Problem1

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```
require(magrittr)
require(dplyr)
require(ggplot2)
require(glmnet)
```

Fit the data in Table 1 to a linear model using:

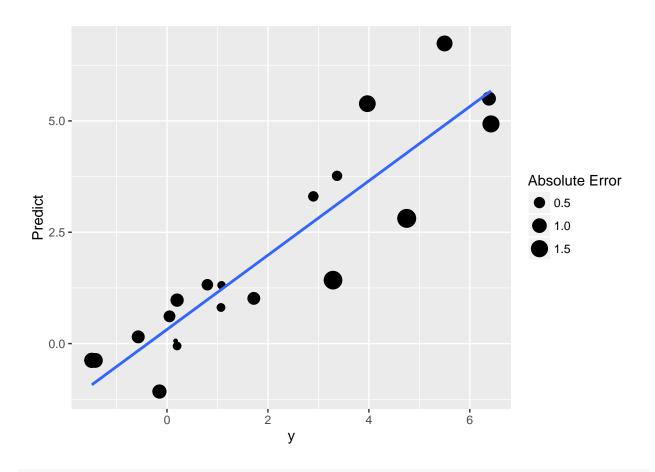
Table1

```
##
        x1
             x2
## 1
      0.99 0.98 6.42
## 2 -0.75 -0.76 0.20
## 3 -0.50 -0.48 0.80
## 4 -1.08 -1.08 -0.57
     0.09 0.09 4.75
## 6 -1.28 -1.27 -1.42
## 7 -0.79 -0.79 1.07
## 8 -1.17 -1.17 0.20
## 9 -0.57 -0.57 1.08
## 10 -1.62 -1.62 -0.15
## 11 0.34 0.35 2.90
## 12 0.51 0.51 3.37
## 13 -0.91 -0.92 0.05
## 14 1.85 1.86 5.50
## 15 -1.12 -1.12 0.17
## 16 -0.70 -0.70 1.72
## 17 1.19 1.18 3.97
## 18 1.24 1.23 6.38
## 19 -0.52 -0.52 3.29
## 20 -1.41 -1.44 -1.49
```

1. Least Squares

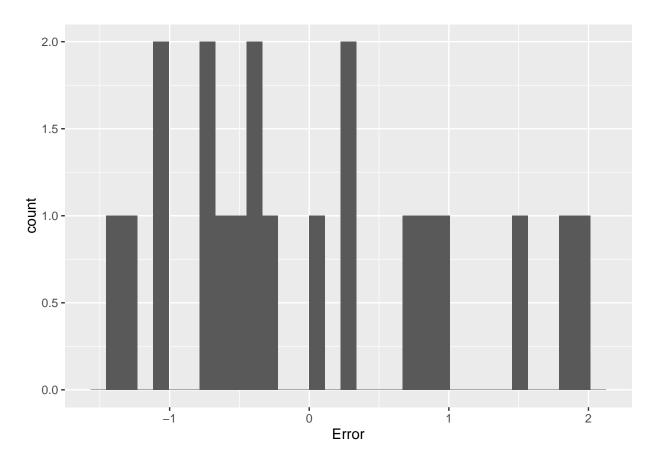
```
LeastFit<-lm(formula = y~x1+x2,data=Table1)
LeastFit</pre>
```

```
plotDF<-Table1
plotDF[,'Type']<-'Train'
plotDF$Predict<-predict(LeastFit,plotDF[,1:2])
plotDF$Error<-plotDF$y-plotDF$Predict
ggplot(plotDF,aes(x=y,y=Predict,size=abs(Error)))+geom_point()+
    scale_size("Absolute Error")+geom_smooth(method="lm",se=F,size=1)</pre>
```



