

Project #2(a): Multilayer Perceptron

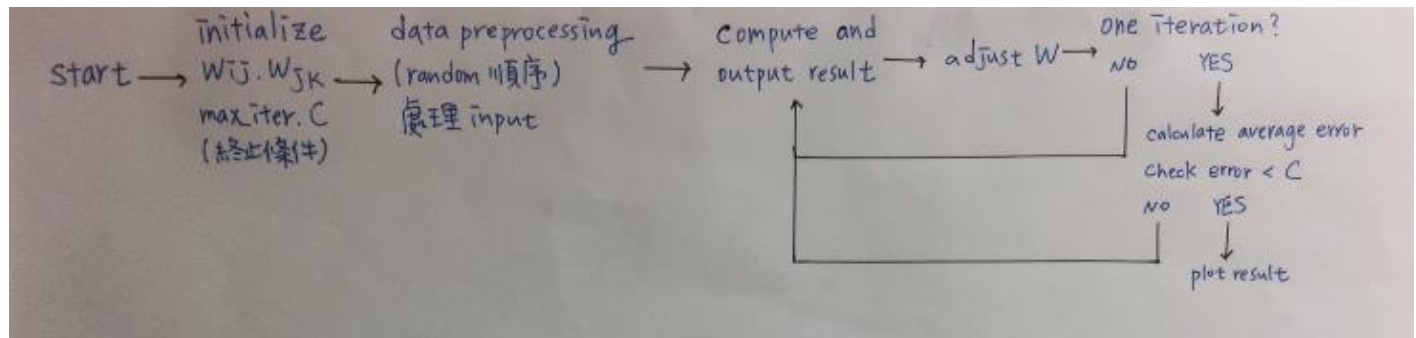
0516222 許芳瑀

Q1:

做法:

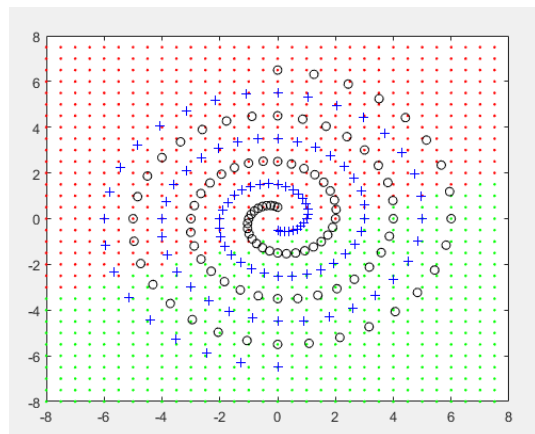
這題我使用了兩層 hidden layer，第一層 20 個 node，第二層 10 個 node，最一開始處理資料時，我先將 x, y 存到 oi (加上常數 1)，並且將資料順序 random，讓同種類的點分散，開始 train 之後，便是接受 input，其中採用的 active function 是 sigmoid function，經過跟 weight 內積之後，output 的輸出是 1 或 0 (因為只有要分兩類)，然後調整 weight，一個 iteration 之後計算 error，直到我指定的最大次數或是 e 小於指定的數字。

Flowchart:

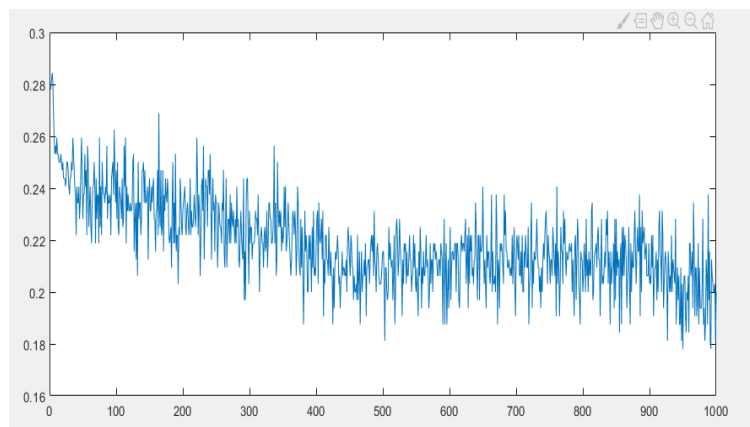


結果:

decision region



average error vs. iteration



討論:

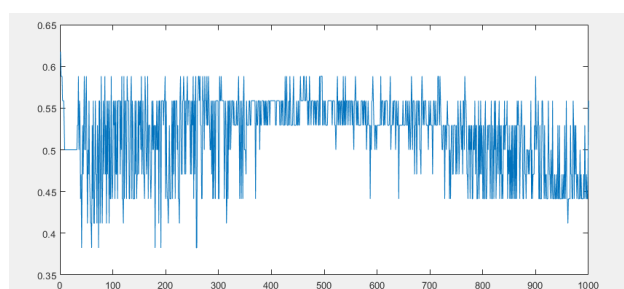
How to determine the hidden node number in each problem?

