

# Multinomial logistic regression

# Agenda

- + No homework this week
- + Extra credit opportunity: department seminar
  - + Dr. Mine Cetinkaya-Rundel
  - + Monday, September 26, 12pm - 1pm in Kirby 120
- + Today: multinomial logistic regression

# Motivation

- + When the response is binary, we use *logistic regression*
- + What happens when the response is categorical, but has MORE than 2 categories?
- + We use *multinomial logistic regression* (aka *multinomial regression*)

# Motivation

**Question:** What is the relationship between age and contraceptive use for women in Indonesia?

**Data:** 1473 Indonesian couples, with variables

- +  $Y_i$  = contraceptive method used (1 = no use, 2 = long-term, 3 = short-term)
- +  $X_i$  = Wife's age (numeric)

## The response variable

Contraception	Freq
Long	511
None	629
Short	333

- +  $n_{None} = 629$  (this is 42.7% of the couples)
- +  $n_{Long} = 511$  (this is 34.7% of the couple)
- +  $n_{Short} = 333$  (this is 22.6% of the couples)

# The response variable

$Y_i$  = contraceptive method used (1 = no use, 2 = long-term, 3 = short-term)

What type of variable is  $Y$ ?

# Parametric model building

What are our two steps in building a parametric model?

# Building a distribution

$Y_i$  = contraceptive method used (1 = no use, 2 = long-term, 3 = short-term)

What notation might we use for the probability of no contraceptive use?



# Building a distribution

$Y_i$  = contraceptive method used (1 = no use, 2 = long-term, 3 = short-term)

$$+ \pi_{i(\\text{None})} = P(Y_i = \\text{None})$$

$$+ \pi_{i(\\text{Short})} = P(Y_i = \\text{Short})$$

$$+ \pi_{i(\\text{Long})} = P(Y_i = \\text{Long})$$

What must be true of the three probabilities?

# The Categorical distribution

**Definition:** Let  $Y_i$  be an **unordered** categorical variable with  $J$  levels  $j = 1, \dots, J$ . Let  $\pi_j = P(Y_i = j)$ , where  $\pi_j \in [0, 1]$  for all  $j$ , and  $\sum_{j=1}^J \pi_j = 1$ .

Then we say  $Y_i \sim \text{Categorical}(\pi_1, \dots, \pi_J)$ .

- + We can use this distribution as the first step in our modeling process!

What distribution does our response (contraceptive use) have?

# Parametric model building

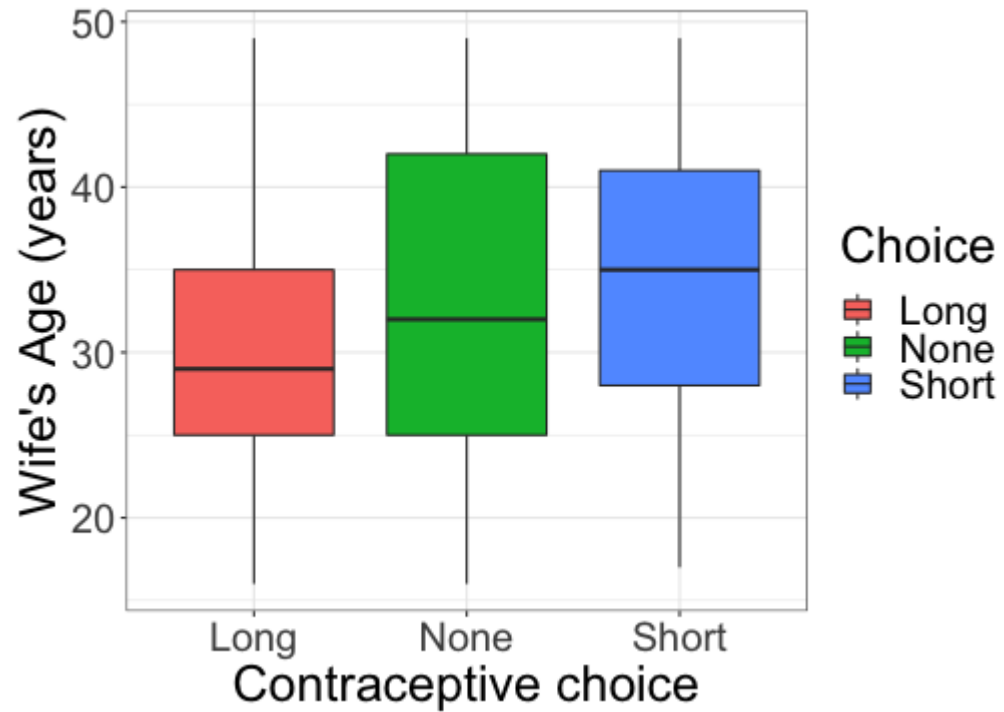
Step 1: Choose a reasonable distribution for  $Y$

$$Y_i \sim \text{Categorical}(\pi_{i(\text{None})}, \pi_{i(\text{Short})}, \pi_{i(\text{Long})})$$

