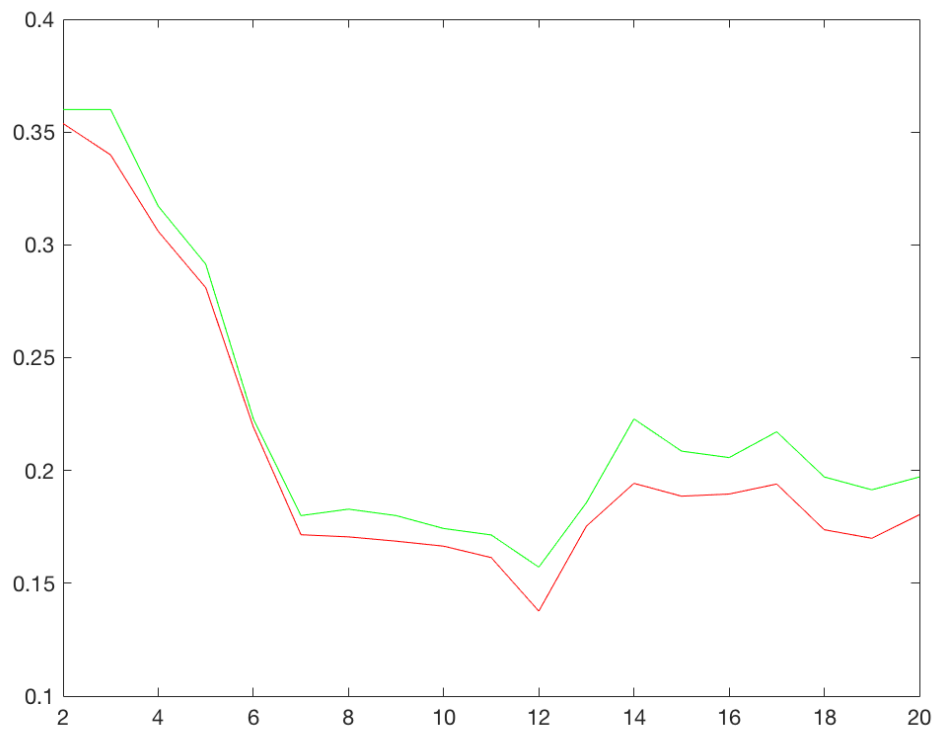
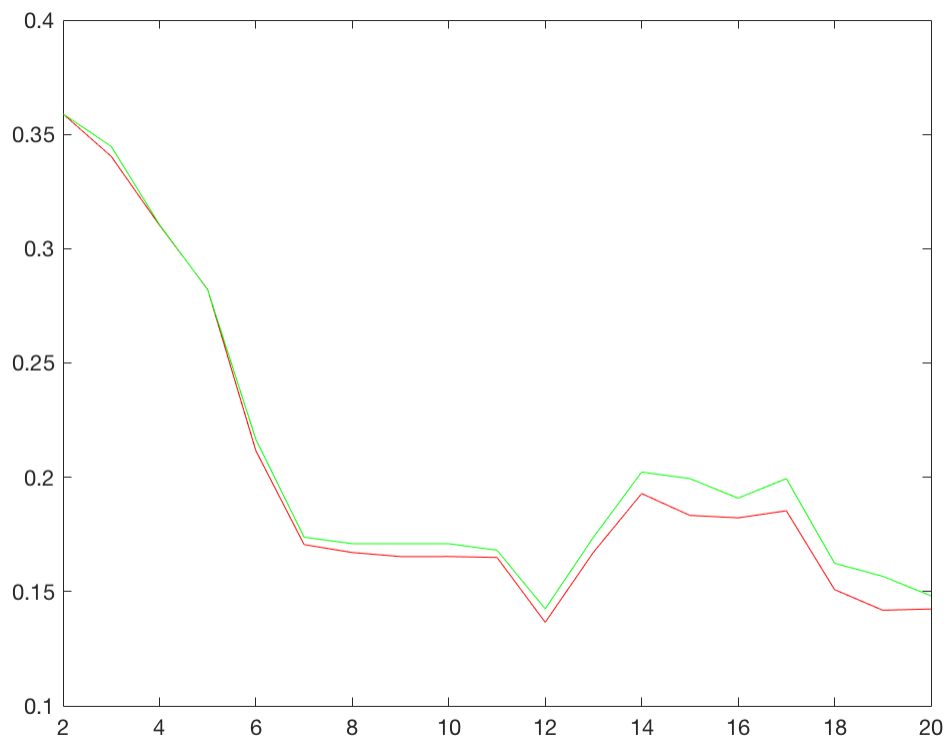


1. For cross validation:



It appears that the optimum number of basis functions is 12.