這題花了蠻久的時間，一開始我只有用一層 layer，50 個 node，發現效果蠻差的，所以多設計一層，以及把 iteration 的次數也加高到 10000，但效果還是不好。

Any experiment

我只用了八成的資料 training，剩下的兩成用於 test，再每次 iteration 之後都算一次答案的正確性。

Accuracy



Code:

```

i=[0:96]
theta_i = i * pi / 16;
theta_j = i * pi / 16;
r_i = 6.5 .* (104 - i) / 104;
r_j = 6.5 .* (104 - i) / 104;
N=250;
theta1 = linspace(-180,180, N)*pi/360;
r = 8 ;
x1 = r_i .* sin(theta_i);
y1 = r_i .* cos(theta_i);
x2 = -x1;
y2 = -y1;
% set ans 0/1 class
oi = [x1.' y1.' zeros(97,1)+1 ; x2.' y2.' zeros(97,1)+1 ];
ans = [zeros(97,1) ;zeros(97,1)+1 ];
% rearrange
total=194;
train_total=160;
test_total = 34;
r = randperm(total);
oi(:,1) = oi(r,1);
oi(:,2) = oi(r,2);
ans = ans(r);

test=oi([train_total+1 :total], :);
test_ans = ans([train_total+1 :total],:);

% set initial value
num_j = 20;
num_k = 10;
wij = rand(3,20);
wjk = rand(21,10);
wkl = rand(11, 1);
i_wij = wij;
i_wjk = wjk;
i_wkl = wkl;
rate = 0.001;
C = 0.01;
max_iter = 15000;
iter = 1;
sj = [];
sk = [];

```

```

sl = [];
oj = [];
ok = [];
ol = [];

% last result
accuracy=[];
e=[];
% regulize
oi(:,1) = (oi(:,1)-mean(oi(:,1))) / std(oi(:,1));
oi(:,2) = (oi(:,2)-mean(oi(:,2))) / std(oi(:,2));
while iter<=max_iter
    disp(iter);
    error = 0;
    for i = 1:1:train_total
        % foward
        sj = oi(i,:)*wij;
        oj = [sj 1];
        sk = oj*wjk;
        ok= [sk 1];
        sl = ok*wkl;
        ol = round(sigmoid(sl,[1,0]));
        prev_wjk = wjk;
        prev_wkl = wkl;
        % update wkl
        f = sigmoid(sl,[1,0]);
        for kk=1:num_k+1
            wkl(kk,1) = wkl(kk,1) + rate * (ans(i)-ol)* f * (1-f)
            * ok(kk);
        end

        % update wjk
        for jj = 1:num_j+1
            for kk = 1:num_k
                f = sigmoid(sk(kk),[1,0]);
                fl = sigmoid(sl,[1,0]);
                sum = (ans(i,1) - ol) * fl * (1-fl) *
                prev_wkl(kk,1);
                wjk(jj,kk) = wjk(jj,kk) + rate* sum * f * (1-
                f) * oj(jj);
            end
        end
    end
end

```

```

% update wij
sum=0;
for ii = 1:3
    for jj = 1:num_j
        fj = sigmf(sj(jj),[1,0]);
        for kk=1:num_k
            fk=sigmf(sk(kk),[1 0]);
            sum = sum + (ans(i,1) - ol)*fk*(1-fk)*wjk(jj,kk);
        end
        wij(ii,jj) = wij(ii,jj) + rate* sum * fj * (1-fj) * oi(i,ii);
    end
end
error = error + (ans(i) - ol).^2/2;
end
e(iter) = error/train_total;
if error < C
    break;
end
acc=0;
for t=1:test_total
    sj = test(t,:)*wij;
    oj = [sj 1];

    sk = oj*wjk;
    ok= [sk 1];

    sl = ok*wkl;
    ol = round(sigmf(sl,[1,0]));
    if ol== test_ans(t)
        acc = acc+1;
    end
end
acc = acc/test_total;
accuracy(iter) = acc;
iter = iter + 1;
end
figure(1);
plot(e);
figure(2);
plot(accuracy);

