

Department of Computer

Science		SCORE/GRADE			
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CODE	CSC167	COURSE	Neural Network	DAT	

ASSIGNMENT TITLE/PAGE

NO.

Assignment 3

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NO. WORKING/STEPS MARK S

self organizing map

Design

I have designed 6 process for SOM. There are som(main process), initialization, competitive_process, cooperative_process, adaptive_process, mapping_process.

Som: main process for som, call each process function one by one in each iteration.

Initialization: For initializing input variable and parameters

Competitive process: Obtain the winning neuron of each sample(input)

Cooperative_process: Calculate distance between winning neuron and each neuron and

topological neighborhood function

Adaptive_process: Calculate change of weight and adjust weight

Mapping process: Show the context map after all iteration. step 1, find the neuron of strongest

responses sample. step 2, fill the unoccupied neuron

parameters of my code

sig0 = 5;

 $t1 = 1000/(\log(\text{sig}0));$

r0 = 0.1;

t2 = 1000;

context map

Eagle	Eagle	Eagle	Cat	Cat	Cat	Cat	Fox	Fox	Dog
Eagle	Eagle	Eagle	Cat	Cat	Cat	Cat	Fox	Wolf	Wolf
Eagle	Eagle	Owl	Owl	Owl	Cat	Cat	Wolf	Wolf	Wolf
Owl	Owl	Owl	Owl	Owl	Owl	Lion	Lion	Wolf	Wolf
Dove	Owl	Owl	Hawk	Owl	Owl	Lion	Lion	Lion	Tiger
Dove	Dove	Owl	Owl	Owl	Owl	Lion	Lion	Lion	Tiger
Dove	Dove	Owl	Owl	Owl	Lion	Lion	Lion	Lion	Tiger
Dove	Dove	Duck	Goose	Goose	Horse	Horse	Horse	Horse	Horse
Hen	Hen	Duck	Goose	Goose	Goose	Horse	Zebra	Horse	Cow
Hen	Duck	Duck	Goose	Goose	Goose	Horse	Horse	Horse	Cow

