



# STOR 320 Modeling III

Lecture 26

Yao Li

Department of Statistics and Operations Research

UNC Chapel Hill

# Tutorial 12

- Instructions
  - Download Tutorial Zip
  - Unzip Folder
  - Required Packages
    - `library(tidyverse)`
    - `library(modelr)`
  - Open .Rmd File and Knit
- Daily Spanish River Data
  - W = Max Water Temperature
  - A = Max Air Temperature
  - L = River Identifier (31 Rivers)

# Part 3: Polynomial Model

- Polynomial Model
  - “Feature Engineering”
  - Generalized Additive Model
  - `Geom_smooth()` Fits a GAM when Fitting a Curve
  - Useful for Approximating Nonlinear Relationships
  - Dependent on Degree “k”
  - Goal: Choose Best “k”

# Part 3: Polynomial Model

- Formula Object in R
  - Special Notation
  - Helpful Table:

| Symbol | Example         | Meaning  |
|--------|-----------------|--|
| +      | +X              | include this variable  |
| -      | -X              | delete this variable   |
| :      | X:Z             | include the interaction between these variables                        |
| *      | X*Y             | include these variables and the interactions between them              |
|        | X   Z           | conditioning: include x given z  |
| ^      | (X + Z + W) ^ 3 | include these variables and all interactions up to three way           |
| I      | I (X*Z)         | as is: include a new variable consisting of these variables multiplied |
| 1      | X - 1           | intercept: delete the intercept (regress through the origin)           |

- We will Use the I() Function to Create New Variables Based Off Variables We Have

# Part 3: Polynomial Model

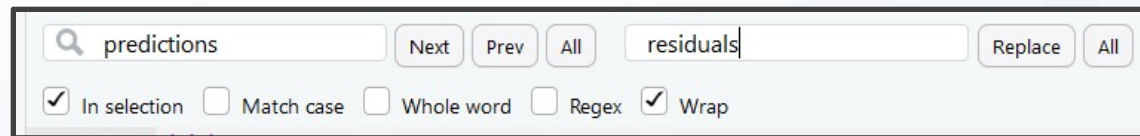
- Run Chunk 1
  - Fits 2<sup>nd</sup> Degree Polynomial
  - Fits 3<sup>rd</sup> Degree Polynomial
  - Fits 4<sup>th</sup> Degree Polynomial
- Run Chunk 2
  - Obtains Predictions Under the Different Polynomial Models

# Part 3: Polynomial Model

- Chunk 3
  - Code Needs Modification
  - Highlight Code

```
TRAIN4 =TRAIN3 %>%  
  add_predictions(poly2mod,var="poly2pred") %>%  
  add_predictions(poly3mod,var="poly3pred") %>%  
  add_predictions(poly4mod,var="poly4pred")  
  
TEST4 =TEST3 %>%  
  add_predictions(poly2mod,var="poly2pred") %>%  
  add_predictions(poly3mod,var="poly3pred") %>%  
  add_predictions(poly4mod,var="poly4pred")
```

- TRAIN3 -> TRAIN4 and etc.
- Use Ctrl+F (Find and Replace)
  - 'predictions' -> 'residuals'
  - 'pred' -> 'res'



Find and Replace dialog box showing the search for 'predictions' and replacement with 'residuals'. The 'In selection' checkbox is checked, and the 'Wrap' checkbox is also checked.

- Run Chunk 3 After Modifying

# Intermission

- Run Code Chunk
  - `save.image()` = Used to Save Workspace into .Rdata File
  - `load()` = Used to Load Workspace from .Rdata File
  - .Rdata = File Extension of R Workspace File (All Objects in Global Environment)

