Exercise: Functions

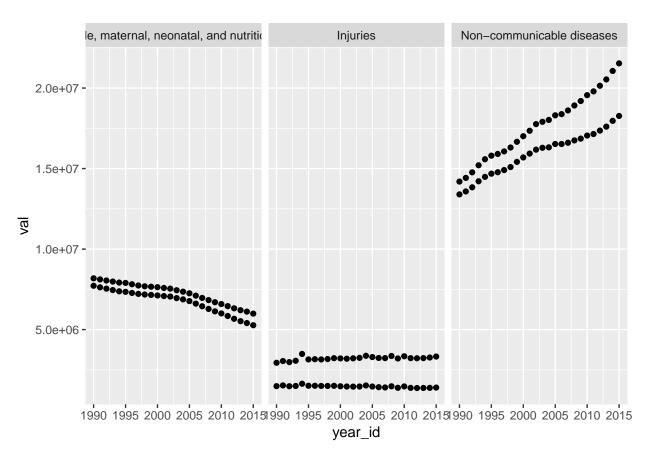
Day 4, Part B

```
> library(foreign)
> library(ggplot2)
> library(reshape2)
```

1. Write a function where the arguments are in_file, x_variable, y_variable, facet_variable, and out_file, that loads the specified input file, creates a scatter plot with the specified x, y, and facetting variables using ggplot, and then saves this plot to the specified output file while also returning the ggplot object. Assume the input file is a csv and the output file is a pdf. Hint: in this particular case, aes_string() is more user-friendly than aes() when calling ggplot.

```
> scatter_plot <- function(in_file, x_variable, y_variable, facet_variable, out_file) {
    # load data
    data <- read.csv(in_file)</pre>
+
    # make plot
+
    gg <- ggplot(data, aes_string(x = x_variable, y = y_variable)) +</pre>
+
      facet_wrap(facet_variable) + geom_point()
+
    # save plot
+
    pdf(out_file)
    print(gg)
    dev.off()
    # return plot
    return(gg)
+ }
```

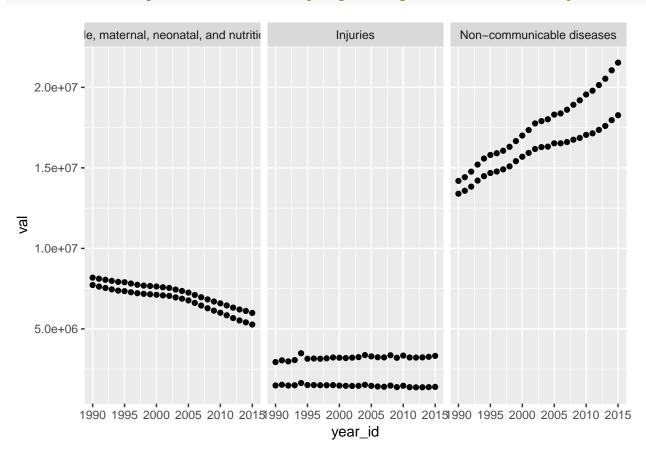
a. Test this function using the data in 'data/gbd2015_global_deaths.csv', with year (year_id) on the x-axis, mean deaths (val) on the y-axis, and cause group (cause_name) as the facetting variable.



b. Add assertions to the function to check that the input and output files are the expected format, and to provide a helpful error message if not. Test this by specifying an output file that is a jpeg, rather than a pdf. Hint: think back to the string functions lecture.

```
> scatter_plot <- function(in_file, x_variable, y_variable, facet_variable, out_file) {
    # check input and output data format
    if (!grepl('.csv$|.CSV$', in file)) stop("'in file' must be a csv file")
    if (!grepl('.pdf$|.PDF$', out_file)) stop("'out_file' must be a pdf file")
    # load data
    data <- read.csv(in_file)</pre>
    # make plot
    gg <- ggplot(data, aes_string(x = x_variable, y = y_variable)) +</pre>
      facet_wrap(facet_variable) + geom_point()
    # save plot
   pdf(out_file)
   print(gg)
    dev.off()
    # return plot
    return(gg)
+ }
> # this should work...
> scatter_plot(in_file = paste0(main_dir, 'data/gbd2015_global_deaths.csv'),
       x_variable = 'year_id', y_variable = 'val', facet_variable = 'cause_name',
```

