

A University of Queensland Advanced Workshop

Session 10: Some Programming Techniques

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1 A simulation example

Find by simulation the relative volume of a sphere of diameter 2 to a square box of side 2?

For dimensions $d = 1, 2, 3, \ldots$

The answer is known mathematically. The volume of the sphere of radius r is

$$V_d^{(S)} = \frac{\pi^{d/2} r^d}{\Gamma(d/2+1)}$$

and the volume of the square box of side 2r is clearly $V_c^{(B)} = 2^d r^d$. Hence the relative size is

$$R = {\text{def.}} = \frac{V_d^{(S)}}{V_d^{(B)}} = \frac{\pi^{d/2}}{\Gamma(d/2 + 1) 2^d}$$

1.1 A simulation strategy

- Generate samples uniformly within the box centred at (0, 0, ..., 0)
- Count the number at a Euclidean distance no further than 1 from the origin.
- Take ratio.

```
mcrvball <- function(d, N = 100000, blocksize = 10000) {</pre>
  n2 <- inside <- 0
  while (n2 < N)
    n1 <- n2
    n2 <- min(n2 + blocksize, N)
    No <- n2 - n1
    samp <- matrix(runif(No * d, -1, 1), No, d)</pre>
    inside <- inside + sum(rowSums(samp^2) < 1)</pre>
  res <- list(dimensions = d, inside = inside,
              total = N, call = match.call())
  class(res) <- "mcrvball"</pre>
  res
```

An alternative implementation using the doParallel family of packages.

To use it:

```
suppressPackageStartupMessages(library(doParallel))
cl <- makePSOCKcluster(detectCores() - 1)
registerDoParallel(cl)  ## not needed in "parallel" package

tst <- mcrvball2(10, N = 1e7, blocksize = 1e5)

stopCluster(cl)
rm(cl)</pre>
```

1.2 S3 methods

```
print.mcrvball <- function(x, ...) {</pre>
  with(x, cat("Dim.:", dimensions,
               "Estimated:", signif(inside/total, 4),
               "Actual:", signif(rvball(dimensions), 4), "\n"))
  invisible(x)
Ops.mcrvball <- function(e1, e2) { ## group generic
  if(inherits(e1, "mcrvball"))
  e1 <- with(e1, inside/total)</pre>
  if(!missing(e2) && inherits(e2, "mcrvball"))
  e2 <- with(e2, inside/total)</pre>
  NextMethod()
```

```
for(i in 4:10) print(mcrvball(i, 1000000))

Dim.: 4 Estimated: 0.3087 Actual: 0.3084

Dim.: 5 Estimated: 0.1645 Actual: 0.1645

Dim.: 6 Estimated: 0.08091 Actual: 0.08075

Dim.: 7 Estimated: 0.03692 Actual: 0.03691

Dim.: 8 Estimated: 0.01587 Actual: 0.01585

Dim.: 9 Estimated: 0.006474 Actual: 0.006442

Dim.: 10 Estimated: 0.002508 Actual: 0.00249

r <- numeric(7)

for(d in 4:10) r[d-3] <- floor(1/mcrvball(d)); r

[1] 3 6 12 27 62 160 390</pre>
```

Having a *call* component allows *update* to be used:

```
p10 <- mcrvball(10)
p10a <- update(p10, N = 1000000, blocksize = 100000)
c(1/p10, 1/p10a)
[1] 411.5226 402.7386
"%+%" <- function(e1, e2)
    UseMethod("%+%")
"%+%.mcrvball" <- function(e1, e2) {
    if(e1$dimensions != e2$dimensions) stop("ball dimensions differ!")
    res <- list(dimensions = e1$dimensions, inside = e1$inside + e2$inside,
                total = e1$total + e2$total, call = e1$call)
    class(res) <- "mcrvball"</pre>
    res
# p10 %+% p10a
floor(1/(p10 %+% p10a))
[1] 403
```

