

Statistics Writing

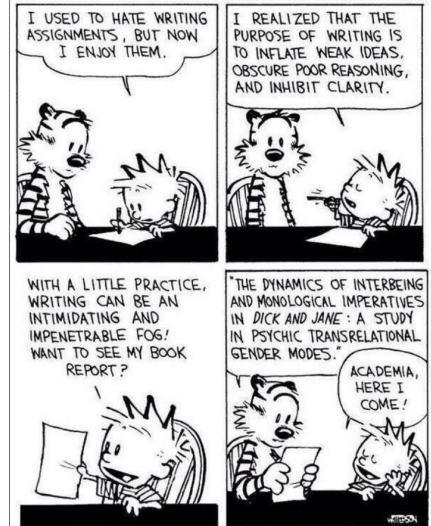
How to write more gooder

Samuel Robinson, Ph.D.

Nov 3, 2023

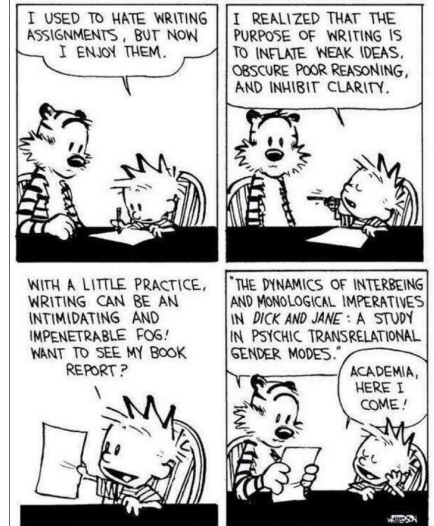
Outline

- Types of scientific writing



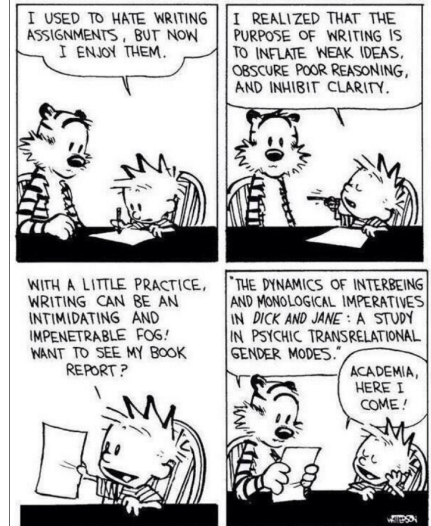
Outline

- Types of scientific writing
 - IMRaD manuscripts



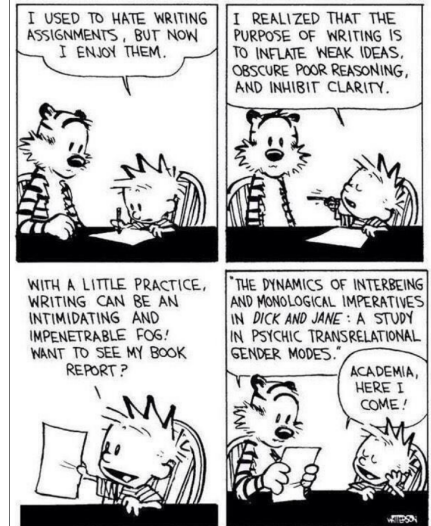
Outline

- Types of scientific writing
 - IMRaD manuscripts
 - Figures and tables



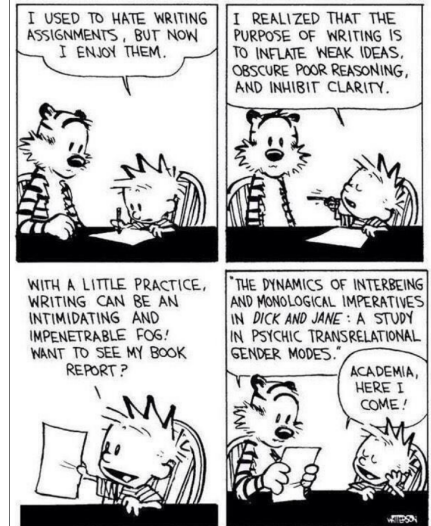
Outline

- Types of scientific writing
 - IMRaD manuscripts
 - Figures and tables
- Writing about statistics



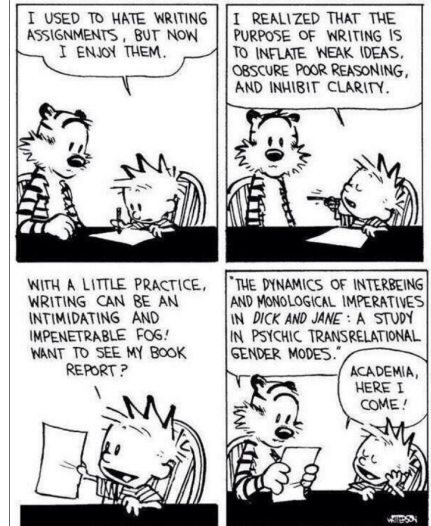
Outline

- Types of scientific writing
 - IMRaD manuscripts
 - Figures and tables
- Writing about statistics
 - Translating model results



Outline

- Types of scientific writing
 - IMRaD manuscripts
 - Figures and tables
- Writing about statistics
 - Translating model results
- How peer review works



Part 1: Types of scientific writing

Where do I start?

- You've finished fitting your models, and the results make sense to you, but. . .

Where do I start?

- You've finished fitting your models, and the results make sense to you, but. . .
- How do I translate all these numbers into “real” English?

Where do I start?

- You've finished fitting your models, and the results make sense to you, but. . .
- How do I translate all these numbers into “real” English?
- Where do I put all these numbers in the paper?

Where do I start?

- You've finished fitting your models, and the results make sense to you, but. . .
- How do I translate all these numbers into “real” English?
- Where do I put all these numbers in the paper?
- Do I need figures and tables?

Answer: “It depends”

What is your story? Who is your audience?

- How do these numbers serve the questions I’m asking?

Answer: “It depends”

What is your story? Who is your audience?

- How do these numbers serve the questions I’m asking?
- Do these numbers help my audience to understand what I found?

Answer: “It depends”

What is your story? Who is your audience?

- How do these numbers serve the questions I’m asking?
- Do these numbers help my audience to understand what I found?
- Would figures or tables help to prove my point more concisely or easily?

Answer: “It depends”

What is your story? Who is your audience?

- How do these numbers serve the questions I’m asking?
- Do these numbers help my audience to understand what I found?
- Would figures or tables help to prove my point more concisely or easily?
- How do these numbers relate to the rest of the literature?

