

AP Research Handbook

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Research Philosophy & Ethics

- No Plagiarism
- Institutional Review Board
- Data privacy standards

This AP Research Handbook is still a work in progress. Please excuse any blank sections and filler text.

Chapter 1

Research Question

You can label chapter and section titles using `{#label}` after them, e.g., we can reference Chapter 1. If you do not manually label them, there will be automatic labels anyway, e.g., Chapter 8.

Figures and tables with captions will be placed in `figure` and `table` environments, respectively.

Chapter 2

Literature Review

List resources for conducting literature review. Show example of literature review with inline citations. Show ways to keep track of sources for bibliography.

- How to Write a Literature Review
 - contains example literature reviews from political science, philosophy, and chemistry.

Consider using a reference management system like Mendeley to organize your sources as you conduct your literature review. In fact, Mendeley has a Literature Search function, so you can manage sources and conduct literature reviews at the same time. See the Bibliography Management Section for more information on managing sources.

- Databases for Literature Reviews
 - Directory of Open Access Journals
 - * Browse by subjects in the humanities and sciences. This can be your starting point if you have not developed a research topic.
 - arXiv
 - * Open-access journal articles in fields such as mathematics, statistics, economics, physics, quantitative biology, quantitative finance, and electrical engineering
 - * arXiv to BibTeX: Outputs automated citations in BibTeX and other formats by typing the arXiv number of the article. For instance, just type in 1905.03758 into the search engine if the article is labeled arXiv: 1905.03758.
 - * Alternatively, use Mendeley Web Importer to import article into Mendeley Desktop for automated citation outputs.
 - Mendeley Literature Search
 - * Download Mendeley Desktop and register for a free account. Mendeley Desktop syncs with your online Mendeley account, but the literature search is currently only available in the desktop version.
 - * Mendeley is primarily a reference managements software, so you can organize your citations as you conduct your literature review.
 - CORE
 - * Search engine with the world's largest collectin of open-access research papers.
 - * For batch searches of metadata and full texts, you may consider requesting a free API key to use the Core API.
 - ScienceOpen
 - * Search for content, authors, collections, and journals in the advanced search, where you have the option to search by discipline or key word.
 - Dimensions
 - * Search for articles in clincial sciences, biochemistry, public health, physical chemistry, and materials engineering.

- EBSCO Open Access
 - * Search open-access journals and dissertations. Note that dissertations can vary in quality, since they have not gone through peer review.
 - * AP Research students should have access to a free EBSCO account from the AP Capstone program.
- SSRN
 - * Many of the social science articles are free access.
- ERIC: Institute of Education Sciences
 - * Search for articles related to education research.
 - * The search engine includes the option to search for full-text articles.
- dblp: Computer Science Bibliography
 - * Index of major computer science publications.
 - * Option to search for open-access articles.
- EconBiz
 - * Search for journal articles, working papers, and conference papers in economics and business.
 - * Option to search for open-access articles.
- MyJSTOR
 - * You can sign up for a free MyJSTOR account to access up to six articles a month for free.
 - * This may be helpful for accessing articles that are not open access.
- Tips for Accessing Paywalled Articles
 - Search for the author's website. Many researchers have draft manuscripts on their websites or research profiles on sites such as ResearchGate.
 - Consult your school's research librarian for other ways to access the article.
 - Send the author an e-mail to request for a digital copy of the article. You should provide context in the e-mail request by including a brief description of your AP Research project and its relevance and connection to the author's article.

Chapter 3

Bibliography Management

While the bibliography is placed at the end of research papers, reference management begins as soon as you begin your literature review.

- Managing Citations in LaTeX
- Mendeley Desktop Download
 - The download will prompt you to create a free account. Mendeley Desktop is synced with your online account.
- Mendeley Web Importer
 - As you search for research articles online, you can use the Mendeley Web Importer to import citations into your Mendeley Desktop. If the importer can recognize the online article's metadata, it will automatically populate the citation entries. If not, you can still enter the citation entries manually and import into the Mendeley Desktop to keep track of your sources.
- Mendeley Tutorials
- Exporting .bib files from Mendeley Desktop
- Install MS Word plugin
- Import Mendeley sources into LyX
- Import .RIS Files into Mendeley

Chapter 4

Paper Guidelines

- AP Research Proposal Guidelines
- AP Research Paper Guidelines (LaTeX)
 - draft in progress

Chapter 5

File Organization

Chapter 6

Research Concepts

6.1 Reliability & Validity

6.2 Accuracy vs. Precision

6.3 Bias vs. Variance Tradeoff

6.4 Curse of Dimensionality

6.5 Correlation vs. Causation

Chapter 7

Research Design

Before you decide how you will conduct your research, read through the list of research designs in the USC library guide.

