```
"""Script for recognizing digits."""
1
2
    import numpy as np
3
    from hopfield import DeterministicHopfieldNetwork
    import pattern_utilities as utils
    def main():
        # Change get_pattern1 to get_pattern2 and get_pattern3 for questions
        # 2 and 3.
10
        pattern_to_feed = get_pattern1()
11
        n_neurons = pattern_to_feed.size
12
        stored_patterns = get_stored_patterns()
13
        n_{epochs} = int(1e3)
14
15
        network = DeterministicHopfieldNetwork()
16
        network.set_diagonal_weights_rule("zero")
17
        network.set_patterns(stored_patterns)
18
        network.generate_weights()
19
20
        updated_pattern = pattern_to_feed.copy()
21
        for epoch in range(n_epochs):
22
            # Update the pattern in typewriter order (since pattern is
23
            # flattened, just go in normal element-order).
24
            for neuron in range(n_neurons):
25
                updated_pattern = network.update_neuron(updated_pattern, neuron)
26
27
        matching_pattern_index = utils.get_index_of_equal_pattern(
28
            updated_pattern, stored_patterns)
29
        inverted_stored_patterns = -1 * stored_patterns
30
        matching_inverted_pattern_index = utils.get_index_of_equal_pattern(
31
            updated_pattern, inverted_stored_patterns)
32
        if matching_pattern_index >= 0:
33
            matching_pattern = stored_patterns[matching_pattern_index, :]
34
            matching_pattern_index += 1
35
        elif matching_inverted_pattern_index >= 0:
36
            matching_pattern = inverted_stored_patterns[
37
                matching_inverted_pattern_index, :]
            matching_pattern_index = -1 * matching_inverted_pattern_index
39
            matching\_pattern\_index -= 1
40
        else:
41
            matching_pattern_index = 6
42
            matching_pattern = np.array([])
43
        print("Converged to pattern {}".format(matching_pattern_index))
45
46
        print("Actual pattern: ")
47
```

```
utils.print_typewriter_pattern(updated_pattern, n_columns=10)
48
49
50
    def get_stored_patterns():
51
        x1 = np.array(
52
                 [ [-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,
                  \rightarrow -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,
                  \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,
                  \rightarrow 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,
                  \hookrightarrow 1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, -1, -1, -1, -1, -1, -1]
                 ).flatten();
54
55
        x2 = np.array(
56
                 [ [-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,
                  \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,
                  \hookrightarrow 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], [
                  \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,
                 \rightarrow 1, 1, -1, -1, -1]
                 ).flatten();
58
59
        x3 = np.array(
60
                 \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, -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, 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, -1,
                  \leftarrow -1],[ 1, 1, 1, 1, 1, 1, 1, 1, -1, -1],[ 1, 1, 1, 1, 1, 1, 1, 1, 1,
                  \hookrightarrow -1, -1]]
                 ).flatten();
62
63
        x4 = np.array(
```

```
[ [ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1], [ -1, -1, 1, 1, 1, 1, 1, 1, 1, 1, 1]
65
                 \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, -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,
                 \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, -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,
                 \hookrightarrow 1, 1, -1, -1]
                ).flatten();
66
67
        x5 = np.array(
68
                [ [-1, 1, 1, -1, -1, -1, -1, 1, 1, -1], [-1, 1, 1, -1, -1, -1, -1, -1, 1, ]
69
                 \hookrightarrow 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,
                 \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, 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], [
                 \rightarrow -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();
70
71
        return np.vstack((x1, x2, x3, x4, x5))
72
73
74
   def get_pattern1():
75
        return np.array(
76
                [[1, 1, 1, -1, -1, -1, -1, 1, 1, 1], [-1, -1, -1, 1, 1, 1, 1, -1, -1,
77
                 \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, -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, -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, -1],
                    \rightarrow -1, -1]]
                ).flatten()
78
79
80
   def get_pattern2():
81
        return np.array(
```

```
[[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1], [1, 1, 1, -1, -1, -1, -1, 1, 1, 1],
83
                \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,
                \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,
                   \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,
                \hookrightarrow 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, -1, -1, -1, -1, -1, -1, -1]
               ).flatten()
84
85
86
87
   def get_pattern3():
       return np.array(
88
               [[-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1,
89
                \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], [-1, -1,
                \rightarrow -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,
                   -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()
90
91
92
   if __name__ == "__main__":
93
       main()
94
```

```
"""Implementation of a Hopfield network using Hebb's rule."""
1
2
    from abc import ABC, abstractmethod
3
    import numpy as np
    from scipy import stats
5
    class HopfieldNetwork(ABC):
8
        """Abstract Hopfield net using Hebb's rule to compute weights."""
9
10
        def set_diagonal_weights_rule(self, diagonal_weights_rule):
11
            """Specify if the diagonal weights should be zero or not.