A bit of history

- (European) Universities are largely offshoots of the Christian monastic tradition



A bit of history

- (European) Universities are largely offshoots of the Christian monastic tradition
- What we now call science started in about the 1600s, largely as offshoots of astrology and alchemy



A bit of history

- (European) Universities are largely offshoots of the Christian monastic tradition
- What we now call science started in about the 1600s, largely as offshoots of astrology and alchemy
 - Biology began slightly later (1700s-1800s), as offshoots of medicine and natural history

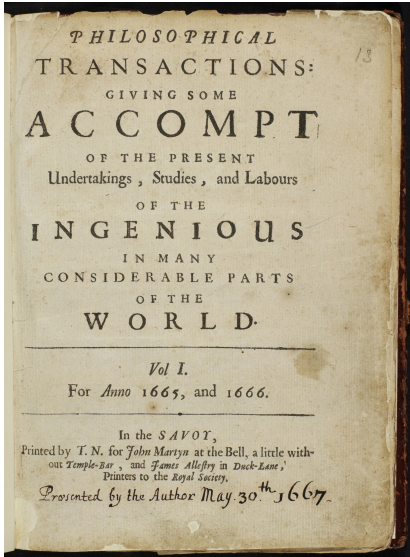


A bit of history

- (European) Universities are largely offshoots of the Christian monastic tradition
- What we now call science started in about the 1600s, largely as offshoots of astrology and alchemy
 - Biology began slightly later (1700s-1800s), as offshoots of medicine and natural history
- “Natural philosophers” (scientists) would write letters to each other about what they were up to

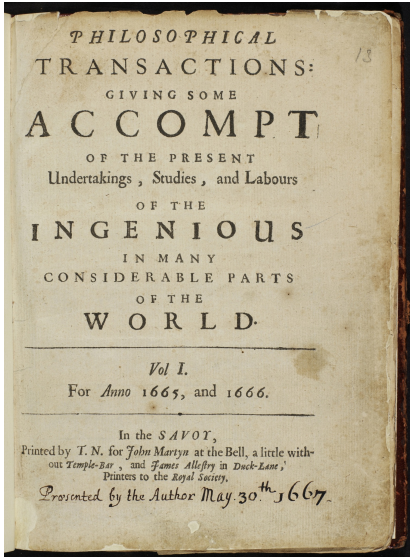


A bit of history (cont.)



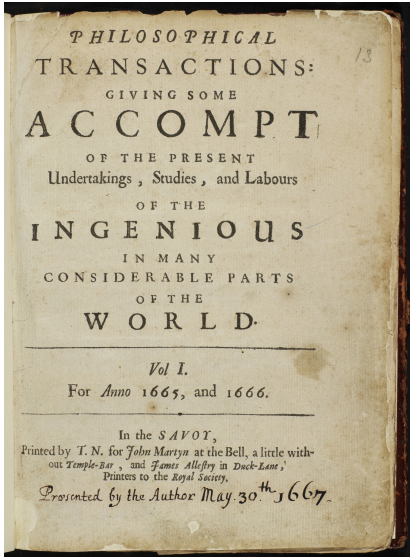
- Eventually, organizations of scientists began publishing research results publicly (e.g. *Philosophical Transactions of the Royal Society*, 1665)

A bit of history (cont.)



- Eventually, organizations of scientists began publishing research results publicly (e.g. *Philosophical Transactions of the Royal Society*, 1665)
- Peer review was sparse, and was usually done by the editor or a board. External peer review wasn't widespread until 1950-1970

A bit of history (cont.)



- Eventually, organizations of scientists began publishing research results publicly (e.g. *Philosophical Transactions of the Royal Society*, 1665)
- Peer review was sparse, and was usually done by the editor or a board. External peer review wasn't widespread until 1950-1970
- Early science writing is *extremely* varied, and is much different from modern science writing

What is science writing for?

- “Recording secret knowledge” (Newton)



What is science writing for?

- “Recording secret knowledge” (Newton)
- “Describing *exactly* how an experiment proceeded” (Bacon)



What is science writing for?

- “Recording secret knowledge” (Newton)
- “Describing *exactly* how an experiment proceeded” (Bacon)
- Modern science writing does mostly the latter:



What is science writing for?

- “Recording secret knowledge” (Newton)
- “Describing *exactly* how an experiment proceeded” (Bacon)
- Modern science writing does mostly the latter:
 - Text should be understood by your peers, not obscured



What is science writing for?

- “Recording secret knowledge” (Newton)
- “Describing *exactly* how an experiment proceeded” (Bacon)
- Modern science writing does mostly the latter:
 - Text should be understood by your peers, not obscured
 - Not *all* details are needed, only those that help make your arguments (e.g. I don’t need to know the brand of pipette tips)



What is science writing for?

- “Recording secret knowledge” (Newton)
- “Describing *exactly* how an experiment proceeded” (Bacon)
- Modern science writing does mostly the latter:
 - Text should be understood by your peers, not obscured
 - Not *all* details are needed, only those that help make your arguments (e.g. I don’t need to know the brand of pipette tips)
- More recent push for *replicability*, with data and code being stored in online repositories



How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches
 - Trade-off between realism and “explainability”

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches
 - Trade-off between realism and “explainability”
- Pushback from some quarters:
One aspect of the ongoing replication crisis

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches
 - Trade-off between realism and “explainability”
- Pushback from some quarters: One aspect of the ongoing replication crisis
 - *Statistics are political*

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches
 - Trade-off between realism and “explainability”
- Pushback from some quarters: One aspect of the ongoing replication crisis
 - *Statistics are political*

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches
 - Trade-off between realism and “explainability”
- Pushback from some quarters: One aspect of the ongoing replication crisis
 - *Statistics are political*

“I have heard from graduate students opting out of academia, assistant professors afraid to come up for tenure, mid-career people wondering how to protect their labs, and senior faculty retiring early, all because of methodological terrorism” - Susan Fiske, APS Past President

How does this relate to statistics?

- Early use of statistics in science was fairly “vibes-based”, at least until computers became more readily available (1950s onward)
 - Not necessarily a bad thing!
- More complex and extensive data collection requires more complex modeling approaches
 - Trade-off between realism and “explainability”
- Pushback from some quarters: One aspect of the ongoing replication crisis
 - *Statistics are political*

“I have heard from graduate students opting out of academia, assistant professors afraid to come up for tenure, mid-career people wondering how to protect their labs, and senior faculty retiring early, all because of methodological terrorism” - [Susan Fiske, APS Past President](#)

“[Fiske is] seeing her professional world collapsing. . . her work and the work of her friends and colleagues is being questioned in a way that no one could’ve imagined ten years ago. It’s scary, and it’s gotta be a lot easier for her to blame some unnamed “terrorists” than to confront the gaps in her own understanding of research methods.” - [Andrew Gelman](#)

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces
- ⑤ Theses

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces
- ⑤ Theses
- ⑥ Proposals

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces
- ⑤ Theses
- ⑥ Proposals
- ⑦ Data papers

Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces
- ⑤ Theses
- ⑥ Proposals
- ⑦ Data papers
- ⑧ Books/book chapters

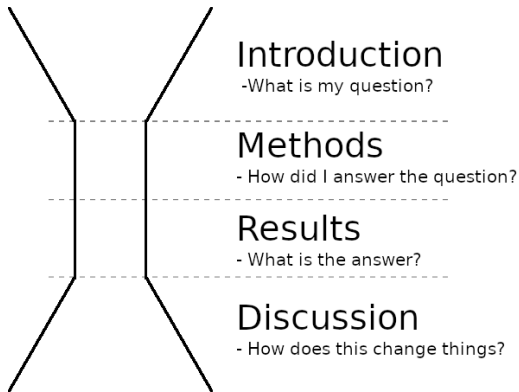
Common types of scientific writing

- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces
- ⑤ Theses
- ⑥ Proposals
- ⑦ Data papers
- ⑧ Books/book chapters
- ⑨ “Grey” or “white” papers

Common types of scientific writing

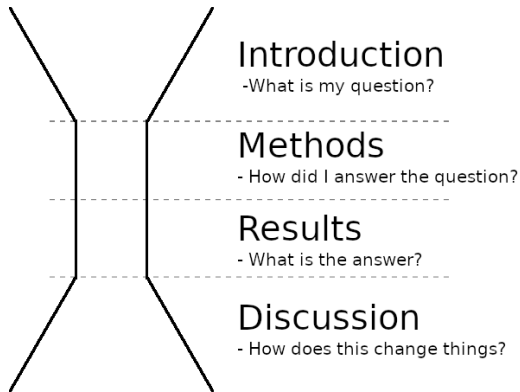
- ① IMRaD papers: “standard” scientific papers
 - Introduction, Methods, Results, and Discussion
- ② Meta-analyses
- ③ Review papers
- ④ Perspective/opinion pieces
- ⑤ Theses
- ⑥ Proposals
- ⑦ Data papers
- ⑧ Books/book chapters
- ⑨ “Grey” or “white” papers
- ⑩ Blogs

