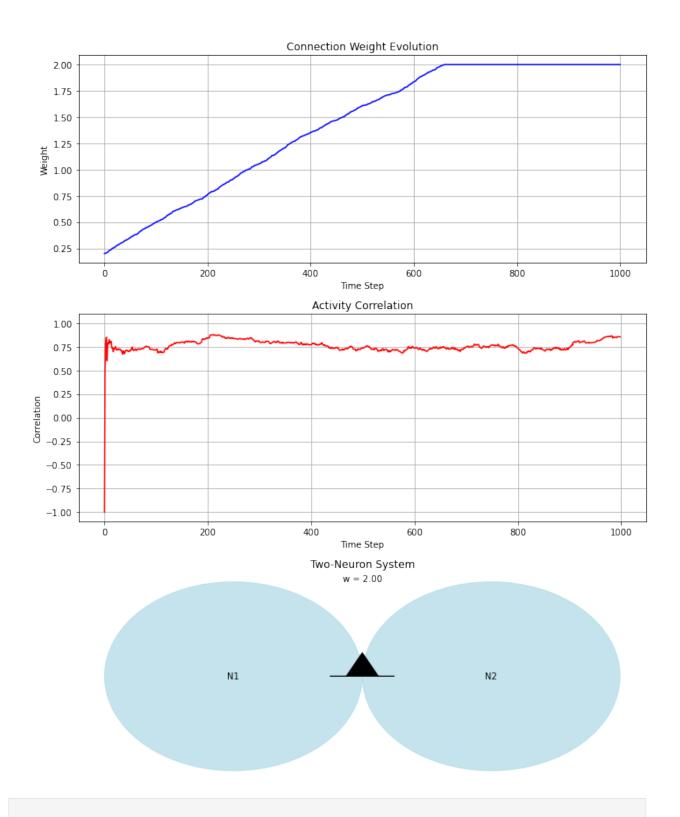
```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.animation import FuncAnimation
# Set random seed for reproducibility
np.random.seed(42)
class TwoNeuronSystem:
    def __init__(self, learning rate=0.01, initial weight=0.2):
        # Initialize connection weight between neurons
        self.weight = initial weight
        self.learning rate = \overline{l}earning rate
        # Store history for visualization
        self.weight history = [initial weight]
        self.activity correlation history = []
# Explanation:
# Defines a class to encapsulate the two-neuron system init is the
constructor that initializes the class when an object is created
# Sets the initial connection weight between neurons to the provided
value (default 0.2)
# Sets the learning rate that controls how quickly weights change
(default 0.01)
# Creates a list to track weight changes over time, starting with the
initial weight
# Creates an empty list to track correlation between neuron activities
    def simulate(self, num timesteps=1000, correlation=0.8):
        Simulate the two-neuron system with correlated/uncorrelated
inputs.
        Args:
            num timesteps: Number of simulation steps
            correlation: Correlation between neuron activities (-1 to
1)
        0.00
# Explanation:
# Defines a method to run the simulation
# Takes parameters for number of time steps and desired correlation
between neurons
        # Storage for neuron activities
        neuron1 activities = np.zeros(num timesteps)
        neuron2 activities = np.zeros(num timesteps)
# Explanation:
# Creates arrays to store the activity values of both neurons at each
time step
# Initializes them with zeros
```

```
# Generate correlated activities
        mean = [0, 0]
        cov = [[1, correlation], [correlation, 1]]
        activities = np.random.multivariate normal(mean, cov,
num timesteps)
# Explanation:
# Sets up parameters for generating correlated random activities
\# mean = [0, 0] specifies that both neurons have a mean activity of 0
# cov is the covariance matrix that determines the correlation:
    # Diagonal values (1) represent variance of each neuron
    # Off-diagonal values (correlation) control how coordinated the
neurons are
#Generates random values from a multivariate normal distribution with
specified correlation
        # Apply sigmoid to constrain activities between 0 and 1
        neuron1 activities = \frac{1}{1} / (\frac{1}{1} + np.exp(-activities[:, 0]))
        neuron2 activities = 1 / (1 + np.exp(-activities[:, 1]))
        # Hebbian learning: "Neurons that fire together, wire
together"
        for t in range(num_timesteps):
            # Current activities
            al = neuron1 activities[t]
            a2 = neuron2 activities[t]
            # Calculate correlation for history
            if t > 0:
                recent corr = np.corrcoef(neuron1 activities[max(0, t-
100):t+1],
                                          neuron2 activities[max(0, t-
100):t+1])[0, 1]
                self.activity correlation history.append(recent corr)
            # Apply Hebbian learning rule (weight change proportional
to product of activities)
            delta w = self.learning rate * a1 * a2
            self.weight += delta w
            # Optional: Weight normalization to prevent unbounded
growth
            self.weight = np.clip(self.weight, 0, 2)
            # Store weight history
            self.weight history.append(self.weight)
        return neuron1 activities, neuron2 activities
    def plot results(self):
```

