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import numpy as np
from PatternRecognition import DeterministicHopfieldNetwork
from PatternUtilities import generate_n_random_patterns, print_pattern
def main():
    n_bits = 120
    n_patterns_vector = [12,24,48,70,100,120]
    diagonal_weights_rule = "non-zero"
    n_{int} = int(1e5)
    one_step_error_probability = np.zeros(len(n_patterns_vector))
    for n_patterns_i, n_patterns in enumerate(n_patterns_vector):
        n_{-}errors = 0
        for i in range(n_iterations):
            network = DeterministicHopfieldNetwork()
            patterns = generate_n_random_patterns(n_patterns, n_bits)
            network . set_patterns ( patterns )
            network.set_diagonal_weights_rule(diagonal_weights_rule)
            network.generate_weights()
            pattern_to_feed_index = np.random.randint(n_patterns)
            original_pattern = patterns[pattern_to_feed_index, :]
            neuron_to_update_index = np.random.randint(n_bits)
            updated_pattern = network.update_neuron(original_pattern ,
                     neuron_to_update_index)
            updated_neuron = updated_pattern[neuron_to_update_index]
            original_neuron = original_pattern[neuron_to_update_index]
            updated_neuron = updated_pattern[neuron_to_update_index]
            if original_neuron != updated_neuron:
                n_{errors} += 1
        one_step_error_probability[n_patterns_i] = n_errors/n_iterations
    print_pattern(one_step_error_probability)
if __name__ == "__main__":
    main()
                Listing 1: Main method for computing the one-step error probability.
import numpy as np
from scipy import stats
from abc import ABC, abstractmethod
class HopfieldNetwork (ABC):
    def set_diagonal_weights_rule(self, diagonal_weights_rule):
        if diagonal_weights_rule == "zero":
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self.diagonal_weights_equal_zero = True
        elif diagonal_weights_rule == "non-zero":
            self.diagonal_weights_equal_zero = False
        else:
            raise KeyError(diagonal_weights_rule +
                    "_not_a_valid_diagonal_weights_rule")
    def set_patterns(self, patterns):
        self.patterns = patterns
    def generate_weights(self):
        _{-}, n_{-}bits = self.patterns.shape
        self.weights = np.zeros((n_bits, n_bits))
        for pattern in self.patterns:
            self.weights += np.outer(pattern, pattern)
        self.weights /= n_bits
        if self.diagonal_weights_equal_zero:
            np.fill_diagonal(self.weights, 0)
    def asynchronous_update(self, pattern, n_updates):
        updated_pattern = pattern.copy()
        for i in range(n_updates):
            updated_pattern = self.update_random_neuron(updated_pattern)
        return updated_pattern
    def update_random_neuron(self, pattern):
        n_bits = pattern.shape
        neuron_index = np.random.randint(n_bits)
        return self.update_neuron(pattern, neuron_index)
    def update_neuron(self, pattern, neuron_index):
        weights_i = self.weights[neuron_index, :]
        local_field = np.inner(weights_i, pattern)
        updated_bit = self.get_state_of_local_field(local_field)
        updated_pattern = pattern.copy()
        updated_pattern[neuron_index] = updated_bit
        return updated_pattern
    @abstractmethod
    def get_state_of_local_field(self, local_field):
        pass
class DeterministicHopfieldNetwork (HopfieldNetwork):
    def get_state_of_local_field(self, local_field):
        return sign_zero_returns_one(local_field)
class StochasticHopfieldNetwork (HopfieldNetwork):
    def get_state_of_local_field(self, local_field):
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p = 1/(1 + np.exp(-2*self.noise_parameter*local_field))
        rand = stats.bernoulli.rvs(p)
        return 1 if rand else -1
    def set_noise_parameter(self, noise_parameter):
        self.noise_parameter = noise_parameter
def sign_zero_returns_one(value):
    return 1 if value >= 0 else -1
                      Listing 2: Classes for creating Hopfield Networks.
import numpy as np
from scipy import stats
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
def get_index_of_equal_pattern(pattern_to_match, patterns):
    for index, pattern in enumerate(patterns):
        n_different_bits = get_n_different_bits(pattern_to_match, pattern)
        if n_different_bits == 0:
            return index
    return -1
def get_n_different_bits(pattern1, pattern2):
    return sum(pattern1 != pattern2)
def vector_to_typewriter(vector, n_columns):
    return np.reshape(vector, (-1, n_columns))
def print_typewriter_pattern(pattern, n_columns):
    print_pattern(vector_to_typewriter(pattern, n_columns))
def print_pattern(pattern):
    np. set_printoptions (formatter={"float_kind": lambda x: "%.4f" % x})
    print(repr(pattern), sep=", ")
def plot_pattern(pattern):
    plt.imshow(pattern, cmap="Greys")
    plt.tick_params(
            axis="both",
            which="both",
            bottom=False,
            top=False,
            left=False,
            labelbottom=False,
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labelleft=False)

Listing 3: Help methods for handling patterns.