IMRaD Paper Structure



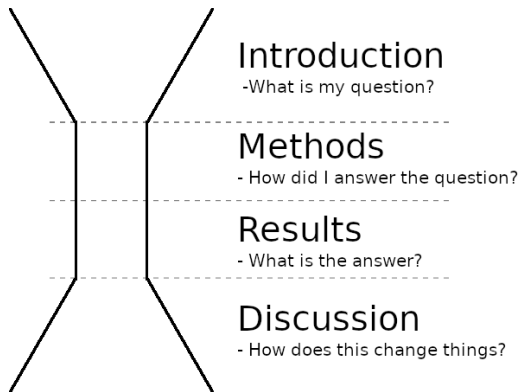
- Most scientific papers follow the IMRaD canon

IMRaD Paper Structure



- Most scientific papers follow the IMRaD canon
- Allows the reader to quickly assess whether this paper is useful and skip to important sections only

IMRaD Paper Structure



- Most scientific papers follow the IMRaD canon
- Allows the reader to quickly assess whether this paper is useful and skip to important sections only
- Generally, statistics are discussed in the *Methods* and *Results* sections only

Group exercise: “pick apart a paper”

- We're going to go through the IMRaD paper you read this week

Group exercise: “pick apart a paper”

- We're going to go through the IMRaD paper you read this week
 - You did read it... didn't you?

Group exercise: “pick apart a paper”

- We're going to go through the IMRaD paper you read this week
 - You did read it... didn't you?
- In each of the sections, we'll identify how the author follows (or doesn't follow) the form described below

Group exercise: “pick apart a paper”

- We're going to go through the IMRaD paper you read this week
 - You did read it... didn't you?
- In each of the sections, we'll identify how the author follows (or doesn't follow) the form described below
- I recommend highlighting, underlining, or annotating the paper for later reference

Introduction

- Set up your research question, using the literature

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)
 - Explain why we should care (“Bats are really cute! Don’t you like cute things?”)

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)
 - Explain why we should care (“Bats are really cute! Don’t you like cute things?”)
- Establish the *knowledge gap* or *question* that your research will address

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)
 - Explain why we should care (“Bats are really cute! Don’t you like cute things?”)
- Establish the *knowledge gap* or *question* that your research will address
 - “Forest have lots of bugs, but nobody has checked whether there are bats there too!”

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)
 - Explain why we should care (“Bats are really cute! Don’t you like cute things?”)
- Establish the *knowledge gap* or *question* that your research will address
 - “Forest have lots of bugs, but nobody has checked whether there are bats there too!”
- Last paragraph: strong statement that sums up what you’re expecting to see

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)
 - Explain why we should care (“Bats are really cute! Don’t you like cute things?”)
- Establish the *knowledge gap* or *question* that your research will address
 - “Forest have lots of bugs, but nobody has checked whether there are bats there too!”
- Last paragraph: strong statement that sums up what you’re expecting to see
 - Hypothesis: “Bats eat bugs, and forests have lots of bugs. Therefore, . . .”

Introduction

- Set up your research question, using the literature
 - Moves from general (“Animals need food”) to specific premises (“Bats need bugs”)
 - Explain why we should care (“Bats are really cute! Don’t you like cute things?”)
- Establish the *knowledge gap* or *question* that your research will address
 - “Forest have lots of bugs, but nobody has checked whether there are bats there too!”
- Last paragraph: strong statement that sums up what you’re expecting to see
 - Hypothesis: “Bats eat bugs, and forests have lots of bugs. Therefore, . . .”
 - Prediction: “. . . we should see more bat foraging activity in forests”

Methods

- Establish how you collected the data, and how you analyzed it

Methods

- Establish how you collected the data, and how you analyzed it
 - This defends against criticism of your model or your data, and makes your results more believable

Methods

- Establish how you collected the data, and how you analyzed it
 - This defends against criticism of your model or your data, and makes your results more believable
- The detail you use depends how “unusual” your model is, which depends on your audience

Methods

- Establish how you collected the data, and how you analyzed it
 - This defends against criticism of your model or your data, and makes your results more believable
- The detail you use depends how “unusual” your model is, which depends on your audience
- Clarify what the dependent, independent variables, and random effects in your models are

Methods

- Establish how you collected the data, and how you analyzed it
 - This defends against criticism of your model or your data, and makes your results more believable
- The detail you use depends how “unusual” your model is, which depends on your audience
- Clarify what the dependent, independent variables, and random effects in your models are
- Sometimes you can just use the actual R model formula:

Methods

- Establish how you collected the data, and how you analyzed it
 - This defends against criticism of your model or your data, and makes your results more believable
- The detail you use depends how “unusual” your model is, which depends on your audience
- Clarify what the dependent, independent variables, and random effects in your models are
- Sometimes you can just use the actual R model formula:
 - “I fit the model using `lm` in R using the following model structure for bat counts (while accounting for unicorns):”

Methods

- Establish how you collected the data, and how you analyzed it
 - This defends against criticism of your model or your data, and makes your results more believable
- The detail you use depends how “unusual” your model is, which depends on your audience
- Clarify what the dependent, independent variables, and random effects in your models are
- Sometimes you can just use the actual R model formula:
 - “I fit the model using `lm` in R using the following model structure for bat counts (while accounting for unicorns):”
 - `lm(batCounts ~ forest + unicorns)`

Results

- Brief summary of what you collected¹

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”
- Present your results as an answer to the questions that you posed in the Introduction.

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”
- Present your results as an answer to the questions that you posed in the Introduction.
 - “Forest cover caused an increase of 3 bats for each 10% of forest ($p < 0.001$), while unicorns had no effect ($p = 0.19$)”

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”
- Present your results as an answer to the questions that you posed in the Introduction.
 - “Forest cover caused an increase of 3 bats for each 10% of forest ($p < 0.001$), while unicorns had no effect ($p = 0.19$)”
 - Try to keep the language as normal and direct as possible

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”
- Present your results as an answer to the questions that you posed in the Introduction.
 - “Forest cover caused an increase of 3 bats for each 10% of forest ($p < 0.001$), while unicorns had no effect ($p = 0.19$)”
 - Try to keep the language as normal and direct as possible
 - Having tons of p-values and other numbers can make the text hard to read

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”
- Present your results as an answer to the questions that you posed in the Introduction.
 - “Forest cover caused an increase of 3 bats for each 10% of forest ($p < 0.001$), while unicorns had no effect ($p = 0.19$)”
 - Try to keep the language as normal and direct as possible
 - Having tons of p-values and other numbers can make the text hard to read
- If something weird happened, just say it and move on. Speculate on *why* in the Discussion.

¹Can sometimes go at the end of the Methods

Results

- Brief summary of what you collected¹
 - “I caught 420 bats at my 69 sampling sites.”
- Present your results as an answer to the questions that you posed in the Introduction.
 - “Forest cover caused an increase of 3 bats for each 10% of forest ($p < 0.001$), while unicorns had no effect ($p = 0.19$)”
 - Try to keep the language as normal and direct as possible
 - Having tons of p-values and other numbers can make the text hard to read
- If something weird happened, just say it and move on. Speculate on *why* in the Discussion.
 - “Surprisingly, frogs had a negative effect on bat counts.”

¹Can sometimes go at the end of the Methods

Discussion

- Relate your results to your research question. Did your results match your expectations?