```

% 畫圖

```

for ix=-16:16
    for iy=-16:16
        dx=0.5*(ix-1);
        dy=0.5*(iy-1);
        oi=[dx dy 1];

        sj = oi*wij;
        oj = [sj 1];
        sk = oj*wjk;
        ok= [sk 1];
        sl = ok*wkl;
        ol = round(sigmf(sl,[1,0]));

        if ol==1
            figure(3);
            axis([-8 8 -8 8]);
            plot(dx,dy,
                'g.',x1,y1,'ko',x2,y2,'b+');
            hold on;
        elseif ol==0
            figure(3);
            axis([-8 8 -8 8]);
            plot(dx,dy,
                'r.',x1,y1,'ko',x2,y2,'b+');
            hold on;
        end
    end
end
end

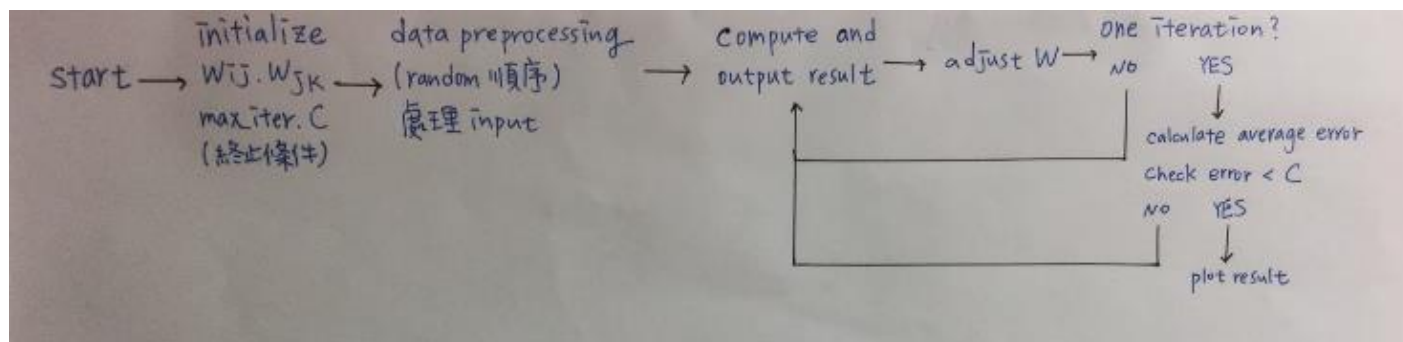
```

Q2:

做法:

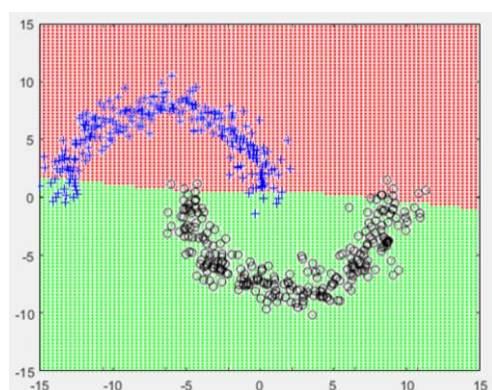
這題我使用了一層 hidden layer、四個 node，最一開始處理資料時，我先將 x y 存到 oi (加上常數 1)，並且將資料順序 random，讓同種類的點分散，開始 train 之後，便是接受 input，其中採用的 active function 是 sigmoid function，經過跟 weight 內積之後，output 的輸出是 1 或 0(因為只有要分兩類)，然後調整 weight，一個 iteration 之後計算 error，直到我指定的最大次數或是 e 小於指定的數字。

Flowchart:

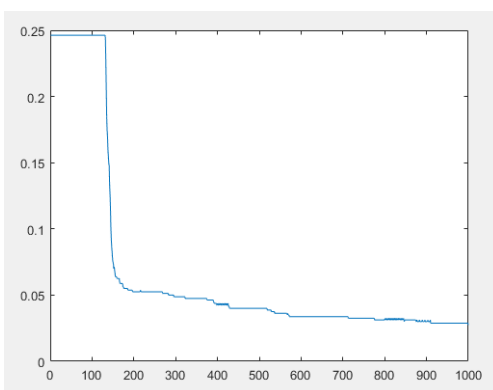


結果:

decision region



average error vs. iteration



討論:

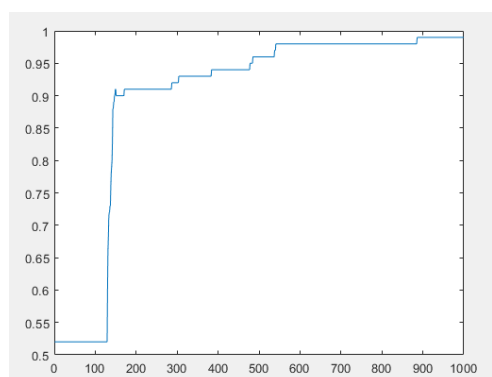
How to determine the hidden node number in each problem?

