# Regression with categorical variables

# Packages for this section

```
library(tidyverse)
library(broom)
```

### The pigs revisited

• Recall pig feed data, after we tidied it:

```
my_url <- "http://ritsokiguess.site/datafiles/pigs2.txt"
pigs <- read_delim(my_url, " ")
pigs</pre>
```

```
# A tibble: 20 x 3
    pig feed weight
  <dbl> <dbl> <dbl>
      1 feed1 60.8
      2 feed1 57
      3 feed1 65
    4 feed1 58.6
5
      5 feed1 61.7
6
      1 feed2 68.7
      2 feed2 67.7
      3 feed2 74
      4 feed2 66.3
10
      5 feed2 69.8
11
      1 feed3
               92.6
```

#### **Summaries**

4 feed4

1 feed1 5 60.6 3.06 2 feed2 5 69.3 2.93 3 feed3 5 94.1 3.61

5 86.2 2.90

### Running through aov and 1m

- What happens if we run this through 1m rather than aov?
- Recall aov first:

```
pigs.1 <- aov(weight ~ feed, data = pigs)
summary(pigs.1)</pre>
```

```
Df Sum Sq Mean Sq F value Pr(>F)

feed 3 3521 1173.5 119.1 3.72e-11 ***

Residuals 16 158 9.9

---

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

#### and now lm

```
pigs.2 <- lm(weight ~ feed, data = pigs)</pre>
summary(pigs.2)
Call:
lm(formula = weight ~ feed, data = pigs)
Residuals:
  Min 10 Median 30 Max
-3.900 -2.025 -0.570 1.845 5.000
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 60.620 1.404 43.190 < 2e-16 ***
feedfeed2 8.680 1.985 4.373 0.000473 ***
feedfeed3 33.480 1.985 16.867 1.30e-11 ***
feedfeed4 25.620 1.985 12.907 7.11e-10 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 3.138 on 16 degrees of freedom

# Understanding those slopes

- Get one slope for each category of categorical variable feed, except for first.
- feed1 treated as "baseline", others measured relative to that.
- Thus prediction for feed 1 is intercept, 60.62 (mean weight for feed 1).
- Prediction for feed 2 is 60.62 + 8.68 = 69.30 (mean weight for feed 2).
- Or, mean weight for feed 2 is 8.68 bigger than for feed 1.
- Mean weight for feed 3 is 33.48 bigger than for feed 1.
- Slopes can be negative, if mean for a feed had been smaller than for feed 1.

### Reproducing the ANOVA

Analysis of Variance Table

Pass the fitted model object into anova:

```
anova(pigs.2)
```

```
Response: weight

Df Sum Sq Mean Sq F value Pr(>F)

feed 3 3520.5 1173.51 119.14 3.72e-11 ***

Residuals 16 157.6 9.85

---

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

- Same as before.
- But no Tukey this way:

```
TukeyHSD(pigs.2)
```

Error in UseMethod("TukeyHSD"): no applicable method for 'TukeyHSD' applied to an o

#### The crickets

- Male crickets rub their wings together to produce a chirping sound.
- Rate of chirping, called "pulse rate", depends on species and possibly on temperature.
- Sample of crickets of two species' pulse rates measured; temperature also recorded.
- Does pulse rate differ for species, especially when temperature accounted for?

#### The crickets data

#### Read the data:

```
my_url <- "http://ritsokiguess.site/datafiles/crickets2.csv"
crickets <- read_csv(my_url)
crickets %>% slice_sample(n = 10) # display sample of rows
```

#### # A tibble: 10 x 3 species temperature pulse\_rate <chr>> <dbl> <dbl> 1 exclamationis 26.2 89.1 2 exclamationis 26.2 87.5 3 exclamationis 30.4 99.3 4 exclamationis 65.1 20.8 5 exclamationis 20.8 67.9 58.5 6 niveus 21 76.1 7 niveus 26.5 8 exclamationis 30.4 102. 51.8 9 niveus 18.9 28.6 84.7 10 niveus

### Fit model with 1m

# Can I remove anything? No:

```
drop1(crickets.1, test = "F")
```

Single term deletions

drop1 is right thing to use in a regression with categorical (explanatory) variables in it: "can I remove this categorical variable as a whole?"

### The summary

summary(crickets.1)

```
Call:
lm(formula = pulse_rate ~ temperature + species, data = crickets)
Residuals:
   Min 10 Median 30
                              Max
-3.0128 -1.1296 -0.3912 0.9650 3.7800
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -7.21091 2.55094 -2.827 0.00858 **
temperature 3.60275 0.09729 37.032 < 2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.786 on 28 degrees of freedom
Multiple R-squared: 0.9896, Adjusted R-squared: 0.9888
F-statistic: 1331 on 2 and 28 DF, p-value: < 2.2e-16
```

#### Conclusions

- Slope for temperature says that increasing temperature by 1 degree increases pulse rate by 3.6 (same for both species)
- Slope for speciesniveus says that pulse rate for niveus about 10 lower than that for exclamationis at same temperature (latter species is baseline).
- R-squared of almost 0.99 is very high, so that the prediction of pulse rate from species and temperature is very good.

### To end with a graph

- Two quantitative variables and one categorical: scatterplot with categories distinguished by colour.
- This graph seems to need a title, which I define first.

```
t1 <- "Pulse rate against temperature for two species of crick
t2 <- "Temperature in degrees Celsius"
ggplot(crickets, aes(x = temperature, y = pulse_rate,
    colour = species)) +
   geom_point() + geom_smooth(method = "lm", se = FALSE) +
   ggtitle(t1, subtitle = t2) -> g
```

# The graph

g

