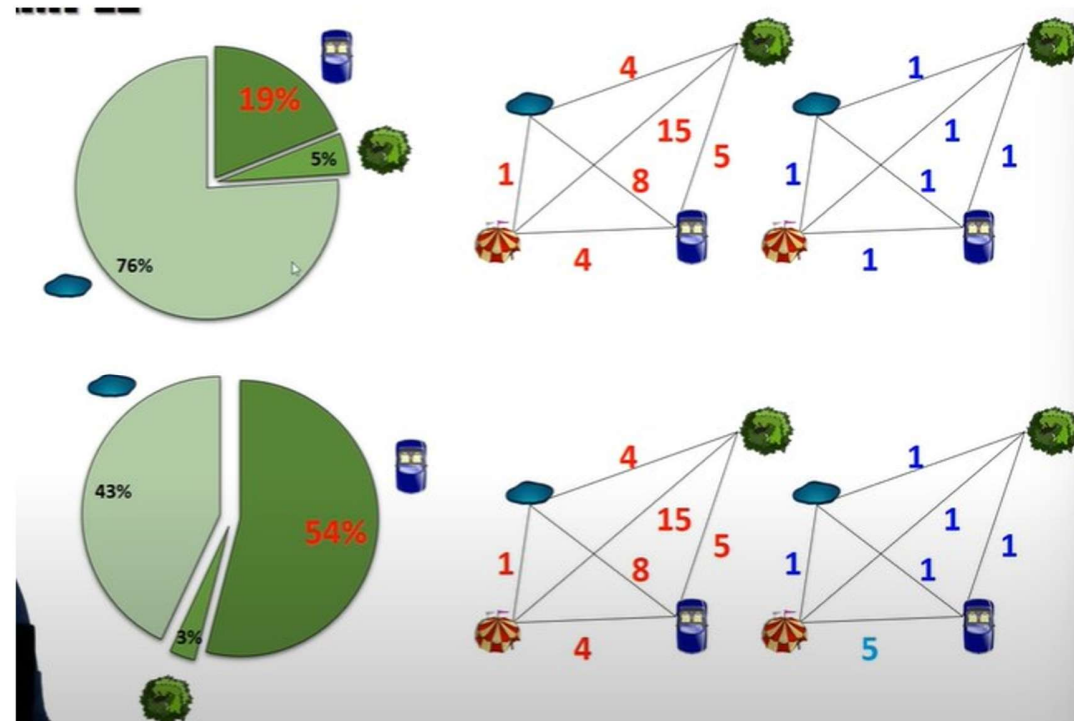


A random number (r) in [0,1]

$$\begin{cases} 0.24 < r \leq 1.00 \\ 0.05 < r \leq 0.24 \\ 0.00 \leq r \leq 0.05 \end{cases}$$



Ant Colony Optimization algorithm



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156

ACO CODE with Python

```

import random
def population_gen(n):
    population=[]
    for i in range(n):
        temp_l = [0, 1, 2, 3, 4]
        random.shuffle(temp_l)
        population.append(temp_l)
    print(population)
    return population
def phermeon_cal():
    fitness=[0]*n

    for i in range(n):
        for j in range(len(l)-1):
            index1=population[i][j]
            index2 = population[i][j+1]
            fitness[i]=fitness[i]+g[index1][index2]
        fitness[i] =1/fitness[i]
    print(fitness)
    for i in range(n):
        for j in range(len(l) - 1):
            index1 = population[i][j]
            index2 = population[i][j + 1]
            phermeon[index1][index2]=
phermeon[index1][index2]+ fitness[i]
    print(phermeon)
    return phermeon
def new_ant_gen():
    for i in range (n):
        old_population=population[i]
        population[i]=[]
        present_city = old_population[0]
        population[i].append(present_city)
        old_population.remove(present_city)
        print(old_population)
        while len(old_population)>0:
            temp_value=[]
            for j in old_population:
temp_value.append(phermeon[present_city][j]*(1/g[present

```

```

_city][j]))
            total_temp=sum(temp_value)
            for j in range (len(temp_value)):
                temp_value[j]=temp_value[j]/total_temp
            present_city =
random.choices(old_population, weights=temp_value) [0]
            old_population.remove(present_city)
            population[i].append(int(present_city))
            print(population[i])
            print('-----')

def best_root():
    fitness = [0] * n
    for i in range(n):
        for j in range(len(l)-1):
            index1=population[i][j]
            index2 = population[i][j+1]
            fitness[i]=fitness[i]+g[index1][index2]
    print(min(fitness))
    index_min=fitness.index(min(fitness))
    print(population[index_min])

g=[ [0,30,4,5,60],
    [2,0,5,8,9],
    [4,5,0,30,19],
    [20,3,5,0,47],
    [70,2,3,4,0],
    ]
print(g)
# A B C D E
city=5
phermeon=[[0]*city]*city
l=[0,1,2,3,4]
n=4
population=population_gen(n)
for i in range(50):
    phermeon=phermeon_cal()
    new_ant_gen()
    best_root()

