

# The dark art of debugging code

Christopher Cahill  
Associate Director  
19 December 2024



Quantitative Fisheries Center  
MICHIGAN STATE UNIVERSITY

# Troubleshooting code as a skill that can be taught, learned, and mastered



# Outline

- The origins of the bug 

# Outline

- The origins of the bug 
- A general workflow for debugging code

# Outline

- The origins of the bug 
- A general workflow for debugging code
- Specifically we want to think about how we might:

# Outline

- The origins of the bug 
- A general workflow for debugging code
- Specifically we want to think about how we might:
  - Write fewer bugs
  - Reproduce bugs
  - Locate bugs
  - Fix bugs

# Outline

- The origins of the bug 
- A general workflow for debugging code
- Specifically we want to think about how we might:
  - Write fewer bugs
  - Reproduce bugs
  - Locate bugs
  - Fix bugs
- Introduce a few useful tools that help with this workflow
  - `styler` library and `browser()`

# Outline

- The origins of the bug 
- A general workflow for debugging code
- Specifically we want to think about how we might:
  - Write fewer bugs
  - Reproduce bugs
  - Locate bugs
  - Fix bugs
- Introduce a few useful tools that help with this workflow
  - `styler` library and `browser()`
- Work through some of examples

# The original sin (bug)

# The first bug



92

9/9

0800 arctan started  
1000 " stopped - arctan ✓  
13"sec (032) MP-MC  $\left\{ \begin{array}{l} 1.2700 \\ 2.15076715 \end{array} \right.$  9.037847025  
(033) PRO 2 2.130476415  
const 2.130676415

Relays 6-2 in 033 failed special speed test  
in relay " 10.000 test .

Relay  
2145  
Relay 3370

1700 Started Cosine Tape (Sine check)  
1525 Started Multi Adder Test.

1545



Relay #70 Panel F  
(moth) in relay.

1630 arctangent started.

1700 closed down .

# **Step 1: Write fewer bugs**

# Follow a consistent style

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference
- The importance of adopting a consistent style cannot be stressed enough

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference
- The importance of adopting a consistent style cannot be stressed enough
- QFC trying to follow the tidyverse style guide

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference
- The importance of adopting a consistent style cannot be stressed enough
- QFC trying to follow the tidyverse style guide
  - Most things lower case and words separated by underscores

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference
- The importance of adopting a consistent style cannot be stressed enough
- QFC trying to follow the tidyverse style guide
  - Most things lower case and words separated by underscores
  - Implemented via the Styler library

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference
- The importance of adopting a consistent style cannot be stressed enough
- QFC trying to follow the tidyverse style guide
  - Most things lower case and words separated by underscores
  - Implemented via the Styler library
- Why care about style?

# Follow a consistent style

- Good coding style is like correct punctuation: you can manage without it, but its sure to make a difference
- The importance of adopting a consistent style cannot be stressed enough
- QFC trying to follow the tidyverse style guide
  - Most things lower case and words separated by underscores
  - Implemented via the Styler library
- Why care about style?
  - Consistent naming conventions, indentation, and spacing help us train our eyes to spot bugs

# An example program that you get from a collaborator

```
1 f = function(pars) {
2   getAll(data, pars);
3   Linfmn = exp(logLinfmn); logLinfsd = exp(loglogLinfsd);
4   Linfs = exp(logLins); K = exp(logK );
5   Sig = exp(logSig);
6   nponds = length(Linfs); nages = length(A);
7   predL = matrix(0, nrow = nages, ncol = nponds);
8   # fill one column (pond) at a time:
9   for (i in 1:nponds) { predL[, i] = Linfs[i] * (1 - exp(-K * (A - t0)));}
10  nll = -sum(dnorm(x = L, mean = predL, sd = Sig, log = TRUE));
11  nprand = -sum(dnorm(x = logLins, mean = logLinfmn, sd = logLinfsd, log = TRUE));
12  jnll = nll + nprand;
13  jnll;
14 }
```

# An example program that you get from a collaborator

```
1 f = function(pars) {  
2   getAll(data, pars);  
3   Linfmn = exp(logLinfmn); logLinfsd = exp(loglogLinfsd);  
4   Linfs = exp(logLins); K = exp(logK );  
5   Sig = exp(logSig);  
6   nponds = length(Linfs); nages = length(A);  
7   predL = matrix(0, nrow = nages, ncol = nponds);  
8   # fill one column (pond) at a time:  
9   for (i in 1:nponds) { predL[, i] = Linfs[i] * (1 - exp(-K * (A - t0)));}  
10  nll = -sum(dnorm(x = L, mean = predL, sd = Sig, log = TRUE));  
11  nprand = -sum(dnorm(x = logLins, mean = logLinfmn, sd = logLinfsd, log = TRUE));  
12  jnll = nll + nprand;  
13  jnll;  
14 }
```

