```
import numpy as np
    from hopfield import DeterministicHopfieldNetwork
2
    import pattern_utilities as utils
    import matplotlib
    matplotlib.use("TkAgg")
    import matplotlib.pyplot as plt
    def main():
        pattern_to_feed = get_pattern1()
10
        n_bits = pattern_to_feed.size
11
        stored_patterns = get_stored_patterns()
12
        n_{epochs} = int(1e3)
13
14
        network = DeterministicHopfieldNetwork()
15
        network.set_diagonal_weights_rule("zero")
16
        network.set_patterns(stored_patterns)
17
        network.generate_weights()
18
19
        updated_pattern = pattern_to_feed.copy()
20
        for epoch in range(n_epochs):
21
            for neuron in range(n_bits):
22
                updated_pattern = network.update_neuron(updated_pattern, neuron)
23
24
        matching_pattern_index = utils.get_index_of_equal_pattern(
25
                updated_pattern, stored_patterns)
26
        inverted_stored_patterns = -1 * stored_patterns
27
        matching_inverted_pattern_index = utils.get_index_of_equal_pattern(
28
                updated_pattern, inverted_stored_patterns)
29
        if matching_pattern_index >= 0:
30
            matching_pattern = stored_patterns[matching_pattern_index, :]
31
        elif matching_inverted_pattern_index >= 0:
32
            matching_pattern_index = -1 * matching_inverted_pattern_index
33
            matching_pattern = inverted_stored_patterns[
34
                    matching_inverted_pattern_index, :]
35
        else:
36
            matching_pattern_index = 5
37
            matching_pattern = np.array([])
39
        if matching_pattern_index > 0:
40
            matching_pattern_index += 1
        else:
42
            matching_pattern_index -= 1
43
        print("Converged to pattern {}".format(matching_pattern_index))
45
46
        print("Actual pattern: ")
47
```

```
n_{columns} = 10
48
        utils.print_typewriter_pattern(updated_pattern, n_columns)
49
50
        plt.subplot(1, 3, 1)
51
        utils_plot_pattern(utils_vector_to_typewriter(pattern_to_feed, n_columns))
52
        plt.title("Fed pattern")
53
        plt.subplot(1, 3, 2)
54
        utils.plot_pattern(utils.vector_to_typewriter(updated_pattern, n_columns))
55
        plt.title("Updated pattern")
56
        if matching_pattern.size > 0:
57
             plt.subplot(1, 3, 3)
58
             utils.plot_pattern(
                      utils.vector_to_typewriter(updated_pattern, n_columns)
60
61
             plt.title("Pattern recognized as")
62
        plt.show()
63
64
65
    def get_stored_patterns():
        x1 = np.array(
67
                 [ [-1, -1, -1, -1, -1, -1, -1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, 1]
68
                  \leftarrow -1, -1, -1],[ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1],[ -1, 1, 1, 1, -1,
                  \leftarrow -1, 1, 1, 1, -1],[-1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[-1, 1, 1,
                  \hookrightarrow 1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1,
                  \rightarrow 1, 1, 1, -1, -1, 1, 1, 1, -1],[-1, 1, 1, 1, -1, -1, 1, 1, 1,
                  \leftarrow -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1,
                  \hookrightarrow 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1,
                  \leftarrow -1, 1, 1, 1, -1],[-1, -1, 1, 1, 1, 1, 1, 1, -1, -1],[-1, -1, -1,
                  \hookrightarrow 1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, -1, -1, -1, -1, -1, -1]
                  \hookrightarrow
                 ).flatten();
69
70
        x2 = np.array(
71
                 [ [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1, -1]
72
                  \leftarrow -1, -1],[ -1, -1, -1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1,
                  \rightarrow 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1,
                  \leftarrow -1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1,
                  \rightarrow -1],[-1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[-1, -1, -1, 1, 1, 1,
                  \rightarrow 1, -1, -1, -1],[-1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[-1, -1, -1,
                  \hookrightarrow 1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[
                  \leftarrow -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[-1, -1, -1, 1, 1, 1, 1, -1,
                  \rightarrow -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1,
                  \hookrightarrow 1, 1, -1, -1]
                 ).flatten();
73
74
        x3 = np.array(
75
```

```
[ [ 1, 1, 1, 1, 1, 1, 1, 1, -1, -1], [ 1, 1, 1, 1, 1, 1, 1, 1, -1,
                  \leftarrow -1],[ -1, -1, -1, -1, -1, 1, 1, 1, -1, -1],[ -1, -1, -1, -1, -1,
                  \rightarrow 1, 1, 1, -1, -1], [ -1, -1, -1, -1, 1, 1, 1, -1, -1], [ -1, -1,
                  \leftarrow -1, -1, -1, 1, 1, 1, -1, -1],[ -1, -1, -1, -1, -1, 1, 1, 1, -1,