Discussion

- Relate your results to your research question. Did your results match your expectations?
- Move from specific (“Bats need bugs”) to general (“Animals need food”); opposite of the Introduction

Discussion

- Relate your results to your research question. Did your results match your expectations?
- Move from specific (“Bats need bugs”) to general (“Animals need food”); opposite of the Introduction
- Put the Results you found into the context of the rest of the literature. If your results contradict other studies, why do you think that occurred?

Discussion

- Relate your results to your research question. Did your results match your expectations?
- Move from specific (“Bats need bugs”) to general (“Animals need food”); opposite of the Introduction
- Put the Results you found into the context of the rest of the literature. If your results contradict other studies, why do you think that occurred?
 - “Barclay et al. (2017) showed that bats don't like forests, but our results may differ because. . .”

Discussion

- Relate your results to your research question. Did your results match your expectations?
- Move from specific (“Bats need bugs”) to general (“Animals need food”); opposite of the Introduction
- Put the Results you found into the context of the rest of the literature. If your results contradict other studies, why do you think that occurred?
 - “Barclay et al. (2017) showed that bats don't like forests, but our results may differ because. . .”
- **So what?** What new things have we learned? How might this affect theory or practice? Should non-bat people pay attention to this paper?

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once
 - Useful for showing information on large numbers of coefficients

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once
 - Useful for showing information on large numbers of coefficients
 - If you have lots of models, `library(broom)` provides summaries of all of them at once

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once
 - Useful for showing information on large numbers of coefficients
 - If you have lots of models, `library(broom)` provides summaries of all of them at once
- Tables and figures (+ captions) should be readable without knowing the rest of the text

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once
 - Useful for showing information on large numbers of coefficients
 - If you have lots of models, `library(broom)` provides summaries of all of them at once
- Tables and figures (+ captions) should be readable without knowing the rest of the text

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once
 - Useful for showing information on large numbers of coefficients
 - If you have lots of models, `library(broom)` provides summaries of all of them at once
- Tables and figures (+ captions) should be readable without knowing the rest of the text

Suggestions:

- ① Choose 2 or 3 figures and tables to be the **Main Characters** in your Results section.

Figures and Tables

- Figures can be excellent tools for telling your story, but. . .
 - Figures take up lots of room, cost \$ in publications, and can overwhelm the reader if there are too many
 - Many resources for good figure design: aim to minimize extra information
- Tables are kind of boring, but are great for conveying lots of numbers at once
 - Useful for showing information on large numbers of coefficients
 - If you have lots of models, `library(broom)` provides summaries of all of them at once
- Tables and figures (+ captions) should be readable without knowing the rest of the text

Suggestions:

- ① Choose 2 or 3 figures and tables to be the **Main Characters** in your Results section.
- ② Use them to illustrate what your models show and move the rest into a supplemental or appendix.

Title and Abstract

- Title: “Advertisement” of your study topic and results

Title and Abstract

- Title: “Advertisement” of your study topic and results
 - *Why should the reader read any further?*

Title and Abstract

- Title: “Advertisement” of your study topic and results
 - *Why should the reader read any further?*
- Abstract: quickly and effectively tells the reader what the paper is about

Title and Abstract

- Title: “Advertisement” of your study topic and results
 - *Why should the reader read any further?*
- Abstract: quickly and effectively tells the reader what the paper is about
 - Usually follows the IMRaD format order

Title and Abstract

- Title: “Advertisement” of your study topic and results
 - *Why should the reader read any further?*
- Abstract: quickly and effectively tells the reader what the paper is about
 - Usually follows the IMRaD format order
 - Not a movie trailer: spoilers are expected!

Title and Abstract

- Title: “Advertisement” of your study topic and results
 - *Why should the reader read any further?*
- Abstract: quickly and effectively tells the reader what the paper is about
 - Usually follows the IMRaD format order
 - Not a movie trailer: spoilers are expected!
- Keywords: extra words that could help search engine results

Give your paper a score!

- How did they do on each section? Did any of them not live up to expectations?

Give your paper a score!

- How did they do on each section? Did any of them not live up to expectations?
- How is the style of the paper?

Give your paper a score!

- How did they do on each section? Did any of them not live up to expectations?
- How is the style of the paper?
 - Did the writing make it difficult to read? Was it too long/short?

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”
 - This is a relevant question, either practically or theoretically

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”
 - This is a relevant question, either practically or theoretically
- ② The question can be answered

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”
 - This is a relevant question, either practically or theoretically
- ② The question can be answered
 - “If you did such-and-such an experiment or observation. . .”

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”
 - This is a relevant question, either practically or theoretically
- ② The question can be answered
 - “If you did such-and-such an experiment or observation. . .”
- ③ You can do the work to answer it

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”
 - This is a relevant question, either practically or theoretically
- ② The question can be answered
 - “If you did such-and-such an experiment or observation. . .”
- ③ You can do the work to answer it
 - “I have a plan to carry out this experiment or observation. . .”

Note on research proposals

- Similar structure to the Introduction and Methods of an IMRaD manuscript
- Results and Discussion are shorter, and focus on *what you think you will find*, and *why that matters*
- In the assignment for this class, the Results and Discussion will include an analysis of simulated or archived data, both aimed at answering the same questions as above

Overall, a proposal must convince the reader that:

- ① The question is important
 - “Other people have asked these questions, but they haven’t answered this one. . .”
 - This is a relevant question, either practically or theoretically
- ② The question can be answered
 - “If you did such-and-such an experiment or observation. . .”
- ③ You can do the work to answer it
 - “I have a plan to carry out this experiment or observation. . .”
 - Formal proposals include a timeline and a budget, as well as a list of personnel

Part 2: Writing about statistics

Models as evidence for arguments

- Scientific discourse can be thought of as a series of logical arguments

Show the *bare minimum* number of statistics needed to convince people. If it's not relevant to your story, move it somewhere else.

Models as evidence for arguments

- Scientific discourse can be thought of as a series of logical arguments
- When making an argument, you bring evidence to support your claims

Show the *bare minimum* number of statistics needed to convince people. If it's not relevant to your story, move it somewhere else.

Models as evidence for arguments

- Scientific discourse can be thought of as a series of logical arguments
- When making an argument, you bring evidence to support your claims
- We use experiments/observations, mathematics, and previous literature to support our claims

Show the *bare minimum* number of statistics needed to convince people. If it's not relevant to your story, move it somewhere else.

Models as evidence for arguments

- Scientific discourse can be thought of as a series of logical arguments
- When making an argument, you bring evidence to support your claims
- We use experiments/observations, mathematics, and previous literature to support our claims
 - None of these are assumption-free: The reader must be convinced that these are appropriate!

Show the *bare minimum* number of statistics needed to convince people. If it's not relevant to your story, move it somewhere else.

Models as evidence for arguments

- Scientific discourse can be thought of as a series of logical arguments
- When making an argument, you bring evidence to support your claims
- We use experiments/observations, mathematics, and previous literature to support our claims
 - None of these are assumption-free: The reader must be convinced that these are appropriate!
- Models also act as a *piece of evidence*, translating raw data into “ammunition” for your claim

Show the *bare minimum* number of statistics needed to convince people. If it's not relevant to your story, move it somewhere else.

Models as evidence for arguments

- Scientific discourse can be thought of as a series of logical arguments
- When making an argument, you bring evidence to support your claims
- We use experiments/observations, mathematics, and previous literature to support our claims
 - None of these are assumption-free: The reader must be convinced that these are appropriate!
- Models also act as a *piece of evidence*, translating raw data into “ammunition” for your claim
 - Model structure and performance checks (residual plots, etc.) should *also* convince the reader that this is believable

Show the *bare minimum* number of statistics needed to convince people. If it's not relevant to your story, move it somewhere else.

