Piping and Functions

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Calculating a Geometric Mean

```
Create some random data
x <- abs(rnorm(100))
head(x)
## [1] 0.8294 0.9943 1.4024 0.4421 0.1621 0.4641
Create a new object after each function
log_x \leftarrow log(x)
mean_logx <- mean(log_x)</pre>
exp(mean_logx)
## [1] 0.412
Nest functions
exp(mean(log(x)))
## [1] 0.412
Overight original object
x \leftarrow log(x)
x \leftarrow mean(x)
exp(x)
Use the pipe \%>\%
library(tidyverse)
## -- Attaching packages -- tidyverse 1.2.1 --
## v ggplot2 3.1.0
                       v purrr
                                   0.2.5
## v tibble 2.0.1
                                   0.7.8
                        v dplyr
            0.8.2
## v tidyr
                        v stringr 1.3.1
             1.3.1
                        v forcats 0.3.0
## v readr
## -- Conflicts ---- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                      masks stats::lag()
x %>%
  log() %>%
  mean() %>%
  exp()
```

Piping always inserts the output into the first argument of the next function

[1] 0.412

```
x[1] \leftarrow NA
x %>%
  log() %>%
  mean(na.rm = TRUE) %>%
  exp()
## [1] 0.4091
exercise solutions
rnorm(n = 100, sd = 10) \%
  sort(decreasing = FALSE) %>%
  diff() %>%
  mean() %>%
  round(digits = 1)
## [1] 0.5
rnorm(100) %>%
  sort()
     [1] -2.90090 -2.11750 -2.00286 -1.81862 -1.81090 -1.56869 -1.44951
##
##
     [8] -1.39331 -1.35129 -1.24939 -1.20011 -1.19408 -1.18180 -1.17203
##
    [15] -1.15507 -1.11632 -1.06064 -1.02052 -1.01217 -0.99334 -0.97741
##
    [22] -0.96525 -0.90063 -0.88789 -0.83108 -0.80094 -0.78991 -0.78040
   [29] -0.77806 -0.77294 -0.68202 -0.67551 -0.59222 -0.58551 -0.54804
##
   [36] -0.53896 -0.51476 -0.49317 -0.48020 -0.45286 -0.43329 -0.36720
##
    [43] -0.34558 -0.30414 -0.26969 -0.20931 -0.19930 -0.13298 -0.06567
   [50] -0.06452 -0.04790 -0.02996 0.01750 0.02230 0.03120
##
                                                                0.05305
   [57] 0.10303 0.11075 0.12665 0.14210 0.16489
                                                      0.17145
##
   [64] 0.19835 0.24050 0.24209 0.28978 0.31234
                                                       0.32742
                                                                0.34449
    [71] 0.52859 0.60359 0.61438 0.64336
                                              0.66301
                                                       0.69345
##
                                                                0.70703
##
   [78] 0.74657 0.74701 0.76130 0.79188
                                              0.81992 0.82273
                                                                0.88927
   [85]
         1.02199 1.03326 1.06103 1.16587
                                              1.18060 1.19263
                                                                1.24619
##
         1.28894 1.38886 1.41441 1.55149 1.59374 1.71939 1.91795
    [92]
   [99] 1.95062 2.32579
sort(rnorm(100))
##
     [1] -2.34726 -2.01772 -1.92093 -1.85237 -1.61404 -1.56923 -1.50231
##
     [8] -1.36368 -1.32001 -1.28425 -1.23541 -1.21947 -1.21817 -1.19260
    [15] -1.18037 -1.16869 -1.16820 -1.11069 -1.10871 -1.10402 -1.09757
##
##
    [22] -1.00617 -0.96740 -0.96708 -0.96093 -0.93438 -0.91852 -0.83635
##
     [29] \ -0.81458 \ -0.81424 \ -0.78171 \ -0.73065 \ -0.72157 \ -0.63521 \ -0.62064 
    [36] -0.62008 -0.56510 -0.56279 -0.50604 -0.46071 -0.45753 -0.45605
     \begin{bmatrix} 43 \end{bmatrix} \ -0.45241 \ -0.44940 \ -0.42420 \ -0.36918 \ -0.35914 \ -0.32952 \ -0.24646 
##
    [50] -0.23534 -0.19531 -0.14924 -0.11776 -0.11415 -0.05792 -0.01594
##
##
    [57] 0.02506 0.03467 0.03710 0.04187 0.11456 0.11533 0.14171
##
   [64] 0.14257 0.14314 0.14798 0.15740 0.26719 0.27243 0.27780
##
    [71] 0.31680 0.40650 0.41574 0.41682
                                              0.45567
                                                       0.49734
                                                                0.50464
##
    [78]
         0.51341 0.55388 0.62305 0.73706
                                              0.82900
                                                       0.88840
                                                                0.96975
##
    [85]
         1.00934 1.04530 1.05521 1.06649
                                              1.15337
                                                       1.16481 1.18453
   [92]
         1.19699 1.48007 1.50020 1.68530 1.73584 1.81377 2.05834
##
   [99] 2.13861 2.59577
```