For loo:



It appears that the optimum number of basis functions is 12.

For cloo:



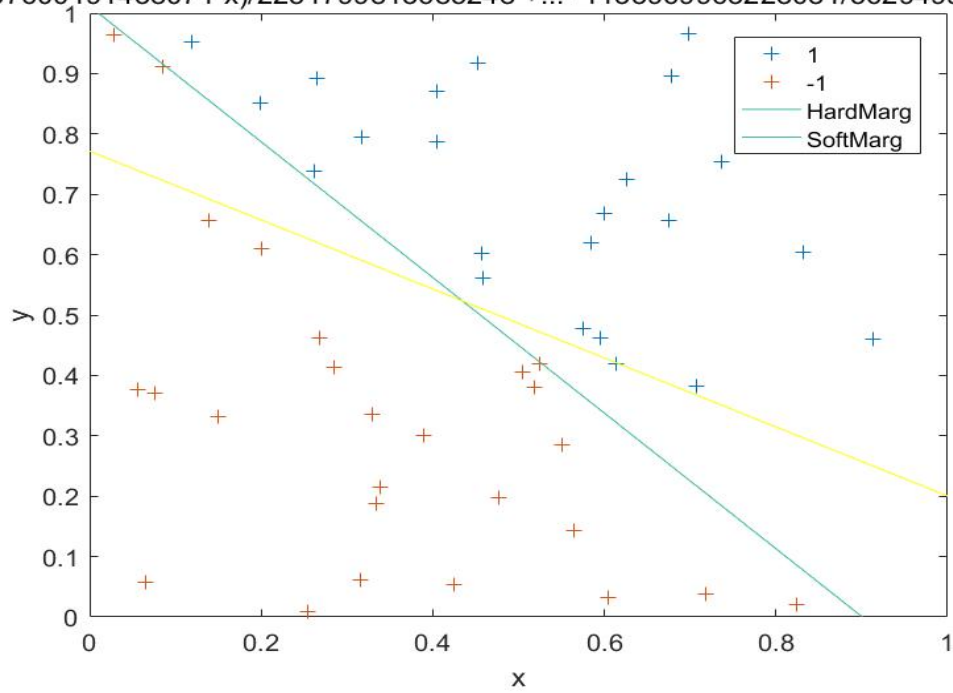
It appears that the optimum number of basis functions is 12.

	Training error	Test error	Number of basis
10-fold-CV	0.1377	0.1571	12
LOO	0.1366	0.1425	12
CLOO		0.1425	12

2.

Linear:

$$(3357600191438071 x)/2251799813685248 + \dots - 1133969963228081/5629499534213$$

