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# Writing clear code Tools for Reproducible Research

#### Karl Broman

Biostatistics & Medical Informatics, UW-Madison

biostat.wisc.edu/~kbroman
 github.com/kbroman
 @kwbroman
Course web: bit.ly/tools4rr

# Basic principles

Code that works

No bugs; efficiency is secondary (or tertiary)

Readable

Fixable; extendible

▶ Reusable

Modular; reasonably general

Reproducible

Re-runnable

Think before you code

More thought ⇒ fewer bugs/re-writes

Learn from others' code

R itself; key R packages

## Break code into small functions

```
sampleone <-
function(vec)
  ifelse(length(vec)==1, vec, sample(vec, 1))
get_grid_index <-
function(vec, step)
  grid <- seq(min(vec), max(vec), by=step)</pre>
  index <- match(grid, vec)</pre>
  if(any(is.na(index)))
    index <- sapply(grid, function(a,b) {</pre>
        d <- abs(a-b);</pre>
         sampleone(which(d == min(d)))
      }, vec)
  index
```

# Another example

# Yet another example

```
# rmvn: simulate from multivariate normal distribution
rmvn <-
function(n, mu=0, V=matrix(1))
{
   p <- length(mu)
   if(any(is.na(match(dim(V),p))))
      stop("Dimension problem!")

   D <- chol(V)
   matrix(rnorm(n*p),ncol=p) %*% D + rep(mu,each=n)
}</pre>
```

# Still more examples

```
# colors from blue to red
revrainbow <-
function(n=256, ...)
  rev(rainbow(start=0, end=2/3, n=n, ...))
# move values above/below quantiles to those quantiles
winsorize <-
function(vec, q=0.006)
  lohi <- quantile(vec, c(q, 1-q), na.rm=TRUE)</pre>
  if(diff(lohi) < 0)
   lohi <- rev(lohi)
  vec[!is.na(vec) & vec < lohi[1]] <- lohi[1]</pre>
  vec[!is.na(vec) & vec > lohi[2]] <- lohi[2]</pre>
  vec
```

# Writing functions

- Break large tasks into small units.
  - Make each discrete unit a function.
- ► If you write the same code more than once, make it a function.
- ► If a line/block of code is complicated, make it a function.

# Don't repeat yourself (or others)

- Avoid having repeated blocks of code.
- Create functions, and call those functions repeatedly.
- This is easier to maintain.
  - If something needs to be fixed/revised, you just have to do it the one time.
- Look at others' libraries/packages.
  - Don't write what others have already written (especially if they've done it better than you would have).

# Don't make things too specific

- Write code that is a bit more general than your specific data
  - Don't assume particular data dimensions.
  - Don't forget about the possibility of missing values (even if your data doesn't have any).
  - Aim for re-use.
- Use function arguments
  - Don't assume particular data file names
  - Don't hard-code tuning parameters
  - R scripts can take command-line arguments:

```
Rscript myscript.R input_file output_file
args <- commandArgs(TRUE)</pre>
```

## No global variables, ever!

- Don't refer directly to objects in your workspace.
- If a function needs something, pass it as an argument.
- ► (But what about really big data sets?)

# No magic numbers

Name numbers and use the names

```
max_iter <- 1000
tol_convergence <- 0.0001</pre>
```

► Even better: include them as function arguments

## Indent!

```
# move values above/below quantiles to those quantiles
winsorize <-
function(vec, q=0.006)
{
lohi <- quantile(vec, c(q, 1-q), na.rm=TRUE)
if(diff(lohi) < 0)
lohi <- rev(lohi)
vec[!is.na(vec) & vec < lohi[1]] <- lohi[1]
vec[!is.na(vec) & vec > lohi[2]] <- lohi[2]
vec
}</pre>
```