- What is Research Design?

- 7.1 Action
- 7.2 Case Study
- 7.3 Causal
- 7.4 Cohort
- 7.5 Cross Sectional
- 7.6 Descriptive
- 7.7 Experimental
- 7.8 Exploratory
- 7.9 Historical
- 7.10 Longitudinal
- 7.11 Meta-Analysis
- 7.12 Mixed Methods
- 7.13 Observational
- 7.14 Philosophical
- 7.15 Sequential
- 7.16 Systematic Review
- 7.17 Quasi-experimental

Chapter 8

Qualitative Research Methods

- Qualitative Research Methods Field Guide

8.1 Case Study

8.2 Narrative

8.3 Phenomenological

8.4 Ethnography

8.5 Grounded Theory

- Grounded Theory as Scientific Method

Chapter 9

Quantitative Methods

Causal Inference

These notes are based on Professor Masten's online course on Causal Inference at the Social Science Research Institute at Duke.

- Causal effect is often easy to detect with simple actions for which the effect immediately follows (e.g., you caused the alarm clock to stop ringing by pressing the snooze button)
- With multiple causes and delayed effects, causality is much harder to detect.
- Measurement:
 - Unit of analysis: countries, city blocks, people, firms, etc.
 - Outcome variable: the characteristic of the unit of analysis that we want to affect
 - Policy/treatment variable: the characteristic that we use to change the outcome variable
- A lot of characteristics cannot readily be quantified, so we often use proxy variables. For example, GDP could be a proxy for economic development.
- Causality: how an intervention in the policy variable affects the outcome variable
- Data:
 - The value of the policy variable has to vary in the dataset. Without this variation, you can't analyze how changes in the policy variable might affect the outcome variable.
 - Larger standard deviation = larger variation
- Correlation vs. Causation
 - If the policy and outcome variables are correlated, this does not necessarily imply a causal relationship.
 - Selection Problem: when units get to choose their policy variable, correlations between policy and outcome variables are unlikely to be causal.
 - * Example: Neighborhoods with a lot of trees tend to have less crime.
 - * If this were a causal relationship, then we could plant more trees in a neighborhood and expect crime to go down. However, this is unlikely. More likely, people who tend to commit less crimes chose to live in neighborhoods with tree-lined streets.
- Average Treatment Effect
 - Causal effects vary among people, so there is a distribution of causal effects in the population.

- Theoretical ideal: you would know the unit level of causal effect for each person and thus make individualized treatment decisions. This is impossible in practice. You can't know the effect of receiving and not receiving treatment for an individual.
- Unit-level causal effect: difference in outcome between treatment & control, holding all other variables fixed
- Avg. treatment effect (ATE): avg. of all values for unit-level causal effects in a population
- Avg. outcome under the policy: avg. outcome when everyone is affected by the policy (i.e., receives treatment)
- Avg. outcome without the policy: avg. outcome when everyone is not affected by policy (i.e., does not receive treatment)
- $ATE = \text{Avg. outcome under policy} - \text{Avg. outcome w/o policy}$