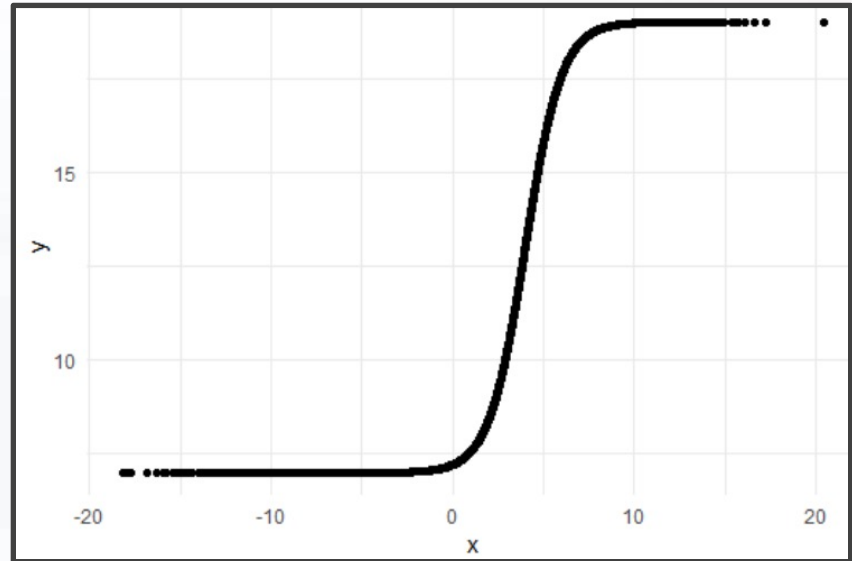
# Part 4: Logistic Model

- Logistic Model
  - “Smart” Model Based On Physical Relationship Between  $A$  and  $W$
  - Four Parameters
    - Controls the Shape of the Relationship
    - $a$  and  $b$
    - $c$  and  $d$
  - What Shape Do You Think This Function Makes?
    - Idea: Precalculus



# Part 4: Logistic Model

- Run Chunk 1
  - Plant that Seed
  - Example Model



- Parameter Investigation
  - What Does 7 Represent?
  - What Does 12 Represent?
  - What Does 4 Represent?
  - What Does 1 Represent?

# Part 4: Logistic Model

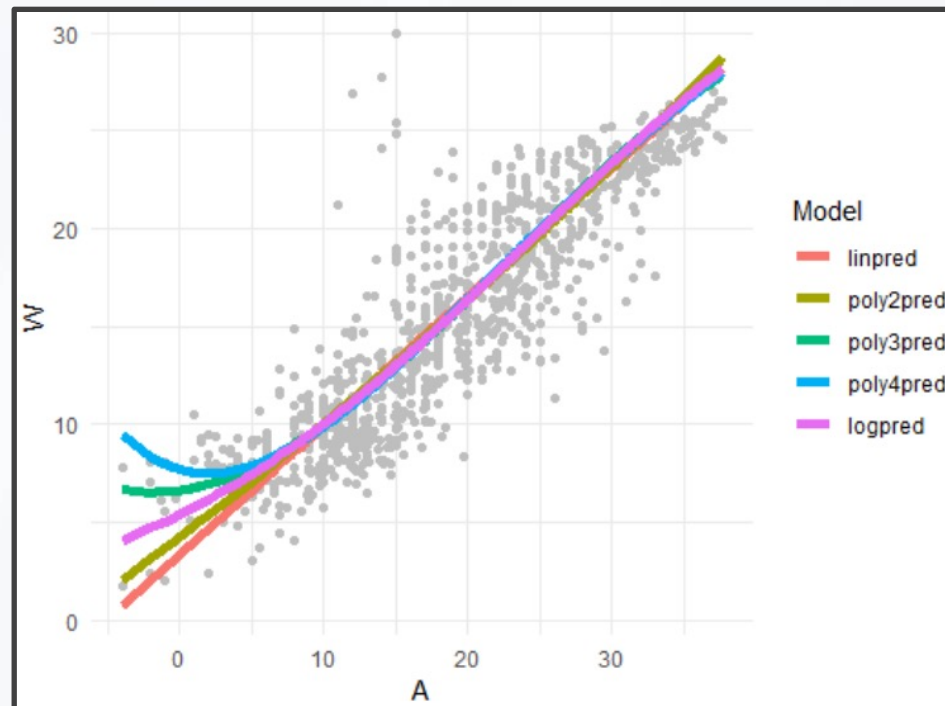
- Run Chunk 2
  - Creation of Modeling Function
  - Creation of MSE Function Specific to this Model
- Run Chunk 3
  - Use `optim()` Function With Smart Starting Values Based on Understanding of The Model
  - Finds Estimates Based on Minimization of MSE

# Part 4: Logistic Model

- Run Chunk 4
  - Use Logistic Model Function and Estimated Parameters from `optim()` to Obtain
    - Predictions
    - Residuals

