



STOR 320 Modeling II&III

Lecture 16

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Tutorial 11

- 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)



Introduction

- Questions About RMarkdown
 - What Does the Following Code Do When Knitted?

```
`r length(unique(DATA$L))`
```

- What Does the following Code Chunk Option Do When Knitted?

```
echo=F
```



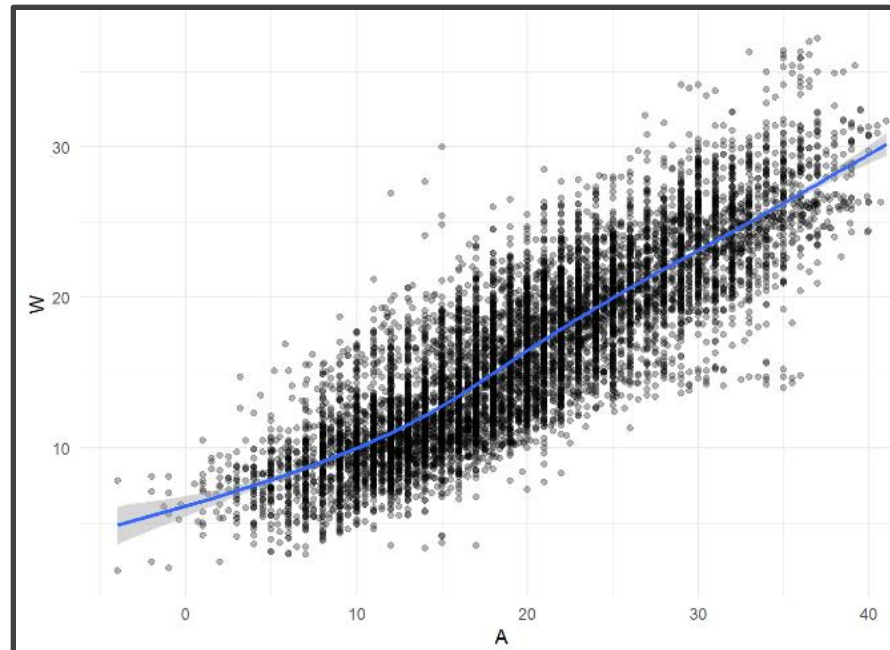
Introduction

- Goal: Build a Model to Predict Max Water Temp Given Max Air Temp
 - What Do You Know About the Relationship of These Variables?
 - Who Would Care About this Relationship?
 - Why Would Someone Want to Predict the Max Water Temp?
 - Why Would this Model Be Useful?



Part 1: Examining the Relationship

- Run Chunk 1
 - What Do You Notice About the Overall Relationship?

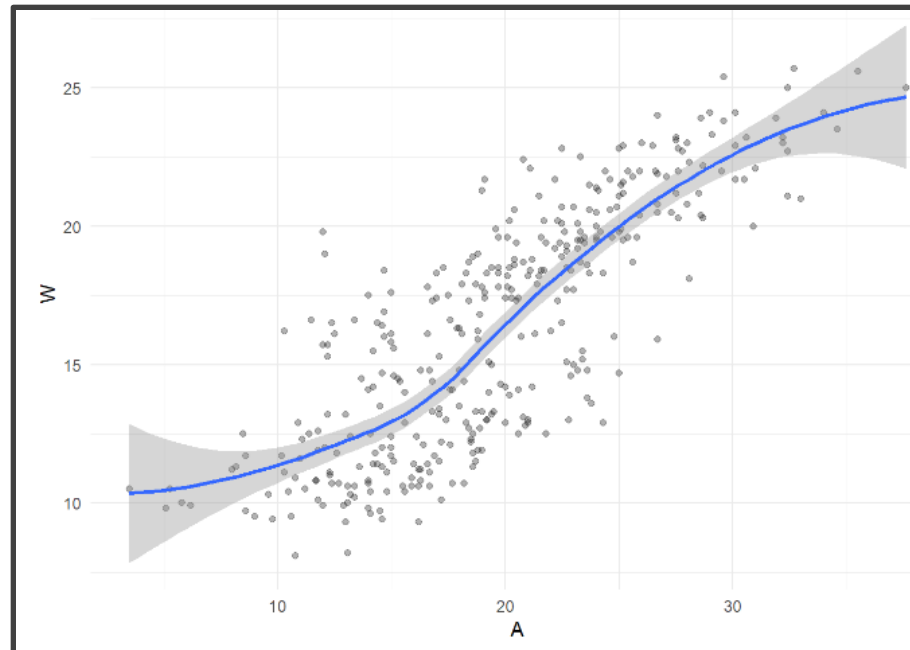


- Do You Think This Relationship is the Same for All Locations?
- Why? `message=F`



Part 1: Examining the Relationship

- Run Chunk 2
 - Location is a Numeric Variable
 - What Do You Notice About the Relationship for $L=103$?