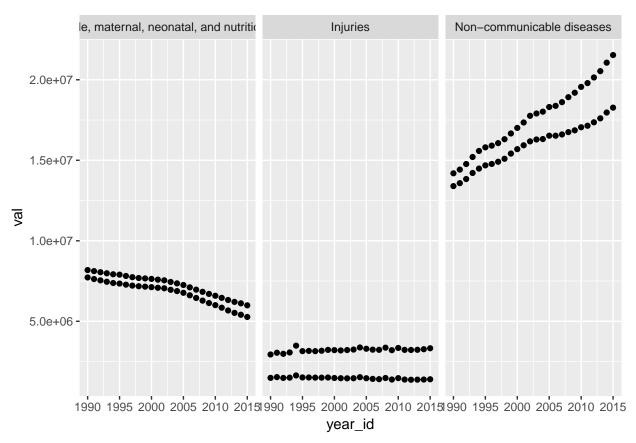




c. Add assertions to the function to check that x_variable, y_variable, and facet_variable all exist as columns in the input data file, and to provide a helpful error message if not. Test this by providing an incorrect variable name.

```
> scatter_plot <- function(in_file, x_variable, y_variable, facet_variable, out_file) {
+    # check input and output data format
+    if (!grepl('.csv$|.CSV$', in_file)) stop("'in_file' must be a csv file")
+    if (!grepl('.pdf$|.PDF$', out_file)) stop("'out_file' must be a pdf file")
+    # load data
+    data <- read.csv(in_file)
+    # check that all variables exist
+    if (!x_variable %in% names(data)) stop(paste(x_variable, 'is not a column in', in_file))
+    if (!y_variable %in% names(data)) stop(paste(y_variable, 'is not a column in', in_file))
+    if (!facet_variable %in% names(data)) stop(paste(facet_variable, 'is not a column in', in_file))
+    # make plot
+    # make plot
+    gg <- ggplot(data, aes_string(x = x_variable, y = y_variable)) +
+    facet_wrap(facet_variable) + geom_point()</pre>
```

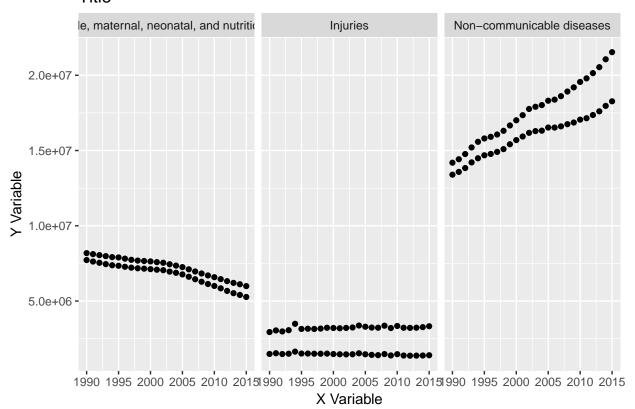
```
+ # save plot
+ pdf(out_file)
+ print(gg)
+ dev.off()
+
+ # return plot
+ return(gg)
+ }
> 
> # this should work...
> scatter_plot(in_file = paste0(main_dir, 'data/gbd2015_global_deaths.csv'),
+ x_variable = 'year_id', y_variable = 'val', facet_variable = 'cause_name',
+ out_file = paste0(main_dir, 'output/gbd2015_global_deaths_scatter.pdf'))
```



