

```

1 import numpy as np
2 from hopfield import DeterministicHopfieldNetwork
3 from pattern_utilities import generate_n_random_patterns, print_pattern
4
5
6 def main():
7     n_bits = 120
8     n_patterns_vector = [12, 24, 48, 70, 100, 120]
9     diagonal_weights_rule = "non-zero"
10    n_iterations = int(1e5)
11    one_step_error_probability = np.zeros(len(n_patterns_vector))
12
13    for n_patterns_i, n_patterns in enumerate(n_patterns_vector):
14        n_errors = 0
15        for i in range(n_iterations):
16            network = DeterministicHopfieldNetwork()
17
18            patterns = generate_n_random_patterns(n_patterns, n_bits)
19            network.set_patterns(patterns)
20            network.set_diagonal_weights_rule(diagonal_weights_rule)
21            network.generate_weights()
22
23            pattern_to_feed_index = np.random.randint(n_patterns)
24            original_pattern = patterns[pattern_to_feed_index, :]
25            neuron_to_update_index = np.random.randint(n_bits)
26
27            updated_pattern = network.update_neuron(
28                original_pattern,
29                neuron_to_update_index
30            )
31            updated_neuron = updated_pattern[neuron_to_update_index]
32            original_neuron = original_pattern[neuron_to_update_index]
33            updated_neuron = updated_pattern[neuron_to_update_index]
34
35            if original_neuron != updated_neuron:
36                n_errors += 1
37
38            one_step_error_probability[n_patterns_i] = n_errors/n_iterations
39
40    print_pattern(one_step_error_probability)
41
42
43 if __name__ == "__main__":
44     main()

```

Listing 1: Main method for computing the one-step error probability.

```

1 import numpy as np
2 from scipy import stats
3 from abc import ABC, abstractmethod
4
5
6 class HopfieldNetwork(ABC):
7     def set_diagonal_weights_rule(self, diagonal_weights_rule):
8         if diagonal_weights_rule == "zero":
9             self.diagonal_weights_equal_zero = True
10        elif diagonal_weights_rule == "non-zero":
11            self.diagonal_weights_equal_zero = False
12        else:
13            raise KeyError(
14                diagonal_weights_rule
15                + " not a valid diagonal_weights_rule"
16            )
17
18    def set_patterns(self, patterns):
19        self.patterns = patterns
20
21    def generate_weights(self):
22        _, n_bits = self.patterns.shape
23        self.weights = np.zeros((n_bits, n_bits))
24        for pattern in self.patterns:
25            self.weights += np.outer(pattern, pattern)
26        self.weights /= n_bits
27
28        if self.diagonal_weights_equal_zero:
29            np.fill_diagonal(self.weights, 0)
30
31    def asynchronous_update(self, pattern, n_updates):
32        updated_pattern = pattern.copy()
33        for i in range(n_updates):
34            updated_pattern = self.update_random_neuron(updated_pattern)
35        return updated_pattern
36
37    def update_random_neuron(self, pattern):
38        n_bits = pattern.shape
39        neuron_index = np.random.randint(n_bits)
40        return self.update_neuron(pattern, neuron_index)
41
42    def update_neuron(self, pattern, neuron_index):
43        weights_i = self.weights[neuron_index, :]
44        local_field = np.inner(weights_i, pattern)
45        updated_bit = self.get_state_of_local_field(local_field)
46        updated_pattern = pattern.copy()
47        updated_pattern[neuron_index] = updated_bit
48        return updated_pattern

```

```

49
50     @abstractmethod
51     def get_state_of_local_field(self, local_field):
52         pass
53
54
55 class DeterministicHopfieldNetwork(HopfieldNetwork):
56     def get_state_of_local_field(self, local_field):
57         return sign_zero_returns_one(local_field)
58
59
60 class StochasticHopfieldNetwork(HopfieldNetwork):
61     def get_state_of_local_field(self, local_field):
62         p = 1/(1 + np.exp(-2*self.noise_parameter*local_field))
63         rand = stats.bernoulli.rvs(p)
64         return 1 if rand else -1
65
66     def set_noise_parameter(self, noise_parameter):
67         self.noise_parameter = noise_parameter
68
69
70 def sign_zero_returns_one(value):
71     return 1 if value >= 0 else -1

```

Listing 2: Classes for creating Hopfield Networks.

```

1  import numpy as np
2  from scipy import stats
3  import matplotlib.pyplot as plt
4
5
6  def generate_n_random_patterns(n_patterns, n_bits):
7      random_0s_and_1s = stats.bernoulli.rvs(0.5, size=(n_patterns, n_bits))
8      random_minus_1s_and_1s = 2*random_0s_and_1s - 1
9      return random_minus_1s_and_1s
10
11
12 def get_index_of_equal_pattern(pattern_to_match, patterns):
13     for index, pattern in enumerate(patterns):
14         n_different_bits = get_n_different_bits(pattern_to_match, pattern)
15         if n_different_bits == 0:
16             return index
17     return -1
18
19
20 def get_n_different_bits(pattern1, pattern2):
21     return sum(pattern1 != pattern2)
22
23

```

```

24 def vector_to_typewriter(vector, n_columns):
25     return np.reshape(vector, (-1, n_columns))
26
27
28 def print_typewriter_pattern(pattern, n_columns):
29     print_pattern(vector_to_typewriter(pattern, n_columns))
30
31
32 def print_pattern(pattern):
33     np.set_printoptions(formatter={"float_kind": lambda x: "%.4f" % x})
34     print(repr(pattern), sep=", ")
35
36
37 def plot_pattern(pattern):
38     plt.imshow(pattern, cmap="Greys")
39     plt.tick_params(
40         axis="both",
41         which="both",
42         bottom=False,
43         top=False,
44         left=False,
45         labelbottom=False,
46         labelleft=False)

```

Listing 3: Help module for handling patterns.