# Simple Linear Regression

Inference & Prediction

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#### **Announcements**

- HW 01 due Wednesday, 9/18 at 11:59p
- Reading 01: ANOVA
- Use Piazza for questions instead of email
  - access it through Sakai
  - feel free to reply if you know the answer to question
  - let me know if you're not on Piazza



### Check in

Any questions from last class?



### Today's Agenda

- Inference for regression
- Prediction
- Cautions



### Packages and Data

```
library(tidyverse)
library(broom)
library(knitr)
library(MASS) #cats dataset
```



#### Cats!

- When veterinarians prescribe heart medicine for cats, the dosage often needs to be calibrated to the weight of the heart.
- It is very difficult to measure the heart's weight, so veterinarians need a way to estimate it.
- One way to estimate it is using a cat's body weight which is more feasible to obtain (though still difficult depending on the cat!).
- **Goal**: Fit a regression model that describes the relationship between a cat's heart weight and body weight.



#### The Data

We will use the **cats** dataset from the MASS package. It contains the following characteristics for 144 cats:

```
Sex: Male (M) or Female (F)
```

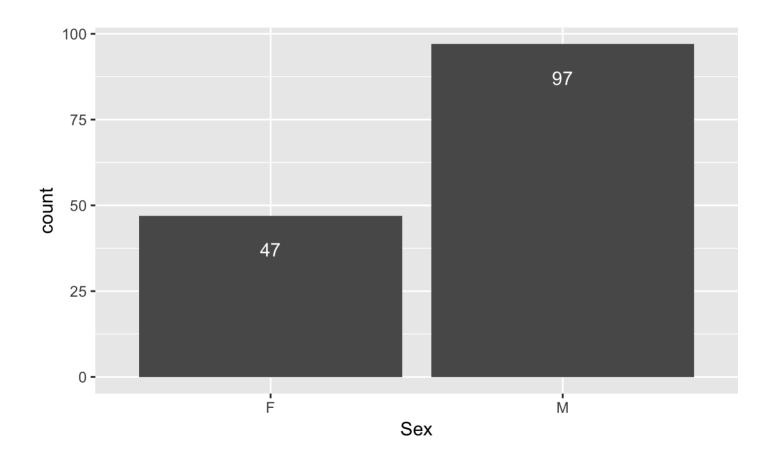
- Bwt: Body weight in kilograms (kg)
- Hwt: Heart weight in grams (g)

```
cats %>% slice(1:10)
```

```
Sex Bwt Hwt
##
## 1
    F 2.0 7.0
    F 2.0 7.4
## 2
## 3
    F 2.0 9.5
## 4
    F 2.1 7.2
## 5
    F 2.1 7.3
     F 2.1 7.6
## 6
## 7
     F 2.1 8.1
## 8
    F 2.1 8.2
    F 2.1 8.3
## 9
##
  10
       F 2.1 8.5
```





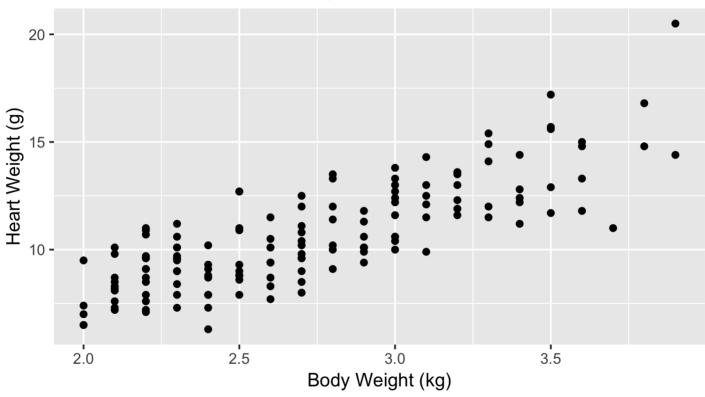




```
## Skim summary statistics
##
   n obs: 144
##
  n variables: 3
##
## — Variable type:factor -
   variable missing complete n n_unique top_counts ordered
##
                    144 144 2 M: 97, F: 47, NA: 0
##
        Sex
                 0
                                                           FALSE
##
## — Variable type:numeric -
   variable missing complete n mean sd p0 p25 p50 p75 p100
##
##
        Bwt
                 0 144 144 2.72 0.49 2 2.3 2.7 3.02 3.9
                 0 144 144 10.63 2.43 6.3 8.95 10.1 12.12 20.5
##
       Hwt
##
       hist
##
##
```



#### Heart Weight vs. Body Weight





### **Application Exercise**

- Make a copy of the cats project on RStudio Cloud.
- Work with your lab groups to complete Part 2: Fit the Model & Check Assumptions
- Put your name at the top of the document. You can put everyone's name on the same document if you're working off of one computer.
- We'll look at one group's Rmd file and discuss as a class after about 10 minutes.