Example arguments:

- Premise 1: Bats eat bugs

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

Example 1:

- Evidence: The model of my data
supports this claim

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

Example 1:

- Evidence: The model of my data **supports this claim**
- Conclusion: This means that our understanding of bugs, bats, and forests is pretty good

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

Example 1:

- Evidence: The model of my data **supports this claim**
- Conclusion: This means that our understanding of bugs, bats, and forests is pretty good

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

Example 1:

- Evidence: The model of my data **supports this claim**
- Conclusion: This means that our understanding of bugs, bats, and forests is pretty good

Example 2:

- Evidence: The model of my data **does not support this claim**

¹Inductive reasoning

Example arguments:

- Premise 1: Bats eat bugs
- Premise 2: Forests have lots of bugs
- Claim: Therefore, bats should prefer forests ²

Example 1:

- Evidence: The model of my data **supports this claim**
- Conclusion: This means that our understanding of bugs, bats, and forests is pretty good

Example 2:

- Evidence: The model of my data **does not support this claim**
- Conclusion: One of these premises is wrong, or we left out an important premise

¹Inductive reasoning

Models as reflections of reality

- Models are meant to reflect an *underlying biological process*

Models as reflections of reality

- Models are meant to reflect an *underlying biological process*
- Things like effect size (mean/SE) reflect the relative strength of the factors involved

Models as reflections of reality

- Models are meant to reflect an *underlying biological process*
- Things like effect size (mean/SE) reflect the relative strength of the factors involved
- Things like R^2 reflect how well the model fits the data *overall*

Models as reflections of reality

- Models are meant to reflect an *underlying biological process*
- Things like effect size (mean/SE) reflect the relative strength of the factors involved
- Things like R^2 reflect how well the model fits the data *overall*
- Causality is implied, but has to be justified

Models as reflections of reality

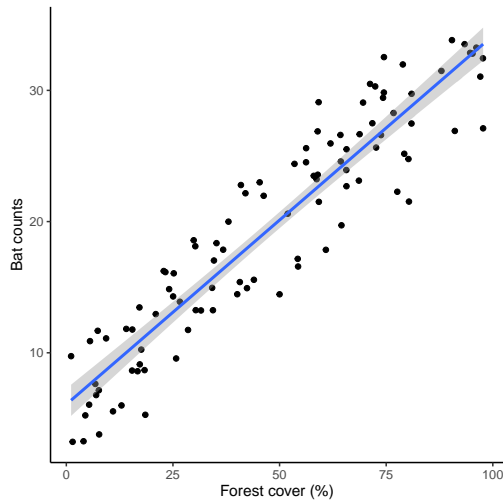
- Models are meant to reflect an *underlying biological process*
- Things like effect size (mean/SE) reflect the relative strength of the factors involved
- Things like R^2 reflect how well the model fits the data *overall*
- Causality is implied, but has to be justified
- “Keep your eye on the biology!”

Models as reflections of reality

- Models are meant to reflect an *underlying biological process*
- Things like effect size (mean/SE) reflect the relative strength of the factors involved
- Things like R^2 reflect how well the model fits the data *overall*
- Causality is implied, but has to be justified
- “Keep your eye on the biology!”

Models as reflections of reality

- Models are meant to reflect an *underlying biological process*
- Things like effect size (mean/SE) reflect the relative strength of the factors involved
- Things like R^2 reflect how well the model fits the data *overall*
- Causality is implied, but has to be justified
- “Keep your eye on the biology!”



What might the underlying physical process be here?

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat
- Interpretation can be:

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat
- Interpretation can be:
 - Yes/no: “Is there any relationship?”

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat
- Interpretation can be:
 - Yes/no: “Is there any relationship?”
 - Directional: “Is the relationship positive?”

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat
- Interpretation can be:
 - Yes/no: “Is there any relationship?”
 - Directional: “Is the relationship positive?”
 - Magnitude: “How big is the slope?”

³For GLMs, slopes are in log or log-odds (logit) units

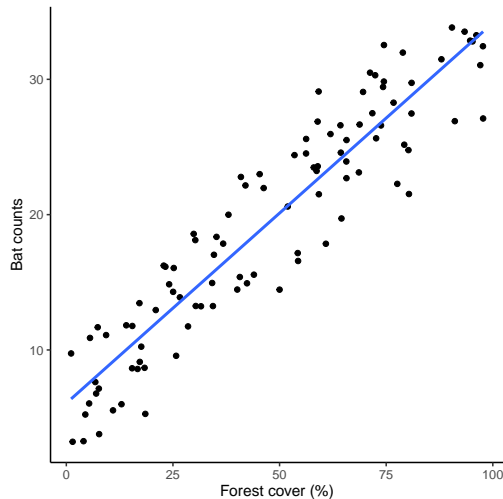
Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat
- Interpretation can be:
 - Yes/no: “Is there any relationship?”
 - Directional: “Is the relationship positive?”
 - Magnitude: “How big is the slope?”

³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 1: coefficients

- Slopes and intercepts have physical interpretations
 - Intercept: How many bats at 0 % forest?
 - Slope³: + 1 % forest = + 1 bat
- Interpretation can be:
 - Yes/no: “Is there any relationship?”
 - Directional: “Is the relationship positive?”
 - Magnitude: “How big is the slope?”



³For GLMs, slopes are in log or log-odds (logit) units

Evidence type 2: variance

- Variance has a physical interpretation

Evidence type 2: variance

- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?

Evidence type 2: variance

- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?
- R^2 relates actual to modeled variance: what % of variance does your model explain?

Evidence type 2: variance

- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?
- R^2 relates actual to modeled variance: what % of variance does your model explain?
- GLMs: different distributions model variance differently

Evidence type 2: variance

- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?
- R^2 relates actual to modeled variance: what % of variance does your model explain?
- GLMs: different distributions model variance differently
- Hierarchical models deal with many levels of variance

Evidence type 2: variance

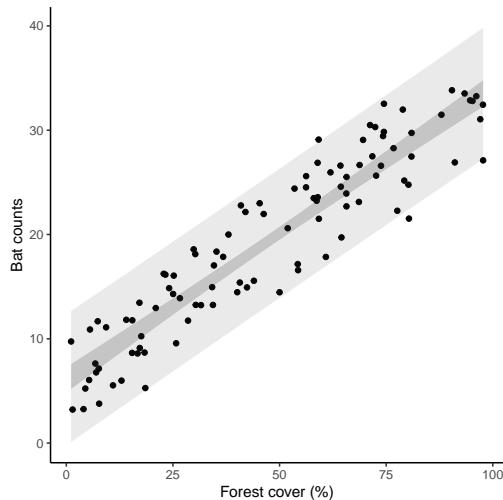
- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?
- R^2 relates actual to modeled variance: what % of variance does your model explain?
- GLMs: different distributions model variance differently
- Hierarchical models deal with many levels of variance
 - Tells you where the variance in your system is coming from

Evidence type 2: variance

- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?
- R^2 relates actual to modeled variance: what % of variance does your model explain?
- GLMs: different distributions model variance differently
- Hierarchical models deal with many levels of variance
 - Tells you where the variance in your system is coming from

Evidence type 2: variance

- Variance has a physical interpretation
 - What is the variation in bat counts at a given level of forest?
- R^2 relates actual to modeled variance: what % of variance does your model explain?
- GLMs: different distributions model variance differently
- Hierarchical models deal with many levels of variance
 - Tells you where the variance in your system is coming from



Example write-up

Say we fit a model of bat counts that looks like this

```
##  
## Call:  
## lm(formula = batAbund ~ forest + unicorns, data = d1)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -6.5932 -2.2268  0.1891  2.7222  6.4240   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept)  5.153103   0.811636   6.349 6.97e-09 ***  
## forest       0.280178   0.010959  25.567 < 2e-16 ***  
## unicorns     0.005906   0.003505   1.685  0.0952 .    
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 3.079 on 97 degrees of freedom  
## Multiple R-squared:  0.8721, Adjusted R-squared:  0.8695   
## F-statistic: 330.7 on 2 and 97 DF,  p-value: < 2.2e-16
```

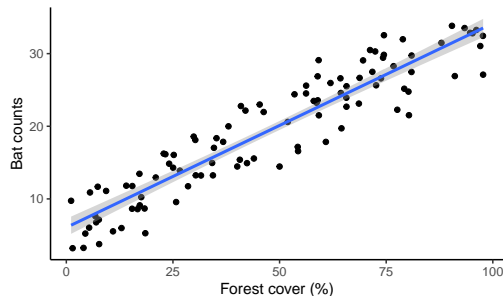
Methods:

"I collected data from 100 sites around Calgary, and recorded..."