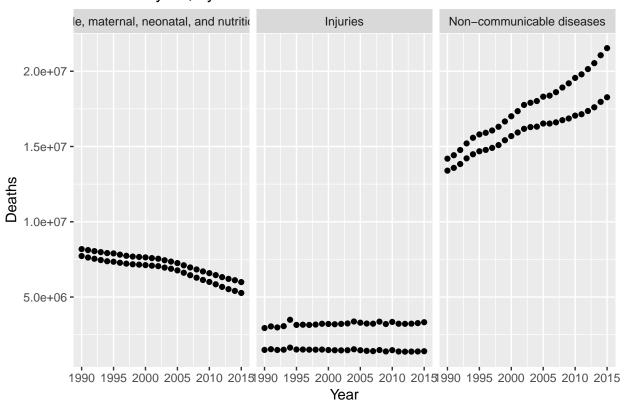
d. Add arguments x_label, y_label, and title to provide labels to the plot. Set them to default to "X Variable", "Y Variable", and "Title", respectively. Test that this works as expected, both when values are supplied for these arguments and when they are not.

```
# load data
    data <- read.csv(in_file)</pre>
    # check that all variables exist
    if (!x_variable %in% names(data)) stop(paste(x_variable, 'is not a column in', in_file))
    if (!y_variable %in% names(data)) stop(paste(y_variable, 'is not a column in', in_file))
    if (!facet_variable %in% names(data)) stop(paste(facet_variable, 'is not a column in', i
    # make plot
    gg <- ggplot(data, aes_string(x = x_variable, y = y_variable)) +</pre>
      facet_wrap(facet_variable) + geom_point() +
      labs(x = x_label, y = y_label, title = title)
    # save plot
   pdf(out_file)
   print(gg)
   dev.off()
   # return plot
    return(gg)
+ }
>
> # using the defaults for labels/titles...
> scatter_plot(in_file = paste0(main_dir, 'data/gbd2015_global_deaths.csv'),
       x_variable = 'year_id', y_variable = 'val', facet_variable = 'cause_name',
       out_file = paste0(main_dir, 'output/gbd2015_global_deaths_scatter.pdf'))
```

Title



Deaths vs year, by Cause



2. Write a function that takes one argument data, a data frame, and returns a report of the number of missing and non-missing values for each variable, as well as the number of unique values, e.g.:

```
> # note that there are many ways to do this, and more than one valid output format
> missing_report <- function(data) {</pre>
    report <- data.frame(vars = names(data),
                           missing = NA,
+
                           non_missing = NA,
+
                           values = NA)
+
    for (v in names(data)) {
+
      report[report$vars == v, 'missing'] <- sum(is.na(data[, v]))</pre>
+
      report[report$vars == v, 'non_missing'] <- sum(!is.na(data[, v]))</pre>
+
      report[report$vars == v, 'values'] <- length(unique(data[, v]))</pre>
+
    }
+
    return(report)
+ }
> data <- read.csv(paste0(main_dir, "data/ebola_polygon_data.csv"))</pre>
> missing_report(data)
vars missing non_missing values
       UNIQ_ID
                                  54
                                          54
```

```
2
           NAME
                        0
                                     54
                                             53
3
                        0
                                     54
                                              6
        Country
4
          Virus
                        0
                                     54
                                              4
5
     CASE_TYPE
                        0
                                     54
                                              3
6
     DATA_TYPE
                        0
                                     54
                                              1
7
                        0
            LAT
                                     54
                                             46
8
           LONG
                        0
                                     54
                                             45
9
     SPR_ORDER
                       18
                                     36
                                              6
10
       SOURCE 1
                       32
                                     22
                                             20
11
       SOURCE_2
                       54
                                      0
                                              1
                                      0
12
       SOURCE_3
                       54
                                              1
13
        STR_DAY
                       42
                                     12
                                             11
14
       STR_MNTH
                       18
                                     36
                                             10
15
       STR_YEAR
                        4
                                     50
                                             15
                                      2
16
        END_DAY
                       52
                                              3
17
       END_MNTH
                       52
                                      2
                                              3
                                      2
                                              3
18
      END_YEAR
                       52
19
       REP_CASE
                                     26
                                             18
                       28
20
     REP_DEATH
                       39
                                     15
                                             12
21
          OB_ID
                        0
                                             23
                                     54
22
    OB_STR_DAY
                        3
                                     51
                                             13
23 OB_STR_MNTH
                        0
                                     54
                                             10
24 OB_STR_YEAR
                        0
                                     54
                                             15
25
    OB END DAY
                       13
                                     41
                                             12
26 OB_END_MNTH
                        0
                                     54
                                             11
27 OB_END_YEAR
                        0
                                     54
                                             14
28
        OB_CASE
                        0
                                     54
                                             18
29
       OB_DEATH
                        0
                                     54
                                             19
```