```

WEEK-11

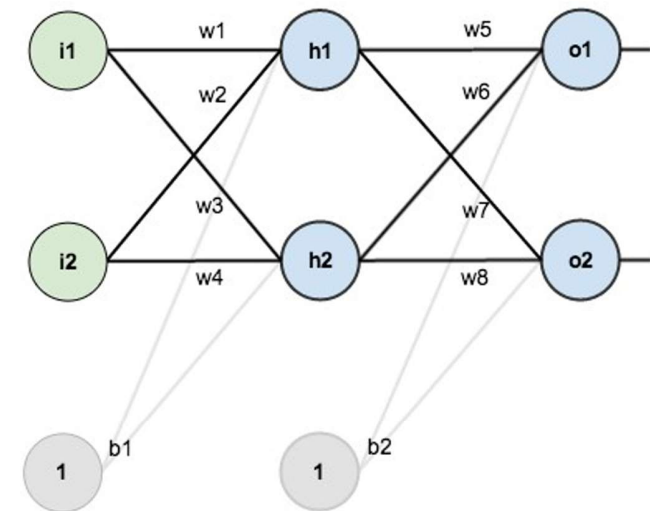
Artificial Neural Networks (ANNs): Forward Pass and Backpropagation

The Forward Pass

- $net_{h1} = (w_1 \times i_1) + (w_2 \times i_2) + (b_1 \times 1)$
- $net_{h2} = (w_3 \times i_1) + (w_4 \times i_2) + (b_1 \times 1)$

Activation Functions

- $out_{h1} = \frac{1}{1+e^{-net_{h1}}}$
- $out_{h2} = \frac{1}{1+e^{-net_{h2}}}$

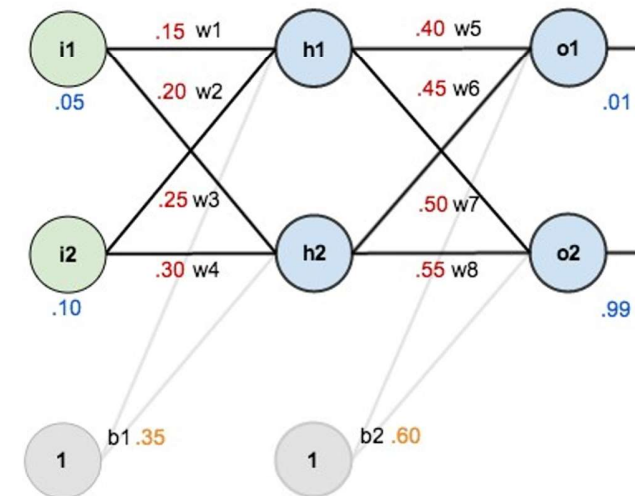


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198

Back propagation Assignment

- The goal of backpropagation is to optimize the weights so that the neural network can learn how to correctly map arbitrary inputs to outputs.
- For the rest of this tutorial, we're going to work with a single training set: given inputs 0.05 and 0.10, we want the neural network to output 0.01 and 0.99 and $\eta=0.1$