# Use white space

```
# move values above/below quantiles to those quantiles
winsorize<-function(vec,q=0.006)
{lohi<-quantile(vec,c(q,1-q),na.rm=TRUE)
if(diff(lohi)<0)lohi<-rev(lohi)
vec[!is.na(vec)&vec<lohi[1]]<-lohi[1]
vec[!is.na(vec)&vec>lohi[2]]<-lohi[2]
vec}</pre>
```

# Don't let lines get too long

```
get_grid_index <-
function(vec, step)
{
  grid <- seq(min(vec), max(vec), by=step)
  index <- match(grid, vec)

  if(any(is.na(index)))
    index <- sapply(grid, function(a,b) { d <- abs(a-b); sampleone(whiindex)}

index
}</pre>
```

# Use parentheses to avoid ambiguity

```
if( (ndraws1==1) && (ndraws2>1) ) {
...
}
leftval <- which( (map - start) <=0 )
```

# Names: meaningful

- Make names descriptive but concise
- ► Avoid tmp1, tmp2, ...
- ► Only use i, j, x, y in the simplest situations
- ▶ If a function is named fv, what might it do?
- ▶ If an object is called nms, what could it be?
- Functions as verbs; objects as nouns

#### Names: consistent

- ► markers VS mnames
- ► camelCase VS. pothole\_case
- ▶ nind VS n.var
- ► If a function/object has one of these, there shouldn't be a function/object with the other.

## Names: avoid confusion

- ► Don't use both total and totals
- ► Don't use both n.cluster and n.clusters
- ▶ Don't use both result and results
- ► Don't use both Mat and mat
- ▶ Don't use both g and gg

## Don't be cute



## Comments

- Comment the tricky bits and the major sections
- Don't belabor the obvious
- Don't comment bad code; rewrite it
- Document the input/output and purpose, not the mechanics
- Don't contradict the code
  - this happens if you revise the code but don't revise the related comments
- Comment code as you are writing it (or before)
- Plan to spend 1/4 of your time commenting

# Error/warning messages

Explain what's wrong (and where)

```
- error("nrow(X) != nrow(Y)")
```

► Suggest corrective action

```
- "You need to first run calc.genoprob()."
```

Give details

```
- "nrow(X) (", nrX, ") != nrow(Y) (", nrY, ")"
```

- ▶ Don't give error/warning messages that users won't understand.
  - X'X is singular.
- Don't let users do something stupid without warning
- ► Include error checking even in personal code.

# Check data integrity

- Check that the input is as expected, or give warnings/errors.
- Write these in the first pass (though they're dull).
  - You may not remember your assumptions later
- ► These are useful for documenting the assumptions.

# Program organization

- ► Break code into separate files (say 300 lines?)
- Each file includes related functions
- Files should be named meaningfully
- Include a brief comment at the top.

# Create an R package!

- Make a personal package with bits of your own code
- ► Mine is R/broman, github.com/kbroman/broman

```
# qqline corresponding to qqplot
qqline2 <- function(x, y, probs = c(0.25, 0.75), qtype = 7, ...)
{
    stopifnot(length(probs) == 2)
    x <- quantile(x, probs, names=FALSE, type=qtype, na.rm = TRUE)
    y <- quantile(y, probs, names=FALSE, type=qtype, na.rm = TRUE)
    slope <- diff(y)/diff(x)
    int <- y[1L] - slope*x[1L]
    abline(int, slope, ...)
    invisible(c(intercept=int, slope=slope))
}</pre>
```

# Complex data objects

- Keep disparate data together in a more complex structure.
  - lists in R
  - I also like to hide things in object attributes
- It's easier to pass such objects between functions
- Consider object-oriented programming

# Avoiding bugs

- Learn to type well.
- ► Think before you type.
- Consider commenting before coding.
- Code defensively
  - Handle cases that "can't happen"
- Code simply and clearly
- Use modularity to advantage
- Think through all special cases
- Don't be in too much of a hurry

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# Summary

- Get the correct answers.
- ► Find a clear style and stick to it.
- Plan for the future.
- Be organized.
- ▶ Don't be too hurried.
- ► Learn from others.