9.0.1 Experiments

- Controlled Experiments
 - Control group does not receive treatment
 - Experimental group receives treatment
 - All possible factors that could affect the outcome are identical for both groups, except for the treatment
 - Difference in the outcome between the two groups is the treatment effect
 - Typically used in hard sciences, but difficult to achieve in social sciences given the myriad of factors, many of which are difficult to measure and control
- Randomized Experiments
 - Split units randomly into two large groups: treatment or control
 - Right after randomization and before the experiment, both groups should be similar (i.e., avg. values of factors should be about the same), because the split was done randomly and the groups are very large
 - Since the two groups are similar in all factors except treatment, changes in the *average* outcomes are due to treatment
 - Complications:
 - * Noncompliance: Even when you randomly assign treatment, people in the treatment group may not all decide to take the treatment. Also, some people in the control group, who should not receive the treatment, might decide to get the treatment.
 - Solution 1: Intent to Treat Analysis
 - The *intent* to provide treatment is by design random regardless of treatment non-compliance.
 - Thus, we can examine the causal effect of the option of providing treatment.
 - Downside: cannot analyze causal effect of treatment itself
 - For example, in the Oregon Health Experiment, while a lottery randomly selected people to receive free Medicaid, there was noncompliance in both the treatment/control groups. Original interpretation (effect of Medicaid on health outcomes) can be revised to effect of Medicaid lottery assignment on health outcomes.
 - Solution 2: Instrumental Variables
 - Advantage: We can analyze the causal effect of the treatment (not just the option of treatment) for a subset of the population.
 - Downside: cannot analyze average treatment effect over the entire population
 - Solution 3: Assume random compliance
 - Assume people comply with their treatment assignment.
 - Just drop the entries of the non-compliers.
 - Advantage: We can analyze the causal effect of the treatment over the entire population.
 - Downside: Decision to not comply is probably not random. We don't observe the reasons for non-compliance.
 - Solution 4: Bounds analysis

- Get lower/upper bounds of average treatment effects using extreme scenarios.
 - Upper bounds: assume maximum value for outcome variable for noncompliers
 - Lower bounds: assume minimum value for outcome variable for noncompliers
- * Survey nonresponse
 - If nonresponse is not random, you cannot interpret the treatment effect as causal.
 - Example: People in the treatment group with negative outcomes responded to surveys at higher rates than those with positive experiences. Data becomes biased toward negative outcomes for the treatment group.
- * Sample Size: Even with a great research design, small sample size limits statistical inference.
- * Control: You may not be able to control the assignment of treatment.
- Issues:
 - * Ethics: Random assignment of treatment may have difficult ethical considerations (e.g., withholding a potentially life-saving drug to a terminally ill patient assigned to a control group in a randomized trial).
 - * Extrapolation: It may be hard to extrapolate the results of a randomized experiment to another study if the treatment conditions and features are different.
- Natural Experiments
 - Researchers not involved in the research design and data collection in natural experiments, unlike in randomized experiments.
 - observational data used instead
 - Example: charter school lotteries
- 1. True Natural Experiments
 - treatment was randomly assigned, just not by researcher
- 2. As-If Natural Experiments
 - treatment not actually randomly assigned, but the treatment/control groups appear randomized as though treatment assignment were random)
 - treatment assignment not related to any variables that could affect outcome
 - balance check: characteristics of all observed variables (other than outcome variable) need to be similar between the treatment/control groups
 - * There could still be differences between groups in unobserved variables.
 - * Thus, we cannot prove treatment assignment is truly random, but balanced observed variables between groups would be part of a convincingly argument that the observational data represents an as-if natural experiment.

9.0.2 Regression as Causality

9.0.3 Instrumental Variables

9.1 Statistical Tests

- Choosing a Statistical Test
- Hypothesis Testing Roadmap
- Choosing the Correct Statistical Test in SAS, Stata, SPSS, and R
- Uses & Misuses of Statistics
- 1 group
 - interval variables
 - * 1-sample t test for the mean
 - * chi-squared test for variance
 - categorical variables
 - * z test for proportions (2 categories)

- * chi-squared goodness-of-fit
 - ordinal or interval
 - * one-sample median test
- 2 groups (independent groups)
 - interval variables
 - * 2 independent sample t-test (equal variances)
 - * 2 independent sample t-test (unequal variances)
 - * F test for difference between 2 variances
 - categorical variables
 - * z test for difference between 2 proportions
 - * chi-squared test for difference between 2 proportions
 - * Fisher’s exact test
- 2 groups (dependent or paired groups)
 - paired t-test (interval variables)
 - McNemar’s test (categorical variables)
 - Wilcoxon signed ranks test (ordinal or interval variables)
- more than 2 groups (independent groups)
 - one-way ANOVA (for interval variables)
 - Kruskal Wallis (for ordinal or interval variables)
 - chi-squared test (for categorical variables)
- more than 2 groups (dependent groups)
 - one-way repeated measures ANOVA (for interval variables)
 - repeated measures logistic regression (for categorical variables)
 - Friedman test (for ordinal or interval)