- What do You Notice Now?



Part 1: Examining the Relationship

- Chunk 2 Modified
 - Modify Chunk 2 to Create a Function Called `WAPlot.func` With 1 Argument Location
 - Function Usage: You Specify the Location as an Integer and the Function Outputs a Figure of the Relationship
 - Use Your Function For Three Different Locations
 - Knit the Document to Observe and Compare



Part 1: Examining the Relationship

- Chunk 2 Discussion
 - What are the Differences in the Relationship Between W and A for the Various Locations?
 - Why do You Think These Differences Exist?
 - How do You Suggest We Handle the Differences?



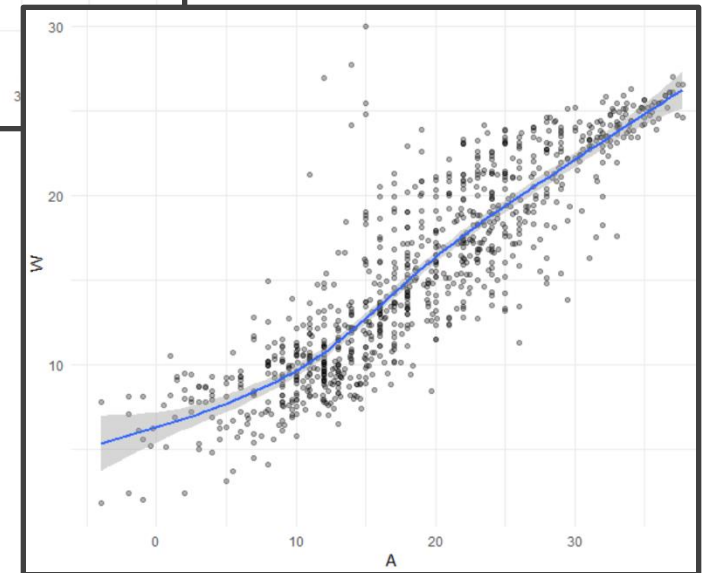
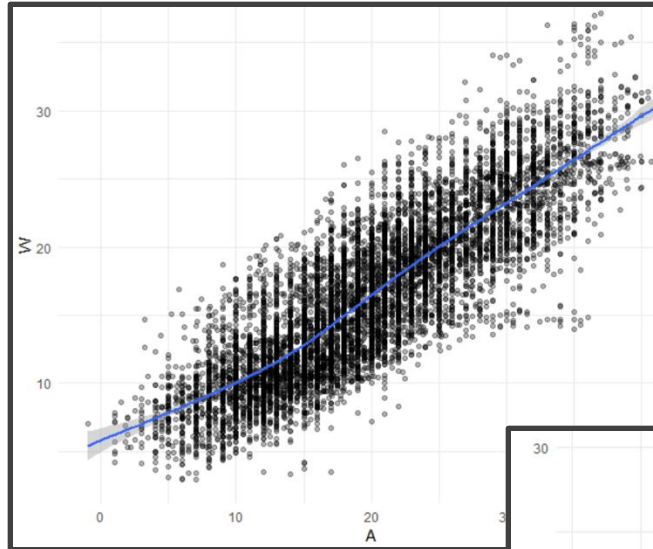
Part 1: Examining the Relationship

- Chunk 3
 - Randomly Samples 3 Locations
 - Plant Your Seed and Run Code
 - Usage:
 - `anti_join()`
 - `semi_join()`
 - Why Don't We Handpick the Three Locations?