"I used a linear model to estimate the effect of forest cover and unicorns on bat abundance. Models were fit using `lm()` in R and were checked for..."

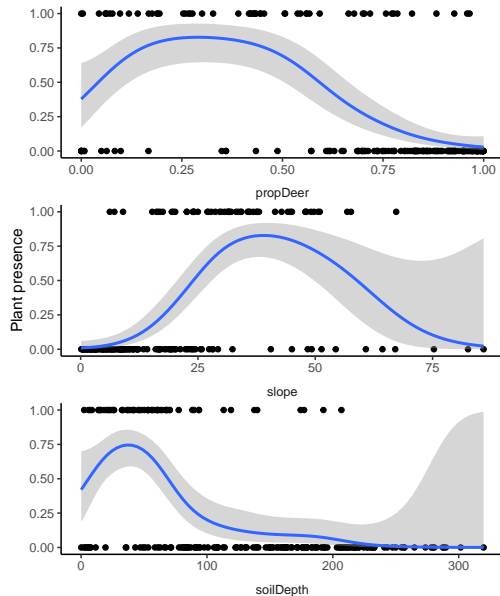
Results:

"My model identified a strong, positive effect of forest cover, with each additional 10% of forest cover adding an additional 3 bats (Figure 1, $p < 0.0001$), while the effect of unicorns was weak ($p = 0.19$). The model also explained $\sim 88\%$ of the variance in bat abundance, further highlighting the importance of forest cover to bats..."



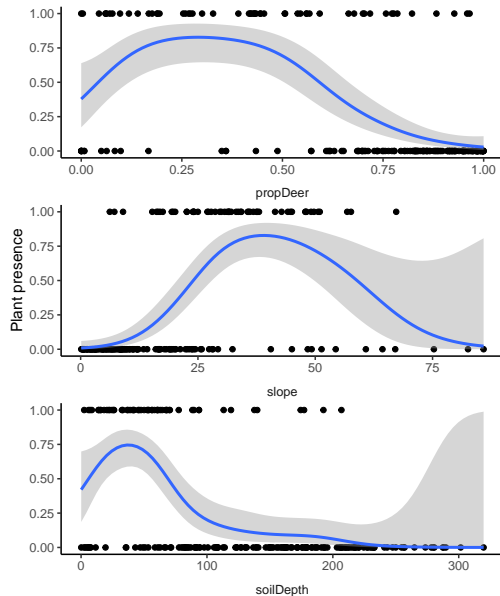
Challenge: analyse and explain

- I have a dataset of plant abundance `plantDat.csv` (see [here](#)) containing records of plant presence/absence, deer browsing, soil depth, and slope



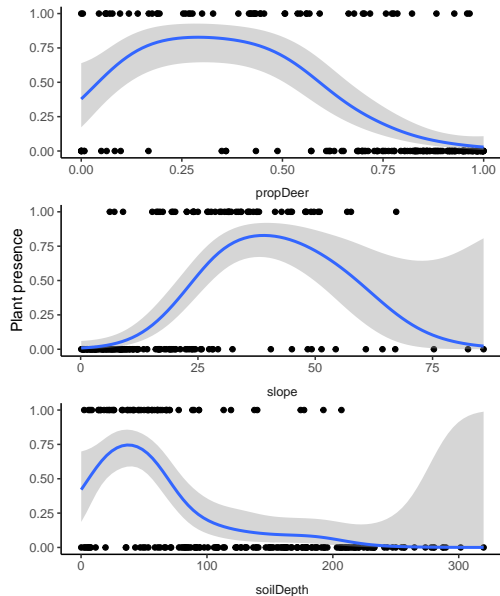
Challenge: analyse and explain

- I have a dataset of plant abundance `plantDat.csv` (see [here](#)) containing records of plant presence/absence, deer browsing, soil depth, and slope
- Come up with a reasonable set of hypotheses about plant presence/absence (given the available data)



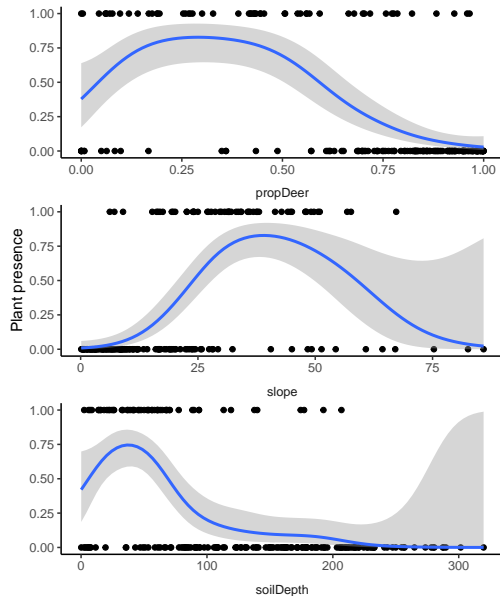
Challenge: analyse and explain

- I have a dataset of plant abundance `plantDat.csv` (see [here](#)) containing records of plant presence/absence, deer browsing, soil depth, and slope
- Come up with a reasonable set of hypotheses about plant presence/absence (given the available data)
- Fit a model that tests those hypotheses, verify the model, and explain the model approach in plain English



Challenge: analyse and explain

- I have a dataset of plant abundance `plantDat.csv` (see [here](#)) containing records of plant presence/absence, deer browsing, soil depth, and slope
- Come up with a reasonable set of hypotheses about plant presence/absence (given the available data)
- Fit a model that tests those hypotheses, verify the model, and explain the model approach in plain English
- Explain the model results, and make an accompanying “Figure 1” to go with the results



My (personal) order of writing a paper

- ① Methods: I usually write this section first, as it gets me “warmed up” for the rest of it⁴

⁴You can even write it before you collect your data!

My (personal) order of writing a paper

- ① Methods: I usually write this section first, as it gets me “warmed up” for the rest of it⁴
- ② Results: I write this section after I write the Methods section

⁴You can even write it before you collect your data!

My (personal) order of writing a paper

- ① Methods: I usually write this section first, as it gets me “warmed up” for the rest of it⁴
- ② Results: I write this section after I write the Methods section
- ③ Discussion: I write this after my model Results. Here you can name-drop all the relevant papers you’ve read (make sure they’re setup in the Introduction first)

⁴You can even write it before you collect your data!

My (personal) order of writing a paper

- ① Methods: I usually write this section first, as it gets me “warmed up” for the rest of it⁴
- ② Results: I write this section after I write the Methods section
- ③ Discussion: I write this after my model Results. Here you can name-drop all the relevant papers you’ve read (make sure they’re setup in the Introduction first)
- ④ Introduction: I find this section the trickiest to write, so I usually write it last

⁴You can even write it before you collect your data!