因為這題的分布看起來沒有很複雜，所以我決定用一層 layer 就好，然後一開始選擇了三個 node，發現結果蠻普通的，所以改用四個 node。

Any experiment

我只用了八成的資料 training，剩下的兩成用於 test，再每次 iteration 之後都算一次答案的正確性，發現結果達到了大於九成的準確率。

Accuracy



如何決定初始的 weight

一開始我的 weight 都是用 random 決定，後來發現這樣有可能會產生很極端的結果，所以我就挑了一次

最好的結果，並改用那次初始的 **weight**。

Code:

```
clear;
N=250;
theta1 = linspace(-180,180, N)*pi/360;
r = 8 ;
x1 = -5 + r*sin(theta1)+randn(1,N);
y1 = r*cos(theta1)+randn(1,N);
x2 = 5 + r*sin(theta1)+randn(1,N);
y2 = -r*cos(theta1)+randn(1,N);

% set ans 0/1 class
oi = [x1.' y1.' zeros(250,1)+1 ; x2.' y2.' zeros(250,1)+1 ];
ans = [zeros(250,1) ;zeros(250,1)+1 ];
% rearrange
total=500;
train_total=400;
test_total = 100;
r = randperm(total);
oi(:,1) = oi(r,1);
oi(:,2) = oi(r,2);
ans = ans(r);

test=oi([train_total+1 :total], :);
test_ans = ans([train_total+1 :total],:);

% set initial value
wij =
[0.749120729449102,0.127013893617675,0.113789807451064,0.492415695453145;0.147735486618356,0.91692135212
5118,0.960227492817496,0.361313401952203;0.00288774509561229,0.672121707253300,0.119383120967445,0.47000
0490547994];
wjk =[0.0587797688209679;0.330011701173118;0.264520477657796;0.245917666364952;0.947676359396337];

rate = 0.0001;
C = 0.0025;
max_iter = 1000;
iter = 1;
sj = [];
sk = [];
oj = [];
ok = [];
% last result
accuracy=[];
e=[];
```

```

while iter<=max_iter
    error = 0;
    for i = 1:1:train_total
        % foward
        sj = oi(i,:)*wij;
        oj = [sigmf(sj,[1,0]) 1];
        sk = oj*wjk;
        ok= round(sigmoid(sj,[1,0]));
        prev_wjk = wjk;
        % update w2
        f = sigmoid(sk,[1,0]);
        for j=1:5
            wjk(j,1) = wjk(j,1) + rate * (ans(i)-ok)* f * (1-f) * oj(j);
        end
        % update w1
        for ii = 1:3
            for jj = 1:4
                f = sigmoid(sj(jj),[1,0]);
                fk = sigmoid(sk,[1,0]);
                sum = (ans(i,1) - ok) * fk * (1-fk) * prev_wjk(jj,1);
                wij(ii,jj) = wij(ii,jj) + rate* sum * f * (1-f) *prev_wjk(jj) * f * (1-f) * oi(i,ii);
            end
        end
        error = error + (ans(i) - ok).^2/2;
    end
    e(iter) = error/train_total;

    if error < C
        break;
    end
    acc=0;
    for t=1:test_total
        sj = test(t,:)*wij;
        oj = [sigmoid(sj,[1,0]) 1];
        sk = oj*wjk;
        ok= round(sigmoid(sk,[1,0]));
        prev_wjk = wjk;
        if ok== test_ans(t)
            acc = acc+1;
        end
    end
    end
    acc = acc/test_total;
    accuracy(iter) = acc;
    iter = iter + 1;

```

```

end
figure(1);
plot(e);
figure(2);
plot(accuracy);
for ix=-50:50
    for iy=-50:50
        dx=0.3*(ix-1);
        dy=0.3*(iy-1);
        oi=[dx dy 1];
        sj = oi*wij;
        oj = sigmf(sj,[1,0]);
        oj = [oj 1];
        sk = oj*wjk;
        ok= round(sigmoid(sk,[1,0]));
        if ok==1
            figure(3);
            axis([-15 15 -15 15]);
            plot(dx,dy, 'g .');
            hold on;
        elseif ok==0
            figure(3);
            axis([-15 15 -15 15]);
            plot(dx,dy, 'r .');
            hold on;
        end
    end
end
end

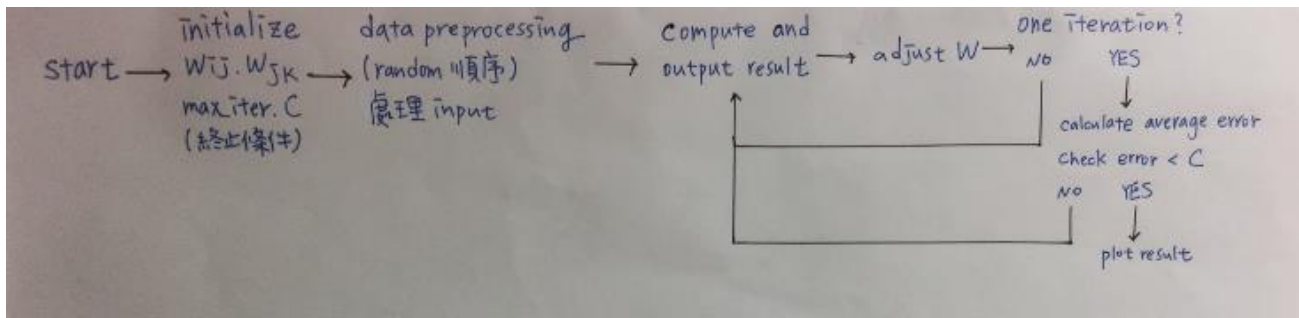
```