9.1.1 1-sample t-test

- Assumptions:
 - data is a simple random sample from population
 - data follows normal distribution
 - by Central Limit Theorem, with sample size $n \geq 30$, the sample mean is normally distributed regardless of the population distribution

- Two-tailed Hypothesis:

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu \neq \mu_0$$

- Test Statistic:

$$T = \frac{\bar{X} - \mu_0}{\frac{S}{\sqrt{n}}} \sim t_{(n-1)}$$

- \bar{X} = sample mean
- μ_0 = hypothesized population mean
- S = sample standard deviation
- $t_{(n-1)}$ = t distribution with $n - 1$ degrees of freedom

9.1.2 chi-squared test for variance

9.1.3 z test for proportions

- Assumptions:
 - sample proportion $p = \frac{X}{n}$ comes from random sample in population, where X is number of events of interest in sample size n .
 - p follows a binomial distribution, but we can assume normality when X and $n - X$ are each at least 5 (old standards) or at least 15 (current standards)

- Two-tailed Hypothesis:

$$H_0 : \pi = \pi_0$$

$$H_1 : \pi \neq \pi_0$$

- Test Statistic:

$$z = \frac{p - \pi_0}{\sqrt{\frac{\pi_0(1-\pi_0)}{n}}} \sim \mathcal{N}(0, 1)$$

- π_0 = hypothesized proportion
- p = sample proportion

9.1.4 t-test for 2 independent samples

- Assumptions:
 - two independent samples are randomly selected from two populations with the same variance
 - if you cannot use the assumption of same variance, use the Welch two-sample t-test
 - * test statistic is the same as below, but degrees of freedom are adjusted
 - *
 - if populations are not normally distributed, the sample sizes n_1 and n_2 from the two populations needs to be at least 30 to ensure that the distribution of the sample means are normal by the Central Limit Theorem

- Two-tailed Hypothesis:

$$H_0 : \mu_1 = \mu_2$$

$$H_1 : \mu_1 \neq \mu_2$$

- μ_1 = population mean of 1st sample
- μ_2 = population mean of 2nd sample

- Test Statistic:

$$\frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{S_p^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim t_{(n_1+n_2-2)}$$

$$S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{(n_1 - 1) + (n_2 - 1)}$$

- S_p = pooled variance
- \bar{X}_1 = mean of 1st sample
- \bar{X}_2 = mean of 2nd sample
- S_1^2 = variance of 1st sample
- S_2^2 = variance of 2nd sample

- More Info

- R Example

- includes examples under both assumptions of equal and unequal variances

- Andrew Heiss provides a brief tutorial with frequentist, simulation-based, and Bayesian approaches to comparing means between two groups. Also see Matti Vuorre’s tutorial for more details.

9.1.5 paired t-test

- Assumptions:
- More Info with R Example

9.1.6 chi-squared test for proportions

- The chi-squared test for 2 x 2 frequency tables is equivalent to the square of the z-test for two proportions. See this link for detailed explanation.

9.1.7 chi-squared test for independence

- Explain connection between chi-squared test for independence and log-linear models, which are Poisson models for categorical data.

9.1.8 ANOVA

9.2 Numerical Methods

In AP Calculus, you mostly encountered problems that can be solved analytically. However, in research, many differential equation models do not have analytical forms and must be solved numerically. Matlab is often used in applied math, engineering, and physical sciences for such cases as well as other modeling applications. Octave is an open-source alternative to Matlab. While R not the first language that comes to mind for numerical methods, many numerical R packages have been developed as well as integration with Matlab, Octave, and Julia.

- Numerical Computing with Matlab
 - This site has PDF versions of Cleve Moler’s textbook on numerical computing alongside a video series with lectures on differential equations and linear algebra by Prof. Gilbert Strang and computational video tutorials by Moler.
- Numerically Solving Differential Equations with R

9.2.1 Root-Finding Algorithms

- Newton-Raphson Method Using R
- Bisection Method Using R
- Secant Method Using R

9.2.2 Numerical Solutions to Differential Equations

- Euler Method Using Matlab
- Runge-Kutta Methods

Resources by Discipline

Biology & Biostatistics

- Handbook of Biological Statistics
- An R Companion for the Handbook of Biological Statistics

Economics & Econometrics

- Introduction to Econometrics with R
- Principles of Econometrics with R
- Introduction to Data Science
- Using R for Introductory Econometrics
- Examples:
 - Annotated Sample Econometrics Paper
 - Microeconomic example of utility maximization constrained by budget lines

