ONE-STEP ERROR PROBABILITY

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Initial setup

```
clear, clc
N = 120;
number_of_trials = 1e5;
p_array = [12,24,48,70,100,120];
```

With zero diagonals

```
error_probability_array_zero_diag = zeros(1,size(p_array,2));
for i = 1:size(p array,2)
    p = p array(i);
    errors = 0;
    for trial = 1:number of trials
        patterns = generate_patterns(p, N);
        weight matrix = generate weight matrix zero diag(patterns);
        pattern_number = randi([1 p]);
        neuron_number = randi([1 N]);
        old_neuron_value = patterns(neuron_number,pattern_number);
        new_neuron_value =
 signum(weight_matrix(neuron_number,:)*patterns(:,pattern_number));
        errors = errors + double(old_neuron_value ~=
 new neuron value);
    error_probability_array_zero_diag(i) = errors / number_of_trials;
end
disp("The one step error probabilities with zero diagonals are:")
disp(num2str(error_probability_array_zero_diag))
The one step error probabilities with zero diagonals are:
0.00035
           0.01147
                        0.05612
                                  0.09402
                                             0.13625
                                                            0.15764
```

Without zero diagonals

```
error_probability_array = zeros(1,size(p_array,2));
for i = 1:size(p_array,2)
    p = p_array(i);
    errors = 0;
    for trial = 1:number_of_trials
```

```
patterns = generate_patterns(p, N);
        weight matrix = generate weight matrix(patterns);
        pattern_number = randi([1 p]);
        neuron number = randi([1 N]);
        old_neuron_value = patterns(neuron_number,pattern_number);
        new neuron value =
 signum(weight_matrix(neuron_number,:)*patterns(:,pattern_number));
        errors = errors + double(old neuron value ~=
 new_neuron_value);
    end
    error_probability_array(i) = errors / number_of_trials;
end
disp("The one step error probabilities without zero diagonals are:")
disp(num2str(error_probability_array))
The one step error probabilities without zero diagonals are:
0.00022
           0.00304
                       0.01242
                                      0.019
                                                0.02119
                                                            0.02308
```

Functions

```
disp('')
function weight_matrix = generate_weight_matrix(patterns)
    weight_matrix = patterns*patterns';
end

function weight_matrix = generate_weight_matrix_zero_diag(patterns)
    weight_matrix = patterns*patterns'/size(patterns,1);
    weight_matrix = weight_matrix - diag(diag(weight_matrix));
end

function sign = signum(x)
    sign = 2*(x >= 0) - 1;
end

function patterns = generate_patterns(p, N)
    % Generates a p by N matrix with p N-bit patterns as columns
    patterns = 2.*randi([0 1],N, p) - 1;
end
```