My (personal) order of writing a paper

- ① Methods: I usually write this section first, as it gets me “warmed up” for the rest of it⁴
- ② Results: I write this section after I write the Methods section
- ③ Discussion: I write this after my model Results. Here you can name-drop all the relevant papers you’ve read (make sure they’re setup in the Introduction first)
- ④ Introduction: I find this section the trickiest to write, so I usually write it last
- ⑤ Title and Abstract: After everything else is done, you can *advertise and summarize*!

⁴You can even write it before you collect your data!

Part 3: Peer review

How do journals work?

- Journals are usually society publications (BES, ESA, IEEE) run out of academic publishing companies (Wiley, Elsevier, Taylor & Francis)

How do journals work?

- Journals are usually society publications (BES, ESA, IEEE) run out of academic publishing companies (Wiley, Elsevier, Taylor & Francis)
- Most journals have a *lead editor* and an *editorial board*. These will be the people who will first see your submitted manuscript

How do journals work?

- Journals are usually society publications (BES, ESA, IEEE) run out of academic publishing companies (Wiley, Elsevier, Taylor & Francis)
- Most journals have a *lead editor* and an *editorial board*. These will be the people who will first see your submitted manuscript
 - Peer review is done for free by working scientists

How do journals work?

- Journals are usually society publications (BES, ESA, IEEE) run out of academic publishing companies (Wiley, Elsevier, Taylor & Francis)
- Most journals have a *lead editor* and an *editorial board*. These will be the people who will first see your submitted manuscript
 - Peer review is done for free by working scientists
- Traditional publishing: costs you nothing, costs the U of C library \$ (depending on subscription)

How do journals work?

- Journals are usually society publications (BES, ESA, IEEE) run out of academic publishing companies (Wiley, Elsevier, Taylor & Francis)
- Most journals have a *lead editor* and an *editorial board*. These will be the people who will first see your submitted manuscript
 - Peer review is done for free by working scientists
- Traditional publishing: costs you nothing, costs the U of C library \$ (depending on subscription)
- Open-access publishing: costs you \$1000-5000 depending on the journal, but then anyone can read it

How do journals work?

- Journals are usually society publications (BES, ESA, IEEE) run out of academic publishing companies (Wiley, Elsevier, Taylor & Francis)
- Most journals have a *lead editor* and an *editorial board*. These will be the people who will first see your submitted manuscript
 - Peer review is done for free by working scientists
- Traditional publishing: costs you nothing, costs the U of C library \$ (depending on subscription)
- Open-access publishing: costs you \$1000-5000 depending on the journal, but then anyone can read it
 - Keep an eye out for predatory or “papermill” journals! Some sets of open-access journals (MDPI) have a *suspiciously fast* peer review process

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)
 - Check that the document conforms to the types of papers they publish

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)
 - Check that the document conforms to the types of papers they publish
 - Some journals are more lenient about the first submissions (e.g. just a pdf with simple formatting)

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)
 - Check that the document conforms to the types of papers they publish
 - Some journals are more lenient about the first submissions (e.g. just a pdf with simple formatting)
 - Double-blind journals require you to remove all identifying info (separate title page that the reviewer never sees)

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)
 - Check that the document conforms to the types of papers they publish
 - Some journals are more lenient about the first submissions (e.g. just a pdf with simple formatting)
 - Double-blind journals require you to remove all identifying info (separate title page that the reviewer never sees)
- Submit the article and wait for a response!

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)
 - Check that the document conforms to the types of papers they publish
 - Some journals are more lenient about the first submissions (e.g. just a pdf with simple formatting)
 - Double-blind journals require you to remove all identifying info (separate title page that the reviewer never sees)
- Submit the article and wait for a response!
 - Think about who you might recommend as a reviewer. Who would you want to read your paper?

OK, you've got a paper written! Now what?

- Identify a journal you'd like to submit it to
 - Which journals do you cite the most in the paper? Maybe one of those? Check their [Aims and Scope](#)
 - Helps to start thinking about it earlier, and have a tier-list
 - **Ask your supervisor!** They will have good experience with this
- Assemble the document in the way that the journal wants. Check their [Guidelines for Authors](#)
 - Check that the document conforms to the types of papers they publish
 - Some journals are more lenient about the first submissions (e.g. just a pdf with simple formatting)
 - Double-blind journals require you to remove all identifying info (separate title page that the reviewer never sees)
- Submit the article and wait for a response!
 - Think about who you might recommend as a reviewer. Who would you want to read your paper?
 - A *cover letter* helps convince the editor they should give your paper a chance

Peer Review Process



What the editor will do

- An editor will skim the paper and make sure that the topic is relevant. If not, your paper gets a *desk reject*

What the editor will do

- An editor will skim the paper and make sure that the topic is relevant. If not, your paper gets a *desk reject*
- If it looks generally OK, the editor will contact peer reviewers and ask them to review the paper

What the editor will do

- An editor will skim the paper and make sure that the topic is relevant. If not, your paper gets a *desk reject*
- If it looks generally OK, the editor will contact peer reviewers and ask them to review the paper
- Once they've gotten the comments back, the editor will assemble the comments, and read the paper a bit more to see if they agree with them

What the editor will do

- An editor will skim the paper and make sure that the topic is relevant. If not, your paper gets a *desk reject*
- If it looks generally OK, the editor will contact peer reviewers and ask them to review the paper
- Once they've gotten the comments back, the editor will assemble the comments, and read the paper a bit more to see if they agree with them
- They will contact you with their decision based on the reviewer's comments: *reject*, *accept with major revisions*, or *accept with minor revisions*

What the editor will do

- An editor will skim the paper and make sure that the topic is relevant. If not, your paper gets a *desk reject*
- If it looks generally OK, the editor will contact peer reviewers and ask them to review the paper
- Once they've gotten the comments back, the editor will assemble the comments, and read the paper a bit more to see if they agree with them
- They will contact you with their decision based on the reviewer's comments: *reject*, *accept with major revisions*, or *accept with minor revisions*
 - They may use *reject and resubmit*, depending on the journal

What the editor will do

- An editor will skim the paper and make sure that the topic is relevant. If not, your paper gets a *desk reject*
- If it looks generally OK, the editor will contact peer reviewers and ask them to review the paper
- Once they've gotten the comments back, the editor will assemble the comments, and read the paper a bit more to see if they agree with them
- They will contact you with their decision based on the reviewer's comments: *reject*, *accept with major revisions*, or *accept with minor revisions*
 - They may use *reject and resubmit*, depending on the journal
 - They may temper the claims from *bad or rude peer reviewers*, or may remove them entirely!

My (personal) style of peer review

- Read the paper once through without writing anything down

My (personal) style of peer review

- Read the paper once through without writing anything down
- Go back through each section and write general “overall” comments

My (personal) style of peer review

- Read the paper once through without writing anything down
- Go back through each section and write general “overall” comments
 - e.g. “Intro needs to be trimmed down”, “Results section is disorganized”, “I don’t understand the relevance of X”

My (personal) style of peer review

- Read the paper once through without writing anything down
- Go back through each section and write general “overall” comments
 - e.g. “Intro needs to be trimmed down”, “Results section is disorganized”, “I don’t understand the relevance of X”
- Write line-by-line comments where needed

My (personal) style of peer review

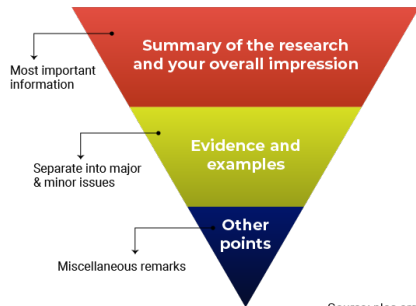
- Read the paper once through without writing anything down
- Go back through each section and write general “overall” comments
 - e.g. “Intro needs to be trimmed down”, “Results section is disorganized”, “I don’t understand the relevance of X”
- Write line-by-line comments where needed
 - e.g. “L40: change insect to arthropod”, “L89: How does this test work, and is it commonly used?”, “L112: Citation needed, perhaps Smith et al. 2020?”