Automatic (lazy) vectorization

```
Mcrvball <- Vectorize(mcrvball, vectorize.args = c("d", "N"), SIMPLIFY = FALSE)</pre>
Mcrvball(4:8, N = 1e6, blocksize = 1e5)
[[1]]
Dim.: 4 Estimated: 0.3082 Actual: 0.3084
[[2]]
Dim.: 5 Estimated: 0.1645 Actual: 0.1645
[[3]]
Dim.: 6 Estimated: 0.08064 Actual: 0.08075
[[4]]
Dim.: 7 Estimated: 0.03692 Actual: 0.03691
[[5]]
Dim.: 8 Estimated: 0.01581 Actual: 0.01585
```

1.3 Some lessons

- Vectorization. (*rvball*) and vectorization tools: (*Vectorize*)
- Taking the "whole object view" of the problem. (mcrvball)
- Object orientation: put all the information you are likely to need into the object and give it a class. (mcrvball)
- Methods for existing generics. (print.mcrvball)
- Group generic functions. (*Ops.mcrvball*)
- Binary operators and new generic functions. (%+%)

2 Intermission: On text processing

Data in the form of text is becoming more common and important.

R has many sophisticated tools and packages for text manipulation—here we just glimpse two elementary ones: grep and (g)sub: the first for finding, the second twin pair for fixing.

2.1 The Authorised Version of King James

The extdata subdirectory of the WWRCourse package folder contains two compressed text files, of the first and last books of AV:

```
subdir <- system.file("extdata", package = "WWRCourse")
dir(subdir, pattern = "\\.txt.gz$") ## NB regular expression
[1] "Genesis.txt.gz" "Revelations.txt.gz"</pre>
```

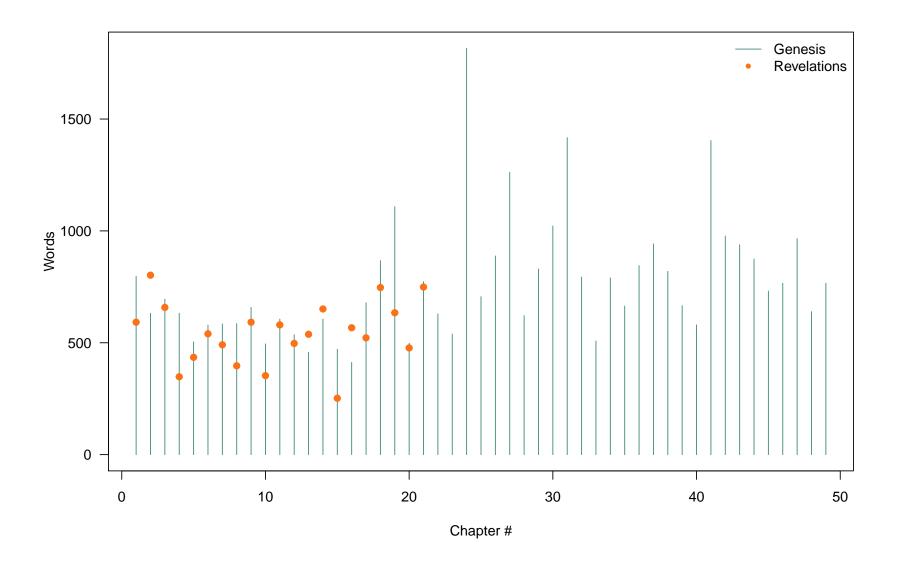
Our task is to compare *chapter length phrase length* and *word length* in these two document, as an aspect of literary style