# Part 5: Evaluation by Visualization

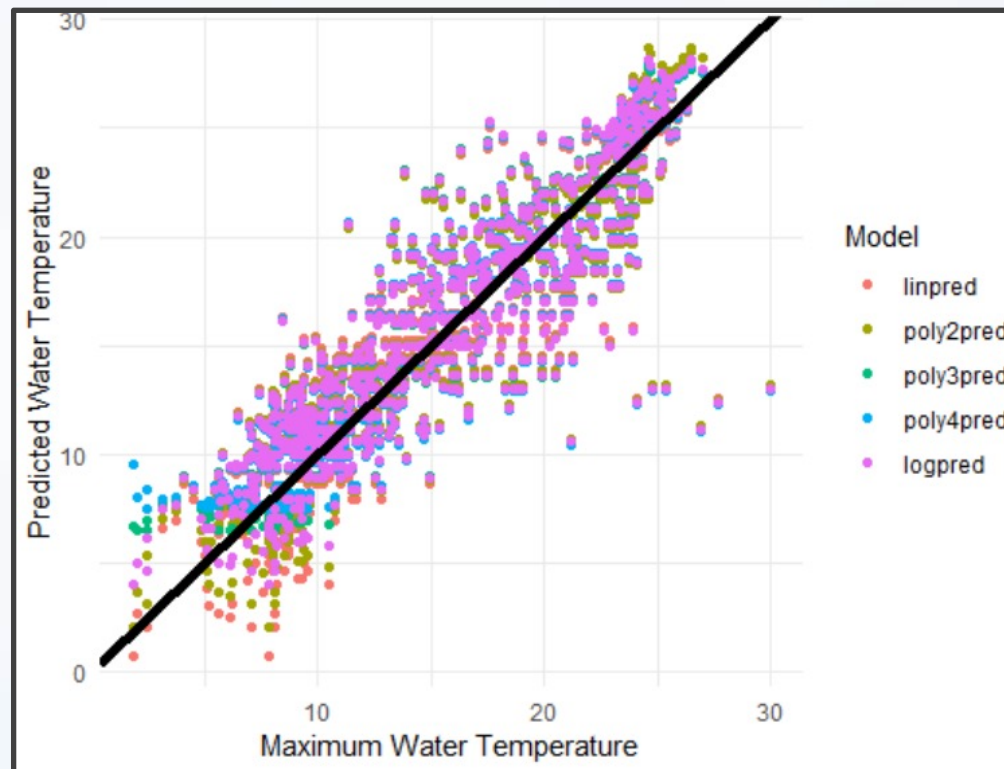
- Run Chunk 1
  - Plots of Different Models
  - What Can We Say About the Different Models?



- Which Model Would You Use?

