

Assignment3

April 5, 2025

0.1 Question1

```
[2]: import numpy as np

def maxnet_network(input_vector, epsilon=-0.15):
    x = np.array(input_vector)
    n = len(x)
    # Construct the weight matrix: off-diagonals are epsilon, diagonals are 1
    # (compensating for epsilon)
    w = np.full((n, n), epsilon) + np.eye(n) * (1 - epsilon)

    iteration = 0
    stop = False
    while not stop:
        iteration += 1
        # Compute input for each node:  $u = x * w^T$ 
        u = np.dot(x, w.T)
        # Apply ReLU activation function
        v = np.maximum(0, u)
        x = v.copy()
        # Count the number of active (non-zero) neurons
        active_neurons = np.count_nonzero(v > 0)

        # Stop if only one neuron is active
        if active_neurons <= 1:
            stop = True
            winner_index = np.where(v > 0)[0]

    if winner_index.size > 0:
        print(f"MAXNET: The winning neuron is neuron {winner_index[0] + 1} with
        a final activation of {v[winner_index[0]]:.4f}, after {iteration} iterations.
        ")
    else:
        print("MAXNET: No active neuron found.")

# Test the MAXNET network
input_vector = [0.1, 0.3, 0.5, 0.7, 0.9]
maxnet_network(input_vector)
```

MAXNET: The winning neuron is neuron 5 with a final activation of 0.4541, after 6 iterations.

```
[3]: def kWTA_network(input_vector, k, deltaT=0.05):
    u = np.array(input_vector)
    n = len(u)
    y = 0.0          # Initialize state variable
    iteration = 0
    stop = False
    # Initialize output vector as zeros
    x = np.zeros(n)

    while not stop:
        iteration += 1
        y = y + deltaT * (np.sum(x) - k)
        # Update outputs using an infinite gain activation function:
        # if u[i] >= y, output 1; otherwise 0.
        for i in range(n):
            if u[i] >= y:
                x[i] = 1
            else:
                x[i] = 0

        active_neurons = int(np.sum(x))

        # Stop when the number of active neurons is exactly equal to k
        if active_neurons == k:
            stop = True
            winners = np.where(x == 1)[0]

        print(f"kWTA (k={k}): Winning neurons are {winners + 1} with threshold y =_{y:.4f}, after {iteration} iterations.")

    # Test the kWTA network for k=1 and k=2
    print("For k = 1:")
    kWTA_network(input_vector, 1)

    print("\nFor k = 2:")
    kWTA_network(input_vector, 2)
```

For k = 1:
kWTA (k=1): Winning neurons are [5] with threshold y = 0.7000, after 8 iterations.

For k = 2:
kWTA (k=2): Winning neurons are [4 5] with threshold y = 0.5500, after 9 iterations.

0.2 Question2

```
[6]: import numpy as np
import matplotlib.pyplot as plt

# Hopfield
def hopfield_network(patterns, probes, iterations=5):
    patterns = np.array(patterns)
    n = patterns.shape[1]

    #
    W = np.zeros((n, n))
    for p in patterns:
        W += np.outer(p, p)
    np.fill_diagonal(W, 0) # 0

    #
    def plot_pattern(pattern, title):
        plt.imshow(pattern.reshape(3,3), cmap='gray')
        plt.title(title)
        plt.axis('off')
        plt.show()

    #
    print("Original Stored Patterns:")
    for i, p in enumerate(patterns):
        plot_pattern(p, f'Original Pattern {i+1}')

    #
    for idx, probe in enumerate(probes):
        print(f"\nRetrieving from Noisy Pattern {idx+1}:")
        s = probe.copy()
        for it in range(iterations):
            s = np.sign(W @ s)
            s[s==0] = 1 # 0 1
        plot_pattern(probe, f'Noisy Input {idx+1}')
        plot_pattern(s, f'Retrieved Pattern {idx+1}')

    # 1 -1
    pattern1 = np.array([-1, -1, -1,
                          1, -1, 1,
                          1, -1, 1])

    pattern2 = np.array([-1, -1, -1,
                          -1, 1, 1,
                          -1, -1, -1])
```

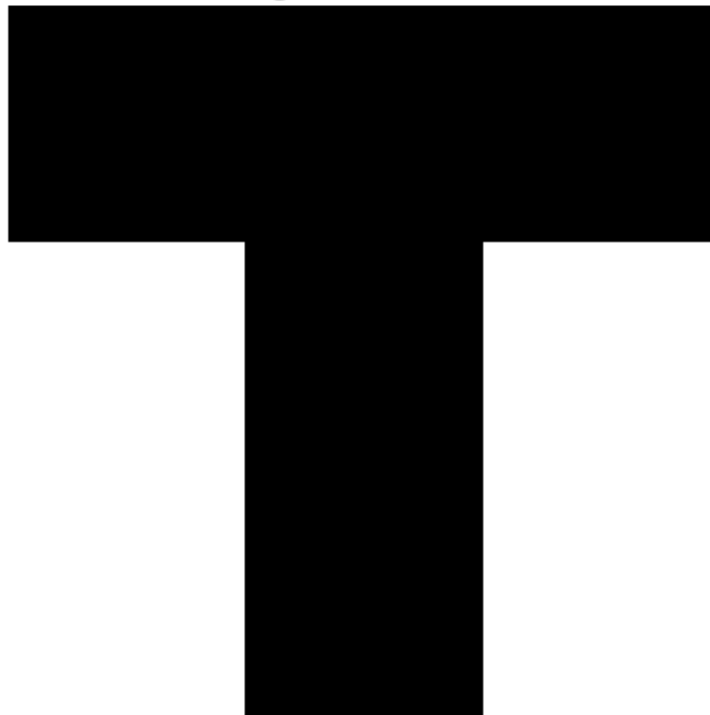
```
#      (      )
probe1 = np.array([-1, -1, -1,
                   1, -1, 1,
                   -1, -1, 1])

probe2 = np.array([-1, -1, -1,
                   -1, 1, 1,
                   -1, -1, 1])

# Hopfield
hopfield_network([pattern1, pattern2], [probe1, probe2])
```

