

# Introduction to R for Epidemiologists

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# Outline

1. Interactions in regression
2. Splines
3. Logistic regression
4. Generalized linear models
5. Survival analysis

# Interactions

Interaction terms allow the association between the covariate and outcome to differ by a third variable

- ▶ Does the association between air pollution and birthweight differ by temperature?
- ▶ Does the association between population and murder rate differ by robbery rate?
- ▶ Does the association between birthweight and gestational age differ by survival status?

# Interactions

Analysis of Covariance (ANCOVA) is the same as linear regression with one categorical covariate and one continuous covariate

$$E(y) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_P x_P$$

where

- ▶  $x_1, x_2, \dots, x_{P-1}$  are indicator variables for whether an observation belongs in that group
  - ▶ Indicator variable is 1 if condition is met, 0 otherwise
  - ▶ When  $x_1 = x_2 = \dots = x_P = 0$ , indicates reference group
- ▶  $x_P$  is the continuous predictor
- ▶  $\beta_0$  is the intercept, or the  $E(y)$  when  $x_1 = x_2 = \dots = x_P = 0$ . In other words, the mean  $y$  for the reference group when  $x_P = 0$
- ▶  $\beta_1$  is the slope for  $x_1$ , or the difference in  $E(y)$  for comparing group 1 to the reference group, controlling for  $x_P$
- ▶  $\beta_0 + \beta_1$  is the mean  $y$  for group 1, when  $x_P = 0$
- ▶  $\beta_P$  is the mean change in  $y$  for a one unit change in  $x_P$ , controlling for the categorical variable

# Logistic regression

Applied when the outcome of interest is binary

- ▶ What is the association between smoking and lung cancer?
- ▶ Is gestational age associated with survival in very low birthweight infants?

$$\text{logit}(E(y|x)) = \beta_0 + \beta_1 x$$

- ▶ logit is the (natural) log odds,  $\log(p/(1-p))$
- ▶  $E(y)$  is the mean  $y$ . Recall that for binary  $y$ , the mean of  $y$  is simply the proportion of 1's
- ▶  $\beta_0$  is the log odds of  $y$  when  $x = 0$
- ▶  $\beta_1$  is the difference in log odds of  $y$  for a one unit change in  $x$ 
  - ▶ The difference in log odds is the same as the log odds ratio
  - ▶  $\beta_1$  quantifies the association between  $x$  and  $y$

# Survival analysis in R

For time-to-event data with censoring

- ▶ What is the time until AIDS for HIV positive individuals?
- ▶ What is the time-to-relapse for smokers?
- ▶ Is a new drug associated with time until lung cancer?

# Survival data in R

Hunger games survival analysis: Do “career” tributes survive longer?

*“which covariates are associated with the odds (or hazard ratios) being ever in your favor?”*

- ▶ <http://www.bdkeller.com/writing/hunger-games-survival-analysis/>

*(source: Brett Keller)*

# Other models

R can also be used to fit more complex models including

- ▶ Ordinal logistic regression
- ▶ Mixed and random effect models
- ▶ Time series models
- ▶ Bayesian modeling