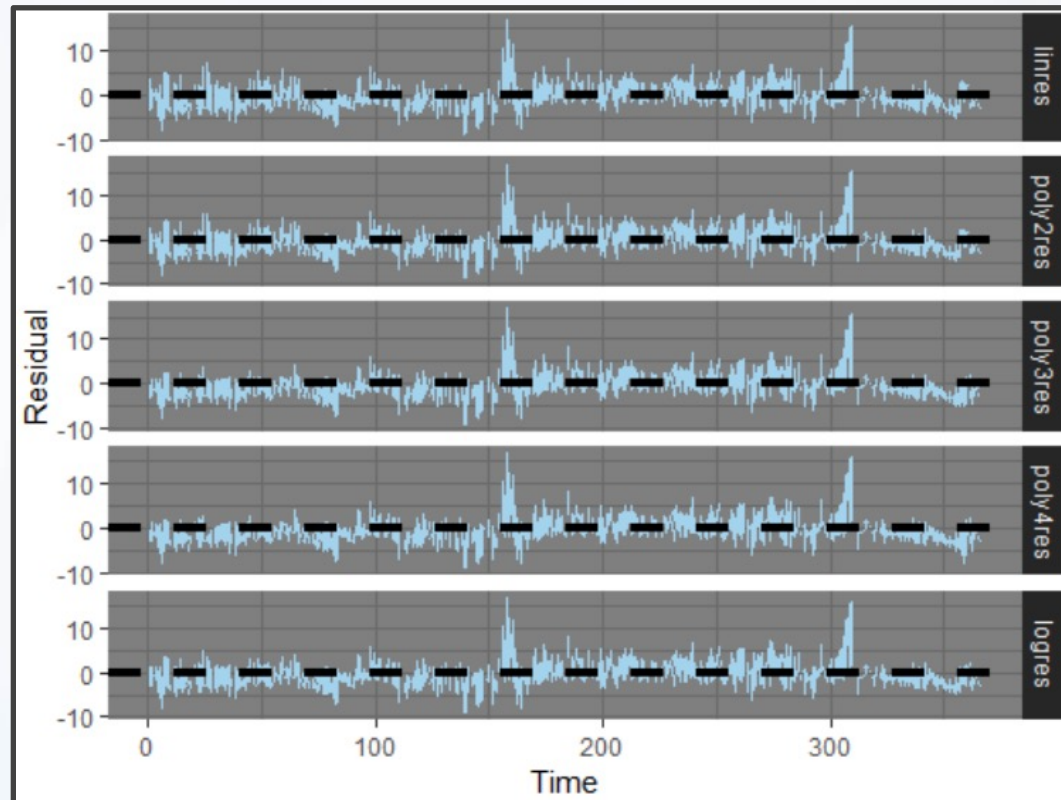
# Part 5: Evaluation by Visualization

- Run Chunk 2
  - Comparing Predictions vs Actual Maximum Water Temperatures
  - Models Give Similar Predictions



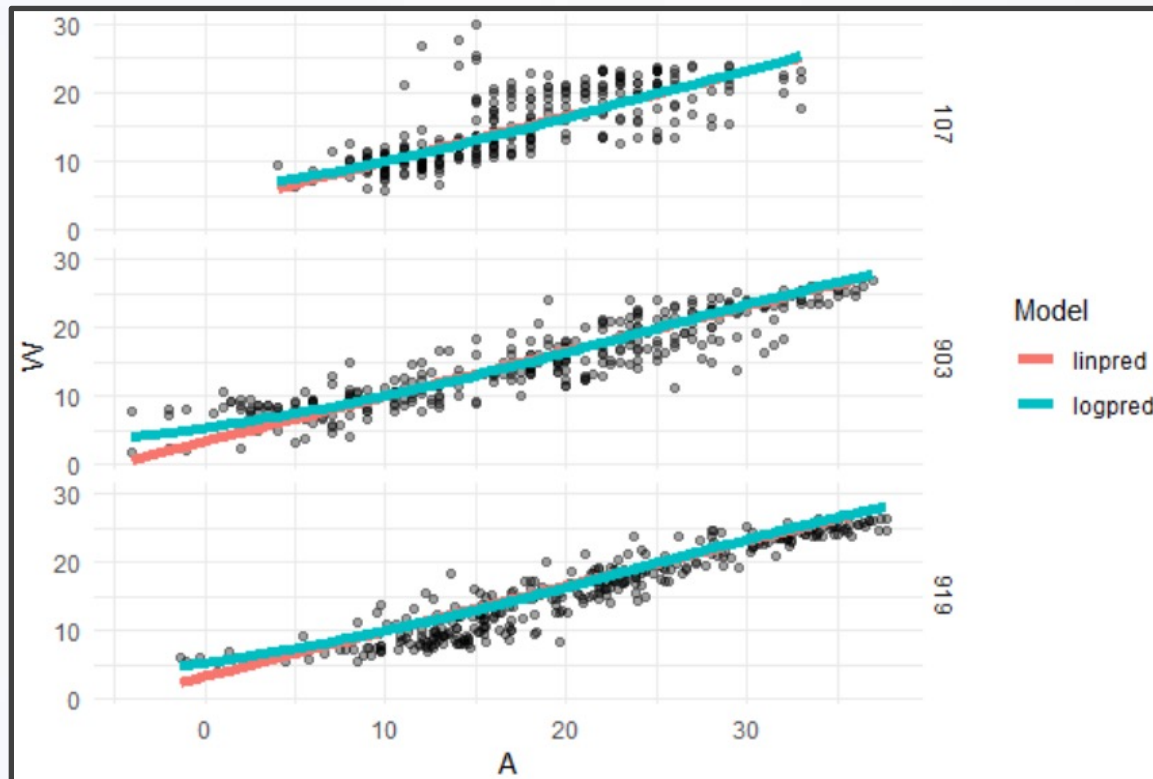
# Part 5: Evaluation by Visualization

- Run Chunk 3
  - Shows Residuals Under the 4 Models Plotted Over Time
  - What is the Problem?



