

# Homework 1: One-step error probability

## Problem description:

Write a computer program implementing asynchronous deterministic dynamics [Eq. (1.9) in the course book] for a Hopfield network. Use Hebb's rule with  $w_{ii} = 0$  [Eq. (2.26) in the course book]. Generate and store  $p = [12, 24, 48, 70, 100, 120]$  random patterns with  $N = 120$  bits. Each bit is either +1 or -1 with probability  $\frac{1}{2}$ .

For each value of  $p$ , estimate the one-step error probability  $P_{\text{error}}^{t=1}$  based on  $10^5$  independent trials.

Here, one trial means that you generate and store a set of  $p$  random patterns, feed one of them, and perform one asynchronous update of a single randomly chosen neuron. If in some trials you encounter  $\text{sgn}(0)$ , simply set  $\text{sgn}(0) = 1$ .

List below the values of  $P_{\text{error}}^{t=1}$  that you obtained in the following form:  $[p_1, p_2, \dots, p_6]$ , where  $p_n$  is the value of  $P_{\text{error}}^{t=1}$  for the  $n$ -th value of  $p$  from the list above. Give four decimal places for each  $p_n$ .

## Hebb's rule

$$w_{ij} = \frac{1}{N} \sum_{\mu=1}^p x_i^{(\mu)} x_j^{(\mu)} \quad \text{for } i \neq j, \quad w_{ii} = 0, \quad \text{and} \quad \theta_i = 0. \quad (2.26)$$

## Asynchronous deterministic updates

$$s_i(t+1) = \begin{cases} g(\sum_j w_{mj} s_j(t) - \theta_m) & \text{for } i = m, \\ s_i(t) & \text{otherwise.} \end{cases} \quad (1.9)$$

## Initial conditions:

```
clear
p = [12 24 48 70 100 120]; % 24 48 70 100 120 amount of stored patterns
N = 120; % amount of neurons
nTrials = 10e5; % numver of trials
```

## Q1: Hebb's rule setting the diagonal weights to zero

```
% loop through all given numbers
diagonal_zero = true;
P_error_list = [];
for i = 1:length(p)
    P_error_list(end+1) = P_err_cal(p(i), N, nTrials, diagonal_zero);
end
```

```
disp("P_error with diagonal elements are 0: ")
```

```
P_error with diagonal elements are 0:
```

```
out=['[' sprintf(' %.4f, ', P_error_list(1:end-1)) num2str(P_error_list(end)) ']'];
```

```
disp(out)
```

```
[ 0.0005, 0.0114, 0.0560, 0.0947, 0.1363, 0.15848]
```

## Q2: Hebb's rule without setting the diagonal weights to zero

```
diagonal_zero = false;  
P_error_list_dz = [];  
for i = 1:length(p)  
    P_error_list_dz(end+1) = P_err_cal(p(i), N, nTrials, diagonal_zero);  
end
```

```
disp("P_error with diagonal elements not set to 0: ")
```

```
P_error with diagonal elements not set to 0:
```

```
out=[' ' sprintf(' %.4f, ', P_error_list_dz(1:end-1) ) num2str(P_error_list_dz(end)) ' '];  
disp(out)
```

```
[ 0.0001, 0.0031, 0.0124, 0.0179, 0.0222, 0.022872]
```

## One-step error calculation function

```
function P_error = P_err_cal(p, N, nTrials, diagonal_zero)  
    error = 0;  
    for trial = 1:nTrials  
        % Generate random patterns 120*p  
        patterns_generated = 2 * randi([0, 1], N, p) - 1;  
        % Initialize weight matrix NxN with Hebb's rule  
        W = zeros(N);  
        for i = 1:p  
            W = W + patterns_generated(:,i)*patterns_generated(:,i)'/N;  
        end  
  
        % set diagonal elements to 0  
        if diagonal_zero == true  
            for i = 1:N  
                W(i,i) = 0;  
            end  
        end  
  
        % choose a random pattern in the stored patterns 120*1  
        pr = patterns_generated(:,randi(p));  
        % choose a bit  
        nr = randi(N);  
  
        b_nr = W(nr,:)*pr;  
        if b_nr == 0  
            x_nr = 1;  
        else
```

```
        x_nr = sign(b_nr);  
    end  
  
    % check if the updated bit equals the origin bit  
    if x_nr ~= pr(nr)  
        error = error + 1;  
    end  
end  
P_error = error/nTrials;  
end
```