```
sqrt(var(data.frame(plotDF %>% filter(Type=="Train") %>% select(Error))))/20
## Error
## Error 0.05088412
```

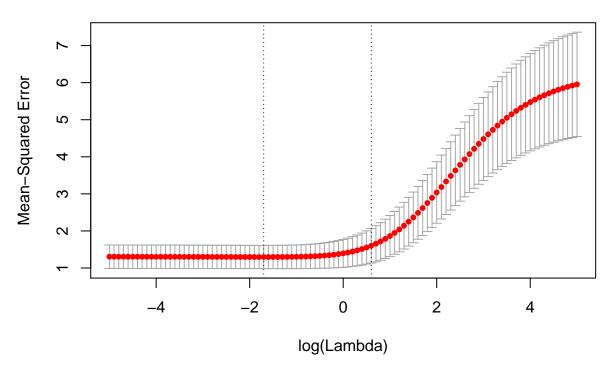
```
ggplot(plotDF,aes(x=Error)) + geom_histogram()
```

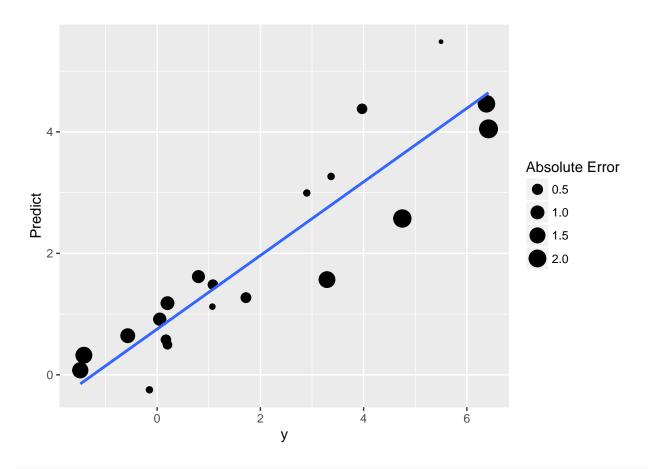


```
sensitivities <- coef(LeastFit)
sensDF <- data.frame(Method = 0, Var = 0, Value=0)
sensDF[1:length(sensitivities), 'Method'] <- "Least-Squares"
sensDF[1:length(sensitivities), 'Var'] <- names(sensitivities)
sensDF[1:length(sensitivities), 'Value'] <- (sensitivities)
rowStart <- length(sensitivities)+1</pre>
```

2. Ridge Regression

Ridge regression sets alpha=0, which adds damping to the coefficients

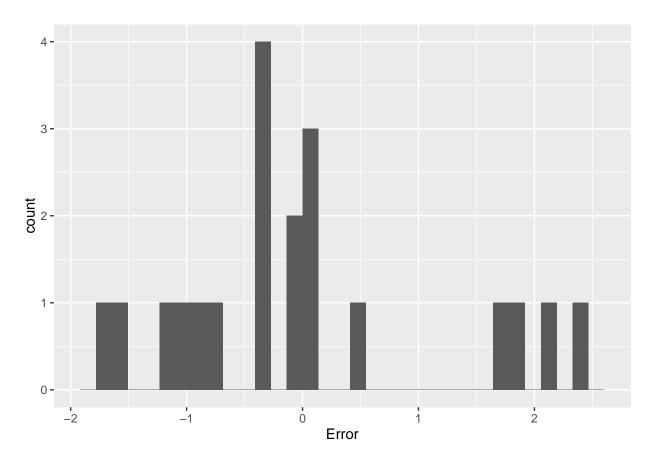