12
13
            diagonal_weights_rule must equal "zero" or "non-zero".
14
15
            if diagonal_weights_rule == "zero":
16
                self.diagonal_weights_equal_zero = True
17
            elif diagonal_weights_rule == "non-zero":
18
```

```
self.diagonal_weights_equal_zero = False
19
            else:
20
                raise KeyError(
                     diagonal_weights_rule
22
                     + " not a valid diagonal_weights_rule"
23
                     )
25
        def set_patterns(self, patterns):
26
             """Specify the stored patterns.
28
            patterns must be a 1D- or 2D-array where each row is a pattern.
29
30
            self.patterns = patterns
31
32
        def generate_weights(self):
33
             """Generate the weights for the network."""
34
            _, n_neurons = self.patterns.shape
35
            self.weights = np.zeros((n_neurons, n_neurons))
36
            for pattern in self.patterns:
37
                 self.weights += np.outer(pattern, pattern)
38
            self.weights /= n_neurons
39
40
            if self.diagonal_weights_equal_zero:
41
                np.fill_diagonal(self.weights, 0)
42
43
        def update_random_neuron(self, pattern):
44
            n_neurons = pattern.shape
45
            neuron_index = np.random.randint(n_neurons)
46
            return self.update_neuron(pattern, neuron_index)
47
48
        def update_neuron(self, pattern, neuron_index):
49
            """Returns updated pattern after update of neuron_index."""
50
            weights_i = self.weights[neuron_index, :]
51
            local_field = np.inner(weights_i, pattern)
52
            updated_neuron = self.get_state_of_local_field(local_field)
            updated_pattern = pattern.copy()
            updated_pattern[neuron_index] = updated_neuron
55
            return updated_pattern
57
        @abstractmethod
58
        def get_state_of_local_field(self, local_field):
59
            pass
60
61
62
    class DeterministicHopfieldNetwork(HopfieldNetwork):
63
        """Concrete Hopfield net with deterministic updating."""
64
65
        def get_state_of_local_field(self, local_field):
            """"Update rule when updating a neuron."""
67
```

```
return 1 if local_field >= 0 else -1
68
69
70
    class StochasticHopfieldNetwork(HopfieldNetwork):
71
        """Concrete Hopfield net with stochastic updating."""
72
73
        def get_state_of_local_field(self, local_field):
74
             """"Update rule when updating a neuron.
75
76
            Returns 1 with probability
77
            1 / (1+exp(-2*noise_parameter*local_field),
78
            else -1.
79
            .....
80
            p = 1/(1 + np.exp(-2*self.noise_parameter*local_field))
81
            rand = stats.bernoulli.rvs(p)
82
            return 1 if rand else -1
83
84
        def set_noise_parameter(self, noise_parameter):
85
            self.noise_parameter = noise_parameter
```

```
"""Functions for working with patterns.
1
2
    A pattern is defined as a 1D-array and several patterns are stored as
    a 2D-array, where each row corresponds to one pattern. Each element can
    take the values -1 or +1 only.
5
    import numpy as np
    from scipy import stats
9
10
11
    def generate_n_random_patterns(n_patterns, n_neurons):
12
        """Returns n_patterns random patterns, each of length n_neurons.
13
14
        If several patterns are generated, each row corresponds to one
15
        pattern.
16
        111111
17
        random_0s_and_1s = stats.bernoulli.rvs(0.5, size=(n_patterns, n_neurons))
18
        random_minus_1s_and_1s = 2*random_0s_and_1s - 1
19
        return random_minus_1s_and_1s
20
21
22
    def get_index_of_equal_pattern(pattern_to_match, patterns):
23
        """Returns the row index of patterns corresponding to the pattern
24
        that is equal to pattern_to_match.
25
26
        Returns 1 if no matching pattern is found.
27
28
```

```
for index, pattern in enumerate(patterns):
29
            n_different_neurons = get_n_different_neurons(pattern_to_match, pattern)
30
            if n_different_neurons == 0:
31
                return index
32
        return -1
33
35
    def get_n_different_neurons(pattern1, pattern2):
36
        return sum(pattern1 != pattern2)
37
38
39
    def vector_to_typewriter(vector, n_columns):
40
        """Returns 2D-array of vector.
41
42
        The first row in the returned array consists of the first n_columns
43
        elements in vector and so on.
        .....
45
        return np.reshape(vector, (-1, n_columns))
46
47
48
    def print_typewriter_pattern(pattern, n_columns):
49
        """Prints a pattern in a typewriter scheme."""
50
        print_pattern(vector_to_typewriter(pattern, n_columns))
51
52
53
    def print_pattern(pattern):
54
        np.set_printoptions(formatter={"float_kind": lambda x: "%.4f" % x})
55
        print(repr(pattern), sep=", ")
56
```