My (personal) style of peer review

- Read the paper once through without writing anything down
- Go back through each section and write general “overall” comments
 - e.g. “Intro needs to be trimmed down”, “Results section is disorganized”, “I don’t understand the relevance of X”
- Write line-by-line comments where needed
 - e.g. “L40: change insect to arthropod”, “L89: How does this test work, and is it commonly used?”, “L112: Citation needed, perhaps Smith et al. 2020?”
- Think about what could improve the paper, and provide a suggested way forward where possible! (e.g. “I suggest moving this paragraph to here. . .”)

My (personal) style of peer review (cont.)

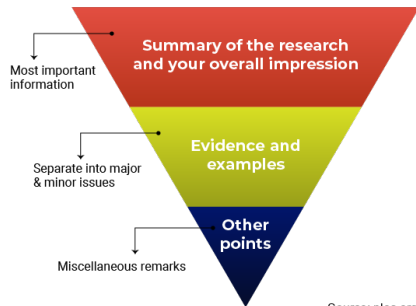
- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.



Source: plos.org

My (personal) style of peer review (cont.)

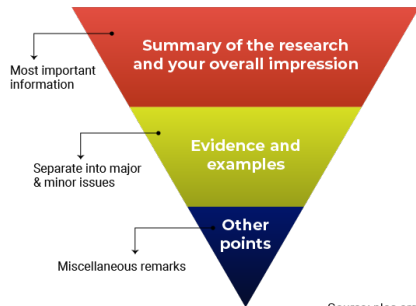
- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)



Source: plos.org

My (personal) style of peer review (cont.)

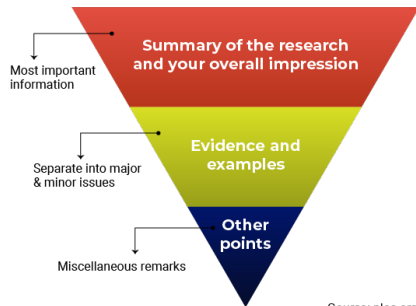
- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)
- Make a reject/accept decision on the paper. Try to be as objective as possible:



Source: plos.org

My (personal) style of peer review (cont.)

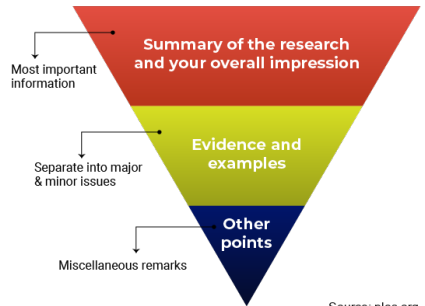
- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)
- Make a reject/accept decision on the paper. Try to be as objective as possible:
 - “This person didn’t do exactly what I would have, but does it matter to the results or overall story?”



Source: plos.org

My (personal) style of peer review (cont.)

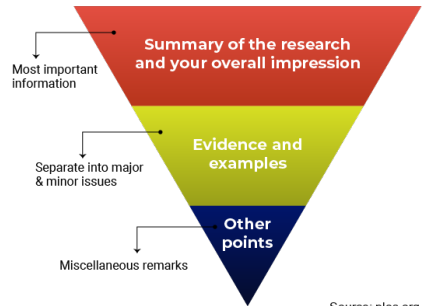
- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)
- Make a reject/accept decision on the paper. Try to be as objective as possible:
 - “This person didn’t do exactly what I would have, but does it matter to the results or overall story?”
 - “Maybe I don’t think these results are very interesting, but are they believable given the evidence?”



Source: plos.org

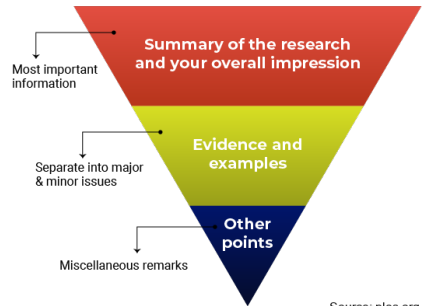
My (personal) style of peer review (cont.)

- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)
- Make a reject/accept decision on the paper. Try to be as objective as possible:
 - “This person didn’t do exactly what I would have, but does it matter to the results or overall story?”
 - “Maybe I don’t think these results are very interesting, but are they believable given the evidence?”
- If the paper is accepted, how much time will it take to do revisions? (Major vs Minor)



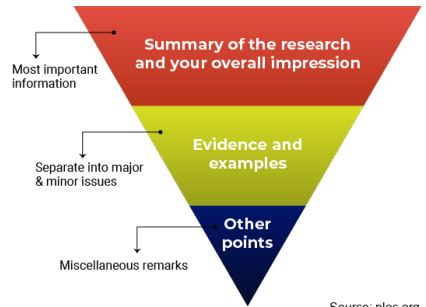
My (personal) style of peer review (cont.)

- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)
- Make a reject/accept decision on the paper. Try to be as objective as possible:
 - “This person didn’t do exactly what I would have, but does it matter to the results or overall story?”
 - “Maybe I don’t think these results are very interesting, but are they believable given the evidence?”
- If the paper is accepted, how much time will it take to do revisions? (Major vs Minor)



My (personal) style of peer review (cont.)

- Put all the comments together in a single document, split into overall and line-by-line, and re-read your comments.
 - *Do you need to tone things down a bit?* (or tone them up)
- Make a reject/accept decision on the paper. Try to be as objective as possible:
 - “This person didn’t do exactly what I would have, but does it matter to the results or overall story?”
 - “Maybe I don’t think these results are very interesting, but are they believable given the evidence?”
- If the paper is accepted, how much time will it take to do revisions? (Major vs Minor)



There are many other approaches to doing peer review: see [here](#), [here](#), or [here](#))

Final remarks

- Good writing is re-writing

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style
 - Read books about science writing, and use techniques from other books you've read

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style
 - Read books about science writing, and use techniques from other books you’ve read
- Use your supervisor and committee members

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style
 - Read books about science writing, and use techniques from other books you’ve read
- Use your supervisor and committee members
 - They have a much wider picture of the field, and have lots of writing and editing experience

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style
 - Read books about science writing, and use techniques from other books you’ve read
- Use your supervisor and committee members
 - They have a much wider picture of the field, and have lots of writing and editing experience
 - This means that they can be a good stand-in for your audience

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style
 - Read books about science writing, and use techniques from other books you’ve read
- Use your supervisor and committee members
 - They have a much wider picture of the field, and have lots of writing and editing experience
 - This means that they can be a good stand-in for your audience

Final remarks

- Good writing is re-writing
 - What is obvious to you may not be obvious to your readers. Revision is annoying and painful, but it **will** help!
 - “[Good writing is:] Telepathy, of course” (Stephen King)
- Use the literature
 - There are tons of poorly-written papers out there, but was there a paper that you found easy to understand? Re-read it, and figure out why!
 - Check out how other scientists display their statistics, and imitate/avoid their style
 - Read books about science writing, and use techniques from other books you’ve read
- Use your supervisor and committee members
 - They have a much wider picture of the field, and have lots of writing and editing experience
 - This means that they can be a good stand-in for your audience

Remember: GOOD WRITING IS GOOD STORYTELLING