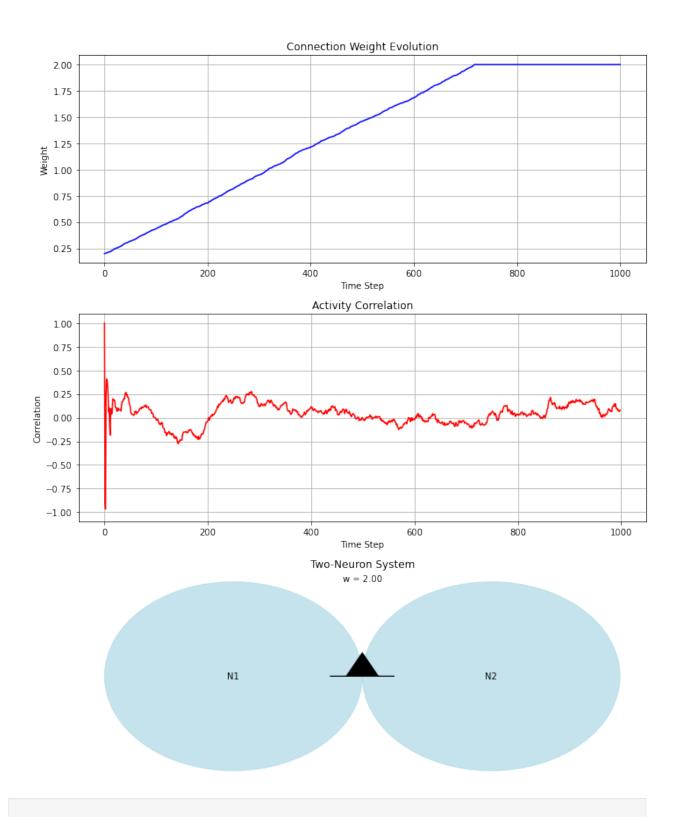
```
"""Plot the simulation results."""
        fig, (ax1, ax2, ax3) = plt.subplots(3, 1, figsize=(10, 12))
        # Plot connection weight over time
        ax1.plot(self.weight history, 'b-')
        ax1.set_title('Connection Weight Evolution')
        ax1.set xlabel('Time Step')
        ax1.set ylabel('Weight')
        ax1.grid(True)
        # Plot correlation history
        ax2.plot(self.activity_correlation_history, 'r-')
        ax2.set_title('Activity Correlation')
        ax2.set xlabel('Time Step')
        ax2.set ylabel('Correlation')
        ax2.set ylim(-1.1, 1.1)
        ax2.grid(True)
        # Visualize neurons and connections
        ax3.axis('off')
        ax3.set title('Two-Neuron System')
        circle1 = plt.Circle((0.3, 0.5), 0.2, color='lightblue',
alpha=0.7)
        circle2 = plt.Circle((0.7, 0.5), 0.2, color='lightblue',
alpha=0.7)
        ax3.add_patch(circle1)
        ax3.add patch(circle2)
        # Draw arrow for connection
        arrow width = self.weight history[-1] * 0.05
        arrow = ax3.arrow(0.5, 0.5, 0, 0, head width=0.05,
                          head length=0.05, fc='black', ec='black',
                          width=arrow width)
        ax3.text(0.3, 0.5, 'N1', ha='center', va='center')
        ax3.text(0.7, 0.5, 'N2', ha='center', va='center')
        ax3.text(0.5, 0.7, f'w = {self.weight history[-1]:.2f}',
ha='center')
        plt.tight layout()
        plt.show()
    def visualize_dynamic_system(self):
        """Create an animation showing the changing connection
strength."""
        fig, ax = plt.subplots(figsize=(8, 6))
        ax.set xlim(0, 1)
        ax.set ylim(0, 1)
        ax.axis('off')
```

```
ax.set title('Two-Neuron System with Changing Connection
Strength')
        # Create the neurons
        circle1 = plt.Circle((0.3, 0.5), 0.2, color='lightblue',
alpha=0.7
        circle2 = plt.Circle((0.7, 0.5), 0.2, color='lightblue',
alpha=0.7)
        ax.add patch(circle1)
        ax.add patch(circle2)
        # Add neuron labels
        ax.text(0.3, 0.5, 'N1', ha='center', va='center')
        ax.text(0.7, 0.5, 'N2', ha='center', va='center')
        # Initial weight text
        weight text = ax.text(0.5, 0.8, f'Weight:
{self.weight_history[0]:.2f}', ha='center')
        step text = ax.text(0.5, 0.9, f'Step: 0', ha='center')
        # Initial arrow for connection
        arrow = ax.arrow(0.5, 0.5, 0, 0, head width=0.05,
                         head length=0.05, fc='black', ec='black',
                         width=self.weight_history[0] * 0.05)
        def update(frame):
            # Update the weight text
            weight text.set text(f'Weight:
{self.weight history[frame]:.2f}')
            step_text.set_text(f'Step: {frame}')
            # Remove old arrow and create a new one
            if 'arrow' in locals():
                arrow.remove()
            # Scale arrow width with weight
            arrow width = self.weight history[frame] * 0.05
            new_arrow = ax.arrow(0.4, 0.5, 0.2, 0, head_width=0.05,
                              head length=0.05, fc='black',
ec='black',
                              width=arrow width)
            return weight_text, step_text, new_arrow
        # Create animation (showing fewer frames for performance)
        ani = FuncAnimation(fig, update, frames=range(0,
len(self.weight_history), 10),
                            blit=True, repeat=False)
```

