3/1/2019 Project1

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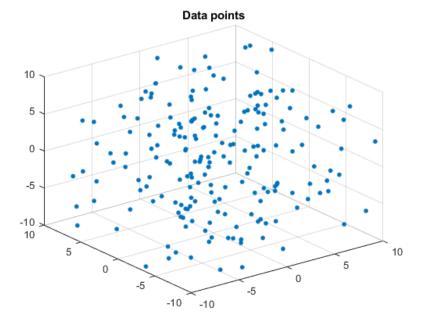
```
clc
clear all
close all
```

#### Part 1 (Data Generation)

### **Data generation**

```
%Inputs
x = -10 + 20*rand(200,3);
%Weights
w = [-0.8; 2.1; 1.5];
%Noise
vr = 10; % variance of the data
mn = 0; % mean of the data
sd = sqrt(vr);
n = 200; % number of points to be generated
noise = sd*randn(200,1) + mn;
%Outputs
y = x*w + 10 + noise;
figure('Name','Generated Data')
scatter3(x(:,1),x(:,2),x(:,3),200,'.')
title('Data points')
save('data.mat','x','y')
clear all
load('data.mat')
```

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#### Part 2 (Ridge Regression)

### Standardizing inputs

```
xtrain = (x(1:100,:) - mean(x(1:100,:)))./std(x(1:100,:));
xtest = (x(101:end,:) - mean(x(1:100,:)))./std(x(1:100,:));
```

#### Normalizing outputs

```
ytrain = y(1:100,:) - mean(y(1:100,:));
ytest = y(101:end,:) - mean(y(1:100,:));
```

### Determining wo (bias term 'b' for ridge)

```
wor = mean(y(1:100,1))
wor =
    9.8933
```

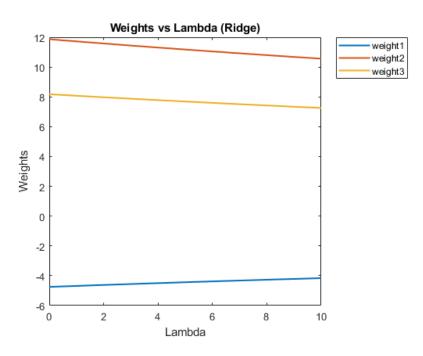
#### Learning weights

```
w = [0 0 0];
for i = 1:101
    lambda = 0.1*(i-1);
    a = ((xtrain'*xtrain + lambda * eye(max(size(xtrain'*xtrain)))) \ (xtrain'*ytrain))';
    w = [w; a];
end
wr = w(2:end,:);
```

# SSE

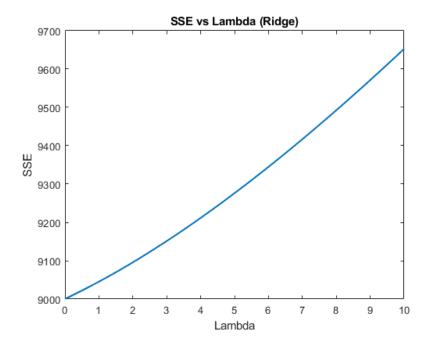
### Plots for Weights vs Lambda (Ridge)

```
lambda = (0:0.1:10)';
figure('Name','Weights vs Lambda (Ridge)')
plot(lambda,wr(:,1),'-','LineWidth',1.5)
hold on
plot(lambda,wr(:,2),'-','LineWidth',1.5)
hold on
plot(lambda,wr(:,3),'-','LineWidth',1.5)
title('Weights vs Lambda (Ridge)'); xlabel('Lambda'); ylabel('Weights');
legend('weight1','weight2','weight3','Location','northeastoutside')
hold off
```



# Plot for SSE vs Lambda (Ridge)

```
figure('Name','SSE vs Lambda (Ridge)')
plot(lambda,SSEr,'-','LineWidth',1.5)
title('SSE vs Lambda (Ridge)'); xlabel('Lambda'); ylabel('SSE');
```



## Part 3 (Lasso Regression)

#### Standardizing inputs

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```
xtrain = (x(1:100,:) - mean(x(1:100,:)))./std(x(1:100,:));
xtest = (x(101:end,:) - mean(x(1:100,:)))./std(x(1:100,:));
```

#### Normalizing outputs

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```
ytrain = y(1:100,:) - mean(y(1:100,:));
ytest = y(101:end,:) - mean(y(1:100,:));
```