- this is misery 💀 😱 💀 😱

# An example program

- Make it pretty via `styler()`

```
1 f <- function(pars) {  
2   getAll(data, pars)  
3   Linfmn <- exp(logLinfmn)  
4   logLinfsd <- exp(loglogLinfsd)  
5   Linfs <- exp(logLinfs)  
6   K <- exp(logK)  
7   Sig <- exp(logSig)  
8   nponds <- length(Linfs)  
9   nages <- length(A)  
10  predL <- matrix(0, nrow = nages, ncol = nponds)  
11  # fill one column (pond) at a time:  
12  for (i in 1:nponds) {  
13    predL[, i] <- Linfs[i] * (1 - exp(-K * (A - t0)))  
14  }  
15  nll <- -sum(dnorm(x = L, mean = predL, sd = Sig, log = TRUE))  
16  nprand <- -sum(dnorm(x = logLinfs, mean = logLinfmn, sd = logLinfsd, log = TRUE))  
17  jnll <- nll + nprand  
18  jnll  
19 }
```

# **Step 2: Reproduce the bug**

# Reproduce the bug

- Reproducing a bug allows us to better understand why the program went wrong

*“Debugging is like being the detective in a crime movie where you are also the murderer.” - Filipe Fortes*

# Reproduce the bug

- Reproducing a bug allows us to better understand why the program went wrong
- Pay attention to versions of R and packages

*“Debugging is like being the detective in a crime movie where you are also the murderer.” - Filipe Fortes*

# Reproduce the bug

- Reproducing a bug allows us to better understand why the program went wrong
- Pay attention to versions of R and packages
- Pay attention to warnings, errors, and other messages

*“Debugging is like being the detective in a crime movie where you are also the murderer.” - Filipe Fortes*

# Reproduce the bug

- Reproducing a bug allows us to better understand why the program went wrong
- Pay attention to versions of R and packages
- Pay attention to warnings, errors, and other messages
- Emphasis here on pay attention

*“Debugging is like being the detective in a crime movie where you are also the murderer.” - Filipe Fortes*

# Reproduce the bug

- Reproducing a bug allows us to better understand why the program went wrong
- Pay attention to versions of R and packages
- Pay attention to warnings, errors, and other messages
- Emphasis here on pay attention
- ISOLATE THE BUG

*“Debugging is like being the detective in a crime movie where you are also the murderer.” - Filipe Fortes*

# An example of isolating a bug

```
1 dgmr(1:10, mu=2:11, Q=Matrix:::Diagonal(10), log=TRUE)
2 #[1] -14.18939
3 dgmr(1:10-2:11, Q=Matrix:::Diagonal(10), log=TRUE)
4 #[1] -14.18939
5 dgmr(1:10, mu=2:11, Q=Matrix:::Diagonal(10)/2/2, log=TRUE)
6 #[1] -17.37086
7 dgmr(1:10-2:11, Q=Matrix:::Diagonal(10)/2/2, log=TRUE)
8 #[1] -17.37086
9 dgmr(1:10-2:11, Q=Matrix:::Diagonal(10), scale=2, log=TRUE)
10 #[1] -17.37086
11 dgmr(1:10, mu=2:11, Q=Matrix:::Diagonal(10), scale=2, log=TRUE)
12 #[1] -96.74586
```

- The last result is wrong

# Determining versions in R

```
1 packageVersion("RTMB")  
[1] '1.6'
```

# Determining versions in R

```
1 sessionInfo()
```

```
R version 4.4.2 (2024-10-31)
Platform: x86_64-pc-linux-gnu
Running under: Linux Mint 21.3

Matrix products: default
BLAS:   /usr/lib/x86_64-linux-gnu/openblas-pthread/libblas.so.3
LAPACK: /usr/lib/x86_64-linux-gnu/openblas-pthread/libopenblas-p0.3.20.so; LAPACK version 3.10.0

locale:
[1] LC_CTYPE=en_US.UTF-8        LC_NUMERIC=C
[3] LC_TIME=en_US.UTF-8        LC_COLLATE=en_US.UTF-8
[5] LC_MONETARY=en_US.UTF-8    LC_MESSAGES=en_US.UTF-8
[7] LC_PAPER=en_US.UTF-8       LC_NAME=C
[9] LC_ADDRESS=C               LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C

time zone: America/New_York
tzcode source: system (glibc)

attached base packages:
[1] stats      graphics   grDevices utils      datasets   methods    base

loaded via a namespace (and not attached):
[1] compiler_4.4.2   fastmap_1.2.0    cli_3.6.3      tools_4.4.2
[5] htmltools_0.5.8.1 yaml_2.3.8     rmarkdown_2.27 knitr_1.47
[9] jsonlite_1.8.8   xfun_0.45      digest_0.6.36  rlang_1.1.4
```