Psychology

- Psychology Research Methods

Public Health & Epidemiology

- Examples:
 - SIR Model Using R

Social Sciences

- Social Science Methods Modules
- Applied Causal Analysis

Chapter 10

Data

10.1 Data Sources by Discipline

10.1.1 Demography and Official Statistics

- U.S. Census Data
 - American Fact Finder
 - IPUMS
 - * U.S. census microdata with social, economic, and health variables.
 - * Create custom data sets or use online tool.
- UK Office for National Statistics
- Statistics Canada

10.1.2 Economics

- Panel Study of Income Dynamics
- University of Michigan Surveys of Consumers

10.1.3 Education

- Institute of Education Sciences: Data Files
- National Assessment of Educational Progress Data Explorer

10.1.4 Law

- Caselaw Access Project
 - Digital access to U.S. state and federal cases from the 1600s to present.

10.1.5 Social Sciences

- ICPSR

10.2 Data Documentation

Cite the source of your data. Provide links to the original data source and accompanying codebook, if any. Your data documentation will document your data analysis from the download of the raw data to the final steps of data analysis.

- Create a Codebook
 - List of codebook creation tools with guides and download links.
- Guide to Writing a Codebook
- How to Use R Codebook Package

Chapter 11

Analysis

- Logical Fallacies

Read about the common fallacies in social research. Summary below:

- fallacies of authority
- fallacies of logic
- fallacies of emotion

- Statistical Biases

- Sampling bias
 - * e.g., 1948 U.S. presidential election (see this case study)
 - * even very large samples could have sampling biases if sampling methods are poor and unrepresentative of the population (e.g., 1936 Literary Digest Poll)
- Omitted variable bias
- Nonresponse bias
- Selection bias
- Survivorship bias
 - * e.g., when bankrupt companies are removed from a stock index and replaced with profitable companies, the index would experience an upward bias. Business failures would not be accounted for in time series data.

Data Programming

- R
 - To download R, choose a CRAN mirror closest to your geographic location.
 - In order to build R packages, you should also download the latest recommended version of Rtools. Currently, the latest recommended version is `Rtools35.exe`.
 - During the installation of Rtools, you may need to add in `"C:\Rtools\mingw_64\bin;"` to the path.
- R Studio
 - R Studio is an integrated development environment (IDE) for R. After downloading R Studio, you should be able to type the following command at the console to download some common R packages for data analysis and visualization.

```
install.packages(c("dplyr", "tidyr", "ggplot2", "esquisse", "stats", "xtable"))
```

11.1 Cleaning and Reshaping Data

```
library(reshape2)
library(tidyr)
library(xtable)
library(stringr)
library(knitr)
options(kableExtra.latex.load_packages = FALSE)
library(kableExtra)
library(pander)

#original data is organized by id/trial (two locations per entry)
game <- data.frame(id = c(rep("X",3), rep("Y",3), rep("Z",3)),
  trial = rep(c(1,2,3), 3),
  location_A = round(rnorm(9, mean = 0, sd = 1), 1),
  location_B = round(rnorm(9, mean = 0, sd = 1), 1))

# reshape data from wide to long (each entry is unique by id/trial/location)
game_long <- melt(game, id = c("id","trial"), value.name = "score")
game_long$variable <- str_sub(game_long$variable,-1,-1)
colnames(game_long)[3] <- "location"

# reshape data back to wide (same as original data)
game_wide <- dcast(game_long, id + trial ~ location, value.var = "score")
# reshape data into even wider form (one entry per id with 6 value columns: 2 locations X 3 trials)
```

```

game_wider <- dcast(game_long, id ~ location + trial, value.var = "score")

# using tidyr and dplyr to reshape data
game_long2 <- game %>% gather(label, score, location_A, location_B) %>%
  separate(label, c("label_p1", "location"), sep = "_") %>%
  dplyr::select(-label_p1)

game_wide2 <- game_long2 %>% spread(location, value = score)

#unite() function creates the location X trial combinations first in long format # then apply the spread
#just like in game_wide, each entry in game_wide2 is unique by id
game_wider2 <- game_long2 %>% unite(location_trial, location, trial) %>%
  spread(location_trial, value = score)

#xtable method
#print(xtable(game, caption = "Wide Data Listed by Person/Trial (Scores by Location)", type="html"))

#kable method
#kable(game, caption = "Wide Data Listed by Person/Trial (Scores by Location)", booktabs = TRUE) %>%
#  kable_styling(latex_options = c("hold_position"))

#pander method (most flexible)
pandoc.table(game, caption = "(\\#tab:wide) Wide Data Listed by Person/Trial (Scores by Location)")

