

Homework 2

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2.

```
data("uswages")
?uswages
summary(uswages)
```

wage		educ		exper		race	
Min.	: 50.39	Min.	: 0.00	Min.	:-2.00	Min.	:0.000
1st Qu.:	308.64	1st Qu.:	12.00	1st Qu.:	8.00	1st Qu.:	0.000
Median :	522.32	Median :	12.00	Median :	15.00	Median :	0.000
Mean :	608.12	Mean :	13.11	Mean :	18.41	Mean :	0.078
3rd Qu.:	783.48	3rd Qu.:	16.00	3rd Qu.:	27.00	3rd Qu.:	0.000
Max.	:7716.05	Max.	:18.00	Max.	:59.00	Max.	:1.000

smsa		ne		mw		so	
Min.	:0.000	Min.	:0.000	Min.	:0.0000	Min.	:0.0000
1st Qu.:	1.000	1st Qu.:	0.000	1st Qu.:	0.0000	1st Qu.:	0.0000
Median :	1.000	Median :	0.000	Median :	0.0000	Median :	0.0000
Mean :	0.756	Mean :	0.229	Mean :	0.2485	Mean :	0.3125
3rd Qu.:	1.000	3rd Qu.:	0.000	3rd Qu.:	0.0000	3rd Qu.:	1.0000
Max.	:1.000	Max.	:1.000	Max.	:1.0000	Max.	:1.0000

we		pt	
Min.	:0.00	Min.	:0.0000
1st Qu.:	0.00	1st Qu.:	0.0000
Median :	0.00	Median :	0.0000
Mean :	0.21	Mean :	0.0925
3rd Qu.:	0.00	3rd Qu.:	0.0000
Max.	:1.00	Max.	:1.0000

```
m <- lm(wage ~ educ + exper, uswages)
summary(m)
```

Call:

```
lm(formula = wage ~ educ + exper, data = uswages)
```

Residuals:

Min	1Q	Median	3Q	Max
-1018.2	-237.9	-50.9	149.9	7228.6

```

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept) -242.7994    50.6816  -4.791 1.78e-06 ***
educ          51.1753     3.3419  15.313 < 2e-16 ***
exper         9.7748     0.7506  13.023 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 427.9 on 1997 degrees of freedom
Multiple R-squared:  0.1351,    Adjusted R-squared:  0.1343
F-statistic: 156 on 2 and 1997 DF,  p-value: < 2.2e-16

```

For every year of education the model estimates a \$51 per week increase in wage. For every year of experience the model estimates only \$10.

```
update(m, log(wage, 10) ~ .) %>% summary()
```

```

Call:
lm(formula = log(wage, 10) ~ educ + exper, data = uswages)

Residuals:
    Min       1Q   Median       3Q      Max
-1.19572 -0.15180  0.04639  0.19025  1.55037

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  2.019608    0.034029   59.35  <2e-16 ***
educ          0.039306    0.002244   17.52  <2e-16 ***
exper         0.007851    0.000504   15.58  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2873 on 1997 degrees of freedom
Multiple R-squared:  0.1749,    Adjusted R-squared:  0.174
F-statistic: 211.6 on 2 and 1997 DF,  p-value: < 2.2e-16

```

Log transformation of the response, means we interpret the coefficients as magnitude (multiplicative) changes. So for each year of education the model expects a 4% increase in wage. For each year of experience < 1%.

3.

```

x <- 1:20
y <- x + rnorm(20)

m <- lm(y ~ I(x^2))
summary(m)

```

Call:

```
lm(formula = y ~ I(x^2))
```

Residuals:

Min	1Q	Median	3Q	Max
-2.7803	-0.5854	-0.0946	0.9263	2.1197

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	4.424344	0.491408	9.003	4.38e-08 ***
I(x^2)	0.043815	0.002585	16.949	1.64e-12 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.441 on 18 degrees of freedom

Multiple R-squared: 0.941, Adjusted R-squared: 0.9378

F-statistic: 287.3 on 1 and 18 DF, p-value: 1.643e-12

```
direct_calc <- function(x, y, degree = 2) {  
  x_mat <- model.matrix(~I(x^degree))  
  solve(crossprod(x_mat), crossprod(x_mat, y))  
}
```

```
direct_calc(x, y)
```

```
      [,1]  
(Intercept) 4.42434396  
I(x^degree) 0.04381472
```

```
map(set_names(3:7), possibly(~ direct_calc(x, y, degree = .), "Error"))
```

```
$'3'
```

```
      [,1]  
(Intercept) 6.004637250  
I(x^degree) 0.002134748
```

```
$'4'
```

```
      [,1]  
(Intercept) 6.9137561565  
I(x^degree) 0.0001051108
```

```
$'5'
```

```
      [,1]  
(Intercept) 7.515077e+00  
I(x^degree) 5.183818e-06
```

```
$'6'
```

```
      [,1]  
(Intercept) 7.945381e+00  
I(x^degree) 2.556065e-07
```