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218

ANN CODE With Python

```
from math import exp
from random import uniform
import numpy as np

def forwardPass(wij,wjk,wkl):
    while(True):
        hiIn = additionBious(np.matmul(input, wij), m) # Hidden Layer 1
        hi = relu(hiIn)
        hjIn = additionBious(np.matmul(hi, wjk), o) # Hidden Layer 2
        hj = sigmoid(hjIn)
        hkIn =additionBious(np.matmul(hj,wkl),p) # Output Layer
        hk = softmax(hkIn)
        if float(format(hk[0],"0.2"))!= y[0] or float(format(hk[1],"0.2"))!= y[1]:
            print (format(hk[0], "0.4"))+" : "+(format(hk[1],"0.4"))
            wij, wjk, wkl=backwardPass(wij,wjk,wkl,hiIn,hjIn,hkIn,hi,hj,hk)
            #print wjk
        else:
            print (format(hk[0], "0.2"))+(format(hk[1],"0.2"))
            break

def backwardPass(wij,wjk,wkl,hiIn,hjIn,hkIn,hi,hj,hk):
    nwkl,oOut,oIn = outputLayerToHiddenlayer2(wkl, hkIn, hk,hj)
    nwjk,h2In,eTotal=hiddenLayer2ToHiddenLayer1(wjk,wkl,hjIn,oOut,oIn,hi)
    nwij, e2Total=hiddenLayer1ToInput(wij, wjk, eTotal, h2In, hi)
    #print nwjk
    return nwij,nwjk,nwkl

def hiddenLayer1ToInput(wij,wjk,eTotal,h2In,hi):
    e2Total = []
    nwij = np.zeros((n, m), dtype='float')
```

```

for i in range(m):
    e2Total.append(0)
    for j in range(o):
        e2Total[i] = e2Total[i] + (eTotal[i] * h2In[i] * wjk[i][j])
for i in range(m):
    for j in range(o):
        nwij[i][j] = wij[i][j] - (learningRate * (e2Total[j] * hi[j] * input[i]))
return nwij,e2Total

def hiddenLayer2ToHiddenLayer1(wjk,wkl,hjIn,oOut,oIn,hi):
    h2In=[]
    eTotal=[]
    nwjk = np.zeros((m, o), dtype='float')
    for i in range(o):
        h2In.append(sigmoid1(hjIn[i])*(1-sigmoid1(hjIn[i])))

    for i in range(o):
        eTotal.append(0)
        for j in range(p):
            eTotal[i]=eTotal[i]+(oOut[j] * oIn[j] * wkl[i][j])
    for i in range(m):
        for j in range(o):
            nwjk[i][j]=wjk[i][j]-(learningRate* (eTotal[j]*h2In[j]*hi[i]))
    return nwjk,h2In,eTotal

def outputLayerToHiddenlayer2(wkl,hkIn,hk,hj):
    oOut = []
    oIn = []
    nwkl = np.zeros((o, p), dtype='float')
    for i in range(p):
        oOut.append(-1 * ((y[i] * (1 / hk[i])) + ((1 - y[i]) * (1 / (1 - hk[i])))))
        oIn.append((exp(hkIn[0]) * exp(hkIn[1])) / ((exp(hkIn[0]) + exp(hkIn[1])) * (exp(hkIn[0]) +
exp(hkIn[1]))))

```

```

for i in range(o):
    for j in range(p):
        nwkl[i][j] =wkl[i][j]-(learningRate* (oOut[j] * oIn[j] * hj[j]))
return nwkl,oOut,oIn

```

```

def relu(hi):
    for i in range(m):
        hi[i]=max(hi[i],0)
    return hi
def sigmoid(hj):
    for i in range(o):
        hj[i]=1/(1+exp(-hj[i]))
    return hj
def sigmoidl(w):
    w=1/(1+exp(-w))
    return w
def softmax(hk):
    sum=0
    for i in range(p):
        sum=sum+exp(hk[i])
    try:
        for i in range(p):
            hk[i] =exp(hk[i])/sum
    except:
        return hk
    return hk
def additionBious(w,n):
    for i in range(n):
        w[i]=w[i]+bious
    return w

```

```

def rnadomWightGenerator():
    wij = np.zeros((n, m), dtype='float')
    wjk = np.zeros((m, o), dtype='float')
    wkl = np.zeros((o,p), dtype='float')
    for i in range(n):
        for j in range(m):
            wij[i][j]=(float(format(uniform(.1, .9), '.3f'))))
    for i in range(m):
        for j in range(o):
            wjk[i][j]=(float(format(uniform(.1, .9), '.3f'))))
    for i in range(o):
        for j in range(p):
            wkl[i][j]=(float(format(uniform(.1, .9), '.3f'))))
    forwardPass(wij,wjk,wkl)

n=13
m=3
o=3
p=2
bious=1
learningRate=-0.01
input=[.1,.2,.7,.8,.45,-.64,.32,.1,.2,-.7,.8,-.45,-.64]
y=[0.0,1.0]
rnadomWightGenerator()

```

WEEK-12

Fuzzy logic control in an automated car system

Parameters of the fuzzy sets

SPEED:

slow [20 to 60 kmph]

Medium [40 to 80 kmph]

fast [60 to 100 kmph]