2.2 Reading in

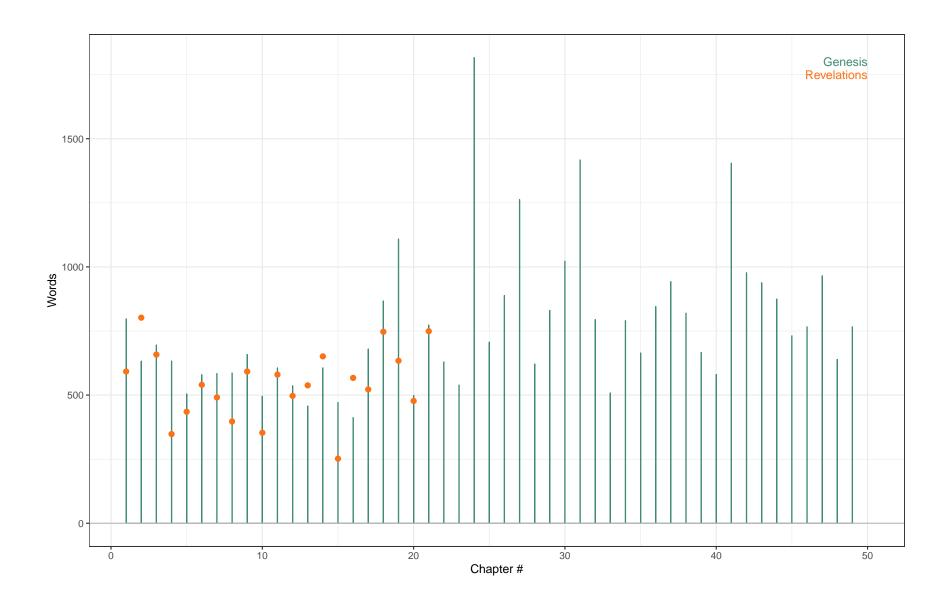
```
gen <- file.path(subdir, "Genesis.txt.gz")
rev <- file.path(subdir, "Revelations.txt.gz")
Genesis <- scan(gzfile(gen), what = "") ## no need to gunzip()
Revelations <- scan(gzfile(rev), what = "")
rbind(head(Genesis), head(Revelations)) %>% noquote
    [,1] [,2] [,3] [,4] [,5] [,6]
[1,] Gen.1 In the beginning God created
[2,] Rev.1 The Revelation of Jesus Christ,
```

Both have chapter markers in them of the form Gen. 1, Rev. 23, which are not part of the text. We can get the chapter lengths from them, though.

Revelations has fewer, and shorter chapters, but the chapter lengths are quite similar to the opening chapter lengths of Genesis:



A (contrived) *ggplot* version



2.3 Phrase length

The first step is to strip out the chapter markers and leave the words, and punctuation, only.

We define a phrase as the words between successive (terminal) punctuation marks. These indicate a pause in the reading.

```
GenP <- diff(c(0, grep("[[:punct:]]$", Gen)))
RevP <- diff(c(0, grep("[[:punct:]]$", Rev)))
rbind(Gen = head(GenP), Rev = head(RevP))

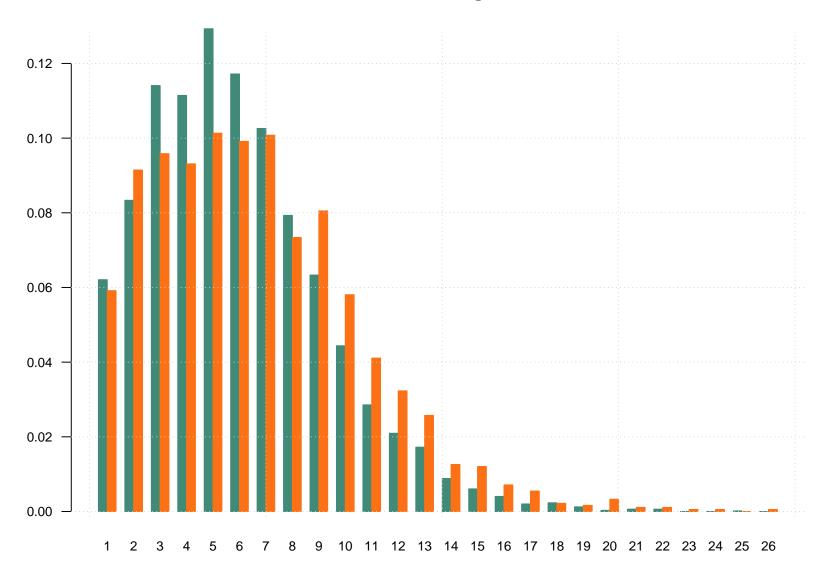
[,1] [,2] [,3] [,4] [,5] [,6]
Gen 10 6 2 9 12 3
Rev 5 5 12 13 8 7</pre>
```