Original Stored Patterns:

Original Pattern 1

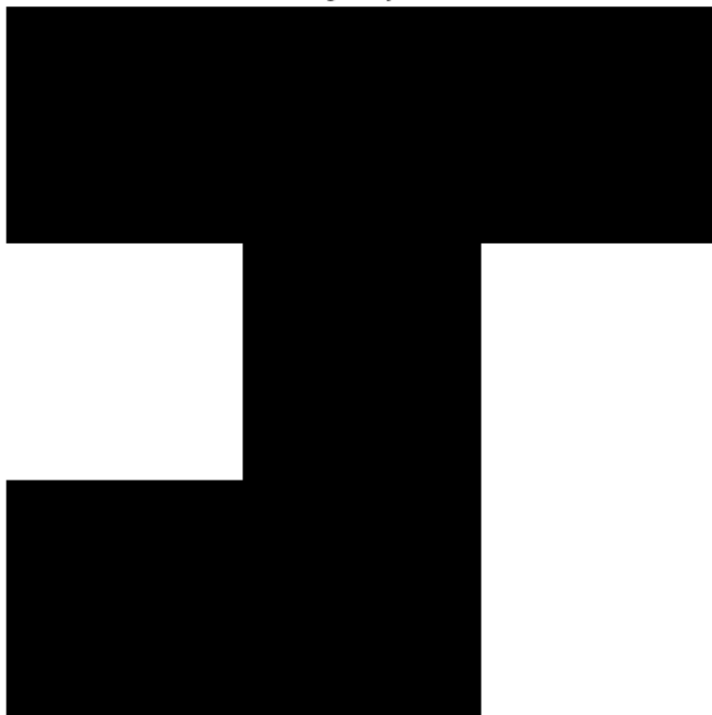


Original Pattern 2

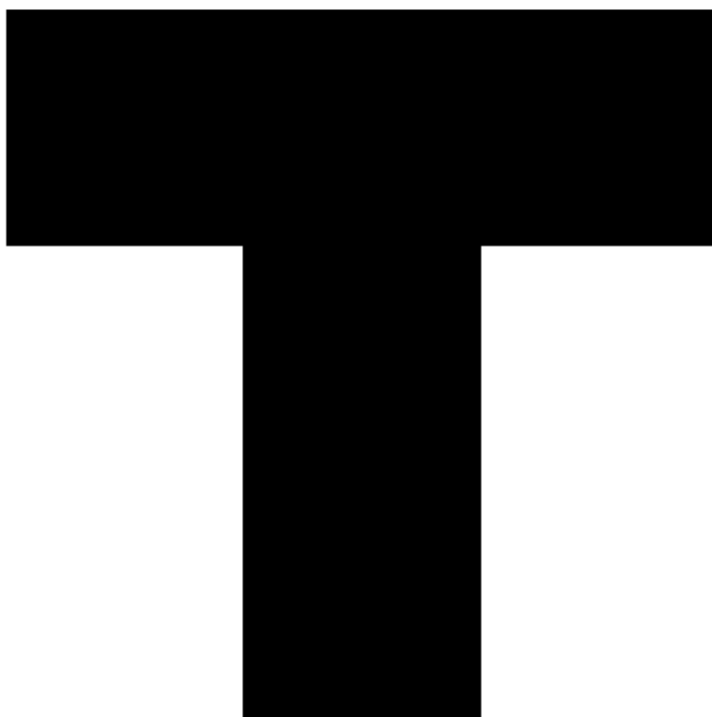


Retrieving from Noisy Pattern 1:

Noisy Input 1

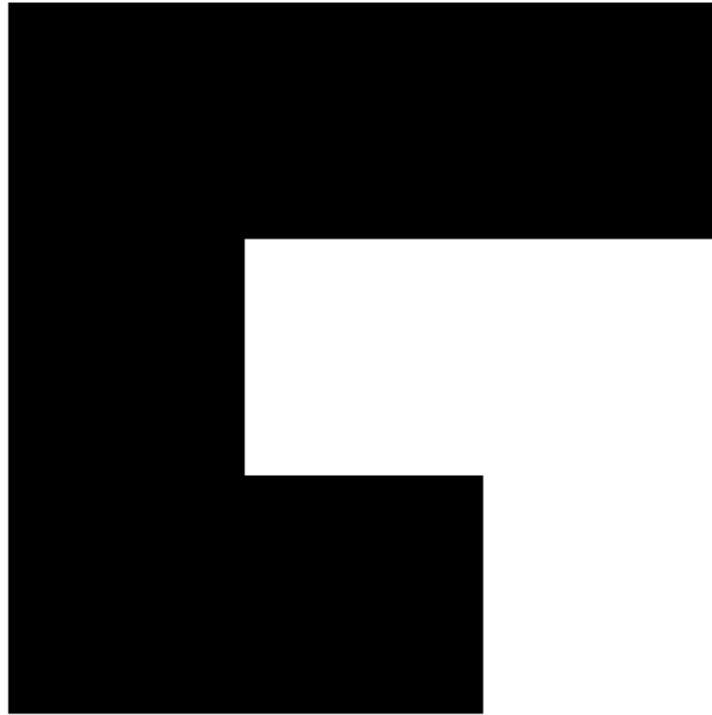


Retrieved Pattern 1



Retrieving from Noisy Pattern 2:

Noisy Input 2



Retrieved Pattern 2