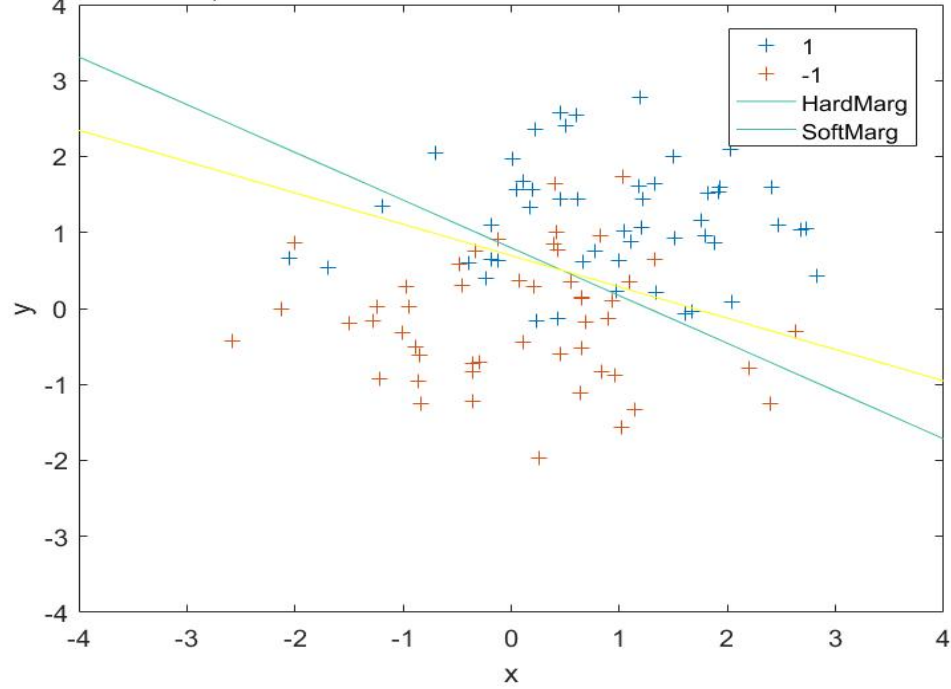


tables:

	train_error	test_error
Hard margin	0	0.0379
Soft margin	0.0588	0.1022

Noisy linear:

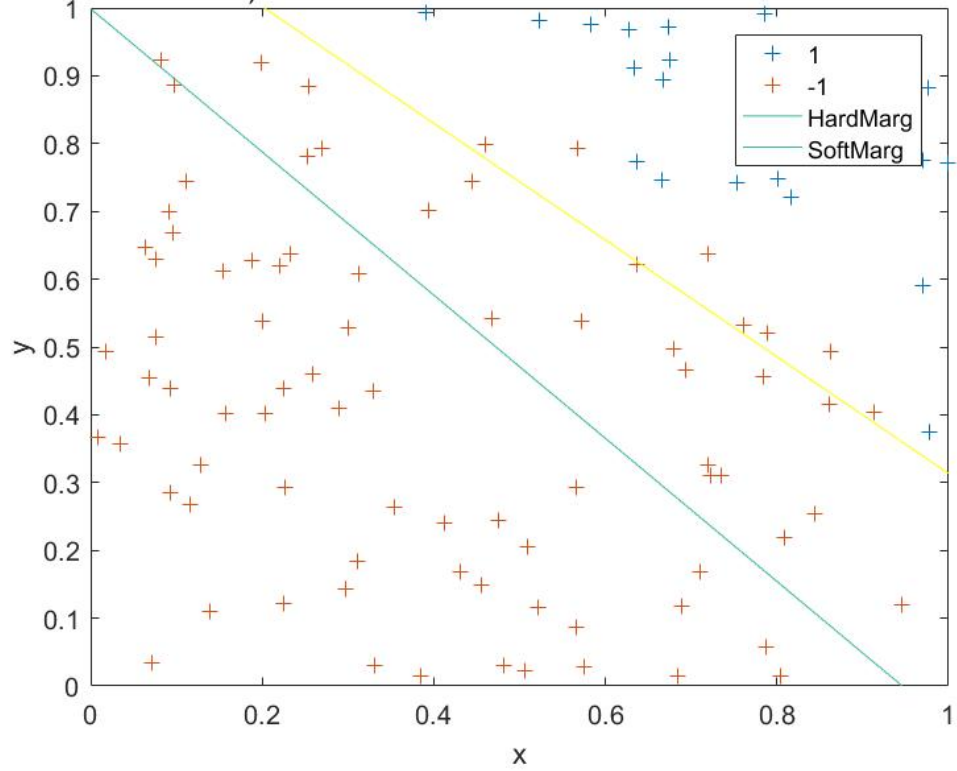
3643128736766055 x)/18014398509481984 +...- 7299583102871809/9007199254740



	train_error	test_error
Hard margin	0.1881	0.2492
Soft margin	0.2079	0.2659

Quadratic:

$$(8427896887220317 x)/4503599627370496 + \dots - 718564201853619/2814749767106$$



	train_error	test_error
Hard margin	0.2673	0.3037
Soft margin	0.0693	0.0723

Code:

Q1

(a)

```
rng(1);
num=5;
[idx, mu]=kmeans(X', num);
mu=mu';
sigma=zeros(num, 1);
phi=zeros(num,size(X,2));
for i=1:size(X,2)
    sigma(idx(i), 1)=sigma(idx(i), 1)+norm(X(:,i)-
mu(:,idx(i)))^2;
end

for i=1:num
    sigma(i,1)=sigma(i,1)/sum(idx==i);
end

for i=1:num
    for j=1:size(X,2)
        phi(i,j)=exp(-norm(X(:,j)-
mu(:,i))^2/(sigma(i,1)+0.000001));
    end
end

W=inv(phi*phi')*phi*y;
y_est=W'*phi;
for i=1:length(y_est)
    if y_est(i)>=0.5
        y_est(i)=1;
    else
        y_est(i)=0;
    end
end
```

```

        end
    end

    error_rate = 1-sum(y_est'==y)/length(y);
(b)
%STANDARD CROSS VALIDATION
num=20; %cluster 1-20 centers
cv_num=10; %divide raw data into 13 groups
a=35;
error_train=zeros(num,1);
error_test=zeros(num,1);
for n=2:num
    rng(150);
    [idx, mu]=kmeans(X', n);
    mu=mu';
    sigma=zeros(n, 1);
    train_error=zeros(cv_num,1);
    test_error=zeros(cv_num,1);

    %calculate sigma
    for l=1:size(X,2)
        sigma(idx(l), 1)=sigma(idx(l), 1)+norm(X(:,l)-
mu(:,idx(l)))^2;
    end
    for i=1:n
        sigma(i,1)=sigma(i,1)/sum(idx==i);
    end

    %cross validation
    for i=1:cv_num

        %split data

```



```

train=[X(:,1:a*(i-1)),X(:,(a*(i)+1):size(X,2))];
test=X(:,(a*(i-1)+1):a*(i));
y_train=[y(1:a*(i-1),1);y((a*(i)+1):size(X,2),1)];
y_test=y((a*(i-1)+1):a*i);

%calculate phi&W&train-error for train data
phi=zeros(n,size(train,2));
for k=1:n
    for j=1:size(train,2)
        phi(k,j)=exp(-norm(train(:,j)-
mu(:,k))^2/(sigma(k,1)+0.000001));
    end
end
W=pinv(phi*phi')*(phi*y_train);
train_est=phi'*W;
for t=1:length(train_est)
    if train_est(t)>=0.5
        train_est(t)=1;
    else
        train_est(t)=0;
    end
end
train_error(i,1)=1-
sum(train_est==y_train)/size(y_train,1);

%calculate phi&W&test-error for test data
phi=zeros(n,a);
for k=1:n
    for j=1:a
        phi(k,j)=exp(-norm(test(:,j)-
mu(:,k))^2/(sigma(k,1)+0.000001));
    end
end

```

```

end

test_est=phi'*W;
for w=1:length(test_est)
    if test_est(w)>=0.5
        test_est(w)=1;
    else
        test_est(w)=0;
    end
end
end
test_error(i,1)=1-
sum(test_est==y_test)/size(y_test,1);
end
error_train(n,1)=sum(train_error)/cv_num;
error_test(n,1)=sum(test_error)/cv_num;
end
plot(2:num,error_train(2:num,1),'r');
hold on;
plot(2:num,error_test(2:num,1),'g');

%STANDARD LOO
num=20; %cluster 1-20 centers
a=1;%size
error_train=zeros(num,1);
error_test=zeros(num,1);
for n=2:num
    rng(150);
    [idx, mu]=kmeans(X', n);
    mu=mu';
    sigma=zeros(n, 1);
    train_error=zeros(cv_num,1);
    test_error=zeros(cv_num,1);

```

```

%calculate sigma
for l=1:size(X,2)
    sigma(idx(l), 1)=sigma(idx(l), 1)+norm(X(:,l)-
mu(:,idx(l)))^2;
end
for i=1:n
    sigma(i,1)=sigma(i,1)/sum(idx==i);
end

%cross validation
for i=1:cv_num

    %split data
    train=[X(:,1:a*(i-1)),X(:,(a*(i)+1):size(X,2))];
    test=X(:,(a*(i-1)+1):a*(i));
    y_train=[y(1:a*(i-1),1);y((a*(i)+1):size(X,2),1)];
    y_test=y((a*(i-1)+1):a*i);

    %calculate phi&W&train-error for train data
    phi=zeros(n,size(train,2));
    for k=1:n
        for j=1:size(train,2)
            phi(k,j)=exp(-norm(train(:,j)-
mu(:,k))^2/(sigma(k,1)+0.000001));
        end
    end

    W=pinv(phi*phi')*(phi*y_train);
    train_est=phi'*W;
    for t=1:length(train_est)
        if train_est(t)>=0.5
            train_est(t)=1;
        end
    end
end