```

Table 11.1: Wide Data Listed by Person/Trial (Scores by Location)

id	trial	location_A	location_B
X	1	0.5	-1.5
X	2	0	-1
X	3	0.6	0.1
Y	1	-0.5	0.2
Y	2	0.4	0.7
Y	3	0.4	1.5
Z	1	-1.4	-1
Z	2	0.2	-0.5
Z	3	-1.2	0

```

pandoc.table(game_wider, caption = "(\\#tab:wider) Wider Data Listed by ID (Scores by Location/Trial)")

```

Table 11.2: Wider Data Listed by ID (Scores by Location/Trial)

id	A_1	A_2	A_3	B_1	B_2	B_3
X	0.5	0	0.6	-1.5	-1	0.1
Y	-0.5	0.4	0.4	0.2	0.7	1.5
Z	-1.4	0.2	-1.2	-1	-0.5	0

```

pandoc.table(game_long, caption = "(\\#tab:long) Long Data")

```

Table 11.3: Long Data

id	trial	location	score
X	1	A	0.5
X	2	A	0
X	3	A	0.6
Y	1	A	-0.5
Y	2	A	0.4
Y	3	A	0.4
Z	1	A	-1.4
Z	2	A	0.2
Z	3	A	-1.2
X	1	B	-1.5
X	2	B	-1
X	3	B	0.1
Y	1	B	0.2
Y	2	B	0.7
Y	3	B	1.5
Z	1	B	-1
Z	2	B	-0.5
Z	3	B	0

- Data Wrangling with dplyr and tidyr

11.2 Regular Expressions

- Regular Expressions in R
- Basic Regular Expressions in R Cheat Sheet

Literate Programming

11.3 LaTeX

- MiKTeX
 - First, download MiKTeX. Choose the version corresponding to your operating system (Windows, Mac, or Linux). Skip this step if you decide to use ShareLaTeX, which is an online LaTeX editor and does not require your computer to have underlying LaTeX packages via MiKTeX.
 - Recommended, download the basic installer, which will download other uninstalled packages on the fly on an as-needed basis. If you want to download all packages, you can choose the Net Installer, but this may take up a lot of space.
- Review of LaTeX Editors
 - Overleaf/ShareLaTeX
 - TeXstudio
 - LyX
- LaTeX Guides
- LaTeX Cheat Sheet
- Q and A:
 - Reference File in Parent Folder

11.4 Beamer

Beamer is a LaTeX class for presentations.

11.5 knitr (R + LaTeX)

- Using knitr in LyX
- Configure Textstudio to use knitr
- Create LaTeX Tables with kable
 - To avoid a incompatibility warning about the LaTeX `xcolor` package, place `options(kableExtra.latex.load_packages = FALSE)` in your R chunk before `library(kableExtra)`. See Hao Zhu's explanation in page 4 of the link above.
- kableExtra Vignettes
 - vignettes for using outputting tables from R into HTML, LaTeX, and Word
- xtable and stargazer Examples

- [pander Tutorial](#)

11.6 R Markdown

- [Markdown Reference](#)
- [R Markdown Cheat Sheet](#)
- [Writing a Reproducible Paper in R Markdown](#)

11.7 R Bookdown

- [Authoring Books with R Bookdown](#)
- [R Markdown: The Definitive Guide](#)
- [Writing Thesis with Bookdown](#)
 - [Section on outputting into Microsoft Word using `bookdown::preview_chapter\(\)`](#)
- [Writing Academic Papers with R Markdown](#)

11.8 Rmd to MS Word

- [Rmd to docx](#)
- [Discussion on Using knitr for Word output](#)

Version Control

- Git

11.9 Github

- Create new repository

```
git init
git add README.md
git commit -m "first commit"
git remote add origin https://github.com/<username>/<repo-name>.git
git push -u origin master
```

- Best Practices Using Github in RStudio
- Tutorial on Git for Behavioral Sciences
- Github and R