Q3:

做法:

這題我使用了一層 hidden layer、五個 node，最一開始處理資料時，我先將 x y 存到 oi (加上常數 1)，並且將資料順序 random，讓同種類的點分散，開始 train 之後，便是接受 input，其中採用的 active function 是 sigmoid function，經過跟 weight 內積之後，output 的輸出是四個數值，哪一個數值為 1 就是哪一類(其他數值為 0)，然後調整 weight，一個 iteration 之後計算 error，直到我指定的最大次數或是 e 小於指定的數字。

Flowchart:



討論:

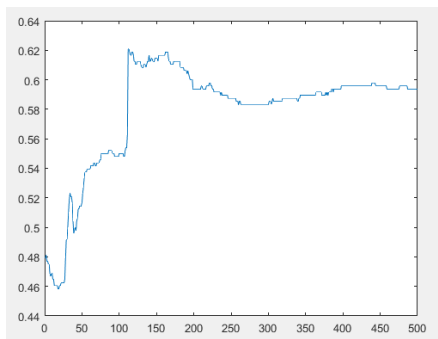
How to determine the hidden node number in each problem?

因為這題的分布看起來沒有很複雜，所以我決定用一層 layer 就好，然後一開始選擇了三個 node，發現結果蠻普通的，所以改用五個 node。

Any experiment

我只用了八成的資料 training，剩下的兩成用於 test，再每次 iteration 之後都算一次答案的正確性。

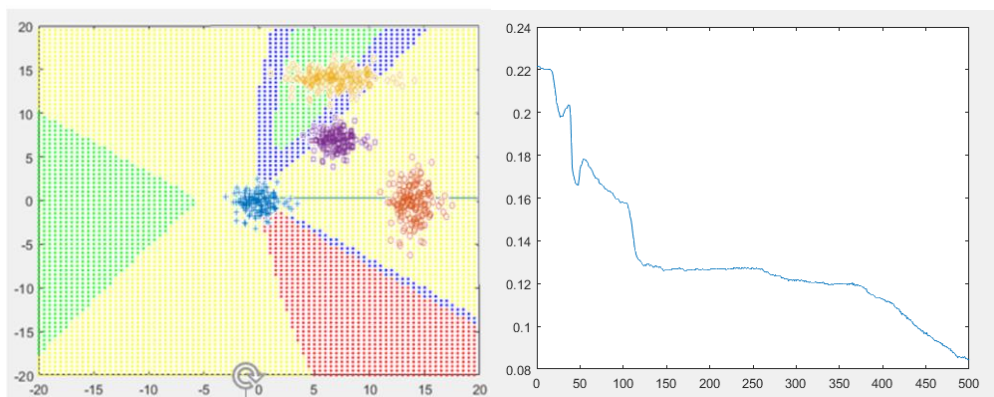
accuracy



結果:

decision region

average error vs. iteration

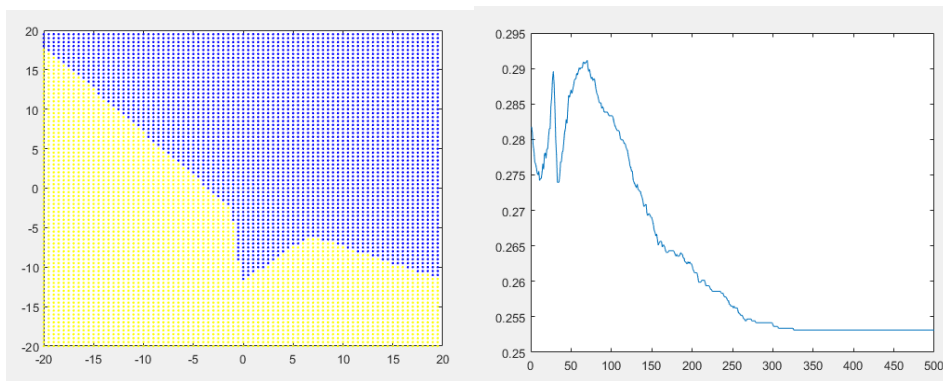


(b) 將 activation function 改為 ReLu function

將 active function 改成 ReLu 之後，我試過很多其他微調(像是 rate 之類)，發現效果都很差，可能這題並沒有很適合 ReLu funcion 當作 active function。