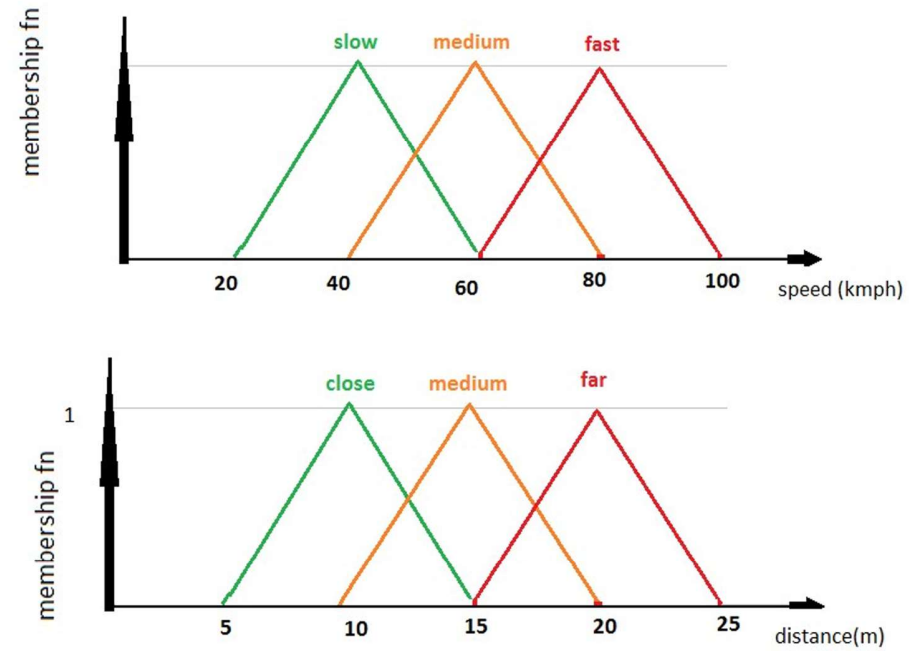
OBSTACLE DISTANCE:

Close [5 to 15 meters]

Medium [10 to 20 meters]

Far [15 to 25 meters]

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Rules

Rule 1: *If (obstacle is far)&(speed is slow) then angle is straight*

Rule 2: *If (obstacle is far)&(speed is medium) then angle is smallT*

Rule 3: *If (obstacle is far)&(speed is fast) then angle is medT*

Rule 4: *If (obstacle is med)&(speed is slow) then angle is smallT*

Rule 5: *If (obstacle is med)&(speed is medium) then angle is medT*

Rule 6: *If (obstacle is med)&(speed is fast) then angle is bigT*

Rule 7: *If (obstacle is close)&(speed is slow) then angle is medT*

Rule 8: *If (obstacle is close)&(speed is medium) then angle is bigT*

Rule 9: *If (obstacle is close)&(speed is fast) then angle is bigT*

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```

import numpy as np
import skfuzzy as fuzz
from skfuzzy import control as ctrl

# Define fuzzy input variables
speed = ctrl.Antecedent(np.arange(0, 121, 1), 'speed') # km/h
distance = ctrl.Antecedent(np.arange(0, 51, 1), 'distance') # meters
brake = ctrl.Consequent(np.arange(0, 101, 1), 'brake') # percentage

# Define membership functions for speed
speed['slow'] = fuzz.trimf(speed.universe, [0, 0, 40])
speed['medium'] = fuzz.trimf(speed.universe, [30, 60, 90])
speed['fast'] = fuzz.trimf(speed.universe, [80, 120, 120])

# Define membership functions for distance
distance['near'] = fuzz.trimf(distance.universe, [0, 0, 15])
distance['medium'] = fuzz.trimf(distance.unive↓ [10, 25, 40])
distance['far'] = fuzz.trimf(distance.universe, [30, 50, 50])

```

```

# Define membership functions for brake intensity
brake['low'] = fuzz.trimf(brake.universe, [0, 0, 40])
brake['medium'] = fuzz.trimf(brake.universe, [30, 50, 70])
brake['high'] = fuzz.trimf(brake.universe, [60, 100, 100])

# Define fuzzy rules
rule1 = ctrl.Rule(speed['slow'] & distance['far'], brake['low'])
rule2 = ctrl.Rule(speed['slow'] & distance['medium'], brake['medium'])
rule3 = ctrl.Rule(speed['slow'] & distance['near'], brake['high'])

rule4 = ctrl.Rule(speed['medium'] & distance['far'], brake['low'])
rule5 = ctrl.Rule(speed['medium'] & distance['medium'], brake['medium'])
rule6 = ctrl.Rule(speed['medium'] & distance['near'], brake['high'])

rule7 = ctrl.Rule(speed['fast'] & distance['far'], brake['medium'])
rule8 = ctrl.Rule(speed['fast'] & distance['medium'], brake['high'])
rule9 = ctrl.Rule(speed['fast'] & distance['near'], brake['high'])

```

```
# Build control system
braking_ctrl = ctrl.ControlSystem([rule1, rule2, rule3,
                                   rule4, rule5, rule6,
                                   rule7, rule8, rule9])

braking = ctrl.ControlSystemSimulation(braking_ctrl)

# Input values
braking.input['speed'] = 45 # km/h
braking.input['distance'] = 12 # meters

# Compute result
braking.compute()

print(f"Brake intensity: {braking.output['brake']:.2f}%")
brake.view(sim=braking)
```