                  \rightarrow -1],[1, 1, 1, 1, 1, 1, 1, -1, -1],[1, 1, 1, 1, 1, 1, 1, 1, 1,
                  \leftarrow -1, -1],[1, 1, 1, -1, -1, -1, -1, -1, -1],[1, 1, 1, -1, -1,
                  \rightarrow -1, -1, -1, -1, -1],[1, 1, 1, -1, -1, -1, -1, -1, -1, -1],[1, 1,
                  \rightarrow 1, -1, -1, -1, -1, -1, -1],[ 1, 1, 1, -1, -1, -1, -1, -1, -1,
                  \rightarrow -1],[1, 1, 1, 1, 1, 1, 1, 1, -1, -1],[1, 1, 1, 1, 1, 1, 1, 1, 1,
                  \hookrightarrow -1, -1]]
                 ).flatten();
77
        x4 = np.array(
79
                 [ [ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1], [ -1, -1, 1, 1, 1, 1, 1, 1, 1, 1,
80
                  \leftarrow -1],[-1, -1, -1, -1, -1, 1, 1, 1, -1],[-1, -1, -1, -1, -1,
                  \hookrightarrow -1, 1, 1, 1, -1],[-1, -1, -1, -1, -1, 1, 1, 1, -1],[-1, -1,
                  \leftarrow -1, -1, -1, -1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, 1, 1, 1,
                  \leftarrow -1],[ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1],[ -1, -1, 1, 1, 1, 1, 1,
                  \rightarrow 1, -1, -1],[ -1, -1, -1, -1, -1, 1, 1, 1, -1],[ -1, -1, -1,
                     -1, -1, -1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, 1, 1, 1, -1], [
                  \hookrightarrow -1, -1, -1, -1, -1, -1, 1, 1, 1, -1],[ -1, -1, -1, -1, -1, -1, 1,
                  \hookrightarrow 1, 1, -1],[ -1, -1, 1, 1, 1, 1, 1, 1, 1, -1],[ -1, -1, 1, 1, 1, 1,
                  \hookrightarrow 1, 1, -1, -1]
                 ).flatten();
81
82
        x5 = np.array(
83
                 [ [-1, 1, 1, -1, -1, -1, -1, 1, 1, -1], [-1, 1, 1, -1, -1, -1, -1, 1,
84
                  \rightarrow 1, -1],[-1, 1, 1, -1, -1, -1, 1, 1, -1],[-1, 1, 1, -1, -1,
                  \rightarrow -1, -1, 1, 1, -1],[ -1, 1, 1, -1, -1, -1, 1, 1, -1],[ -1, 1,
                  \rightarrow 1, -1, -1, -1, -1, 1, 1, -1], [-1, 1, 1, -1, -1, -1, -1, 1, 1,
                  \leftarrow -1],[-1, 1, 1, 1, 1, 1, 1, 1, -1],[-1, 1, 1, 1, 1, 1, 1, 1,
                  \hookrightarrow 1, -1],[-1, -1, -1, -1, -1, -1, 1, 1, -1],[-1, -1, -1, -1,
                  \leftarrow -1, -1, -1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, 1, 1, 1, -1], [
                  \hookrightarrow -1, -1, -1, -1, -1, -1, 1, 1, -1],[ -1, -1, -1, -1, -1, -1,
                  \rightarrow -1, 1, 1, -1],[-1, -1, -1, -1, -1, -1, 1, 1, -1],[-1, -1,
                  \rightarrow -1, -1, -1, -1, 1, 1, -1]
                 ).flatten();
85
86
        return np.vstack((x1, x2, x3, x4, x5))
87
88
89
    def get_pattern1():
90
        return np.array(
91
```

```
[[1, 1, 1, -1, -1, -1, -1, 1, 1, 1], [-1, -1, -1, 1, 1, 1, 1, -1, -1,
                  1, 1, 1, 1, -1, -1, -1, [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1],
                  [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1,
                  -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1,
                  -1, 1, 1, 1, -1, -1, -1, -1, -1, -1, -1, 1, 1, 1, -1, -1, -1, -1],
               \hookrightarrow [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1,
               \rightarrow -1, -1]]
               ).flatten()
93
94
95
   def get_pattern2():
96
       return np.array(
97
               [[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1], [1, 1, 1, -1, -1, -1, -1, 1, 1, 1],
                  [1, 1, -1, -1, -1, -1, -1, -1, 1, 1], [1, -1, -1, -1, 1, 1, -1,
               \leftarrow -1, -1, 1], [1, -1, -1, -1, 1, 1, -1, -1, -1, 1], [1, -1, -1, -1,
               \hookrightarrow 1, 1, -1, -1, -1, 1], [1, -1, -1, -1, 1, 1, -1, -1, -1, 1], [1,
                  -1, -1, -1, 1, 1, -1, -1, -1, 1], [1, -1, -1, -1, 1, -1, 1, 1, 1,
                  -1], [-1, 1, 1, 1, -1, -1, 1, 1, 1, -1], [-1, 1, 1, 1, -1, -1, 1,
                  1, 1, -1], [-1, 1, 1, 1, -1, -1, 1, 1, 1, -1], [-1, 1, 1, 1, -1,
                  -1, 1, 1, 1, -1], [-1, -1, 1, 1, 1, 1, 1, -1, -1], [-1, -1, -1,
               \hookrightarrow 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, -1, -1, -1, -1, -1, -1, -1]]
               ).flatten()
99
100
101
   def get_pattern3():
102
103
       return np.array(
               [[-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1,
104
               \leftarrow -1, -1], [-1, -1, -1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1,
                 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1,
                  -1, 1, 1, 1, -1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, -1, -1, -1],
                  [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1,
                  -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1,
                  -1, 1, 1, 1, -1, -1, -1, -1, [-1, -1, -1, 1, 1, 1, -1, -1, -1],
               \hookrightarrow [1, 1, 1, -1, -1, -1, -1, 1, 1, 1], [1, 1, 1, -1, -1, -1, -1, 1,
               ).flatten()
105
106
107
   if __name__ == "__main__":
108
       main()
109
```