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RECOGNISING DIGITS

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Initial setup, loads patterns

```
clear, clc
1, -1, -1, -1],[ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1],[ -1, 1, 1, 1, -1,
-1, 1, 1, 1, -1],[-1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[-1, 1, 1, 1,
-1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1,
1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1, 1,
1, 1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1,
1, 1, 1, -1, -1, 1, 1, 1, -1],[ -1, 1, 1, 1, -1, -1, 1, 1, 1, -1],
[-1, -1, 1, 1, 1, 1, 1, 1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, -1, -1,
-1], [-1, -1, -1, -1, -1, -1, -1, -1, -1, -1];
x2=[ [-1, -1, -1, 1, 1, 1, 1, -1, -1, -1], [-1, -1, -1, 1, 1, 1, 1, 1]
-1, -1, -1, [ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1, [ -1, -1, -1, 1,
1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[ -1, -1,
-1, 1, 1, 1, -1, -1, -1, -1, [ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],
-1, -1],[-1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[-1, -1, -1, 1, 1, 1,
1, -1, -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1,
1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1],[ -1, -1,
-1, 1, 1, 1, 1, -1, -1, -1],[ -1, -1, -1, 1, 1, 1, 1, -1, -1, -1] ];
x3=[ [ 1, 1, 1, 1, 1, 1, 1, 1, -1, -1],[ 1, 1, 1, 1, 1, 1, 1, 1, -1,
-1],[-1, -1, -1, -1, -1, 1, 1, 1, -1, -1],[-1, -1, -1, -1, -1, 1,
-1, -1, 1, 1, -1, -1],[-1, -1, -1, -1, -1, 1, 1, 1, -1, -1],[1,
1, 1, 1, 1, 1, 1, 1, -1, -1],[ 1, 1, 1, 1, 1, 1, 1, 1, -1, -1],[ 1,
1, 1, -1, -1, -1, -1, -1, -1, -1, [ 1, 1, 1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1]
-1],[ 1, 1, 1, -1, -1, -1, -1, -1, -1, -1],[ 1, 1, 1, -1, -1, -1, -1,
-1, -1, -1],[ 1, 1, 1, -1, -1, -1, -1, -1, -1],[ 1, 1, 1, 1, 1,
1, 1, 1, -1, -1],[ 1, 1, 1, 1, 1, 1, 1, -1, -1] ];
x4=[ [ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1],[ -1, -1, 1, 1, 1, 1, 1, 1, 1,
-1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, 1, 1, 1, -1], [-1, -1, -1,
-1, -1, -1, 1, 1, 1, -1],[ -1, -1, -1, -1, -1, 1, 1, 1, -1],[ -1,
-1, 1, 1, 1, 1, 1, -1, -1],[-1, -1, 1, 1, 1, 1, 1, -1, -1],
-1, 1, 1, 1, -1],[ -1, -1, -1, -1, -1, 1, 1, 1, -1],[ -1, -1, 1,
1, 1, 1, 1, 1, -1],[ -1, -1, 1, 1, 1, 1, 1, 1, -1, -1] ];
```

```
x5=[[-1, 1, 1, -1, -1, -1, -1, 1, 1, -1], [-1, 1, 1, -1, -1, -1, -1, -1]
1, 1, -1],[ -1, 1, 1, -1, -1, -1, 1, 1, -1],[ -1, 1, 1, -1, -1,
-1, -1, 1, 1, -1], [-1, 1, 1, -1, -1, -1, 1, 1, -1], [-1, 1, 1,
-1, -1, -1, -1, 1, 1, -1, [-1, 1, 1, -1, -1, -1, -1, 1, 1, -1, [-1,
1, 1, 1, 1, 1, 1, 1, 1, -1],[ -1, 1, 1, 1, 1, 1, 1, 1, 1, -1],[ -1,
-1, -1, -1, -1, -1, -1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, -1, 1, 1,
-1],[-1, -1, -1, -1, -1, -1, -1, 1, 1, -1],[-1, -1, -1, -1, -1, -1,
-1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, 1, 1, -1], [-1, -1, -1,
-1, -1, -1, -1, 1, 1, -1],[ -1, -1, -1, -1, -1, -1, 1, 1, -1] ];
stored_patterns = [ x1', x2', x3', x4', x5' ];
fed pattern 1 = [[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1], [1, 1, 1, -1, -1, -1, -1]
-1, 1, 1, 1], [1, 1, -1, -1, -1, -1, -1, -1, 1], [1, -1, -1, -1,
1, 1, -1, -1, -1, 1], [1, -1, -1, -1, 1, 1, -1, -1, 1], [1, -1,
-1], [-1, 1, 1, 1, -1, -1, 1, 1, 1, -1], [-1, 1, 1, 1, -1, -1, 1, 1, 1]
1, -1], [-1, 1, 1, 1, -1, -1, 1, 1, 1, -1], [-1, 1, 1, 1, -1, -1, 1,
1, 1, -1], [-1, -1, 1, 1, 1, 1, 1, -1, -1], [-1, -1, -1, 1, 1, 1,
1, -1, -1, -1], [-1, -1, -1, -1, -1, -1, -1, -1, -1]]';
fed_pattern_2 = [[1, 1, -1, -1, -1, -1, -1, 1, 1], [-1, -1, 1, 1]
-1, -1, -1, -1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, 1, 1, 1, 1, -1],
[-1, -1, -1, -1, -1, -1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, 1,
1, 1, -1], [-1, -1, 1, 1, 1, 1, 1, -1, -1], [-1, -1, 1, 1, 1, 1,
-1, -1, -1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, 1, 1, 1, -1], [-1,
-1, -1, -1, -1, -1, 1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, 1, 1, 1,
-1], [-1, -1, 1, 1, 1, 1, 1, 1, -1], [-1, -1, 1, 1, 1, 1, 1, 1, 1,
-1, -1]]';
fed_pattern_3 = [[1, -1, -1, 1, 1, 1, 1, -1, -1, 1], [-1, 1, 1, -1, ]
-1, -1, -1, 1, 1, -1], [-1, 1, 1, -1, -1, -1, -1, 1, 1, -1], [-1, 1,
1, -1, -1, -1, -1, 1, 1, -1, [-1, 1, 1, -1, -1, -1, -1, 1, 1, -1],
1, -1], [-1, 1, 1, 1, 1, 1, 1, 1, -1], [-1, 1, 1, 1, 1, 1, 1, 1, 1,
-1, -1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, 1, 1, -1], [-1, -1,
-1, -1, -1, -1, 1, 1, -1, [-1, -1, -1, -1, -1, -1, 1, 1,
-1], [-1, -1, -1, -1, -1, -1, -1, 1, 1, -1], [-1, -1, -1, -1, -1, -1, -1,
-1, 1, 1, -1]]';
```