Put into a data frame and look at the relative frequencies.^a

```
Phrase <- rbind(data.frame(Book = "Gen", Length = GenP),
                data.frame(Book = "Rev", Length = RevP))
tab <- with(Phrase, table(Book, Length)); t(tab)</pre>
      Book
Length Gen Rev
    1 400 108
    2 537 167
    23
         0 1
    24
        0 1
    25
        1 0
    26
         0
           1
colours <- c(Genesis = "#418A78", Revelations = "#FC7115")
par(cex.axis = 0.8)
barplot(tab/rowSums(tab), beside = TRUE, main = "Phrase Length",
        fill = colours, colour = colours)
grid()
```

^aExercise: Assuming Length-1 is Poisson, test for a difference in mean phrase length between the two books

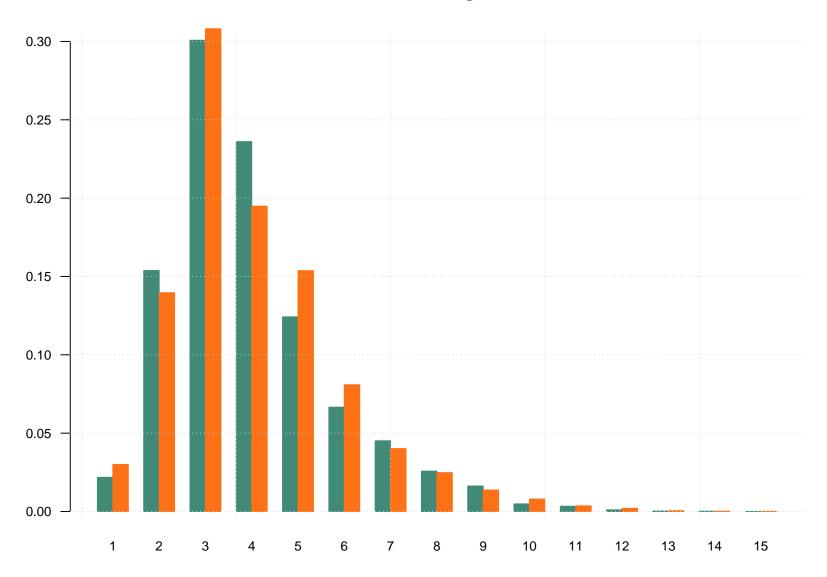
Phrase Length



Finally we come to word length distributions. For this we strip out any non-letter character and count the string lengths of what is left.

```
GenW <- nchar(gsub("[^[:alpha:]]", "", Gen))</pre>
RevW <- nchar(gsub("[^[:alpha:]]", "", Rev))</pre>
noquote(rbind(head(Rev), head(RevW))) ## word 5 comma excluded
     [,1] [,2] [,3] [,4] [,5] [,6]
[1,] The Revelation of Jesus Christ, which
[2,] 3 10 2 5 6
Words <- rbind(data.frame(Book = "Gen", Length = GenW),
              data.frame(Book = "Rev", Length = RevW))
tab <- with(Words, table(Book, Length)); t(tab)
     Book
Length Gen
            Rev
. . . .
   14 7 2
            1
   15
par(cex.axis = 0.8)
barplot(tab/rowSums(tab), beside = TRUE, main = "Word Length",
       fill = colours, colour = colours)
grid()
```

Word Length



3 Convolutions

Given two sequences of numbers, a_i , $i=0,1,\ldots$ and b_j , $j=0,1,\ldots$ their convolution is defined by

$$(ab)_k = \sum_{i+j=k} a_i b_j$$

(this is the operation involved in polynomial multiplication).