Part 1: Examining the Relationship

- Run Chunk 4
 - Train Plot
 - Test Plot





Part 2: Linear Model

- Linear Model
 - Simplest Relationship that is Easily Explained
 - For every 1 Degree Change in A , W changes by b Degrees
 - When $A=0$ Degrees, the Expected Water Temperature is a Degrees



Part 2: Linear Model

- Run Chunk 1
 - Fits Linear Model to Train Data
 - What is Your Intercept?
 - What is Your Slope?
- Run Chunk 2
 - Saves Predictions to Train/Test

```
add_predictions(MODEL,var="NAME")
```

- Run Chunk 3
 - Saves Residuals to Train/Test

```
add_residuals(MODEL,var="NAME")
```



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 2nd Degree Polynomial
 - Fits 3rd Degree Polynomial
 - Fits 4th Degree Polynomial
- Run Chunk 2
 - Obtains Predictions Under the Different Polynomial Models

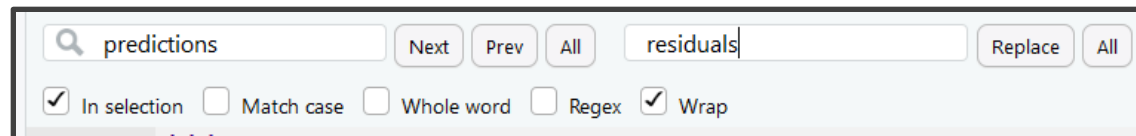


Part 3: Polynomial Model

- 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 %>%  
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```

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



- 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)



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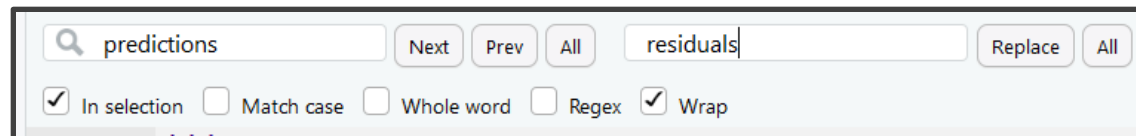


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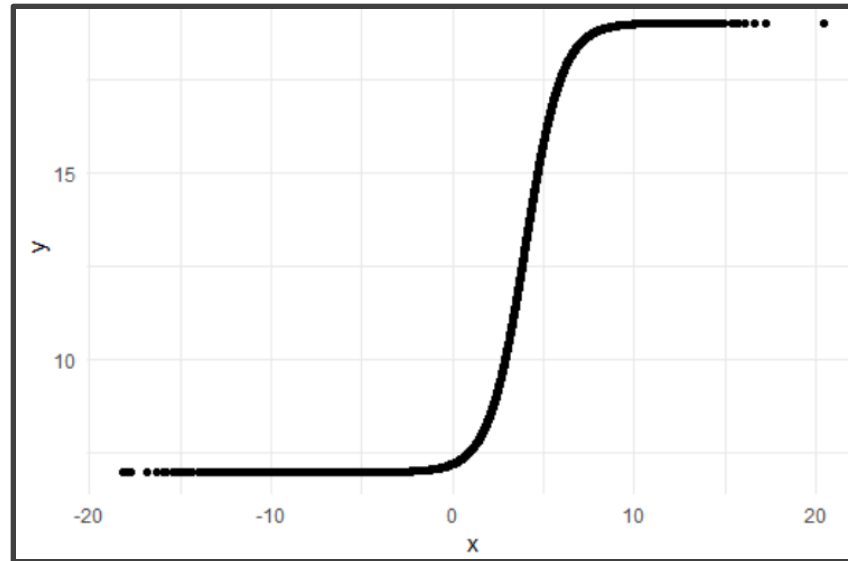
Part 4: Logistic Model