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Source code
som.m
%Main process of Self-organizing map
clear all
close all
% initialization
[sig0,t1,r0,t2,x,sam\ name,x\ nor,w,n] = initialization();
for n1=1:n % for n interations; n1: current iterations; n: the number of iterations
  %show the process precentage
  if mod(n1.10) == 0
   floor((n1/n)*100)
  end
  for m=1:16 % m for each animal's type (each input)
     % competitive process
     [i]=competitive process(m,x nor,w);
     % cooperative process & adaptive process
     [w,d,h]=cooperative process(i,sig0,t1,w,n1,r0,t2,x nor,m);
  end
end
%mapping process
[map.result]=mapping process(x nor,w,sam name);
initialization.m
function [sig0,t1,r0,t2,x,sam\ name,x\ nor,w,n] = initialization()
% paramter
sig0 = 5; % exponential decay at the initiation; should be the radious of map size
t1 = 1000/(\log(\text{sig}0)); % time constant, relative to sig0
r0 = 0.1; % learning rate; 0.1 is recommended
t2 = 1000; % tor 2; 1000 is recommended
% x: input data
x=[1\ 0\ 0\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 1\ 0; \%Dove
  1 0 0 1 0 0 0 0 1 0 0 0 0; %Hen
  1 0 0 1 0 0 0 0 1 0 0 0 1; %Duck
  1 0 0 1 0 0 0 0 1 0 0 1 1; %Goose
  1001000011010; %Owl
  1 0 0 1 0 0 0 0 1 1 0 1 0; %Hawk
  0 1 0 1 0 0 0 0 1 1 0 1 0; %Eagle
 0 1 0 0 1 1 0 0 0 1 0 0 0; %Fox
  0 1 0 0 1 1 0 0 0 0 1 0 0; %Dog
 0 1 0 0 1 1 0 1 0 1 1 0 0; %Wolf
  1 0 0 0 1 1 0 0 0 1 0 0 0; %Cat
  0 0 1 0 1 1 0 0 0 1 1 0 0; %Tiger
 0 0 1 0 1 1 0 1 0 1 1 0 0; %Lion
 0 0 1 0 1 1 1 1 0 0 1 0 0; %Horse
 0 0 1 0 1 1 1 1 0 0 1 0 0; %Zebra
 0 0 1 0 1 1 1 0 0 0 0 0 0; %Cow
];
sam_name={'Dove';'Hen';'Duck';'Goose';'Owl';'Hawk';'Eagle';'Fox';'Dog';'Wolf';'Cat';'Tiger';'Li
on';'Horse';'Zebra';'Cow';};
% x nor: normolization of input data
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x \text{ nor}=[];
for m=1:16
  x \text{ row}=0;
  for n=1:13
    x row = x row + (x(m,n))^2;
  end
  for n=1:13
    x \text{ nor(m,n)} = x(m,n)/x \text{ row}^{(1/2)};
  end
end
% w: synaptic-weight vector of each neuron
w = randn(100, 13);
% n: the number of iterations
n=1000;
% up to now, initialization finish
end
competitive process.m
function [i]=competitive process(m,x nor,w)
 % i: inedx which the x nor and w is most similiar with each other; winning neuron
 temp=[];
 for n=1:100
  temp(n)=0;
  for k=1:13
   temp(n) = temp(n) + abs(x nor(m,k)-w(n,k));
  end
 end
 [M,Min i] = min(temp);
 i=Min i;
end
cooperative process.m
function [w,d,h]=cooperative process(i,sig0,t1,w,n1,r0,t2,x nor,m)
 % i: winning neuron
 % j: each neuron of 10*10
 % ai: x-axis of i
 % bi: y-axis of i
 % aj: x-axis of j
 % bj: y-axis of j
 % d: distance between j and i
 % n: the number of iterations
 % n1: current iterations
 % n1: the number of iterations in current process
 % sig(n): exponential decay
 % h: topological neighborhood function
 % r0: adaptive process usage
 % t2: adaptive process usage
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% w: adaptive process usage
 % x nor: adaptive process usage
 % m: sample number m; adaptive process usage
 ai = mod(i-1,10);
 bi=floor((i-1)/10);
 for j=1:100
  aj = mod(j-1,10);
  bj= floor((j-1)/10);
  d(i)=((ai-ai)^2+(bi-bi)^2)^(1/2);
  sig(n1)=sig0*exp(-n1/t1);
  h(j)=\exp((-(d(j))^2)/(2*(sig(n1))^2));
  % adaptcive process
  [w(j,:)]=adaptive process(r0,t2,i,h(j),w(j,:),x nor,m,n1);
end
adaptive process.m
function [w]=adaptive process(r0,t2,i,h,w,x nor,m,n1)
 % deltaw: change of the weight vector in neuron j
 % r(n): learning rate at n iterations
 % n1: current iterations
 r(n1)=r0*exp(-n1/t2);
 deltaw=r(n1)*h.*(x nor(m,:)-w);
 w=w+deltaw;
end
mapping process.m
function [map,result]=mapping process(x nor,w,sam name)
 % map: map of animal in number formula
 % result: map of animal in name formula
 % X: x-axis of lattice map
 % Y: y-axis of lattice map
 %plot the grid
 [X,Y]=meshgrid(0:10);
 figure; hold on;
 plot(X,Y,'k');
 plot(Y,X,'k');
 axis off;
 grid on;
 % step 1, find the neuron of strongest responses sample
 for m=1:16
   for j=1:100
```

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temp(j)=0;
    for k=1:13
       temp(j) = temp(j) + abs(x nor(m,k)-w(j,k));
    end
   end
   [M,temp win j] = min(temp);
   result(temp win j,:)=sam name(m,:);
   occupied(temp win j)=1; % 1 mean that neuron is occupied
 end
 % step 2, fill the unoccupied neuron
 temp=[];
 for j=1:100
  if (occupied(j) \sim = 1)
    occupied(i)=1;
    for m=1:16
       temp(m)=0;
       for k=1:13
         temp(m) = temp(m) + abs(x nor(m,k)-w(j,k));
       end
    end
    [M,map(j)] = min(temp);
    result(j,:)=sam name(map(j),:);
  text(mod(j-1,10)+0.1,floor((j-1)/10)+0.5,result(j,:)); % show the content of map in lattice
 end
 % save the result figure
 saveas(gcf,'Result som.png')
end
```