Consider some methods for doing this in R.

```
convolve0 <- function(a, b) {</pre>
  ab <- rep(0, length(a) + length(b) - 1)
  for(i in 1:length(a))
    for(j in 1:length(b))
      ab[i+j-1] \leftarrow ab[i+j-1] + a[i]*b[j]
  ab
###
convolve1 <- function(a, b) {</pre>
  ab <- rep(0, length(a) + length(b) - 1)
  ind <- 1:length(a)</pre>
  for(j in 1:length(b)) {
    ab[ind] \leftarrow ab[ind] + a*b[j]
    ind <- ind + 1
  ab
```

```
convolve1a <- function(a, b) {</pre>
  if(length(a) < length(b)) Recall(b,a) else {</pre>
    ab <- rep(0, length(a) + length(b) - 1)
    ind <- 1:length(a)</pre>
    for(j in 1:length(b)) {
      ab[ind] \leftarrow ab[ind] + a*b[j]
      ind <- ind + 1
    ab
###
convolve2 <- function(a, b) {</pre>
 p <- outer(a, b)
  as.vector(tapply(p, row(p) + col(p), sum))
### Uses much less memory than convolve2
convolve2a <- function(a, b)</pre>
  as.vector(tapply(outer(a, b),
             outer(seq(along = a), seq(along = b), "+"), sum))
```

The young geek's version.

A C code version

```
File VR_convolve.c has the code:
void VR_convolve(double *a, int *na,
                  double *b, int *nb,
                  double *ab)
  int i, j, nab = *na + *nb - 1;
  for(i = 0; i < nab; i++) ab[i] = 0.0;
  for(i = 0; i < *na; i++)
    for(j = 0; j < *nb; j++)
      ab[i + j] += a[i] * b[j];
}
To compile:
$ R CMD SHLIB VR_convolve.c
```

```
convolve3 <- function(a, b) {</pre>
  if(!is.loaded("VR_convolve")) {
    path <- file.path("SharedObjects",</pre>
                        paste("VR_convolve",
                               .Platform$dynlib.ext, sep=""))
    dyn.load(path)
  storage.mode(a) <- "double"</pre>
  storage.mode(b) <- "double"</pre>
  .C("VR_convolve",
     a,
     length(a),
     b, length(b),
     ab = double(length(a) + length(b) - 1))$ab
```

The Rcpp revolution

- Makes it easy to write compiled code without the need for an R-side interface function (such as the above).
- Can interact directly with R-objects, using R to make most of the mode conversions
- Good tools for creating packages using such code, (Rcpp.package.skeleton, in particular).
- Drawbacks:
 - The code has to be written specifically for Rcpp, using the extensive C++ header files,
 - There can be a small performance overhead using "RcppSugar",
 (but quicker programming and fewer bugs).

Use Rcpp to extend the R system; Use the *dyn.load* to use of existing code, or minimally modified.

An Rcpp convolution function: file src/convolve3a.Cpp

```
#include <Rcpp.h>
using namespace Rcpp;
// [[Rcpp::export]]
NumericVector convolve3a(NumericVector x, NumericVector y)
    int nx = x.size(), ny = y.size(), nz = nx + ny - 1;
    Numeric Vector z(nz); // set to 0 on creation. NB z() not z[] here!
    for(int i = 0; i < nx; ++i) {</pre>
        for(int j = 0; j < ny; ++j) {</pre>
            z[i+j] += x[i]*y[j];
    return z;
```

To make the code available as an R function, you need to have the necessary tools installed, but then:

```
library(Rcpp)
sourceCpp("src/convolve3a.cpp")
```

The R version of the function looks like:

```
convolve3a
function (x, y)
.Call(<pointer: 0x7f60987ca630>, x, y)
```

Alterntively, small functions can be compiled directly as a text string. The headers are assumed:

```
Rcpp::cppFunction('
NumericVector convolve3a(NumericVector x, NumericVector y)
{
   int nx = x.size(), ny = y.size(), nz = nx + ny - 1;
   NumericVector z(nz); // set to 0 on creation. NB z() not z[] here!
   for(int i = 0; i < nx; ++i) {
      for(int j = 0; j < ny; ++j) {
        z[i+j] += x[i]*y[j];
      }
   }
   return z;
}</pre>
```

