Types, vectors, and functions in R

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Vectors and Types

Vectors

```
c(1, 3, 5)
c(TRUE, FALSE, TRUE, TRUE)
c("red", "blue")
```

Vectors

Vectors have 1 dimension

Vectors have a length

length(c("blue", "red"))

Some vectors have names.

$$names(c("x" = 1, "y" = 1))$$

Vectors have types

Types

Packages to work with types

Strings/character: stringr

Factors: forcats

Dates: lubridate

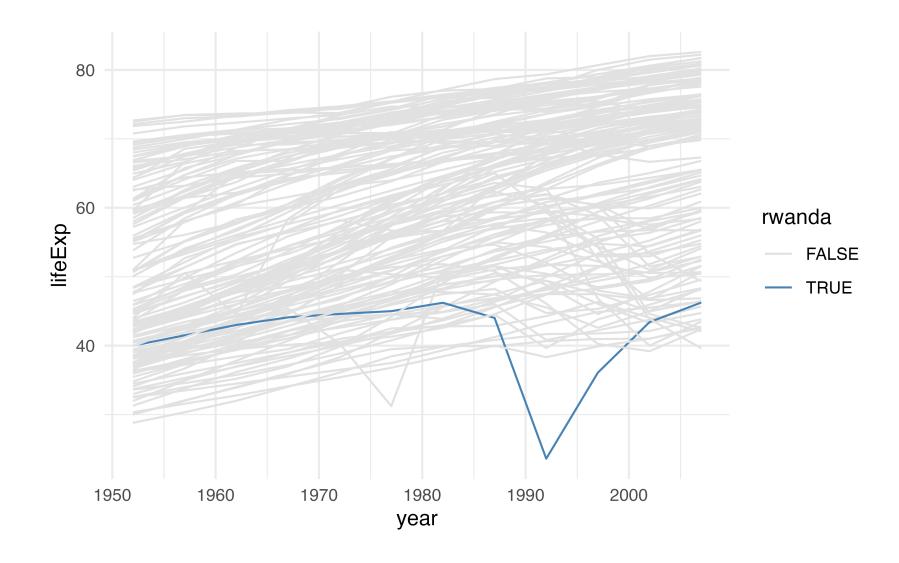
Making vectors

```
1 1:3
1 c(1, 2, 3)
1 \text{ rep}(1, 3)
1 seq(from = 1, to = 3, by = .5)
```

Create a character vector of colors using c(). Use the colors "grey90" and "steelblue". Assign the vector to a name.

Use the vector you just created to change the colors in the plot below using scale_color_manual(). Pass it using the values argument.

```
1 cols <- c("grey90", "steelblue")
2
3 gapminder |>
4  mutate(rwanda = country == "Rwanda") |>
5  ggplot(aes(year, lifeExp, color = rwanda, group = country == country == color_manual(values = cols) + color_manual()
```



Working with vectors

 $1 \times < -c(1, 5, 7)$ $2 \times [2]$ $1 \times [2]$ 1 x[c(FALSE, TRUE, FALSE)] 1 x[[c(FALSE, TRUE, FALSE)]]

Working with vectors Modify elements

```
1 x
```

[1] 1 5 7

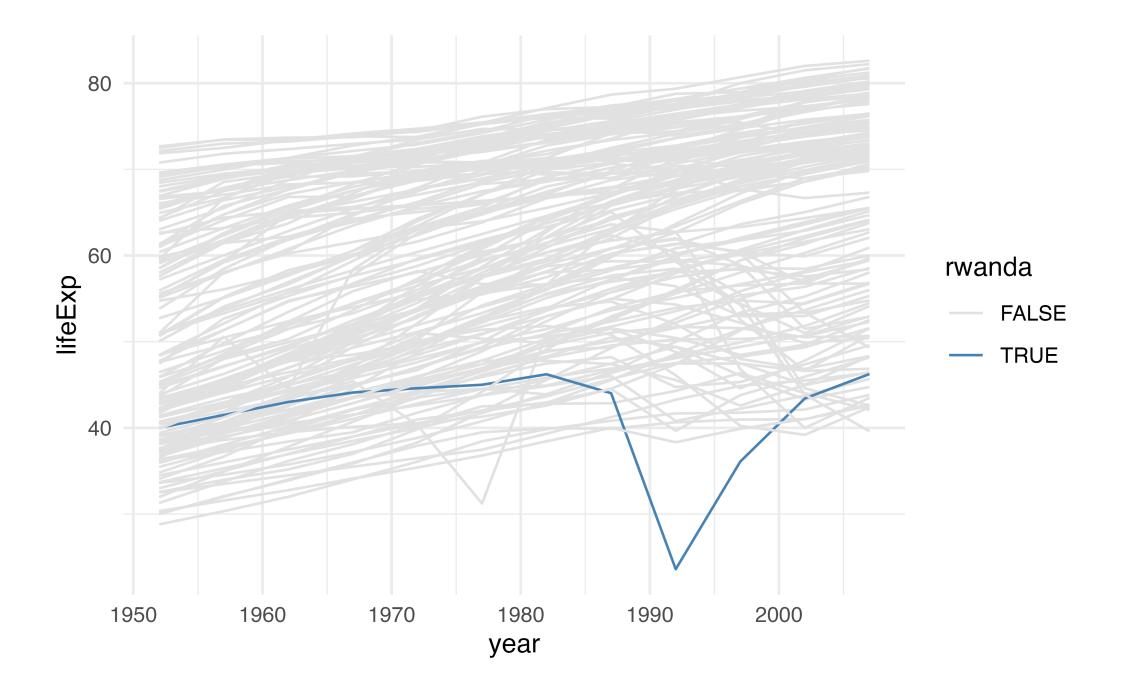
Working with vectors Modify elements

```
1 x[[2]] <- 100
2 x
```