```

```

        else
            train_est(t)=0;
        end
    end
    train_error(i,1)=1-
sum(train_est==y_train)/size(y_train,1);

    %calculate phi&W&test-error for test data
    phi=zeros(n,a);
    for k=1:n
        for j=1:a
            phi(k,j)=exp(-norm(test(:,j)-
mu(:,k))^2/(sigma(k,1)+0.000001));
        end
    end
    test_est=phi'*W;
    for w=1:length(test_est)
        if test_est(w)>=0.5
            test_est(w)=1;
        else
            test_est(w)=0;
        end
    end
    test_error(i,1)=1-
sum(test_est==y_test)/size(y_test,1);
    end
    error_train(n,1)=sum(train_error)/cv_num;
    error_test(n,1)=sum(test_error)/cv_num;
end
plot(2:num,error_train(2:num,1),'r');
hold on;
plot(2:num,error_test(2:num,1),'g');

```

```

%STANDARD CROSS VALIDATION without iteration
num=20; %cluster 1-20 centers
rng(150);
y_error=zeros(num/2,1);
for n=2:num
    %cluster data using Kmeans
    [idx,mu]=kmeans(X',n);
    mu=mu';
    sigma=zeros(n,1);
    for j=1:size(X,2)
        sigma(idx(j),1)=sigma(idx(j),1)+norm(X(:,j)-
mu(:,idx(j)))^2;
    end
    for i=1:n
        sigma(i,1)=sigma(i,1)/sum(idx==i);
    end

    %calculate phi&W&train-error for train data
    phi=zeros(n,size(X,2));
    for i=1:n
        for j=1:size(X,2)
            phi(i,j)=exp(-norm(X(:,j)-
mu(:,i))^2/(sigma(i,1)+0.0001));
        end
    end
    w=pinv(phi*phi')*(phi*y);
    y_est=phi'*w;
    H=phi'*pinv(phi*phi')*phi;
    for t=1:length(y_est)
        if y_est(t)>=0.5
            y_est(t)=1;
        else
            y_est(t)=0;
        end
    end
end

```

```

        end
    end
    y_error(i,1)=1-sum(y_est==y)/size(y,1);
end

plot(2:num,y_error(2:num,1),'g');

y_error(13,1)

```

Q2

**%harmargin**

```

function [b,b0]=hardmarg(X,y)
[d,n]=size(X);
Y=diag(y)
H=(X*Y)'*(X*Y);
f=-1*ones(n,1);
Aeq=Y';
beq=0;
lb=zeros(n,1);
alpha=quadprog(H,f,[],[],Aeq,beq,lb);
b=X*Y*alpha;
[argVal,argMax]=max(alpha);
b0=1/y(argMax,1) - b'* X(:, argMax);
end

```

**%softmargin**

```

function [b,b0] = SoftMarg(X,y,gamma)
[d,n]=size(X);
Y=diag(y)
H=(X*Y)'*(X*Y);
f=-1*ones(n,1);
Aeq=Y';
beq=0;

```

```

lb=zeros(n,1);
ub=gamma*ones(n,1);
alpha = quadprog(H, f, [], [], A, bb, lb, ub);
b=X*Y*alpha;
while alpha(arg,1) < 0.01 || alpha(arg,1) > gamma
arg = arg+1;
end
b0s = 1/y(arg,1) - b'* X(:, arg);
end

%classifier
function[yhat]=classify(Xtest,b,b0)
[d,n] = size(X);
yhat=(b'*Xtest+b0)';
for i=1:n
    if yhat(i)>=0
        yhat(i)=1;
    else
        yhat(i)=-1;
    end
end
end

%error
count = 0;
for i = 1:size(Xtest, 2)
if yhat(i,1) ~= ytest(i,1)
count = count+1;
end
end
test_error = count/size(X, 2);

```