# Is there truly a linear relationship between the response and predictor variables?



### Recall: Outline of Hypothesis Test

- 1. State the hypotheses
- 2. Calculate the test statistic
- 3. Calculate the p-value
- 4. State the conclusion in the context of the problem



### 1. State the hypotheses

- We are often interested in testing whether there is a significant linear relationship between the predictor and response variables
- If there is truly no linear relationship between the two variables, the population regression slope,  $\beta_1$ , would equal 0
- We can test the hypotheses:

$$H_0: \beta_1 = 0$$

$$H_a:\beta_1\neq 0$$

■ This is the test conducted by the lm() function in R



#### 2. Calculate the test statistic

$$H_0: \beta_1 = 0$$

$$H_a: \beta_1 \neq 0$$

#### **Test Statistic:**

test statistic = 
$$\frac{\text{Estimate} - \text{Hypothesized}}{SE}$$

$$=\frac{\hat{\beta}_1 - 0}{SE(\hat{\beta}_1)}$$



### 3. Calculate the p-value

**p-value** is calculated from a t distribution with n-2 degrees of freedom

p-value = 
$$P(t \ge |\text{test statistic}|)$$



#### 4. State the conclusion

Magnitude of p-value	Interpretation	
p-value < 0.01	strong evidence against $H_{ m 0}$	
0.01 < p-value < 0.05	moderate evidence against $H_0$	
0.05 < p-value < 0.1	weak evidence against $H_{ m 0}$	
p-value > 0.1	effectively no evidence against $H_0$	

**Note:** These are general guidelines. The strength of evidence depends on the context of the problem.



### Cats!: Hypothesis test for $\beta_1$

- Refer back to the cats application exercise to answer the following questions:
- a. State the hypotheses in (1) words and (2) statistical notation.
- b. What is the meaning of the test statistic in the context of the problem?
- c. What is the meaning of the p-value in the context of the problem?
- d. State the conclusion in context of the problem.



### **Predictions**



#### **Predictions for New Observations**

• We can use the regression model to predict for a response at  $x_0$ 

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x_0$$

■ Because the regression models produces the mean response for a given value of  $x_0$ , it will produce the same estimate whether we want to predict the mean response at  $x_0$  or an individual response at  $x_0$ 



### Predicting Mindy's heart weight

My cat Mindy weighs about 3.18 kg (7 lbs).

What is her predicted heart weight?



$$hwt = -0.3567 + 4.0341 \times 3.18$$
  
= 12.472 grams



### Uncertainty in predictions

- There is uncertainty in our predictions, so we need to calculate an a standard error (SE) to capture the uncertainty
- The SE is different depending on whether you are predicting an average value or an individual value
- SE is larger when predicting for an individual value than for an average value



### Standard errors for predictions

#### Predicting the mean response

$$SE(\hat{\mu}) = \hat{\sigma} \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{\sum_{i=1}^{n} (x_i - \bar{x})^2}}$$

#### Predicting an individual response

$$SE(\hat{y}) = \hat{\sigma} \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{\sum_{i=1}^{n} (x_i - \bar{x})^2}}$$



### Predicting Mindy's heart weight

 Calculate a 95% confidence interval for Mindy's predicted heart weight.

```
## fit lwr upr
## 1 12.47166 12.14269 12.80063
```

■ Calculate a 95% prediction interval for the predicted mean heart weight for the subset of cats who weigh 3.18 kg.



```
## fit lwr upr
## 1 12.47166 9.581804 15.36151
```

# **Cautions**



### **Caution: Extrapolation**

- The regression is only useful for predictions for the response variable *y* in the range of the predictor variable *x* that was used to fit the regression
- It is risky to predict far beyond that range of x, since you don't have data to tell whether or not the relationship continues to follow a straight line



## **Caution: Extrapolation**

My cat Andy weighs about 8.60 kg (10 lbs).

Should we use this regression model to predict his heart weight?



min	q1	median	q3	max
2	2.3	2.7	3.025	3.9

The heaviest cat in this dataset weighs 3.9 kg (8.6 lbs). We should <u>not</u> use this model to predict Andy's heart weight, since that would be a case of **extrapolation**.



### Caution: Correlation $\neq$ Causation

- The regression model is <u>not</u> a statement of causality
  - The regression model provides a description of the averages of *Y* for different values of *X*
- The regression model alone <u>cannot</u> prove causality. You need either
  - Randomized experiment
  - Observational study in which all relevant confounding variables are controlled for adequately