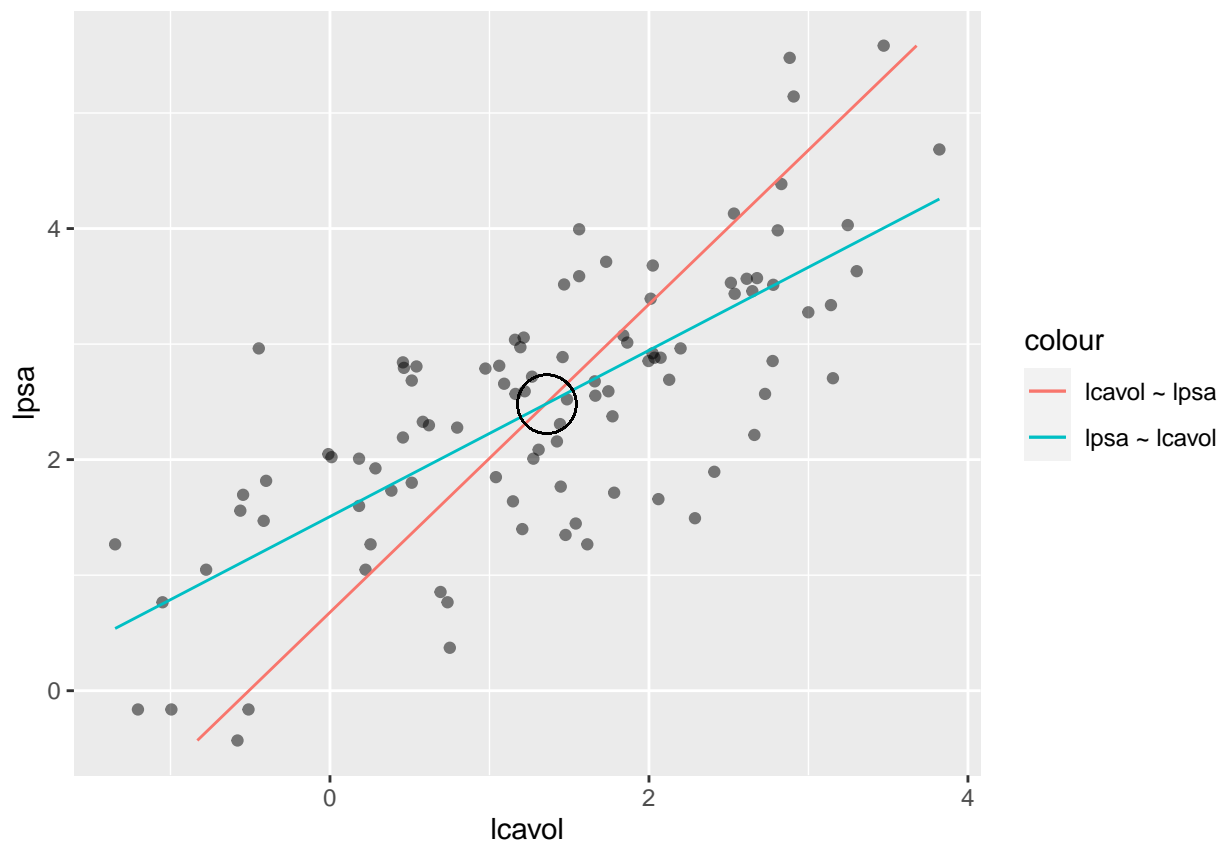
```
$'7'
```

```
[1] "Error"
```

5.

```
m <- lm(lcavol ~ lpsa, prostate)
m2 <- lm(lpsa ~ lcavol, prostate)

ggplot(prostate, aes(lcavol, lpsa)) +
  geom_point(alpha = .5) +
  geom_line(aes(x = predict(m), color = "lcavol ~ lpsa")) +
  geom_line(aes(y = predict(m2), color = "lpsa ~ lcavol")) +
  geom_point(aes(y = 2.48, x = 1.36), shape = 1, size = 10)
```



Algebra not shown but you can calculate the intersection if you solve for either `lcavol` or `lpsa` in the system of equations represented by the model coefficients.

6.

```
data("cheddar")
summary(cheddar)
```

taste	Acetic	H2S	Lactic
Min. : 0.70	Min. : 4.477	Min. : 2.996	Min. : 0.860

1st Qu.:13.55	1st Qu.:5.237	1st Qu.: 3.978	1st Qu.:1.250
Median :20.95	Median :5.425	Median : 5.329	Median :1.450
Mean :24.53	Mean :5.498	Mean : 5.942	Mean :1.442
3rd Qu.:36.70	3rd Qu.:5.883	3rd Qu.: 7.575	3rd Qu.:1.667
Max. :57.20	Max. :6.458	Max. :10.199	Max. :2.010

a.

```
m <- lm(taste ~ ., cheddar)
summary(m)
```

Call:

```
lm(formula = taste ~ ., data = cheddar)
```

Residuals:

Min	1Q	Median	3Q	Max
-17.390	-6.612	-1.009	4.908	25.449

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-28.8768	19.7354	-1.463	0.15540
Acetic	0.3277	4.4598	0.073	0.94198
H2S	3.9118	1.2484	3.133	0.00425 **
Lactic	19.6705	8.6291	2.280	0.03108 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 10.13 on 26 degrees of freedom

Multiple R-squared: 0.6518, Adjusted R-squared: 0.6116

F-statistic: 16.22 on 3 and 26 DF, p-value: 3.81e-06

b.

```
cor(m$fitted.values, cheddar$taste)^2
```

```
[1] 0.6517747
```

R-Squared, the coefficient of determination or percentage of variance explained

c.

```
m2 <- update(m, . ~ -1 + .)
cor(m2$fitted.values, cheddar$taste)^2
```

```
[1] 0.6244075
```

d.

```
m_mat <- model.matrix(m)

qr_decomp <- qr(m_mat)

backsolve(
  qr.R(qr_decomp), # upper-right
  t(qr.Q(qr_decomp)) %*% cheddar$taste
)
```

```
      [,1]
[1,] -28.8767696
[2,]  0.3277413
[3,]  3.9118411
[4,] 19.6705434
```