

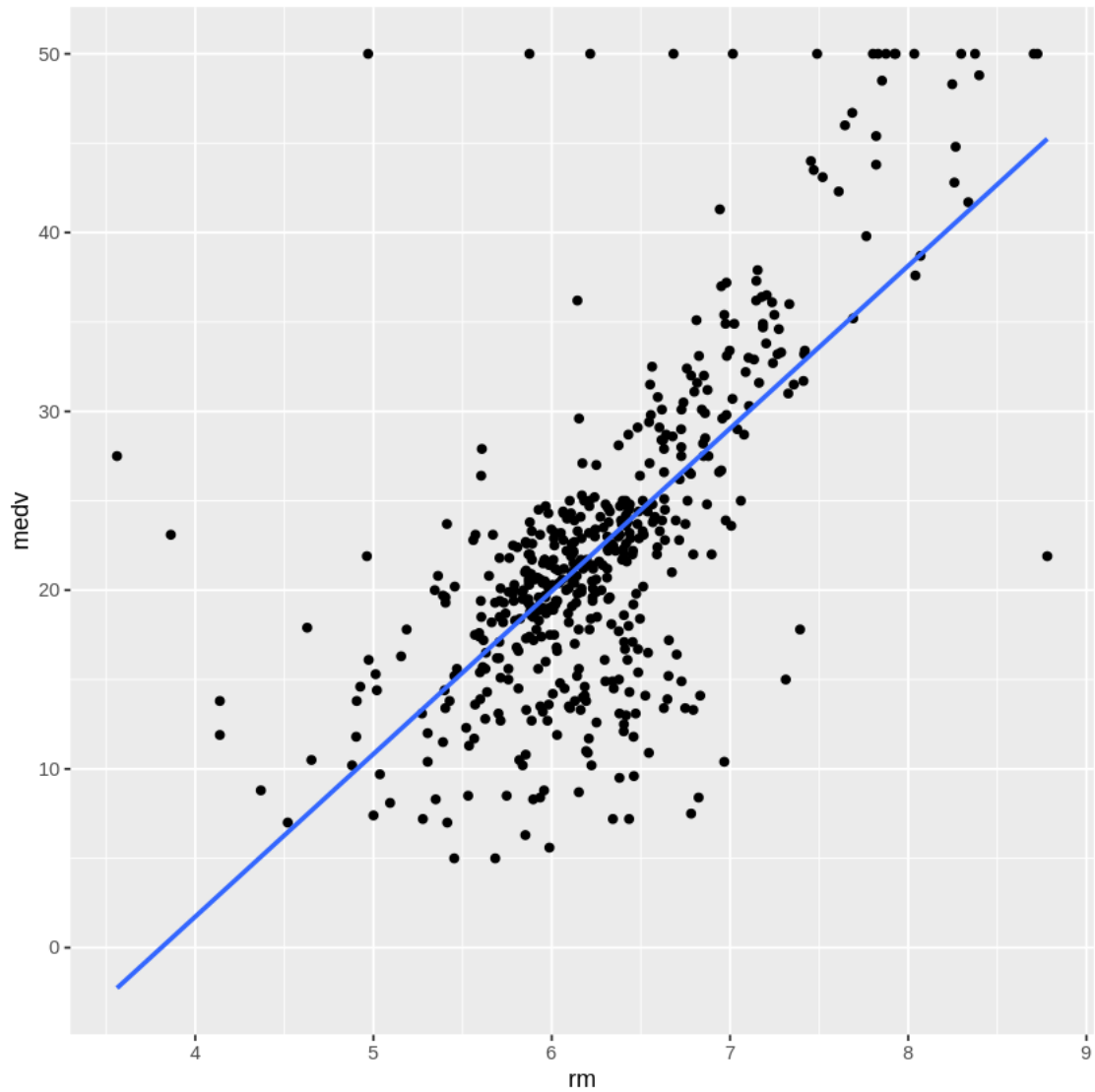
# Ezra Cohen lab 8

August 1, 2021

task 1

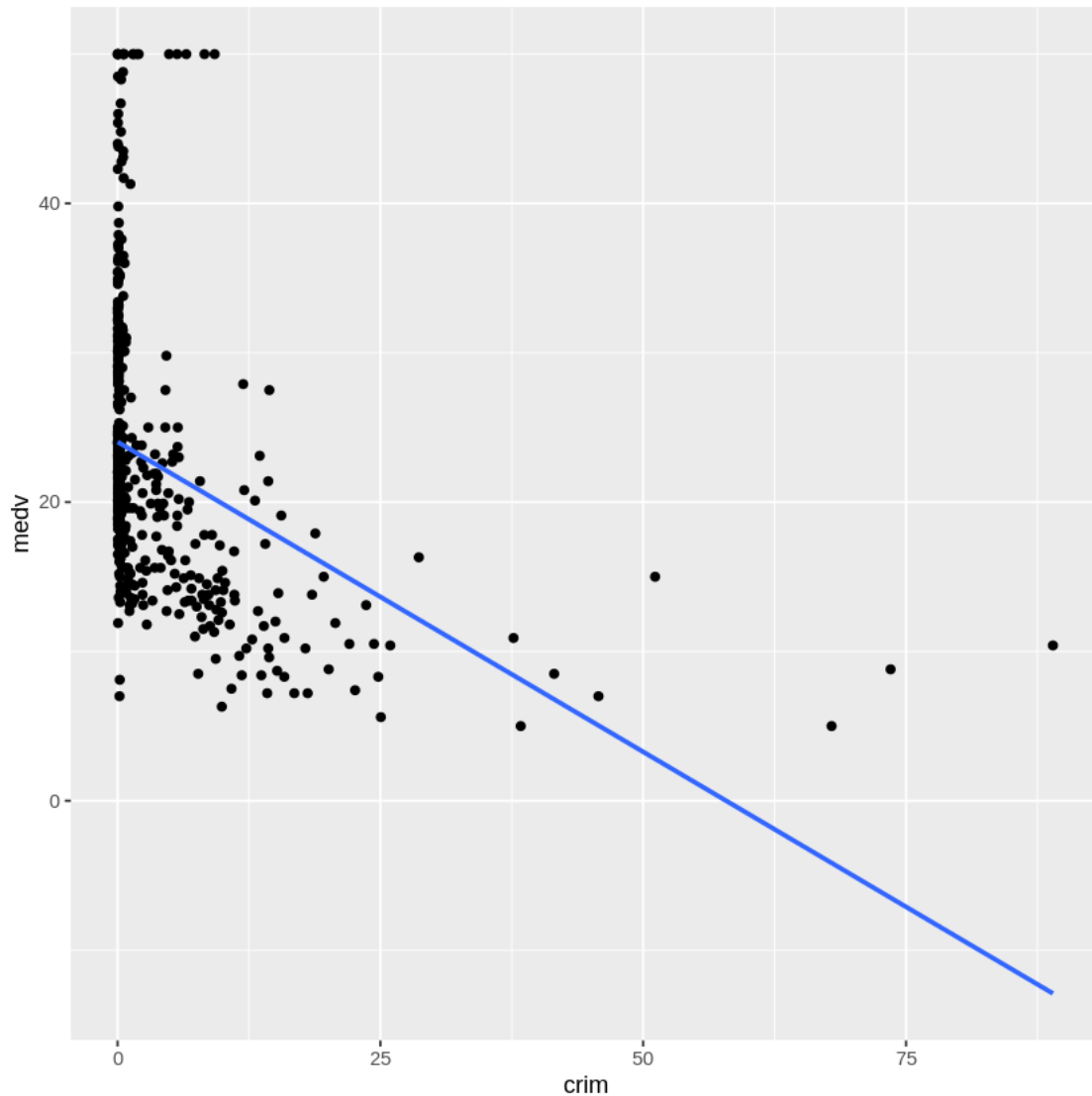
```
[5]: library(ggplot2)
library(MASS)
ggplot(data=Boston) + aes(x=rm, y=medv) + geom_point() +
  ↪geom_smooth(method="lm", se=FALSE)
```

