# Chapter 8: Interpreting a Linear Regression Model

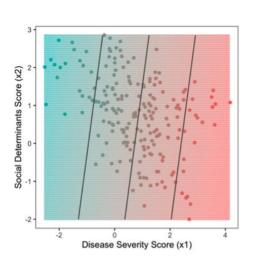
Modern Clinical Data Science Chapter Guides Bethany Percha, Instructor



# How to Use this Guide

- Read the corresponding notes chapter first
- Try to answer the discussion questions on your own
- Listen to the chapter guide (should be 15 min, max) while following along in the notes

What are the number of samples, n, and the number of predictors, p, for this dataset?



```
Call:
lm(formula = y \sim x1 + x2, data = df)
```

Residuals:

Min Median 3Q Max -11.9218 -3.1032 0.2891 2.8316 12.5813

#### Coefficients:

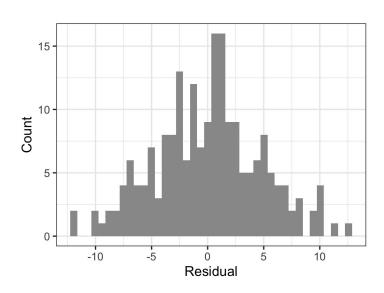
Estimate Std. Error t value Pr(>|t|)

(Intercept) 49.8600 0.5370 92.844 < 2e-16 \*\*\* 0.2855 36.555 < 2e-16 \*\*\* x1 10.4372 -1.8824 0.3609 -5.215 4.63e-07 \*\*\*

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

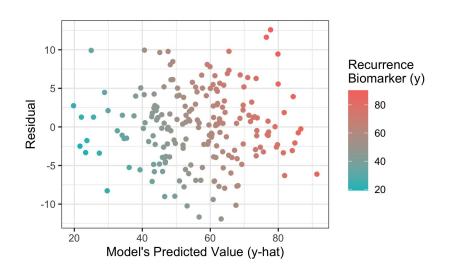
Residual standard error: 4.769 on 197 degrees of freedom Multiple R-squared: 0.9026, Adjusted R-squared: 0.9016 F-statistic: 912.4 on 2 and 197 DF, p-value: < 2.2e-16

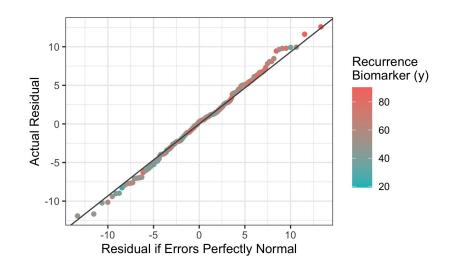
Estimate the maximum, minimum, and mean residuals from this graph. Do they match what is in the model output?



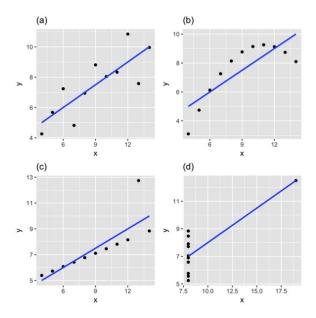
$$y = \beta_0 + \beta_1 x_1 + \cdots + \beta_p x_p + \varepsilon$$

Errors have constant variance (assume normal).

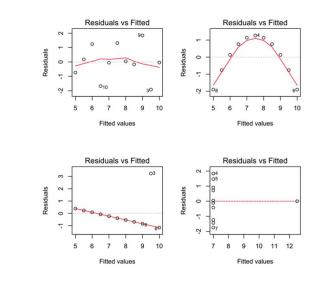




The four plots below show a famous dataset called **Anscombe's quartet**. The regression lines produced by fitting a linear regression model to each dataset are identical, but only one dataset actually fulfills the assumptions of a linear regression model.



We can check these assumptions by examining plots of the residuals vs. fitted values of the model (here the "fitted value" of point i means  $\hat{y}^{(i)}$ ).

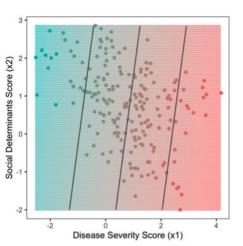


Which of the four datasets fulfills the assumption of a linear regression model that the error has constant variance? How can you tell?

Looking at the form of the linear regression model

$$y = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \varepsilon$$

what does the value of each of the  $\beta$ s mean? What is  $\beta_i$  telling us about how y varies with the predictor *j*, all else being equal?



```
Call:
lm(formula = y \sim x1 + x2, data = df)
```

Residuals: Min Median 30

Max -11.9218 -3.1032 0.2891 2.8316 12.5813

Coefficients:

Estimate Std. Error t value Pr(>|t|) (Intercept) 49.8600 0.5370 92.844 < 2e-16 \*\*\* 10.4372 0.2855 36.555 < 2e-16 \*\*\* -1.8824 0.3609 -5.215 4.63e-07 \*\*\*