Step 2: Choose a model for any parameters

+ Need to relate our probabilities to  $X = \text{Age}$

# EDA



# EDA

- + Boxplots show there may be some differences with age, but don't let us model the relationship
- + We want something like an empirical log odds plot

Can we use the log odds here?

# Relative risk

- + If  $Y_i$  is *binary*, the odds  $\frac{\pi_i}{1 - \pi_i}$  compare the probabilities of the two possible outcomes
- + If  $Y_i$  has more than two outcomes, we need to generalize the odds
- + The *relative risk* compares the probabilities of two potential outcomes

**Relative risk of long term vs. no contraceptive use:**

**Relative risk of short term vs. no contraceptive use:**

## Example

Consider the 48 twenty-one year old wives in our data:

- + Long: 23
- + Short: 6
- + None: 19

For a 21 year old, what is the *empirical* relative risk of using long term vs. short term contraceptives?

# Relative risk

**Definition:** Let  $Y_i$  be a categorical variable with  $J$  levels  $j = 1, \dots, J$ . Let  $\pi_j = P(Y_i = j)$ . Then the relative risk of level  $j$  vs. level  $k$  is

$$\frac{\pi_{ij}}{\pi_{ik}}$$



# Class activity, Part I

[https://sta214-f22.github.io/class\\_activities/ca\\_lecture\\_14.html](https://sta214-f22.github.io/class_activities/ca_lecture_14.html)

## Class activity

Speed Range	Slow	Good	Fast	Total
(50, 51)	5	1	0	6
(51, 52)	5	5	3	13
(52, 53)	6	12	2	20
(53, 54)	5	31	4	40

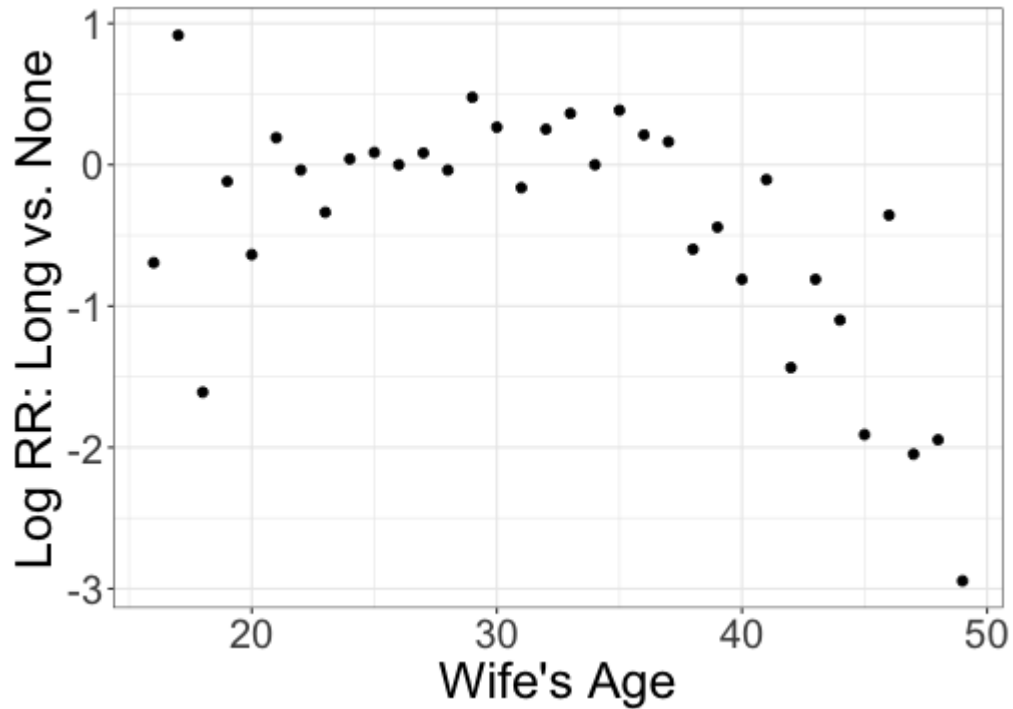
What is the relative risk of Good vs. Slow for the (52, 53) speed group?