`geom\_smooth()` using formula 'y ~ x'



```
[6]: ggplot(data=Boston) + aes(x=crim, y=medv) + geom_point() + └
↪geom_smooth(method="lm", se=FALSE)
```

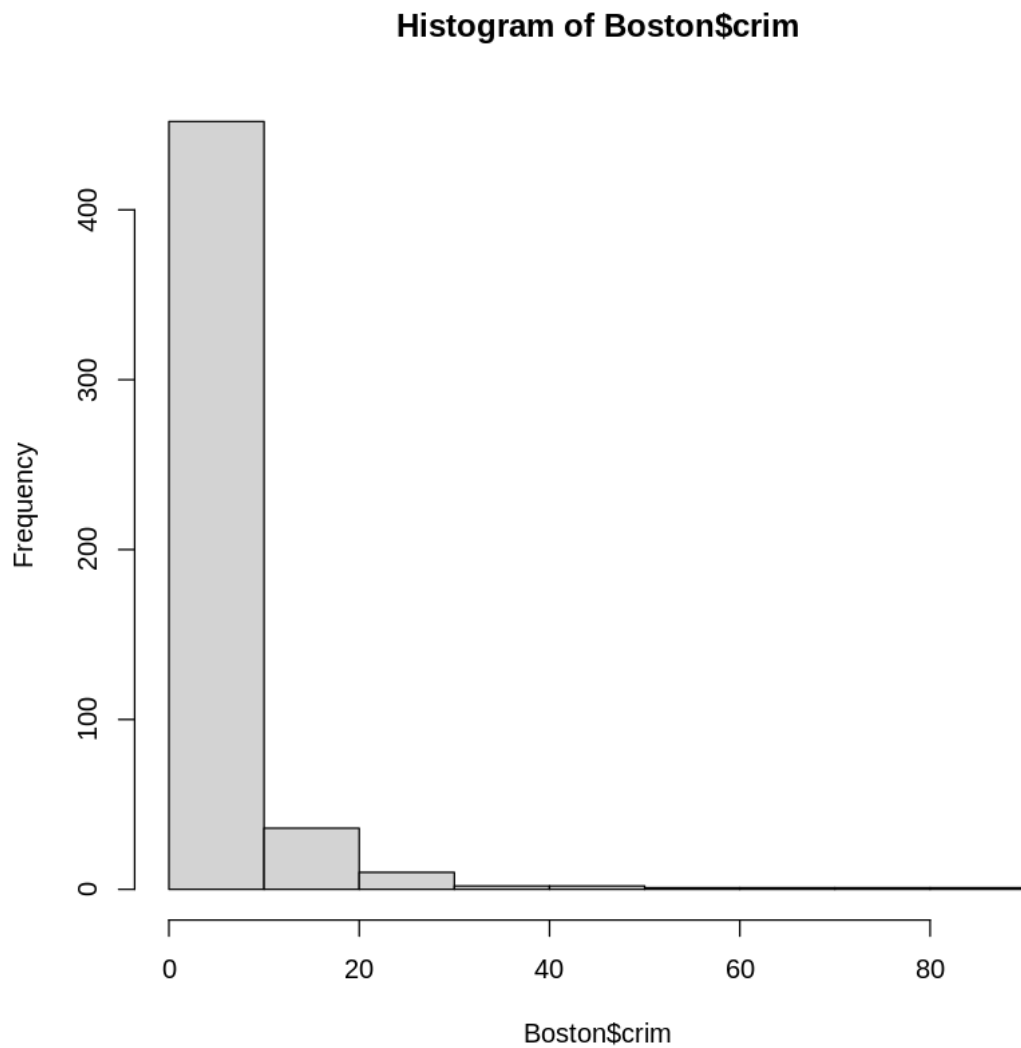
``geom_smooth()`` using formula 'y ~ x'