Calculate and classify steady states of patterns

```
clc
weight_matrix = generate_weight_matrix_zero_diag(stored_patterns);
steady_state_pattern_1 =
  update_pattern_until_steady_state(weight_matrix, fed_pattern_1);
```

```
col_index_pattern_1 = classify_pattern(steady_state_pattern_1,
 stored patterns);
disp(strcat("The steady state of pattern 1 corresponds to digit index
 ", int2str(col_index_pattern_1), "."))
steady state pattern 2 =
 update_pattern_until_steady_state(weight_matrix, fed_pattern_2);
col_index_pattern_2 = classify_pattern(steady_state_pattern_2,
 stored patterns);
disp(strcat("The steady state of pattern 2 corresponds to digit index
 ", int2str(col_index_pattern_2), "."))
steady state pattern 3 =
 update_pattern_until_steady_state(weight_matrix, fed_pattern_3);
col_index_pattern_3 = classify_pattern(steady_state_pattern_3,
 stored_patterns);
disp(strcat("The steady state of pattern 3 corresponds to digit index
 ", int2str(col_index_pattern_3), "."))
The steady state of pattern 1 corresponds to digit index 2.
The steady state of pattern 2 corresponds to digit index 4.
The steady state of pattern 3 corresponds to digit index 5.
```

Functions

```
disp('')
function col_index = classify_pattern(pattern, stored_patterns)
    for i = 1:size(stored_patterns, 1)
        if isequal(pattern, stored patterns(:,i))
            col index = i;
            break
        elseif isequal(-pattern, stored_patterns(:,i))
            col index = -i;
            break
        end
    end
end
function steady_state_pattern =
 update_pattern_until_steady_state(weight_matrix, fed_pattern)
    old_pattern = fed_pattern;
   new_pattern = update_state_asynchronously(weight_matrix,
 old_pattern);
    while ~isequal(old_pattern, new_pattern)
       old pattern = new pattern;
       new_pattern = update_state_asynchronously(weight_matrix,
 old pattern);
    end
    steady_state_pattern = new_pattern;
end
function state = update_state_asynchronously(weight_matrix, state)
    for neuron_number = 1:size(state, 1)
```

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```
state(neuron_number) =
signum(weight_matrix(neuron_number,:)*state);
end
end

function weight_matrix =
generate_weight_matrix_zero_diag(stored_patterns)
    weight_matrix = stored_patterns*stored_patterns'/
size(stored_patterns,1);
    weight_matrix = weight_matrix - diag(diag(weight_matrix));
end

function sign = signum(x)
    sign = 2*(x >= 0) - 1;
end
```

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STOCHASTIC HOPFIELD NETWORK

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Initial setup

```
clear, clc
beta = 2;
N = 200;
T = 2e5;
number_of_updates = T/N;
number_of_tests = 100;
```

For p = 7

```
p = 7;
m=zeros(1,number_of_tests);
for n_test = 1:number_of_tests
    patterns = generate_patterns(p,N);
    weight_matrix = generate_weight_matrix_zero_diag(patterns);
    initial_pattern = patterns(:,1);
    state = initial_pattern;
    sum_t = 0;
    for t = 1:number_of_updates
        for neuron_number = 1:N
            state = update_neuron_asynchronously(weight_matrix, state,
 neuron number);
            sum_t = sum_t + 1/N*sum(state.*initial_pattern);
        end
    end
    m(n_{test}) = sum_t/T;
end
m_average = mean(m);
disp(strcat("The average m_1 for p = 7 is ", num2str(m_average)))
The average m_1 for p = 7 is 0.85115
```

For p = 45

```
p = 45;
m=zeros(1,number_of_tests);
for n_test = 1:number_of_tests
    patterns = generate_patterns(p,N);
```

STOCHASTIC HOPFIELD NETWORK

```
weight_matrix = generate_weight_matrix_zero_diag(patterns);
    initial pattern = patterns(:,1);
    state = initial pattern;
    sum t = 0;
    for t = 1:number_of_updates
        for neuron number = 1:N
            state = update_neuron_asynchronously(weight_matrix, state,
 neuron number);
            sum_t = sum_t + 1/N*sum(state.*initial_pattern);
        end
    end
    m(n_{test}) = sum_t/T;
end
m_average = mean(m);
disp(strcat("The average m_1 for p = 45 is ", num2str(m_average)))
The average m_1 for p = 45 is 0.11416
```

Functions

```
disp('')
function g = g(b)
   beta = 2;
   g = 1./(1+exp(-2*beta*b));
end
function neuron_value = generate_neuron_value(b)
   x = rand;
   if x < g(b)
        neuron_value = 1;
    else
        neuron_value = -1;
    end
end
function weight_matrix = generate_weight_matrix_zero_diag(patterns)
    weight_matrix = patterns*patterns'/size(patterns,1);
   weight_matrix = weight_matrix - diag(diag(weight_matrix));
end
function patterns = generate_patterns(p, N)
    % Generates a p by N matrix with p N-bit patterns as columns
   patterns = 2.*randi([0\ 1],N,p) - 1;
end
function state = update_neuron_asynchronously(weight_matrix, state,
neuron_number)
   b = weight_matrix(neuron_number,:)*state;
    state(neuron_number) = generate_neuron_value(b);
end
```

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