- 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



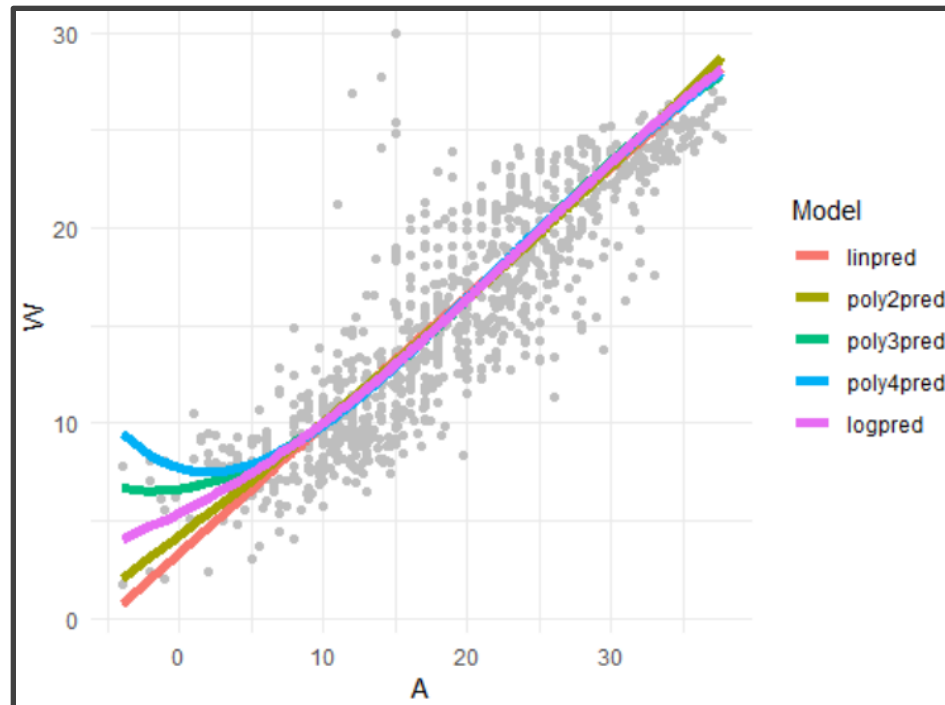
Part 4: Logistic Model

- 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?

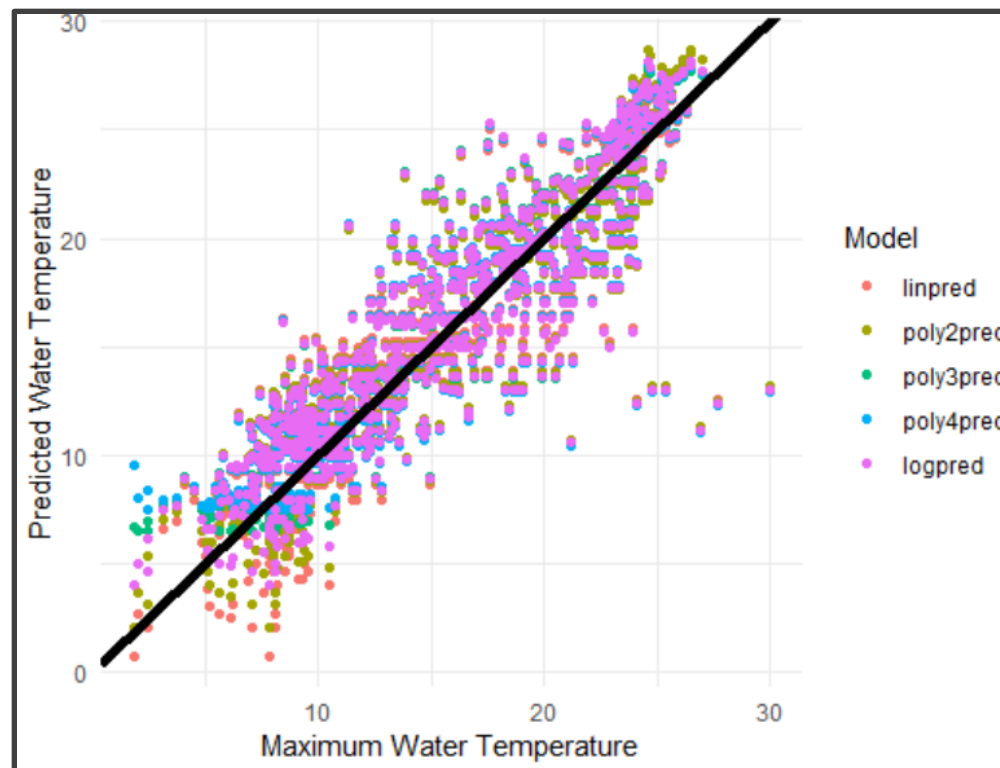


- Which Model Would You Use?