# Step 3: Locate the bug



# Manual debugging

[Wikipedia link to Shotgun Debugging](#)

# Manual debugging

- What most people are used to, links back to Jim's bit about understanding the math

[Wikipedia link to Shotgun Debugging](#)

# Manual debugging

- What most people are used to, links back to Jim's bit about understanding the math
- There are smart and effective ways to do this

[Wikipedia link to Shotgun Debugging](#)

# Manual debugging

- What most people are used to, links back to Jim's bit about understanding the math
- There are smart and effective ways to do this
- There are also stupid ways to do this

[Wikipedia link to Shotgun Debugging](#)

# Formal debugging tools in R

- There are several, but I am going to teach you `browser()`

# Formal debugging tools in R

- There are several, but I am going to teach you `browser()`
- `browser()` gives us a way to interrupt the execution of an expression and allow the inspection of the environment

# Formal debugging tools in R

- There are several, but I am going to teach you `browser()`
- `browser()` gives us a way to interrupt the execution of an expression and allow the inspection of the environment
  - When I say environment here, think local environment inside a function

# The downside of `browser()`

- Have to manually add it and then remove it from your code later on

# **Step 4: Fix the bug**

# Exorcise the bug (demon)



Image reference

# My strategy for debugging code

1. Format the code in a sensible and consistent way
2. Reproduce and isolate the bug, noting software versions
  - 2b. Locate the bug using debugging tools
3. Explore the unexpected behavior
4. Conduct an exorcism to remove the demon
5. Make the code more robust for future users

# Simulation as a critical debugging tool

- All of the bugs we have discussed so far are the ones that cause errors
  - **These are not the only bugs**
    - The worst bugs allow code to run but unknowingly return incorrect answers
  - Maximum likelihood methods have a theoretical property of *consistency*, which you can use to your advantage
  - A word to the wise: most of the bugs Lisa Chong (QFC postdoc) is finding in Great Lakes assessment models are of this type
- 

# **Three more RTMB specific bugs and tricks**

# Avoid struggles with JIT

- R's just-in-time compiler kicks in at unpredictable times
- There are a few conflicts with RTMB, because JIT does not preserve specific definitions made by RTMB
- Can usually be avoided with the following added to the top of your RTMB function

```
1 f <- function(par) {  
2   getAll(data, par, warn = FALSE)  
3   "[<-" <- ADoverload("[<-") # hey Jim look here =)  
4   ... more code below...
```

- The forboding sign: phantom errors creeping about in the shadows

# Isolating RTMB bugs during optimization

- This is sometimes where happiness goes to die
- Remember that the optimization routine for hierarchical models involves an inner and an outer optimization
- We can inspect both the inner and outer optimization components of a model fit to find bugs in our implementation (coding bugs)

# Outer optimization despair



```
1 obj = MakeADFun(f, par, random = c("a", "b"), silent = TRUE)
2 opt <- nlmnb(obj$par, obj$fn, obj$gr)
3 sdr <- sdreport(obj)
4 sdr
```

```
sdreport(.) result
  Estimate Std. Error
mua     8.467355      NaN
sda     3.463551      NaN
mub    29.044188      NaN
sdb    10.541116      NaN
sdc    1.000000      NaN
sdeps 12.890654      NaN
Warning:
Hessian of fixed effects was not positive definite.
Maximum gradient component: 1.117405e-05
```

# Inner optimization despair



```
1 obj = MakeADFun(f, par, random = c("a", "b"))
2 obj$fn()
```

```
iter: 1 value: 61833.54 mgc: 34813 ustep: 0.9999
iter: 2 value: 61833.54 mgc: 0.004086614 ustep: 0.99995
iter: 3 value: 61833.54 mgc: 4.813844e-08 ustep: 0.999975
iter: 4 Error in newton(par = c(a = 0, a =
0, :
  Newton failed to find minimum.
[1] NaN
```

# References

- Kasper Kristensen, Anders Nielsen, Casper W. Berg, Hans Skaug, Bradley M. Bell. 2016. TMB: Automatic Differentiation and Laplace Approximation. *Journal of Statistical Software*, 70(5), 1-21. doi:10.18637/jss.v070.i05
- Kristensen K. 2023. RTMB: R Bindings for TMB. R package version 1.0, <https://github.com/kaskr/RTMB>