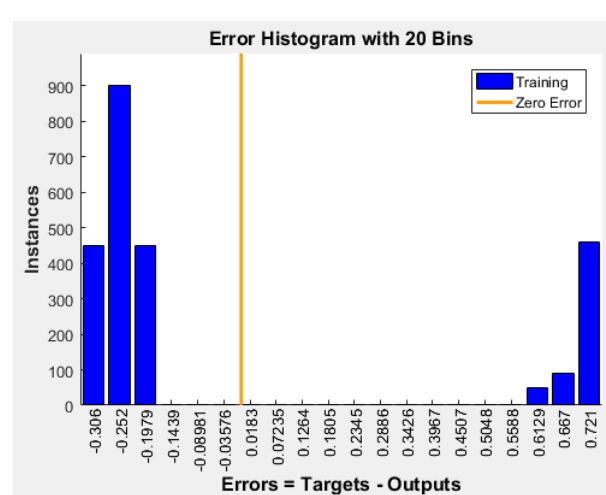
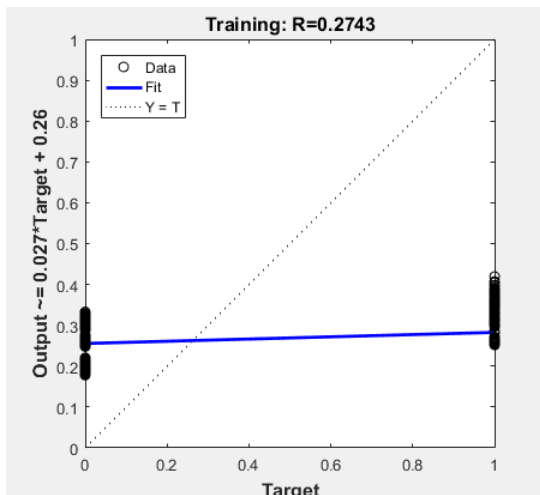
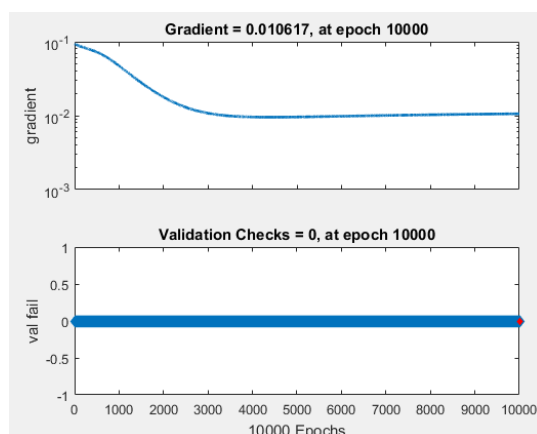
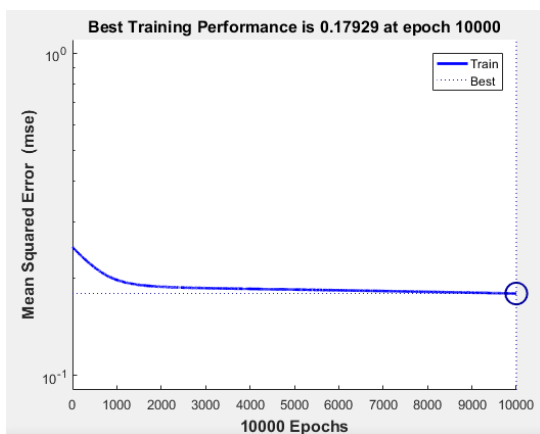
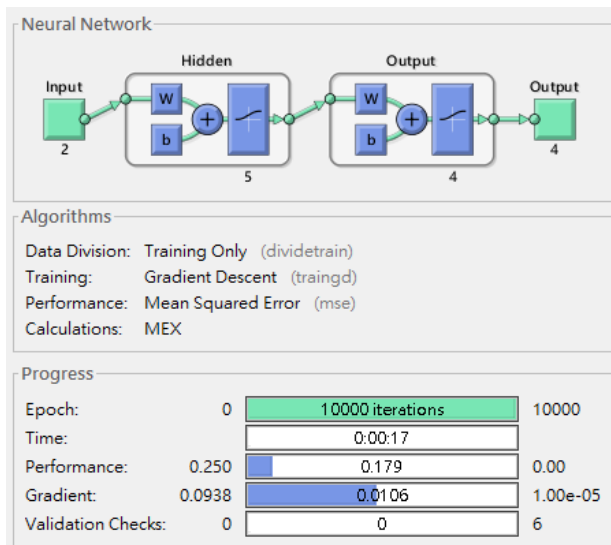
decision region