## Class activity

How would you interpret the relative risk of Good vs. Slow for the (52, 53) speed group?

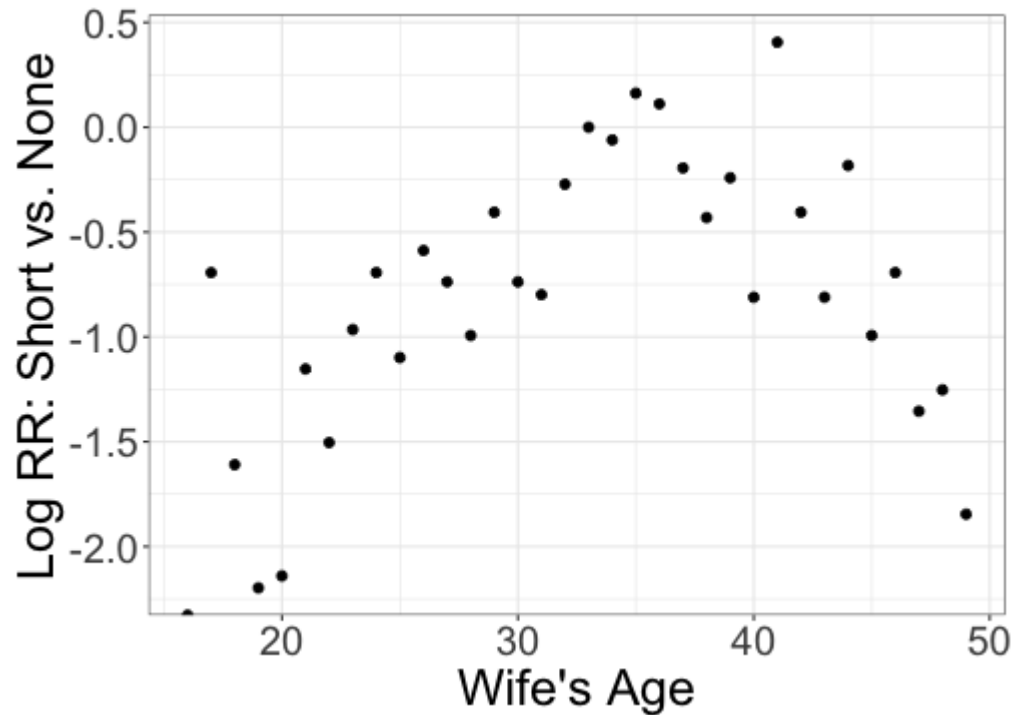
## Log relative risk

Instead of modeling the log odds, we can model the *log relative risk*



## Log relative risk

Instead of modeling the log odds, we can model the *log relative risk*



# Multinomial regression model

Step 1: Choose a reasonable distribution for  $Y$

$$Y_i \sim \text{Categorical}(\pi_{i(\text{None})}, \pi_{i(\text{Short})}, \pi_{i(\text{Long})})$$

Step 2: Choose a model for any parameters

$$\log\left(\frac{\pi_{i(\text{Long})}}{\pi_{i(\text{None})}}\right) = \beta_{0(\text{Long})} + \beta_{1(\text{Long})} \text{Age}_i$$

$$\log\left(\frac{\pi_{i(\text{Short})}}{\pi_{i(\text{None})}}\right) = \beta_{0(\text{Short})} + \beta_{1(\text{Short})} \text{Age}_i$$

- + Pick a *reference* or *baseline* category to compare to (here it is None)