Part 5: Evaluation by Visualization

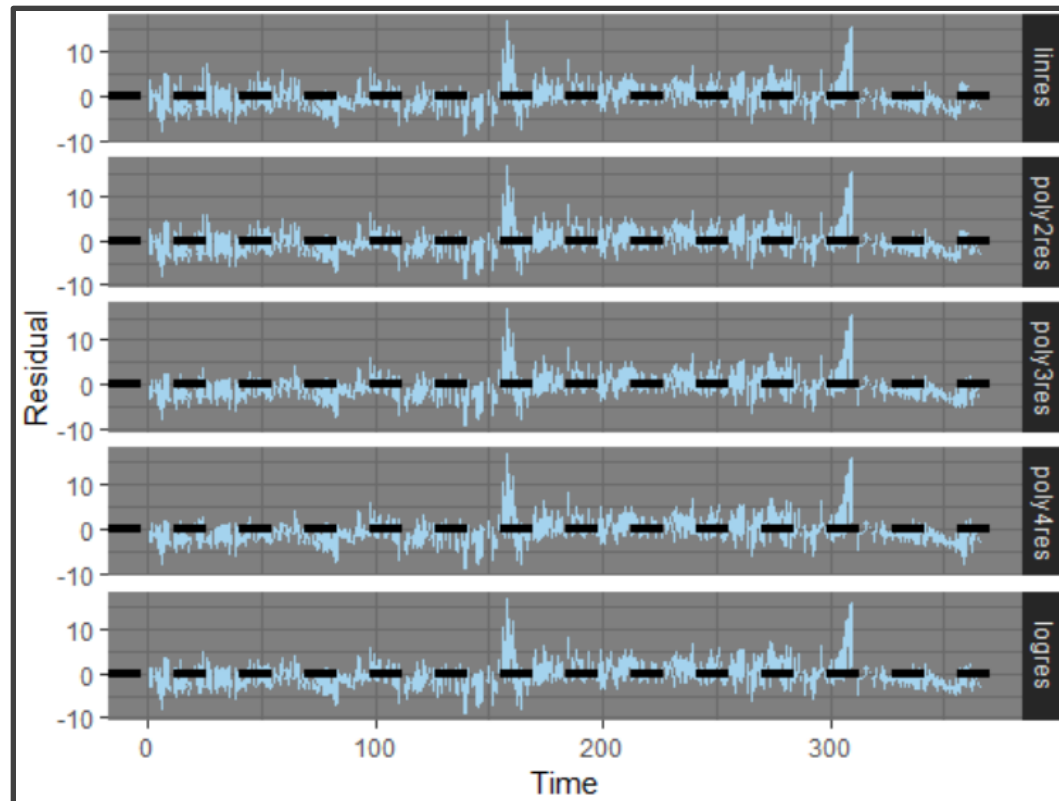
- Run Chunk 2
 - Comparing Predictions vs Actual Maximum Water Temperatures
 - Models Give Similar Predictions





Part 5: Evaluation by Visualization

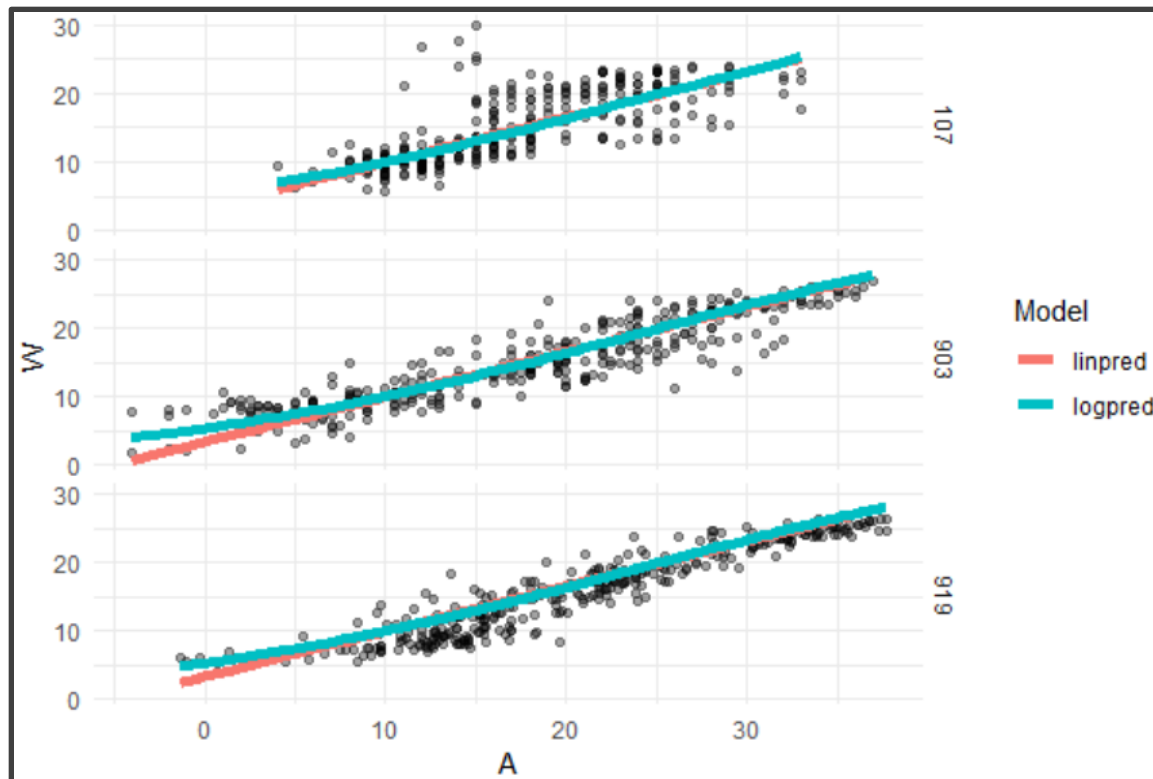
- Run Chunk 3
 - Shows Residuals Under the 4 Models Plotted Over Time
 - What is the Problem?