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

Residual standard error: 4.769 on 197 degrees of freedom Multiple R-squared: 0.9026, Adjusted R-squared: 0.9016 F-statistic: 912.4 on 2 and 197 DF, p-value: < 2.2e-16

$$\hat{\sigma}^2 = \frac{1}{n-p-1} \sum_{i=1}^{n} (y^{(i)} - \hat{y}^{(i)})^2$$

Call:

 $lm(formula = y \sim x1 + x2, data = df)$ 

The sum of the squared residuals for our model is 4480.678. There are n=200 datapoints, and the number of predictors, p, is 2. Calculate  $\hat{\sigma}$  for this model. Do you see this number anywhere in the model output? What is it called?

```
Residuals:

Min 1Q Median 3Q Max
-11.9218 -3.1032 0.2891 2.8316 12.5813

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 49.8600 0.5370 92.844 < 2e-16 ***
x1 10.4372 0.2855 36.555 < 2e-16 ***
x2 -1.8824 0.3609 -5.215 4.63e-07 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error 4.769 on 197 degrees of freedom
Multiple R-squared: 0.9026, Adjusted R-squared: 0.9016
```

F-statistic: 912.4 on 2 and 197 DF, p-value: < 2.2e-16

Call:

 $lm(formula = y \sim x1 + x2, data = df)$ 

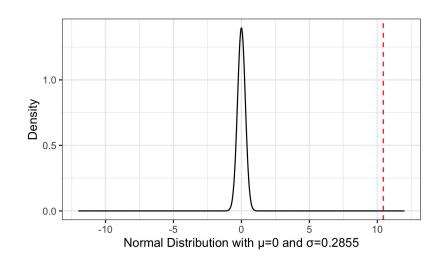
The standard errors attempt to capture how much we expect our estimates of the model coefficients to vary if we were to refit the model using a different dataset, provided that the new dataset is similar to (i.e., sampled from the same population distribution as) the one used to fit the model. On average, approximately how much would we expect  $\beta_0$  (the intercept) to deviate from its fitted value of 49.8600? How much would we expect  $\beta_1$  and  $\beta_2$  to deviate from their fitted values?

Residual standard error: 4.769 on 197 degrees of freedom Multiple R-squared: 0.9026, Adjusted R-squared: 0.9016 F-statistic: 912.4 on 2 and 197 DF, p-value: < 2.2e-16

Sketch the null distributions for the hypothesis tests of our three regression coefficients,  $\beta_0$ ,  $\beta_1$ , and  $\beta_2$ . Do you see why the *p*-values for these tests are so low?

```
Call:
lm(formula = y \sim x1 + x2, data = df)
Residuals:
    Min
              10 Median
                                       Max
-11.9218 -3.1032 0.2891 2.8316 12.5813
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 49.8600
                        0.5370 92.844 < 2e-16 ***
            10.4372
                        0.2855 36.555 < 2e-16 ***
x1
                        0.3609 -5.215 4.63e-07 ***
            -1.8824
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 4.769 on 197 degrees of freedom
```

Residual standard error: 4.769 on 197 degrees of freedom Multiple R-squared: 0.9026, Adjusted R-squared: 0.9016 F-statistic: 912.4 on 2 and 197 DF, p-value: < 2.2e-16



Interpret the values of each of these coefficients. Based on the coefficient values and their standard errors, which predictor(s) do you think have the greatest impact on mortality?

MORT	Total age-adjusted mortality from all causes,
	in deaths per 100,000 population
PRECIP	Mean annual precipitation (in inches)
EDUC	Median number of school years completed
	for persons of age 25 years or older
NONWHITE	Percentage of the 1960 population that is nonwhite
NOX	Relative pollution potential of oxides of nitrogen
SO2	Relative pollution potential of sulfur dioxide

```
Call:
lm(formula = MORT ~ PRECIP + EDUC + NONWHITE + NOX + SO2, data = d)
Residuals:
  Min
          10 Median
                            Max
-91.38 -18.97 -3.56 16.00 91.83
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 995.63646 91.64099 10.865 3.35e-15 ***
PRECIP
            1.40734 0.68914 2.042 0.046032 *
EDUC
           -14.80139 7.02747 -2.106 0.039849 *
NONWHITE
         3.19909
                       0.62231 5.141 3.89e-06 ***
NOX
            -0.10797
                       0.13502 -0.800 0.427426
S02
            0.35518
                       0.09096
                                3.905 0.000264 ***
```

Residual standard error: 37.09 on 54 degrees of freedom Multiple R-squared: 0.6746, Adjusted R-squared: 0.6444 F-statistic: 22.39 on 5 and 54 DF, p-value: 4.407e-12

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

In this model, is the effect of one predictor (say, PRECIP) impacted by the value(s) of any of the other predictor(s)? How does this differ from the other regression algorithms we've seen (KNN and decision trees)? What are the advantages and disadvantages of this choice?

#### **Question 8.10**

Is a normal distribution the right distribution to model an outcome of ageadjusted mortality (MORT)? Why or why not? Look back at our discussion of the normal distribution in Chapter 4 if you need a refresher.