3. Repeat question 1 from lecture 4a using lapply() instead of a for loop.

```
> data <- lapply(1997:2015, function(year) {</pre>
    if (year < 2004) {
+
      sub <- read.csv(paste0(main_dir, "/data/wa_income_", year, ".csv"))</pre>
+
+
    } else {
+
      sub <- read.dta(paste0(main_dir, "/data/wa_income_", year, ".dta"))</pre>
      sub <- plyr::rename(sub, c("FIPS" = "fips", "median_income" = "income_median"))</pre>
+
    }
+
    return(sub)
+
+ })
> data <- do.call('rbind', data)</pre>
> summary(data)
fips
                            income_median
                 year
                                                poverty
                                          :27453
Min.
        :53001
                  Min.
                          :1997
                                  Min.
                                                   Min.
                                                           : 6.60
 1st Qu.:53019
                  1st Qu.:2001
                                  1st Qu.:36992
                                                   1st Qu.:11.50
Median :53039
                  Median :2006
                                  Median :42369
                                                   Median :14.10
Mean
        :53039
                  Mean
                         :2006
                                  Mean
                                          :43726
                                                   Mean
                                                           :14.26
 3rd Qu.:53059
                  3rd Qu.:2011
                                  3rd Qu.:48693
                                                   3rd Qu.:16.40
Max. :53077
                  Max. :2015
                                  Max. :81816
                                                   Max. :32.30
```

Bonus:

4. Repeat the first part of question 3 from lecture 4a using sapply() or tapply() instead of a for loop. Then do the same again, but calculate the ratio of the maximum to the minimum value instead of the mean.

```
> # mean with sapply...
> sapply(1997:2015, function(yy) {
    mean(data[data$year == yy, 'income_median'])
+ })
[1] 36051.05 37450.33 37636.90 39212.51 38255.59 38989.44 39935.90 41323.79 41837.21
[10] 43795.28 46171.38 48219.13 47017.18 46621.33 47544.08 48141.85 49234.31 50873.38
[19] 52487.00
> # mean with tapply...
> tapply(data$income_median, data$year, mean)
         1998
                  1999
                            2000
                                     2001
                                              2002
                                                       2003
                                                                 2004
                                                                          2005
36051.05 37450.33 37636.90 39212.51 38255.59 38989.44 39935.90 41323.79 41837.21
             2007
                      2008
                                2009
                                         2010
                                                  2011
                                                            2012
                                                                     2013
43795.28 46171.38 48219.13 47017.18 46621.33 47544.08 48141.85 49234.31 50873.38
    2015
52487.00
> # max/min with sapply...
> sapply(1997:2015, function(yy) {
    x <- data[data$year == yy, 'income_median']</pre>
    \max(x)/\min(x)
+ })
[1] 1.868648 1.872010 1.816942 1.852431 1.857221 1.819798 1.761449 1.822441 1.842353
[10] 1.928604 1.906808 2.005580 1.945463 1.853792 1.985815 2.136076 2.010447 2.004605
[19] 2.034768
> # max/min with tapply...
> tapply(data$income_median, data$year, function(x) max(x) / min(x))
         1998
                  1999
                            2000
                                     2001
                                              2002
                                                       2003
                                                                 2004
                                                                          2005
1.868648 1.872010 1.816942 1.852431 1.857221 1.819798 1.761449 1.822441 1.842353
             2007
                      2008
                                2009
                                         2010
                                                  2011
                                                            2012
                                                                     2013
1.928604 1.906808 2.005580 1.945463 1.853792 1.985815 2.136076 2.010447 2.004605
    2015
2.034768
```

5. Load the US cigarette data ('data/us_state_cigarette_data.rdata') and put all four of the data frames into a single list. Use the Reduce() function to merge all of these data sets in one go, rather than by calling merge directly three separate times. Do this once retaining only the rows where there are matches, and a second time retaining all rows.

```
> # load data and combine in a list
> load(paste0(main_dir, "/data/us_state_cigarette_data.rdata"), verbose = T)
Loading objects:
    cig_tax
    cig_csmp
    pop
    locs
> cig_data_list <- list(cig_tax, cig_csmp, pop, locs)
> # for only matching rows, we are totally fine with defaults
> cig_data <- Reduce(merge, cig_data_list)
> dim(cig_data)
[1] 204 6
```

```
> # for all rows, we need to set `all=T`, which requires defining a function that calls
> # `merge()` with this setting
> cig_data <- Reduce(function(x, y) merge(x, y, all = T), cig_data_list)
> dim(cig_data)
[1] 306 6
```