Part 5: Evaluation by Visualization

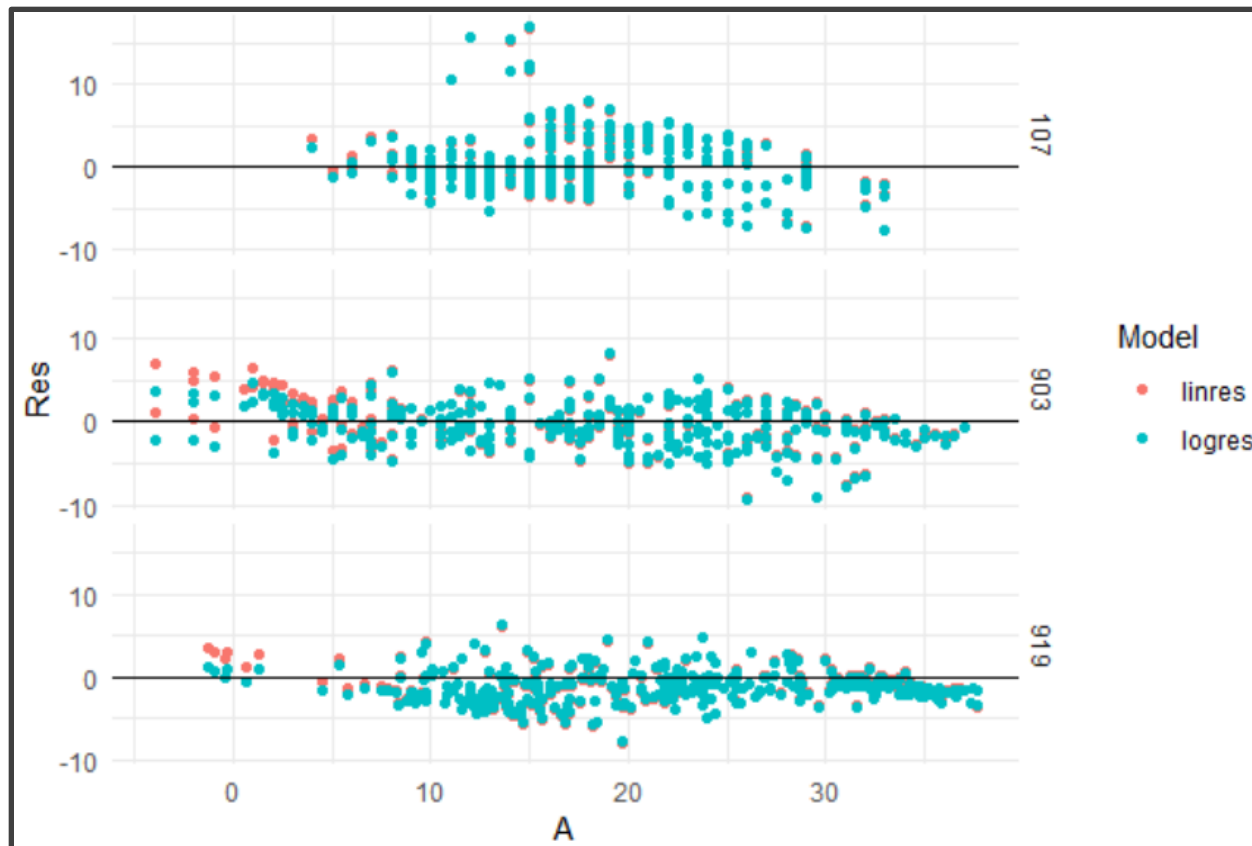
- Run Chunk 4
 - Evaluate Models For the Three Locations Separately