average error vs. iteration



(c) 利用 MATLAB 的 neural networks 的 toolbox

比較:用函式很快就跑出結果了，但是 performance 和 gradient 都沒有表現的很好，我也試了其他 train function，發現用其他的效果會比 gradient decent 好很多，像試 trainscg 之類的



討論

Code:

```
clear;

% set data1
mu = [0;0];
sigma = [1 0; 0 1];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
temp = zeros(150,1)+1;
oi = [r,temp];

% set data2
mu = [14;0];
sigma = [1 0; 0 4];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [ oi ;r temp];

% set data3
mu = [7;14];
sigma = [4 0; 0 1];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [ oi ;r temp];

% set data4
mu = [7;7];
sigma = [1 0; 0 1];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [ oi ;r temp];

% set ans
ans = [zeros(150,1)+1 ;zeros(150,1)+2 ;zeros(150,1)+3; zeros(150,1)+4];

% rearrange
total=600;
train_total=total*0.8;
test_total = total*0.2;
r = randperm(total);
oi(:,1) = oi(r,1);
oi(:,2) = oi(r,2);
ans = ans(r);
t=[1:total];
final_ans = zeros(600,4);
for i=1:600
    final_ans(i,ans(i)) = 1;
```

```

end

test=oi([train_total+1 :total], :);
test_ans = final_ans([train_total+1 :total],:);

% set initial value

wij = rand(3,5).*2-1;
wjk = rand(6,4).*2-1;
i_wij=wij;
i_wjk=wjk;

rate = 0.0001;
C = 0.005;
max_iter = 500;
iter = 1;
sj = [];
sk = [];
oj = zeros(600,6);
ok = [];
% last result
accuracy=[];
e=[];
% regulize
oi(:,1) = (oi(:,1)-mean(oi(:,1))) / std(oi(:,1));
oi(:,2) = (oi(:,2)-mean(oi(:,2))) / std(oi(:,2));
while iter<=max_iter
    error = 0;
    for i = 1:1:train_total
        % foward
        sj = oi(i,:)*wij;
        oj = [sigmf(sj,[1,0]) 1];
        sk = oj*wjk;
        ok= round(sigmoid(sk,[1,0]));
        prev_wjk = wjk;
        % update w2
        for k=1:4
            f = sigmoid(sk(k),[1,0]);
            for j=1:6
                wjk(j,k) = wjk(j,k) + rate * (final_ans(i,k)-ok(k))* f * (1-f) * oj(j);
            end
        end
    end
    % update w1
    for ii = 1:3
        for jj = 1:5

```

```

        f = sigmf(sj(jj),[1,0]);
        sum=0;
        for kk=1:4
            fk = sigmf(sk(kk),[1,0]);
            sum = sum + (final_ans(i,k) - ok(k) ) * fk * (1-fk) * prev_wjk(jj,kk);
        end
        wij(ii,jj) = wij(ii,jj) + rate*    sum * f * (1-f) * oi(i,ii);
    end
end
for tt=1:4
    error = error + (final_ans(i,tt) - ok(tt)).^2/8;
end

end
e(iter) = error/train_total;

if error < C
    break;
end
acc=0;
for t=1:test_total
    sj = test(t,:)*wij;
    oj = [sigmf(sj,[1,0]) 1];
    sk = oj*wjk;
    ok= round(sigmoid(sk,[1,0]));
    prev_wjk = wjk;
    for temp=1:4
        if ok(temp) == test_ans(t,temp)
            acc = acc+1;
        end
    end
end
end
acc = acc/test_total/4;
accuracy(iter) = acc;
iter = iter + 1;

end
figure(1);
plot(e);
figure(2);
plot(accuracy);
for ix=-40:40 %% ±q-15~15
    for iy=-40:40
        dx=0.5*(ix-1);
        dy=0.5*(iy-1);

