Homework 1

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1 Boolean functions

dimensions	samples	separable functions
2	16	14
3	256	104
4	10000	276
5	10000	0

Table 1: Number of linearly separable functions of a n-dimensional Boolean function under the constraint samples $\leq 10^4$

For dimensions ≤ 3 the number of possible functions i less then 10^4 since number of unique Boolean functions k is given by: $k=2^{2^n}$ where n denotes the number of dimensions. Therefore the functions were sampled $\min(k, 10^4)$ times. From our results its evident that the number of linearly separable functions does not grow in proportion to the number of unique Boolean functions. I have not come across any general formula that describes the number of linearly separable functions given Boolean function of n-dimensions. The results given by the computer program are not surprising given the rapid divergence between a set of Boolean functions and the number of linearly separable functions of the same set from the get go. The fact that we restrict the number of sampled functions also makes it less probable that we sample a linearly separable one as we increase n. This is due to the discrepancy that arises in the partition of linearly separable functions of a set of unique n-dimensional Boolean functions.

```
t matplotlib.pyplot as plt
    binary_inputspace = np.array(list(itertools.product([0, 1], repeat=dimensions)))
    input_space = generate_binary_inputspace(number_of_dimensions)
    number_of_samples = min(number_of_samples, np.power(2, pow(2,number_of_dimensions)))
number_of_entries = input_space.shape[0]
    sampled functions = np.zeros((number_of_samples, number_of_entries, number_of_dimensions+1))
    sampled_outputs = np.zeros((number_of_samples, number_of_entries,1))
         if not any((random_output == sample).all() for sample in sampled_outputs):
    sampled_functions[j] = random_function
         last_element = sampled_functions.flatten()[-1]
def initiate weights(number of weights):
    distribution mean = 0
weights_2D = initiate_weights(2)
weights 3D = initiate weights(3)
weights_4D = initiate_weights(4)
weights_5D = initiate_weights(5)
threshold = 0
perceptron_3D = perceptron(weights_3D, threshold)
perceptron_4D = perceptron(weights_4D, threshold)
perceptron_5D = perceptron(weights_5D, threshold)
boolean_funtion_space_2D = boolean_function_sampler(nr_of_samples, 2)
boolean_funtion_space_3D = boolean_function_sampler(nr_of_samples, 3)
boolean_funtion_space_3D
                               boolean_function_sampler(nr_of_samples, 3)
nr_linearly_seperable_2D = perceptron_2D.evaluate_function_space(boolean_funtion_space_2D, weights_2D, threshold, eta)
nr_linearly_seperable_3D = perceptron_3D.evaluate_function_space(boolean_funtion_space_3D, weights_3D, threshold, eta)
nr_linearly_seperable_4D = perceptron_4D.evaluate_function_space(boolean_funtion_space_4D, weights_4D, threshold, eta)
nr_linearly_seperable_5D = perceptron_5D.evaluate_function_space(boolean_funtion_space_5D, weights_5D, threshold, eta)
print(nr_linearly_seperable_2D)
```

print (nr_linearly_seperable_4D)
print (nr_linearly_seperable_5D)

```
self.weights = weights
    size_input_space = len(inputs)-1
b = self.weights * inputs[0:size_input_space]
    if b == 0:
def evaluate_perceptron_list(self, inputs):
    outputs = np.zeros(len(inputs))
    for input in inputs:
         signum b = self.evaluate perceptron(input)
def update_weights(self, pattern, output, eta):
    size_input_space = len(pattern)-1
    error = pattern[-1] - output
dw = eta * error * inputs
def evaluate_function_space(self, function_space, weights, threshold, eta):
    linearly_separable_functions = 0
    for function in function space:
         self.train network(function, eta, eta, 20)
         outputs = self.evaluate_perceptron_list(function)
             linearly_separable_functions = linearly_separable_functions + 1
    return linearly_separable_functions
    error = (pattern[-1] - output)
d_theta = -eta * error
self.threshold = self.threshold + d_theta
             output = self.evaluate_perceptron(loc_input)
             self update thresholds (loc_input, output, eta)
```

```
matplotlib.pyplot as plt
  port numpy as np
training_patterns = [x1, x2, x3, x4, x5]
question_patterns = [q1, q2, q3]
    patterns = format patterns(patterns)
    weight_matrix = np.matmul(patterns.transpose(), patterns) * 1/NUMBER_OF_NEURONS
    return weight matrix
        patterns[pattern_index] = np.array(pattern).flatten()
    patterns = np.array(patterns)
    state_t = np.array(state_t).flatten()
state_t = state_t.transpose()
    b_t = np.matmul(weight_matrix, state_t)
b_t = np.where(b_t == 0, 1, b_t)
    state_t_plus_1 = np.sign(b_t)
    return state_t_plus_1
    plt.show()
        temporary_state = calculate_next_state(weight_matrix, new_state)
        temporary_state = temporary_state.reshape((16, 10))
        new_state[i] = temporary_state[i]
weight_matrix = hebb_weight_matrix(training_patterns)
    results.append(type_writer_update(weight_matrix, pattern))
```

```
-1, -1, -1, -1],[ -1,
                     -1, -1, 1, 1, 1, 1,
```