Part 5: Evaluation by Visualization

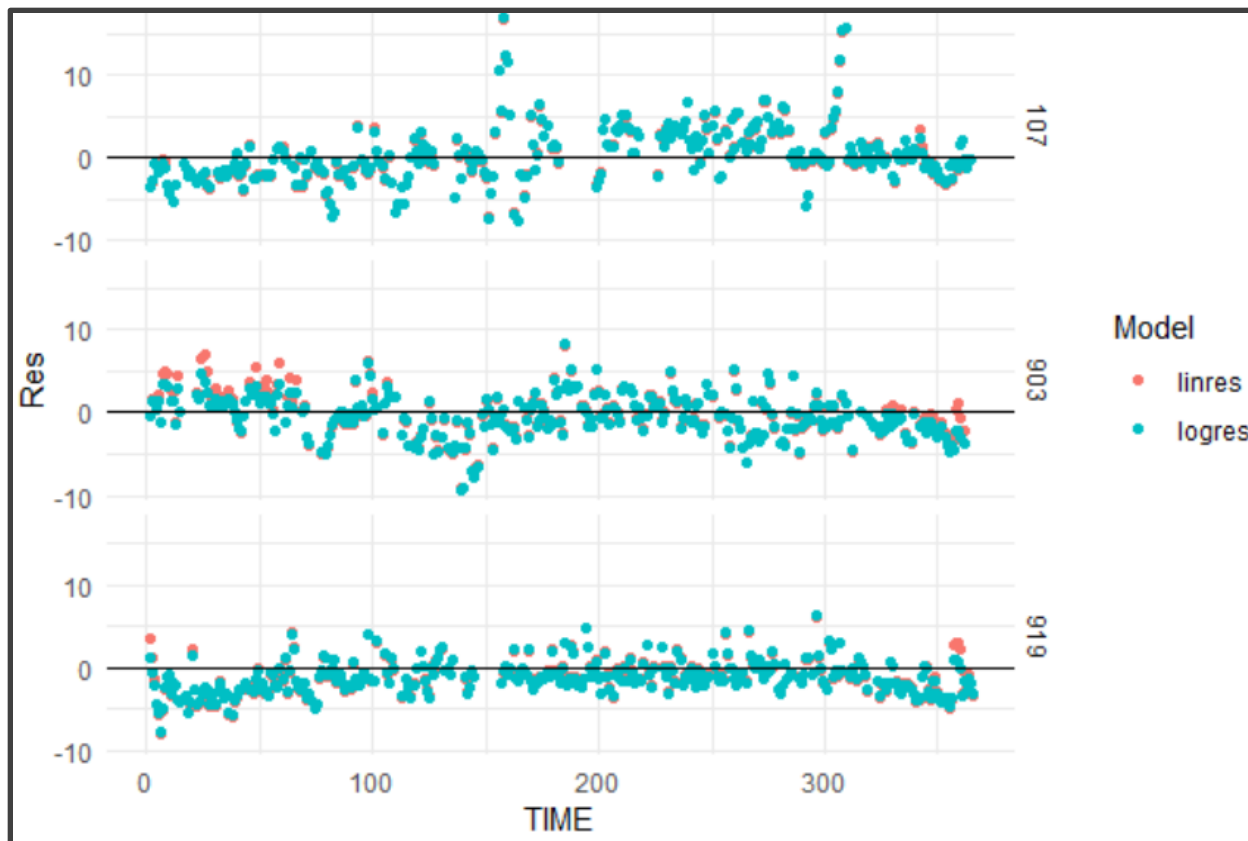
- Run Chunk 5
 - Evaluate Error For the Three Locations Separately (by A)





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 <fctr>	MB <dbl>	MAE <dbl>	RMSE <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