# Week 5: Linear regression

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### By the end of this lab you should know

- How to use lm to get estimated coefficients and  $R^2$
- How to calculate an estimated outcome (Y) based on a particular value of an independent variable (X) and estimated regression coefficients
- How to plot a fitted SLR line on a scatter plot

### Read in, prepare, plot the data

We will be using the country indicators dataset again, and exploring the relationship between the total fertility rate (TFR) and child mortality in 2017.

```
library(tidyverse)
library(here)
country_ind <- read_csv(here("data/country_indicators.csv"))

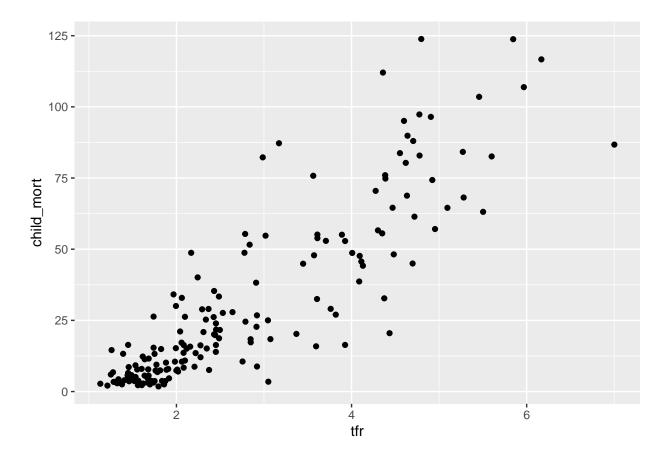
# NOTE if you are having trouble with the 'here' package
# don't use it and just type in the whole file path.</pre>
```

### Filter to just be 2017

```
country_ind_2017 <- country_ind %>% filter(year==2017)
```

### Look at the observed relationship between TFR and child mortality

```
ggplot(country_ind_2017, aes(tfr, child_mort)) +
  geom_point()
```



### Question

Alter the code above to make a plot title and make the X and Y axes more readable.

## Estimating SLR using 1m

We don't have to calculate the regression coefficients or  $\mathbb{R}^2$  'by hand', we can just use the  $\operatorname{lm}$  function. ('lm' stands for 'linear models').

The main arguments are

- The formula, which is written in the form y~x
- The data frame that contains the variables

Fit our SLR:

```
childmort_tfr_model <- lm(formula = child_mort~tfr, data = country_ind_2017)</pre>
```

Print out the summary:

```
summary(childmort_tfr_model)
```

```
##
## Call:
## lm(formula = child_mort ~ tfr, data = country_ind_2017)
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
  -43.937 -7.093 -0.558
                            5.404 52.029
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -25.8338
                           2.6135
                                   -9.885
                                            <2e-16 ***
                           0.8579 23.730
               20.3581
                                            <2e-16 ***
## tfr
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 14.73 on 174 degrees of freedom
## Multiple R-squared: 0.7639, Adjusted R-squared: 0.7626
## F-statistic: 563.1 on 1 and 174 DF, p-value: < 2.2e-16
```

Confirm that the resulting values are the same as the ones you obtained by doing the calculations 'by hand'.

#### Extract results

To extract coefficients from the model output, use the coef() function

Can assign these to variables by indexing the relevant number:

```
beta_0 <- coef(childmort_tfr_model)[1] # the [1] means get the first item
beta_1 <- coef(childmort_tfr_model)[2]</pre>
```

To extract the value of  $\mathbb{R}^2$ , use

```
summary(childmort_tfr_model)[["r.squared"]]
```

```
## [1] 0.7639494
```

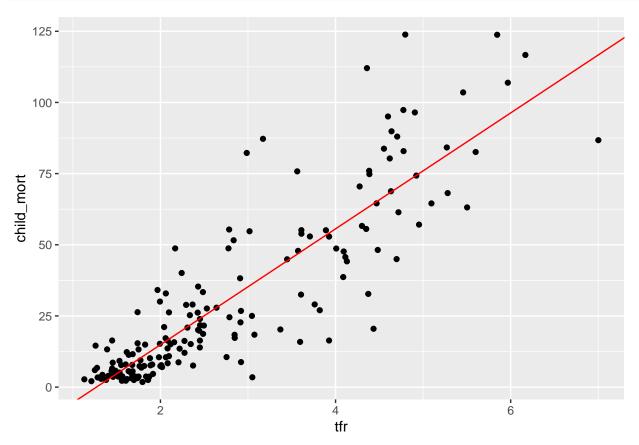
### Question

What is the estimated child mortality for a country with a TFR or 5?

## Plotting the fitted line on a scatter plot

To visualize our fitted line we can add to our plot from before using geom\_abline

```
ggplot(country_ind_2017, aes(tfr, child_mort)) +
  geom_point() +
  geom_abline(intercept = beta_0, slope = beta_1, color = "red")
```



## Multiple linear regression

Running MLR in R is an easy extension of SLR. Here are some practice questions using the lego\_towers dataset. This dataset shows observations of the time (in seconds) it took my toddler to build a lego tower, the number of blocks given to him, and the number of other distractions present.

Reading in the data:

```
lego <- read_csv(here("data/lego_towers.csv"))</pre>
```

### Questions

- 1. Make a scatter plot of time versus blocks
- 2. Make a scatter plot of time versus distractions
- 3. Based on 1 and 2, what do you expect the magnitude and sign (direction) of  $\hat{\beta}_1$  and  $\hat{\beta}_2$  to be?
- 4. Fit the above model using 1m
- 5. Interpret  $\hat{\beta}_1$  and  $\hat{\beta}_2$
- 6. Using mutate create a new variable called blocks\_3 which is the number of blocks minus 3
- 7. Refit the model using lm where  $X_{i1}$  is now  $blocks_3$
- 8. Interpret  $\hat{\beta_0}$