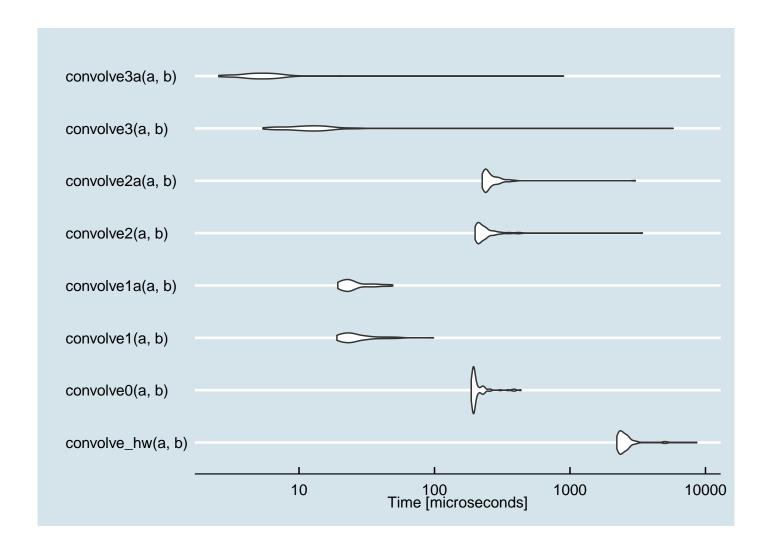
Such functions *must* be re-made the first time they are needed in an R session, in order to integrate them with the R program. They cannot be <code>save()</code>d in one session and <code>load()</code>ed in another, for example.

Permanent versions *can* be made in packages, however, with the system integration automatically happening when the package is loaded.

Some checks:

```
a <- 1:3; b <- 4:7
rbind(convolve0(a,b), convolve1(a,b), convolve1a(a,b),
      convolve2(a,b), convolve2a(a,b), convolve3(a,b),
      convolve3a(a,b), convolve_hw(a, b))
     [,1] [,2] [,3] [,4] [,5] [,6]
[1,]
        4
            13
                 28
                      34
                           32
                                21
[2,]
            13
                 28
                      34
                           32
       4
                                21
[3,]
       4
            13
                 28
                      34
                           32
                                 21
[4,]
       4
            13
                 28
                      34
                           32
                                21
[5,]
            13
                      34
                           32
                                 21
       4
                 28
[6,]
       4
            13
                 28
                      34
                           32
                                21
[7,]
            13
                 28
                           32
                                 21
       4
                      34
[8,]
        4
            13
                 28
                      34
                           32
                                 21
```

```
library(microbenchmark)
a <- 1:300; b <- 4:7
(b <- microbenchmark(convolve_hw(a, b),</pre>
                    convolve0(a,b),
                    convolve1(a,b),
                    convolve1a(a,b),
                    convolve2(a,b),
                    convolve2a(a,b),
                    convolve3(a,b),
                    convolve3a(a,b))) %>% summary() %>% arrange(median) %>%
                               .[, cs(expr, min, median, mean, max, cld)]
                             median
                                                     max cld
                        min
              expr
                                            mean
  convolve3a(a, b) 2.555
                             5.3060
                                       14.72706 897.728 a
1
2
    convolve3(a, b) 5.417
                              12.1750
                                       70.08196 5776.490 ab
  convolve1a(a, b) 19.331
                              23.7105
                                       26.52728 49.217 a
   convolve1(a, b) 19.043
                              24.2050
                                       28.04480 98.359 a
4
5
   convolve0(a, b) 186.723 195.9835
                                       212.63153 434.143 bc
   convolve2(a, b) 199.330 221.9620
6
                                       270.35601 3422.914
  convolve2a(a, b) 224.943
                                       288.69014 3034.108
                             246.1995
                                                           C
8 convolve_hw(a, b) 2216.659 2459.1560 2660.45093 8618.315
                                                            d
suppressMessages(autoplot(b)) + theme_economist()
```