task 2

```
[7]: hist(Boston$crim)
```

*#In most places the crime rate is very low from 0 to 10, there are a couple of ↵*  
*↵places with the crime rate being from 10 to 30, but after that from 30 to 80 ↵*  
*↵there are incredibly few places with crime rates that high and these are the ↵*  
*↵outliers, especially those on the high end from 50 to 80 where there are ↵*  
*↵even less places then there are from 30 to 50 and the crime rate is ↵*  
*↵incredibly high in those places*



task 3

```
[27]: lmout<-lm(formula=medv ~ crim,data=Boston)
      summary(lmout)
```

Call:

```
lm(formula = medv ~ crim, data = Boston)
```

Residuals:

Min	1Q	Median	3Q	Max
-16.957	-5.449	-2.007	2.512	29.800

Coefficients:

```

              Estimate Std. Error t value Pr(>|t|)
(Intercept) 24.03311    0.40914   58.74  <2e-16 ***
crim         -0.41519    0.04389   -9.46  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 8.484 on 504 degrees of freedom
Multiple R-squared:  0.1508,    Adjusted R-squared:  0.1491
F-statistic: 89.49 on 1 and 504 DF,  p-value: < 2.2e-16

```

task 4

```
[28]: lmout2<-lm(formula=medv ~ crim + rm + dis,data=Boston)
      summary(lmout2)
```

Call:

```
lm(formula = medv ~ crim + rm + dis, data = Boston)
```

Residuals:

```

      Min       1Q   Median       3Q      Max
-21.247  -2.930  -0.572   2.390   39.072

```

Coefficients:

```

              Estimate Std. Error t value Pr(>|t|)
(Intercept) -29.45838    2.60010  -11.330  < 2e-16 ***
crim         -0.25405    0.03532   -7.193 2.32e-12 ***
rm           8.34257    0.40870   20.413  < 2e-16 ***
dis          0.12627    0.14382    0.878    0.38
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 6.238 on 502 degrees of freedom
Multiple R-squared:  0.5427,    Adjusted R-squared:  0.5399
F-statistic: 198.6 on 3 and 502 DF,  p-value: < 2.2e-16

```

task 5

```
[17]:
```

#The adjusted r-squared value is only about .54 which is about 55%, so that  
 ↳ indicates that there is some correlation but not necessarily that it is a  
 ↳ very high correlation send it is only just above 50%, but the r-squared  
 ↳ value needs to be examined on a case-by-case basis since there is no set  
 ↳ amount that is a good value, the P value was 2.2e-16 Which is quite good (I  
 ↳ did look up on Google if this was a good P value since I didn't know exactly  
 ↳ what it meant by 2.2e - 16) and it is below .05 which shows that there is  
 ↳ correlation and that this would be unlikely for the data to have been like  
 ↳ this if we knew that there was no factor that was causing them to be this  
 ↳ correlated, the coefficients are the intercept, crim rm and dis, we can see  
 ↳ that the P value for all of them except distance the last 1 are extremely  
 ↳ significant meaning that we can reject the null hypothesis, the book really  
 ↳ didn't go over the rest of the stuff for the coefficients so I Googled it,  
 ↳ the estimate gives the intercept and slopes, The standard error is about how  
 ↳ much the predicted values differ from the actual values and it seems like  
 ↳ that should be low if we want to say there is probably correlation, and all  
 ↳ the standard errors other than the intercept Are quite low so it again  
 ↳ supports the fact that there is a likely correlation, the t value is used to  
 ↳ calculate the P value and while I don't fully understand how to interpret it  
 ↳ since I know that the P values are very significant for each of them I think  
 ↳ I can safely say that the T values are all good enough to reject the null  
 ↳ hypothesis again with the exclusion of distance which has a bad p value

task 6

```
[29]: predDF <- data.frame(crim = 0.26, dis=3.2, rm=6.2)
```

task 7

```
[31]: predict(lmout2, predDF)
```

1: 22.6035475000185