Functions and Function Creation

A very basic function

```
add_two <- function(a, b) {</pre>
  ## tons of code goes here
  c <- a + b
 return(c)
add_two(a = 2, b = 5)
## [1] 7
The book example
df <- data.frame(</pre>
 a = rnorm(100),
 b = rnorm(100),
 c = rnorm(100),
  d = rnorm(100)
head(df)
##
           a
                   b
                            С
## 1 1.8291 -1.1056 -1.74077 1.02329
## 2 -0.7598  0.4003  1.14011  0.04564
## 3 -1.0405 1.5169 -0.01836 0.23579
## 4 -0.9783 -0.9541 1.99346 0.18172
## 5 -0.1687 0.3324 0.66745 0.60737
## 6 0.3156 2.7165 -0.79239 -0.94581
```

Suppose the goal is to transform these variables so that all elements are between 0 and 1

Copy code

- Obnoxious
- Subject to error
- If you have to change later, it is obnoxious and subject to error

```
df$a <- ((df$a - min(df$a))) / (max(df$a) - min(df$a))
df$b <- ((df$b - min(df$b))) / (max(df$a) - min(df$b))
df$c <- ((df$c - min(df$c))) / (max(df$c) - min(df$c))
df$d <- ((df$d - min(df$d))) / (max(df$d) - min(df$d))</pre>
```

Create a function!

```
x <- df$a

rescale1 <- function(x) {

## Calculate minimum ------
min_x <- min(x, na.rm = TRUE)

## Rescale ------
rescale_vec <- (x - min_x) /
   (max(x, na.rm = TRUE) - min_x)

return(rescale_vec)</pre>
```

```
rescale1(c(1, 2, 3))
## [1] 0.0 0.5 1.0
df$a <- rescale1(df$a)
df$b <- rescale1(df$b)
df$c <- rescale1(df$c)</pre>
df$d <- rescale1(df$d)
rescale1(c(NA, 1, 3, 4))
## [1]
            NA 0.0000 0.6667 1.0000
x \leftarrow c(NA, NA, 2, 3, NA)
y < -c(NA, NA, 5, 6, 7)
both_na <- function(x, y) {</pre>
  na_x <- is.na(x)</pre>
  na_y <- is.na(y)</pre>
  index_vector <- 1:length(x)</pre>
  sum(na_x & na_y)
}
sum(c(TRUE, TRUE, TRUE))
## [1] 3
```

If-then statements

```
x <- 6
if (x > 5) {
    ## code
    "hurray!"
} else {
    ## other code
    "oh well"
}
```

[1] "hurray!"

The has-name function

Check if a vector has any names attached if not then it will return FALSE for all elements

If yes, then it will return TRUE in the positions where the vector has a name

```
x \leftarrow c(a = 1, 2, 3)
names(x)
```

```
## [1] "a" ""
has_name <- function(x) {
  nms <- names(x)
  if (is.null(nms)) {
    rep(FALSE, length(x))
} else {
  !is.na(nms) & nms != ""</pre>
```

```
}

x1 <- c(1,2,3)
x2 <- c(a = 1, 2,3)
has_name(x1)</pre>
```

[1] FALSE FALSE FALSE

Multiple conditional execution

```
x <- -5
if (x > 5) {
   "Hurray!"
} else if (x > 0) {
   "oh well"
} else if (x > -10) {
   "oh boy"
} else {
   "oh no"
}
```

[1] "oh boy"