Some timings (the old fashioned way):

```
a <- 1:1000; b <- 1:10000
library(rbenchmark)
benchmark(convolve0 (a,b), convolve1 (a,b), convolve1 (b,a),
         convolve1a(a,b), convolve1a(b,a), convolve2 (a,b),
         convolve2a(a,b), convolve3 (a,b), convolve3a(a,b),
         columns = c("test", "replications", "elapsed", "relative"),
         order = "relative", relative = "elapsed", replications = 20)
             test replications elapsed relative
9 convolve3a(a, b)
                            20
                                0.096
                                          1.000
 convolve3(a, b)
                            20
                               0.099
                                         1.031
  convolve1(b, a)
                               2.804 29.208
                            20
4 convolve1a(a, b)
                            20
                               2.810 29.271
5 convolve1a(b, a)
                               2.948 30.708
                            20
2 convolve1(a, b)
                            20
                               3.068 31.958
  convolve2(a, b)
                            20
                               16.996 177.042
7 convolve2a(a, b)
                            20 17.492 182.208
1 convolve0(a, b)
                            20 22.478 234.146
```

Session information

Date: 2021-01-29

• R version 4.0.3 (2020-10-10), x86_64-pc-linux-gnu

• Running under: Ubuntu 20.04.1 LTS

• Matrix products: default

• BLAS: /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.9.0

• LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.9.0

• Base packages: base, datasets, graphics, grDevices, methods, parallel, stats, utils

- Other packages: doParallel 1.0.16, dplyr 1.0.3, english 1.2-5, forcats 0.5.1, foreach 1.5.1, ggplot2 3.3.3, ggthemes 4.2.4, gridExtra 2.3, iterators 1.0.13, knitr 1.31, lattice 0.20-41, microbenchmark 1.4-7, patchwork 1.1.1, purrr 0.3.4, Rcpp 1.0.6, readr 1.4.0, scales 1.1.1, stringr 1.4.0, tibble 3.0.5, tidyr 1.1.2, tidyverse 1.3.0, WWRCourse 0.2.3, WWRData 0.1.0, WWRGraphics 0.1.2, WWRUtilities 0.1.2, xtable 1.8-4
- Loaded via a namespace (and not attached): assertthat 0.2.1, backports 1.2.1, broom 0.7.3, cellranger 1.1.0, cli 2.2.0, codetools 0.2-18, colorspace 2.0-0, compiler 4.0.3, crayon 1.3.4, DBI 1.1.1, dbplyr 2.0.0, digest 0.6.27, ellipsis 0.3.1, evaluate 0.14, fansi 0.4.2, farver 2.0.3, fractional 0.1.3, fs 1.5.0, generics 0.1.0, glue 1.4.2, grid 4.0.3, gtable 0.3.0, haven 2.3.1, highr 0.8, hms 1.0.0, httr 1.4.2, jsonlite 1.7.2, labeling 0.4.2, lazyData 1.1.0, lifecycle 0.2.0, lubridate 1.7.9.2, magrittr 2.0.1, MASS 7.3-53, Matrix 1.3-2, modelr 0.1.8, multcomp 1.4-15, munsell 0.5.0, mvtnorm 1.1-1, PBSmapping 2.73.0, pillar 1.4.7, pkgconfig 2.0.3, R6 2.5.0, randomForest 4.6-14, readxl 1.3.1, reprex 1.0.0, rlang 0.4.10, rpart 4.1-15, rstudioapi 0.13, rvest 0.3.6, sandwich 3.0-0, SOAR 0.99-11, splines 4.0.3, stringi 1.5.3, survival 3.2-7, TH.data 1.0-10, tidyselect 1.1.0, tools 4.0.3, vctrs 0.3.6, withr 2.4.1, xfun 0.20, xml2 1.3.2, zoo 1.8-8