```
sqrt(var(data.frame(plotDF %>% filter(Type=="Train") %>% select(Error))))/20
```

Error 0.05976789

ggplot(plotDF,aes(x=Error)) + geom_histogram()

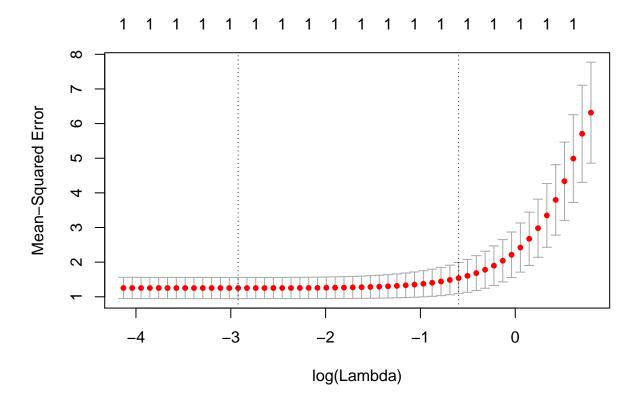


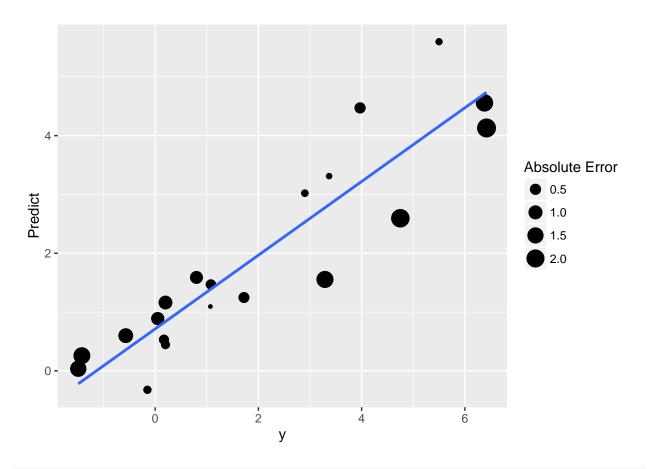
```
sensDF[rowStart:(rowStart + length(sensitivities)-1),'Method'] <- "Ridge"
sensDF[rowStart:(rowStart+length(sensitivities)-1),'Var']<-t(t(rownames(sensitivities)))
sensDF[rowStart:(rowStart +length(sensitivities)-1),'Value']<-as.numeric(sensitivities)
rowStart <- rowStart + length(sensitivities)</pre>
```

3. Lasso Regression

Lasso regression sets alpha=1

```
crossValid <- cv.glmnet(as.matrix(Table1[,1:2]),as.matrix(Table1$y),alpha = 1)
plot(crossValid)</pre>
```

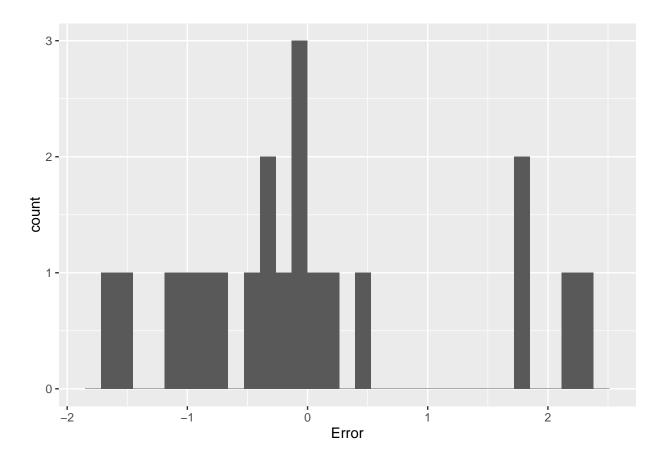




```
sqrt(var(data.frame(plotDF %>% filter(Type=="Train") %>% select(Error))))/20
```

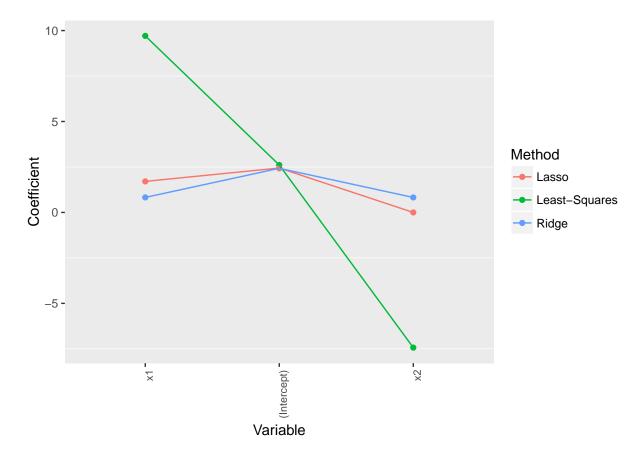
```
## Error 0.05832321
```

```
ggplot(plotDF,aes(x=Error)) + geom_histogram()
```



```
sensDF[rowStart:(rowStart + length(sensitivities)-1),'Method'] <- "Lasso"
sensDF[rowStart:(rowStart+length(sensitivities)-1),'Var']<-t(t(rownames(sensitivities)))
sensDF[rowStart:(rowStart +length(sensitivities)-1),'Value']<-as.numeric(sensitivities)
rowStart <- rowStart + length(sensitivities)</pre>
```

Compare Methods



The Lasso and Ridge are both more bounded in their coefficents.