# Statistical Inference with Linear & Logistic Regression

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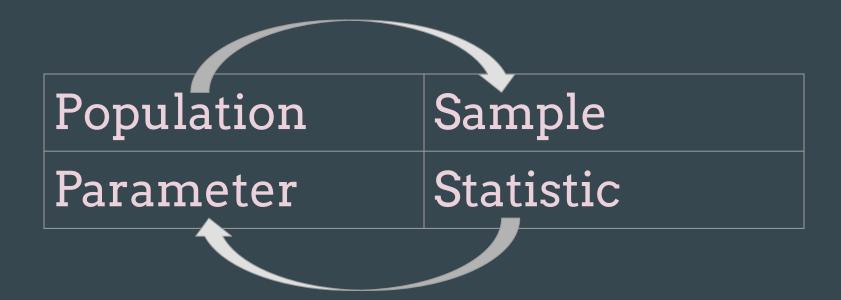
GW Libraries Workshop

Dan Kerchner ~ October 2022

go.gwu.edu/rstats

# Super-Brief Review of Inference for Regression

# **High-Level Objective**



# Regression Model

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ... + \beta_n X_n$$

(and an interaction term might look like  $oldsymbol{eta}_{12} X_1 X_2$  )

#### <u>Interpretation</u>

$$\beta_0$$
 = Mean Y when  $X_i$  values are 0

 $\beta_i$  = Mean change in Y for a 1-point increase in  $X_i$ , adjusting for other X variables

# Correlation between dependent & independent variables

**Pearson** correlation ( $\varrho$ ) measures strength and direction (+/-) of linear association between  $X_i$  and Y. Ranges from -1 to 1.

If relationship looks non-linear (but monotonic) then **Spearman** correlation should be considered.

$$H_0: \varrho = 0$$

$$H_1: \varrho \neq 0$$

Valid if joint distribution of X, Y is bivariate normal

#### **Linear Regression Model**

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ... + \beta_n X_n$$

where Y is a <u>continuous</u> outcome

#### <u>Interpretation</u>

$$\beta_0$$
 = Mean Y when  $X_i$  values are 0

 $\beta_i$  = Mean change in Y for a 1-point increase in  $X_i$ , adjusting for other X variables

### **Linear Regression Assumptions**

- Observations are independent
- Linearity
- Homoscedasticity
- Normality

#### **GLMs - Generalized Linear Models**

$$g(\mu) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ... + \beta_n X_n$$
  
where  $\mu = E(Y)$ 

#### Common link functions

$$g(\mu) = \mu$$

$$g(\mu) = \log(\mu)$$

$$g(\mu) = \log(\mu / (1 - \mu)) = \log(\mu)$$

# GLMs: (Binary) Logistic Regression Model

odds 
$$log(\frac{p}{1-p}) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ... + \beta_n X_n$$
 where p = Probability of Y = 1 1-p = Probability of Y = 0

#### Interpretation

 $\beta_i$  = log OR (Odds Ratio) for having Y = 1 for a 1-point increase in  $X_i$ , adjusting for other predictors

 $e^{\beta_i}$  = OR for having Y = 1 for a 1-point increase in  $X_i$ 

### Binary Logistic Regression Assumptions

- Binary outcome (0, 1)
- Each predictor is linearly related to the <u>log odds</u> of the outcome

# Inference for Regression Modeling

Confidence Interval

95%CI for 
$$\beta$$
 = (0.44, 0.49)

Hypothesis Testing

$$H_0$$
:  $\beta = \beta_0 \leftarrow \text{Null Hypothesis}$ 

$$H_A: \beta \neq \beta_0 \leftarrow Alternative Hypothesis$$

p-value: Chance that we are rejecting H<sub>0</sub> when we should not be



# Today's Goal

- Learn to use R to read in data and conduct regression analysis and associated inference tests
  - Checking assumptions
  - Visualizing
  - Computing p-values, regression coefficients, confidence intervals, and odds ratios (for logistic models)

### Today: 2 Scenarios

- Linear Regression (continuous outcome)
- Logistic Regression (categorical outcome)

(with both continuous and categorical predictors)

#### Today's Data Set #1: Siddarth et al., 2018

Siddarth P, Burggren AC, Eyre HA, Small GW, Merrill DA (2018) Sedentary behavior associated with reduced medial temporal lobe thickness in middle-aged and older adults. PLoS ONE 13(4): e0195549.