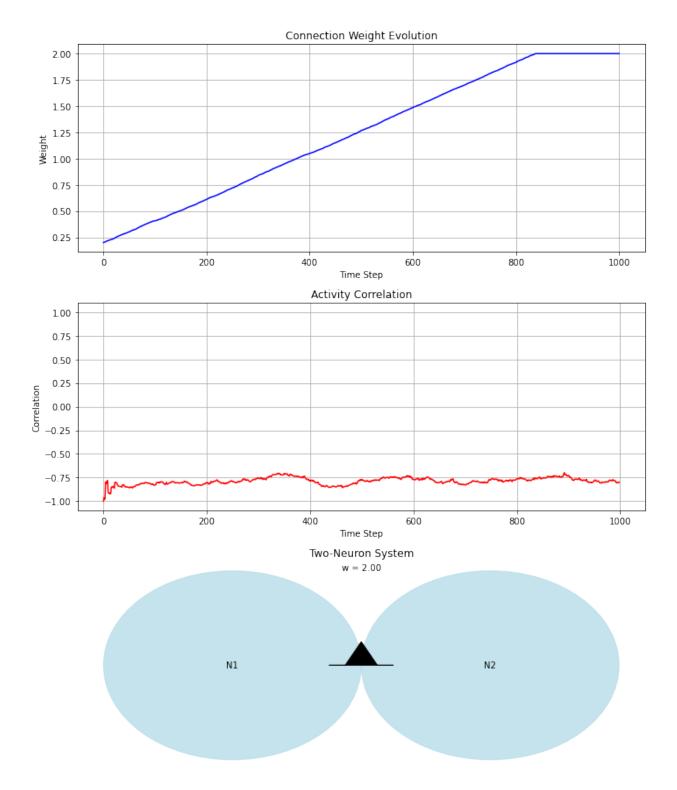
```
plt.tight layout()
        plt.show()
# Run simulations with different correlations
def run demonstration():
   # 1. Positively correlated activity
   print("Simulation 1: Positively Correlated Neurons")
    positive sys = TwoNeuronSystem(learning rate=0.01,
initial weight=0.2)
    positive sys.simulate(num timesteps=1000, correlation=0.8)
   positive_sys.plot results()
   # 2. Uncorrelated activity
   print("\nSimulation 2: Uncorrelated Neurons")
   uncorrelated sys = TwoNeuronSystem(learning rate=0.01,
initial weight=0.2)
   uncorrelated sys.simulate(num timesteps=1000, correlation=0.0)
   uncorrelated sys.plot results()
   # 3. Negatively correlated activity
   print("\nSimulation 3: Negatively Correlated Neurons")
   negative sys = TwoNeuronSystem(learning rate=0.01,
initial weight=0.2)
   negative sys.simulate(num timesteps=1000, correlation=-0.8)
   negative sys.plot results()
    return positive sys, uncorrelated sys, negative sys
if name == " main ":
   positive sys, uncorrelated sys, negative sys = run demonstration()
   # Optional: Show animation for positive correlation
   positive_sys.visualize_dynamic_system()
Simulation 1: Positively Correlated Neurons
```



Simulation 2: Uncorrelated Neurons



Simulation 3: Negatively Correlated Neurons



Two-Neuron System with Changing Connection Strength

Step: 0

Weight: 0.20