Listing 1: Main method for recognizing digits.

```
import numpy as np
from scipy import stats
```

```
from abc import ABC, abstractmethod
3
4
    class HopfieldNetwork(ABC):
6
        def set_diagonal_weights_rule(self, diagonal_weights_rule):
            if diagonal_weights_rule == "zero":
                self.diagonal_weights_equal_zero = True
            elif diagonal_weights_rule == "non-zero":
10
                self.diagonal_weights_equal_zero = False
            else:
12
                raise KeyError(
13
                         diagonal_weights_rule
14
                         + " not a valid diagonal_weights_rule"
15
16
17
        def set_patterns(self, patterns):
18
            self.patterns = patterns
19
20
        def generate_weights(self):
21
            _, n_bits = self.patterns.shape
22
            self.weights = np.zeros((n_bits, n_bits))
23
            for pattern in self.patterns:
24
                self.weights += np.outer(pattern, pattern)
25
            self.weights /= n_bits
26
27
            if self.diagonal_weights_equal_zero:
28
                np.fill_diagonal(self.weights, 0)
29
30
        def asynchronous_update(self, pattern, n_updates):
31
            updated_pattern = pattern.copy()
32
            for i in range(n_updates):
33
                updated_pattern = self.update_random_neuron(updated_pattern)
            return updated_pattern
35
36
        def update_random_neuron(self, pattern):
            n_bits = pattern.shape
38
            neuron_index = np.random.randint(n_bits)
39
            return self.update_neuron(pattern, neuron_index)
41
        def update_neuron(self, pattern, neuron_index):
42
            weights_i = self.weights[neuron_index, :]
            local_field = np.inner(weights_i, pattern)
            updated_bit = self.get_state_of_local_field(local_field)
45
            updated_pattern = pattern.copy()
46
            updated_pattern[neuron_index] = updated_bit
            return updated_pattern
48
49
        @abstractmethod
        def get_state_of_local_field(self, local_field):
51
```

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

Listing 2: Classes for creating Hopfield Networks.

```
import numpy as np
    from scipy import stats
2
    import matplotlib.pyplot as plt
    def generate_n_random_patterns(n_patterns, n_bits):
        random_0s_and_1s = stats.bernoulli.rvs(0.5, size=(n_patterns, n_bits))
        random_minus_1s_and_1s = 2*random_0s_and_1s - 1
        return random_minus_1s_and_1s
10
11
    def get_index_of_equal_pattern(pattern_to_match, patterns):
12
        for index, pattern in enumerate(patterns):
13
            n_different_bits = get_n_different_bits(pattern_to_match, pattern)
14
            if n_different_bits == 0:
15
                return index
16
        return -1
17
18
19
    def get_n_different_bits(pattern1, pattern2):
20
        return sum(pattern1 != pattern2)
21
22
23
    def vector_to_typewriter(vector, n_columns):
24
        return np.reshape(vector, (-1, n_columns))
25
26
```

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

Listing 3: Help module for handling patterns.