#### doi.org/10.1371/journal.pone.0195549

- Examined associations between sedentary behavior and medial temporal lobe (MTL) subregion integrity
- 35 non-demented middle-aged and older adults
- Measured physical activity levels w/questionnaire
- Measured MTL thickness w/MRI scan
- Adjusted for age

Sedentary behavior associated with reduced medial temporal lobe thickness in middleaged and older adults

Prabha Siddarth<sup>1\*</sup>, Alison C. Burggren<sup>2</sup>, Harris A. Eyre<sup>3</sup>, Gary W. Small<sup>1</sup>, David A. Merrill<sup>1</sup>

1 Semel Institute for Neuroscience and Human Behavior, UCLA, Los Angeles, CA, United States of America 2 Center for Cognitive Neurosciences, UCLA, Los Angeles, CA, United States of America, 3 Discipline of Psychiatry, University of Adelaide, Adelaide, Australia

\* psiddarth@mednet.ucla.edu

#### Abstract

Atrophy of the medial temporal lobe (MTL) occurs with aging, resulting in impaired episodic memory. Aerobic fitness is positively correlated with total hippocampal volume, a heavily studied memory-critical region within the MTL. However, research on associations between sedentary behavior and MTL subregion integrity is limited. Here we explore associations between thickness of the MTL and its subregions (namely CA1, CA23DG, fusiform gyrus, subiculum, parahippocampal, perirhinal and entorhinal cortex.), physical activity, and sedentary behavior. We assessed 35 non-demented middle-aged and older adults (25 women, 10 men; 45–75 years) using the International Physical Activity Questionnaire for older adults, which quantifies physical activity levels in MET-equivalent units and asks about the average number of hours spent sitting per day. All participants had high resolution MRI scans performed on a Siemens Allegra 3T MRI scanner, which allows for detailed investigation of the MTL. Controlling for age, total MTL thickness correlated inversely with hours of sitting/day (r = -0.37, p = 0.03). In MTL subregion analysis, parahippocampal (r = -0.45, p = 0.045) or official and subiculum (r = -0.38, p = .045) inclinated and subiculum (r = -0.38, p = .045) inclinated and subiculum (r = -0.36, p = .045) inclinated and subiculum (r = -0.36, p = .047) thicknesses correlated

### Today's Data Set #2: Framingham Heart Study

- <u>framinghamheartstudy.org</u>
- Long-term prospective study of the etiology of cardiovascular disease among a population of subjects in Framingham, MA
- Began in 1948 with 5,209 subjects
- Is the source of the term "risk factor"
- Over 3,000 peer-reviewed papers published based on this study
- Participants were each followed for a total of 24 years for cardiovascular events (heart attack, stroke, death, etc.)

# Some Handy R Links

#### Some R packages & functions for regression++

- **lme4** Linear mixed-effects models
- MASS Model selection
- ts(), arima() Time series object, model (part of stats)
- forecast Forecasting for time series and linear models
- mice imputation of missing data
- medflex mediation analysis
- splines2 regression spline basis functions
- **survival** survival analysis; JM joint modeling
- AND MANY, MANY, MANY MORE

#### **Tutorials**

- RStudio R paths: <u>education.rstudio.com/learn/</u>
- Data Carpentry & Software Carpentry:
  - o datacarpentry.org and software-carpentry.org
- Linkedin Learning @ GW: go.gwu.edu/linkedinlearning
- <u>r-tutor.com/r-introduction</u> & <u>r-tutor.com/elementary-statistics</u>
- UCLA Data Analysis Examples: <u>stats.idre.ucla.edu/other/dae/</u>
- Visualizing regression results: worldbank.github.io/r-econ-visual-library/RegressionCoef.html
- R Graph Gallery (w/code): r-graph-gallery.com

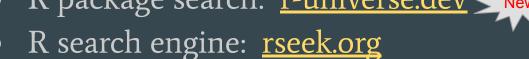
#### Books you can access for free

- Free books online Hadley Wickham:
  - R for Data Science <u>r4ds.had.co.nz</u>
  - Advanced R <u>adv-r.hadley.nz/</u>
- Through your GW library privileges:



#### **Reference Links**

- R language (CRAN): <u>r-project.org</u>
- R package search: <u>r-universe.dev</u> New!



- rstudio.com
  - Cheat Sheets! <u>rstudio.com/resources/cheatsheets</u>
- stackoverflow.com

# Statistics+R help @ GW

R-Statistics Appointments: go.gwu.edu/dataconsulting

Also...

Appointments with me: <u>calendly.com/kerchner</u>

Coding consultations

(Python, R, git, etc.): <u>calendly.com/gwul-coding/</u>

#### Thanks!

Dan Kerchner

kerchner@gwu.edu