```

```

input=[dx dy 1];
sj = input*wij;
oj = sigmf(sj,[1,0]);
oj = [oj 1];
sk = oj*wjk;
ok= round(sigmoid(sk,[1,0]));
figure(3);
if ok(1)==1
    axis([-20 20 -20 20]);
    plot(dx,dy, 'g .');
    hold on;
end
if ok(2)==1
    axis([-20 20 -20 20]);
    plot(dx,dy, 'r .');
    hold on;
end
if ok(3)==1
    axis([-20 20 -20 20]);
    plot(dx,dy, 'b .');
    hold on;
end
if ok(4)==1
    axis([-20 20 -20 20]);
    plot(dx,dy, 'y .');
    hold on;
end

end

end

```

```

%%% relu
wij = rand(3,5)*2-1;
wjk = rand(6,4)*2-1;
i_wij=wij;
i_wjk=wjk;
rate = 0.0001;
iter=1;
temp2=[];
while iter<=max_iter
    error = 0;
    for i = 1:1:train_total

```

```

% foward
sj = oi(i,:)*wij;
oj = [max(sj,0) 1];
sk = oj*wjk;
temp2(i,:) = sk;
ok= max(sk,0) > 0;
prev_wjk = wjk;
% update w2
for k=1:4
    f=1;
    if max(sj(k),0)==0
        f=0;
    end
    for j=1:6
        wjk(j,k) = wjk(j,k) + rate * (final_ans(i,k)-ok(k))* f * (1-f) * oj(j);
    end
end
% update w1
for ii = 1:3
    for jj = 1:5
        f=1;
        if max(sj(jj),0)==0
            f=0;
        end
        sum=0;
        for kk=1:4
            fk=1;
            if max(sk(kk),0)==0
                fk=0;
            end
            sum = sum + (final_ans(i,k) - ok(k) ) * fk * prev_wjk(jj,kk);
        end
        wij(ii,jj) = wij(ii,jj) + rate* sum * f * oi(i,ii);
    end
end
for tt=1:4
    error = error + (final_ans(i,tt) - ok(tt)).^2/8;
end

```

end

e(iter) = error/train_total;

if error < C

break;

```

end
acc=0;
for t=1:test_total
    sj = test(t,:)*wij;
    oj = [max(sj,0) 1];
    sk = oj*wjk;
    ok= max(sk,0)>0;
    prev_wjk = wjk;
    for temp=1:4
        if ok(temp) == test_ans(t,temp)
            acc = acc+1;
        end
    end
end
end
acc = acc/test_total/4;
accuracy(iter) = acc;
iter = iter + 1;
end
figure(4);
plot(e);
figure(5);
plot(accuracy);
for ix=-40:40 %% ±q-15~15
    for iy=-40:40
        dx=0.5*(ix-1);
        dy=0.5*(iy-1);
        oi=[dx dy 1];
        sj = oi*wij;
        oj = [max(sj,0) 1];
        sk = oj*wjk;
        ok= max(sk,0)>0;
        figure(6);
        if ok(1)==1
            axis([-20 20 -20 20]);
            plot(dx,dy, 'g .');
            hold on;
        end
        if ok(2)==1
            axis([-20 20 -20 20]);
            plot(dx,dy, 'r .');
            hold on;
        end
        if ok(3)==1
            axis([-20 20 -20 20]);

```

```

        plot(dx,dy, 'b .');
        hold on;
    end
    if ok(4)==1
        axis([-20 20 -20 20]);
        plot(dx,dy, 'y .');
        hold on;
    end

end

end
end

```

```

% use toolbox
clear;

% set data1
mu = [0;0];
sigma = [1 0; 0 1];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [r];

% set data2
mu = [14;0];
sigma = [1 0; 0 4];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [ oi ;r];

% set data3
mu = [7;14];
sigma = [4 0; 0 1];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [ oi ;r];

% set data4
mu = [7;7];
sigma = [1 0; 0 1];
rng default % For reproducibility
r = mvnrnd(mu,sigma,150);
oi = [ oi ;r ];

% set ans
ans = [zeros(150,1)+1 ;zeros(150,1)+2 ;zeros(150,1)+3; zeros(150,1)+4];

% rearrange
total=600;
train_total=total*0.8;

```



```

test_total = total*0.2;
r = randperm(total);
oi(:,1) = oi(r,1);
oi(:,2) = oi(r,2);
ans = ans(r);
t=[1:total];
final_ans = zeros(600,4);
for i=1:600
    final_ans(i,ans(i)) = 1;
end
test=oi([train_total+1 :total], :);
test_ans = final_ans([train_total+1 :total],:);

net = feedforwardnet(5);
net.layers{2}.size = 4;
net.inputs{1}.size=2;
net.layers{1}.transferFcn = 'logsig';
net.layers{2}.transferFcn = 'logsig';
net.divideFcn = 'dividetrain';
net.trainFcn='traingd';
net=train(net,oi.',final_ans.');
```