#### Determining wo (bias term 'b' for Lasso)

```
wol = mean(y(1:100,1))

wol = 9.8933
```

#### Overall least square estimate

```
wp = xtrain\ytrain;
```

#### First value of GE

```
GE = sign(wp)';
```

### Learning weights using TIBSHIRANI algorithm

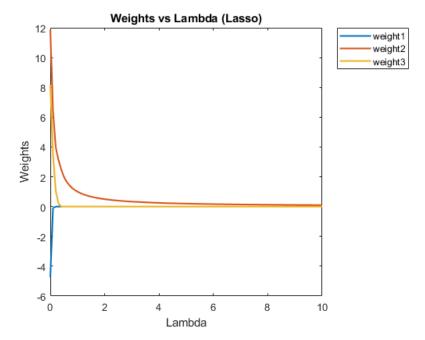
```
eps = 10^-4;
w = [0 0 0];
for i = 1:101
    lambda = 0.1*(i-1);
    t = 1/lambda;
        while sum(abs(wp)) > t+eps
        options = optimoptions(@quadprog, 'Display', 'off');
        [wp] = quadprog(2*xtrain'*xtrain, -2*xtrain'*ytrain, GE, t*ones(size(GE,1),1),[],[],[],[],wp,options);
        GE = [GE; sign(wp)'];
        end
        w = [w; wp'];
end
wl = w(2:end,:);
```

## SSE

# Plots for Weights vs Lambda (Lasso)

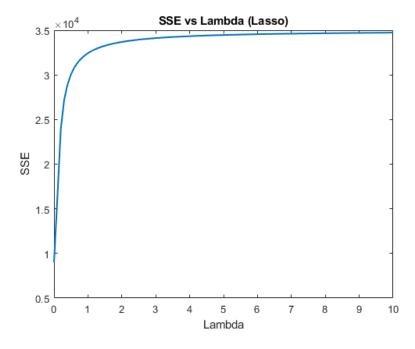
```
lambda = (0:0.1:10)';
figure('Name','Weights vs Lambda (Lasso)')
plot(lambda,wl(:,1),'-','LineWidth',1.5)
hold on
plot(lambda,wl(:,2),'-','LineWidth',1.5)
hold on
plot(lambda,wl(:,3),'-','LineWidth',1.5)
title('Weights vs Lambda (Lasso)'); xlabel('Lambda'); ylabel('Weights');
legend('weight1','weight2','weight3','Location','northeastoutside')
hold off
```

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### Plot for SSE vs Lambda (Lasso)

```
figure('Name','SSE vs Lambda (Lasso)')
plot(lambda,SSE1,'-','LineWidth',1.5)
title('SSE vs Lambda (Lasso)'); xlabel('Lambda'); ylabel('SSE');
```

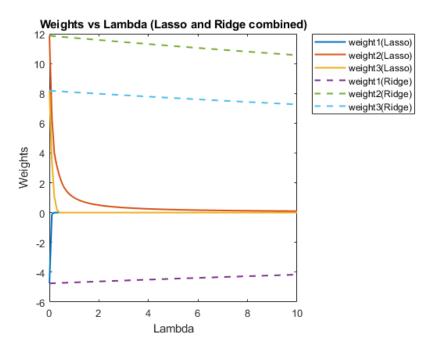


#### Part 4 (Miscellaneous plots for understanding purpose only)

### Plot for Weights vs Lambda (Lasso and Ridge combined)

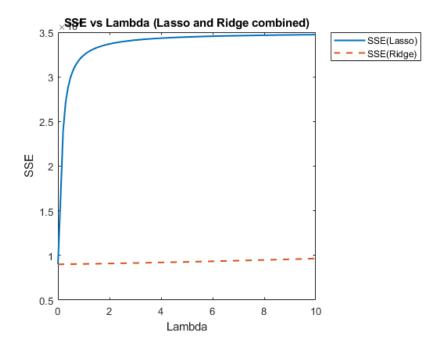
```
figure('Name','Plot for Weights vs Lambda (Lasso and Ridge combined)')
plot(lambda,wl(:,1),'-','LineWidth',1.5)
hold on
plot(lambda,wl(:,2),'-','LineWidth',1.5)
hold on
plot(lambda,wl(:,3),'-','LineWidth',1.5)
hold on
plot(lambda,wr(:,1),'--','LineWidth',1.5)
hold on
plot(lambda,wr(:,2),'--','LineWidth',1.5)
hold on
```

```
plot(lambda,wr(:,3),'--','LineWidth',1.5)
title('Weights vs Lambda (Lasso and Ridge combined)'); xlabel('Lambda'); ylabel('Weights');
legend('weight1(Lasso)','weight2(Lasso)','weight3(Lasso)','weight1(Ridge)','weight2(Ridge)','weight3(Ridge)','Location','northeastoutside')
hold off
```



### Plot for SSE vs Lambda(Lasso and Ridge combined)

```
figure('Name','SSE vs Lambda (Lasso and Ridge combined)')
plot(lambda,SSEl,'-','LineWidth',1.5)
hold on
plot(lambda,SSEr,'--','LineWidth',1.5)
title('SSE vs Lambda (Lasso and Ridge combined)'); xlabel('Lambda'); ylabel('SSE');
legend('SSE(Lasso)','SSE(Ridge)','Location','northeastoutside')
hold off
```



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