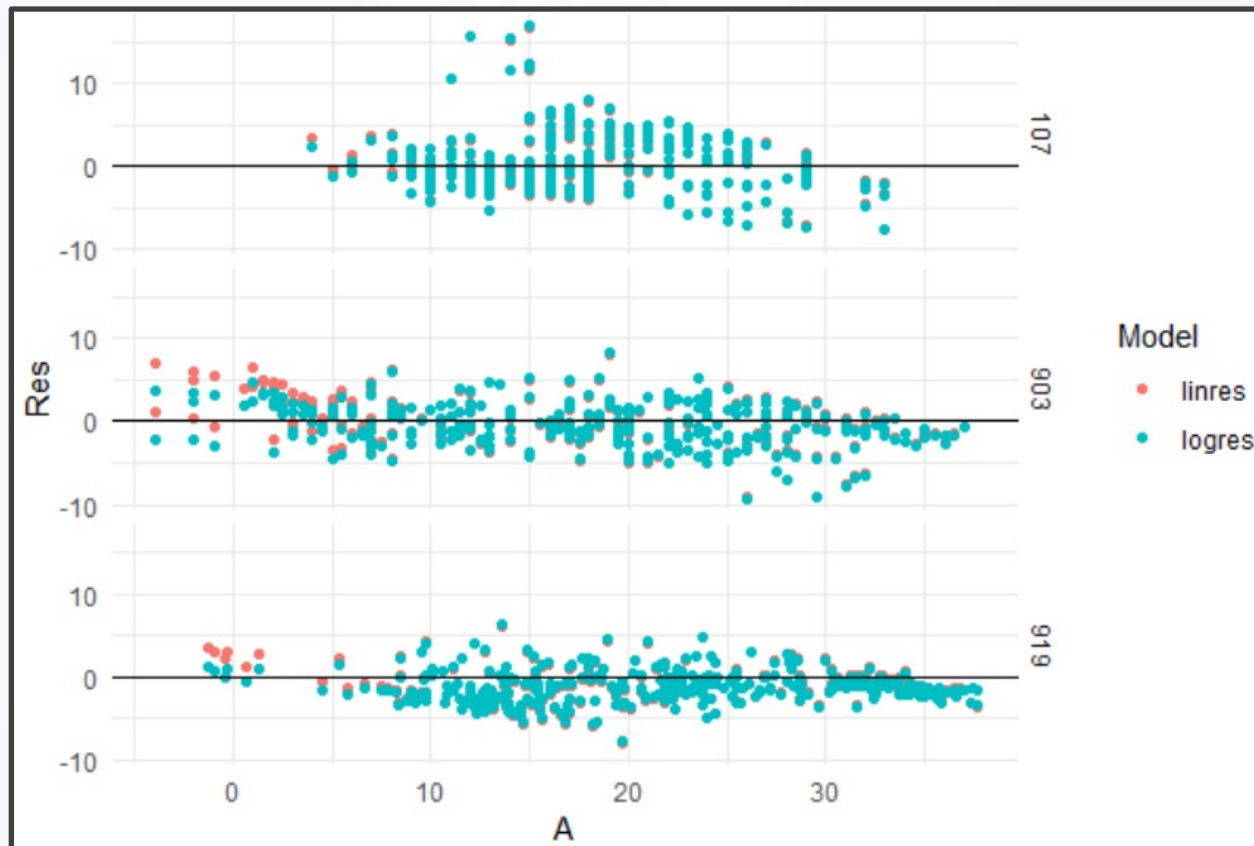
# Part 5: Evaluation by Visualization

- Run Chunk 4
  - Evaluate Models For the Three Locations Separately



# Part 5: Evaluation by Visualization

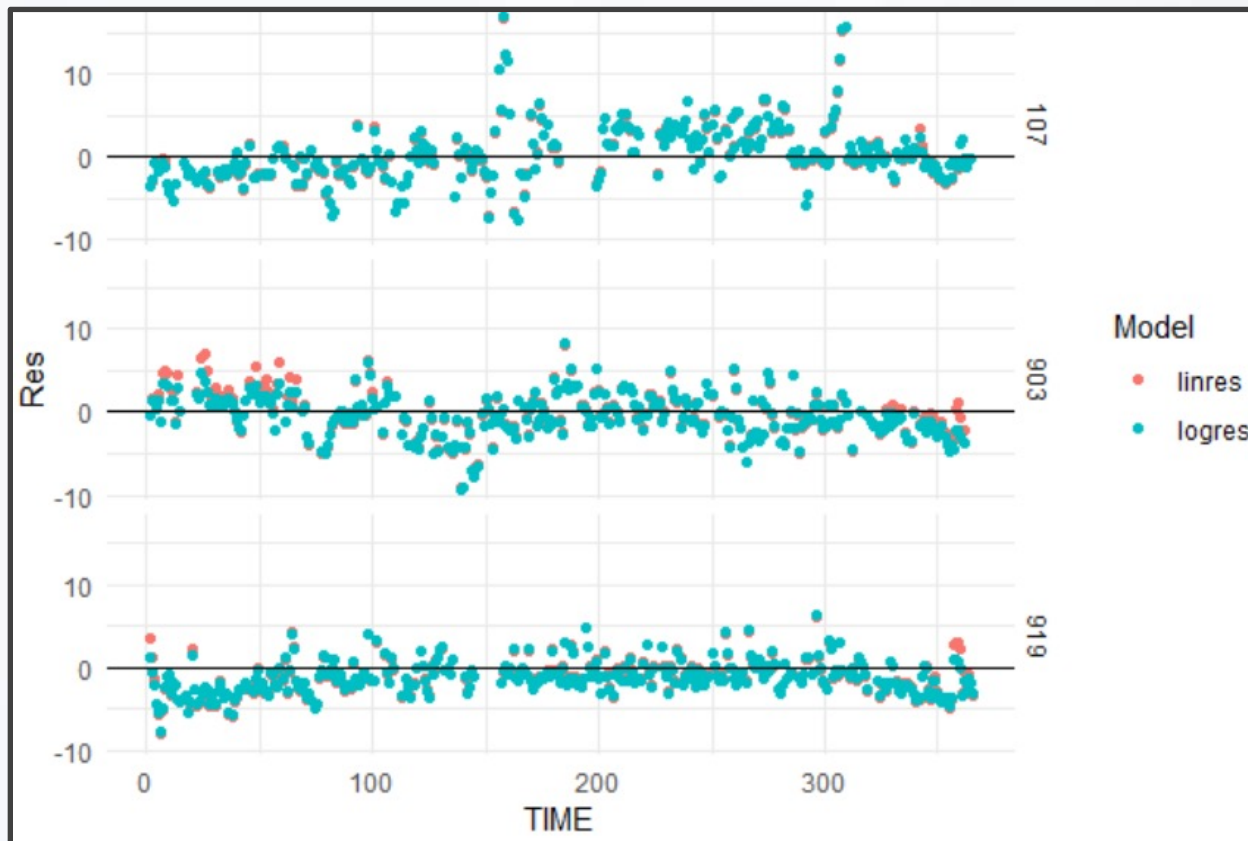
- Run Chunk 5
  - Evaluate Error For the Three Locations Separately (by A)





# Part 5: Evaluation by Visualization

- Run Chunk 6
  - Evaluate Error For the Three Locations Separately (by Time)



# Part 6: Evaluation by Numerical Summary

- Run Chunk 1
  - Mean Bias

$$\text{MB} = \frac{1}{N} \sum \hat{\epsilon}_k$$

- Mean Absolute Error

$$\text{MAE} = \frac{1}{N} \sum |\hat{\epsilon}_k|$$

- Root Mean Squared Error

$$\text{RMSE} = \sqrt{\frac{1}{N} \sum \hat{\epsilon}_k^2}$$

- MB, MAE, and RMSE are in Degrees Celsius

# Part 6: Evaluation by Numerical Summary

- Summarizing Table
  - Evaluate MB, MAE, and RMSE on Test Data to Choose Best Model Going Forward
  - Sketch of Table We Want

| Model    | MB | MAE | RMSE |
|----------|----|-----|------|
| Linear   |    |     |      |
| Poly(2)  |    |     |      |
| Poly(3)  |    |     |      |
| Poly(4)  |    |     |      |
| Logistic |    |     |      |

- Before Writing Code, Have a Plan for the Output

# Part 6: Evaluation by Numerical Summary

- Chunk 2
  - Run Line-By-Line
  - Think About Ways to Quickly Apply All 3 Functions to All Residuals
- Run Chunk 3
  - Combine `rename()`, `gather()`, `group_by()`, and `summarize()`
- Chunk 4
  - Change `eval=F` to `eval=T` and Knit the File (What is Seen?)

# Part 6: Evaluation by Numerical Summary

- My Results Based on My Seed

| Model<br><fctr> | MB<br><dbl> | MAE<br><dbl> | RMSE<br><dbl> |
|-----------------|-------------|--------------|---------------|
| Linear          | 0.9534126   | 2.750323     | 3.351594      |
| Poly(2)         | 0.9742415   | 2.732399     | 3.344867      |
| Poly(3)         | 0.9903951   | 2.706833     | 3.328889      |
| Poly(4)         | 0.9920042   | 2.715366     | 3.338710      |
| Logistic        | 0.2613184   | 3.135313     | 3.711664      |

- When Results Are This Close, Always Consider the Most Simple Model