# Multinomial regression model

Step 1: Choose a reasonable distribution for  $Y$

$$Y_i \sim \text{Categorical}(\pi_{i(One)}, \pi_{i(Short)}, \pi_{i(Long)})$$

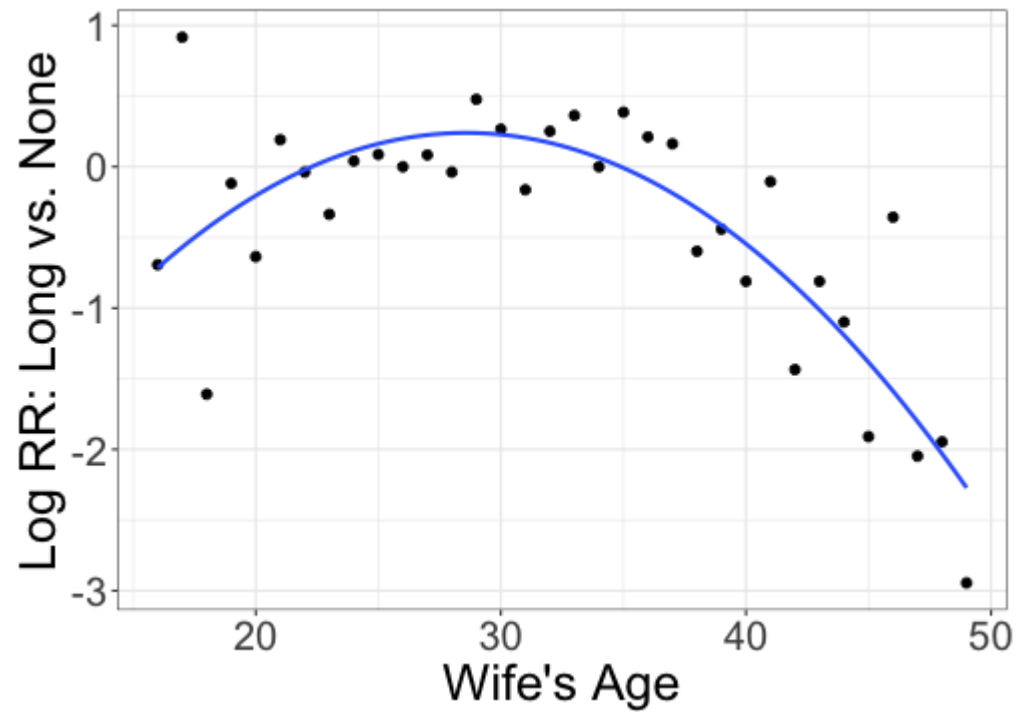
Step 2: Choose a model for any parameters

$$\log\left(\frac{\pi_{i(Long)}}{\pi_{i(One)}}\right) = \beta_{0(Long)} + \beta_{1(Long)} \text{Age}_i$$

$$\log\left(\frac{\pi_{i(Short)}}{\pi_{i(One)}}\right) = \beta_{0(Short)} + \beta_{1(Short)} \text{Age}_i$$

From the empirical log relative risk plots, did it look like the log relative risk was a linear function of Age?

## Log relative risk





# Multinomial regression model

Step 1: Choose a reasonable distribution for  $Y$

$$Y_i \sim \text{Categorical}(\pi_{i(Nothing)}, \pi_{i(Short)}, \pi_{i(Long)})$$

Step 2: Choose a model for any parameters

$$\log\left(\frac{\pi_{i(Long)}}{\pi_{i(Nothing)}}\right) = \beta_{0(Long)} + \beta_{1(Long)}Age_i + \beta_{2(Long)}Age_i^2$$

$$\log\left(\frac{\pi_{i(Short)}}{\pi_{i(Nothing)}}\right) = \beta_{0(Short)} + \beta_{1(Short)}Age_i + \beta_{2(Short)}Age_i^2$$

## Estimated model

$$\log\left(\frac{\hat{\pi}_{i(Long)}}{\hat{\pi}_{i(None)}}\right) = -5.07 + 0.37Age_i - 0.0063Age_i^2$$

$$\log\left(\frac{\hat{\pi}_{i(Short)}}{\hat{\pi}_{i(None)}}\right) = -8.21 + 0.46Age_i - 0.0065Age_i^2$$

What is the predicted relative risk of long term vs. none for a woman age 30?

# Class activity, Part II

[https://sta214-f22.github.io/class\\_activities/ca\\_lecture\\_14.html](https://sta214-f22.github.io/class_activities/ca_lecture_14.html)

## Class activity

Write down the population multinomial regression model, using Slow as the reference category, and assuming that the log relative risk is a linear function of Speed.

## Class activity

$$\log\left(\frac{\hat{\pi}_{i(Good)}}{\hat{\pi}_{i(Slow)}}\right) = -39.68 + 0.77 \text{ Speed}_i$$

Calculate the predicted relative risk of Good vs. Slow for a race where the winning speed was 52.5 mph.

## Class activity

From this information, can you calculate the predicted *probability* that the condition was Good? If not, what more information do you need?