# Week 3: Basic regression

## 2. Introduction to linear model

Stat 140 - 04

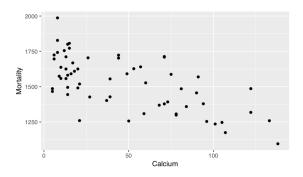
Mount Holyoke College

- 1. Line of best fit
- 2. Finding the least square line in R
- 3. Prediction
- 4. How useful is the model?

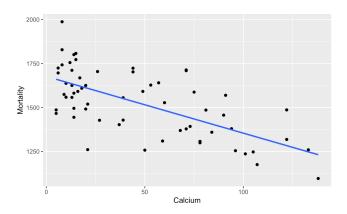
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Scientists believe that water with high concentrations of calcium and magnesium is beneficial for health.

We have recordings of the mortality rate (deaths per 100,000 population) and concentration of calcium in drinking water (parts per million) in 61 large towns in England and Wales.



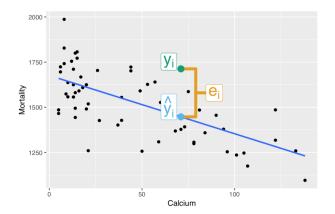
- Response variable: variable whose behavior or variation you are trying to understand, on the y-axis (dependent variable)
- ▶ **Explanatory variables**: other variables that you want to use to explain the variation in the response, on the x-axis (independent variables); these are also referred to as predictors or features.



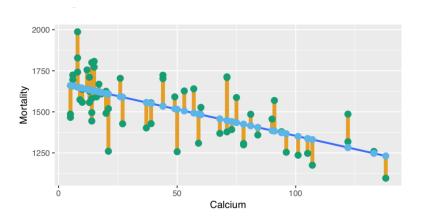
What is the best line of fit? What qualities are important here?

### Residual = Observed - Predicted

- 1. **Predicted value**  $\hat{y}$ : estimate made from a model
- 2. **Observed value** *y*: value in the dataset



The line of best fit is the line for which the sum of the squared residuals is smallest, the **least squares line**.



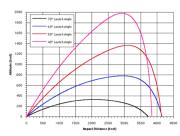
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The algebraic equation for a line is

$$Y = b_0 + b_1 X$$

The use of coordinate axes to show functional relationships was invented by Rene Descartes (1596-1650). He was an artillery officer, and probably got the idea from pictures that showed the trajectories of cannonballs.





General form of 'lm' command:

 $Im(y_variable \sim x_variable, data = data_frame)$ 

Use this to estimate the intercept and the slope of line in the Mortality/Health data.

 ${\sf linear\_fit} \leftarrow {\sf Im}({\sf Mortality} \sim {\sf Calcium}, \, {\sf data} = {\sf mortality\_water})$ 

### Coefficients:

(Intercept) Calcium 1676.356 -3.226 From the coefficients, we can write the regression line in the Mortality/Health data as

$$\widehat{\text{Mortality}} = 1676 - 3 \text{ Calcium}$$

Abstractly,

$$\hat{y} = 1676 - 3x$$

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By hand:

$$\widehat{\text{Mortality}} = 1676 - 3 \times 71 = 1463$$

The predicted value for the mortality rate in that town is 1463 deaths per 100,000 population.

## By R:

The general form of 'predict' command:

```
predict(linear_model, newdata = data_frame)
```

### Code:

```
linear_fit <- lm(Mortality ~ Calcium, data = mortality_water)
predict_data <- data.frame( Calcium = 71)
predict(linear_fit, newdata = predict_data)</pre>
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## Output:

1

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The outputs by hand and by R are different because of rounding errors.

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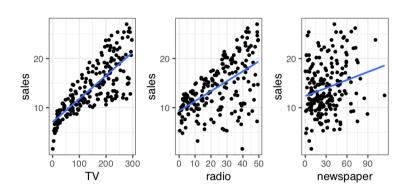
We have a data set of 200 markets, and we are interested in the relationship between sales and advertising budget. We look at the following variables.

- **sales** is a measure of sales volume in thousands of units
- ▶ **TV** is TV advertising budget
- ▶ radio is radio advertising budget
- newspaper is newspaper advertising budget

### Poll question

Which of these models would you prefer to use for predicting sales?

- TV
- 6 radio
- newspaper

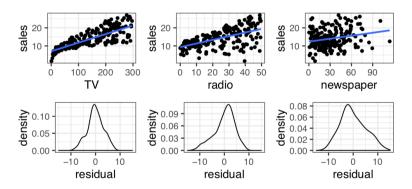


Being as specific and concrete as possible, write down a rule for selecting your preferred model

- 1. based only on **visual characteristics** of the plot.
- based only on a quantitative summary of the data. You
  can describe how you would calculate your numeric
  summary of the data in a general sense; if you'd like you
  can write down a formula.

### Residuals:

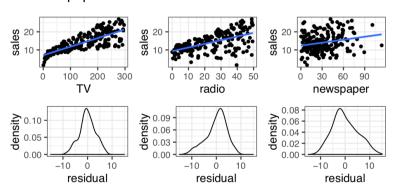
- $ightharpoonup e_i = y_i \hat{y}_i$  (vertical distance between point and line)
- ▶ Smaller residuals mean the predictions were better.
- ▶ The key is to measure the spread of residuals.



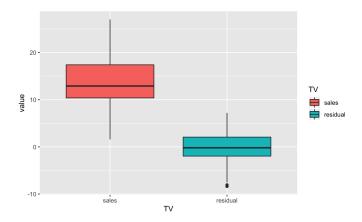
Measure spread of residuals with the standard deviation. We call this the **residual standard error**,  $s_{\mathsf{RES}}$ .

TV: 3.26radio: 4.28

newspaper: 5.09



The variability in the residuals describes how much variation remains after using the model



Let's compute the reduction in variation.

$$\frac{s_{\mathsf{sales}}^2 - s_{\mathsf{RES}}^2}{s_{\mathsf{sales}}^2} = 0.61$$

This number describes the amount of variation in the y-variable that is explained by the least squares line.

An value of 61% indicates that 61% of the variation in sales can be accounted for by the TV advertisement budget.

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How do we get 61%?

Statisticians found this value is  $R^2$ , the **square of Correlation**.

Square of the correlation coefficient R: between 0 and 1, closer to 1 is better.

 ${\cal R}^2$  describes the amount of variation in the y-variable that is explained by the least squares line.

► TV: 0.61

▶ radio: 0.33

▶ newspaper: 0.05

meaning, 61% of the variation in sales can be accounted for by the TV advertisement budget; 33% of the variation in sales can be accounted for by the radio advertisement budget; 5% of the variation in sales can be accounted for by the newspaper advertisement budget.

```
##
## Call:
## lm(formula = Mortality ~ Calcium, data = mortality water)
##
## Residuals:
##
      Min
                10 Median
                                3Q
                                       Max
## -348.61 -114.52 -7.09 111.52 336.45
##
                   b0 intercept
## Coefficients:
                                                      Useful later
##
                Estimate Std. Error t value Pr(>|t|)
                                     57.217 < 2e-16 ***
## (Intercept) 1676.3556
                            29.2981
## Calcium
                 -3.2261
                             0.4847 -6.656 1.03e-08 ***
## ---
                b1 slope
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
                                R squared
## Residual standard error: 143 on 59 degrees of freedom
## Multiple R-squared: 0.4288, Adjusted R-squared: 0.4191
## F-statistic: 44.3 on 1 and 59 DF, p-value: 1.033e-08
```

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Interpretation is the most important thing in this module.

- slope
- intercept
- residual
- $ightharpoonup R^2$