Working with vectors Modify elements

```
1 x[x > 10] <- NA
2 x
```

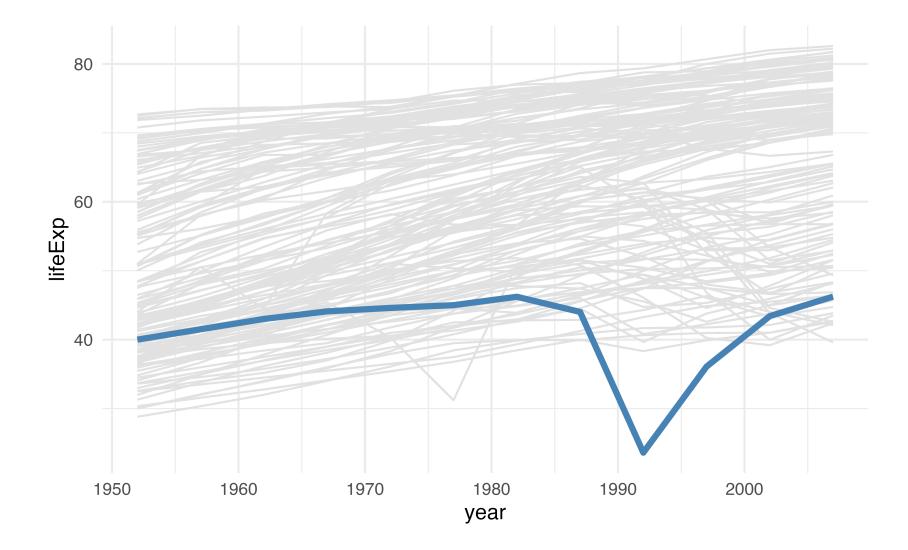
```
[1] 1 NA 7
```



```
1 cols <- c("grey90", "steelblue")
2
3 gapminder |>
4    mutate(rwanda = ifelse(country == "Rwanda", TRUE, FALS
5    ggplot(aes(year, lifeExp, color = rwanda, group = coundations geom_line() +
7    scale_color_manual(values = cols) +
8    theme_minimal()
```

```
1 cols <- c("grey90", "steelblue")
2
3 gapminder |>
4    mutate(rwanda = ifelse(country == "Rwanda", TRUE, FALS
5    ggplot(aes(year, lifeExp, group = country)) +
6    geom_line(
7     data = function(x) filter(x, !rwanda),
8    color = cols[[1]]
9    ) +
10    theme_minimal()
```

```
1 cols <- c("grey90", "steelblue")</pre>
   gapminder |>
     mutate(rwanda = ifelse(country == "Rwanda", TRUE, FAL
 5
     ggplot(aes(year, lifeExp, color = rwanda, group = coul
     geom line(
       data = function(x) filter(x, !rwanda),
       color = cols[[1]]
 9
     ) +
10
   geom line(
11
       data = function(x) filter(x, rwanda),
12 color = cols[[2]],
       linewidth = 1.5
13
14 ) +
15 theme minimal()
```



Create a numeric vector that has the following values: 3, 5, NA, 2, and NA.

Try using sum(). Then add na.rm = TRUE.

Check which values are missing with is.na();
save the results to a new object and take a look

Change all missing values of x to -9999.

Try sum() again without na.rm = TRUE.

21

```
1 x <- c(3, 5, NA, 2, NA)
2 sum(x)
```

[1] NA

```
1 sum(x, na.rm = TRUE)
```

[1] 10

```
1 x_missing <- is.na(x)
2 x_missing

[1] FALSE FALSE TRUE FALSE TRUE

1 x[x_missing] <- -9999
2 x

[1] 3 5 -9999 2 -9999</pre>
```

Functions that return vectors

```
gdpPercap <- gapminder |>
filter(year == 1952, continent == "Americas") |>
pull(gdpPercap)

sum(gdpPercap)
```

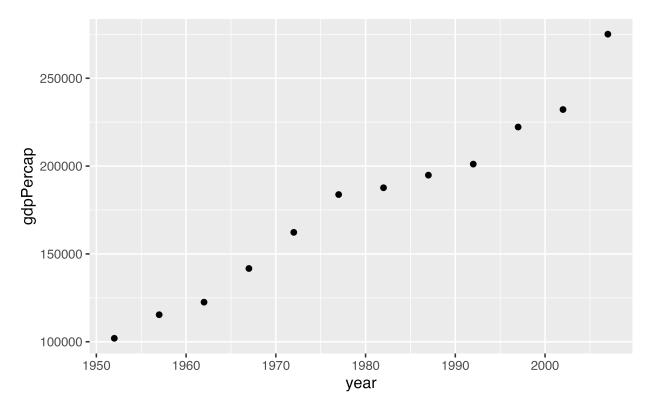
```
[1] 101976.6
```

Functions that return data frames

```
gapminder >
             group by(year) |>
             filter(continent == "Americas") |>
             summarize(gdpPercap = sum(gdpPercap))
# A tibble: 12 \times 2
   year qdpPercap
  <int> <dbl>
   1952 101977.
   1957 115401.
   1962 122539.
   1967 141706.
   1972 162283.
   1977 183800.
   1982 187668.
   1987 194835.
 9
   1992 201123.
   1997 222233.
10
```

Functions that make plots

```
gapminder |>
group_by(year) |>
filter(continent == "Americas") |>
summarize(gdpPercap = sum(gdpPercap)) |>
ggplot(aes(year, gdpPercap)) +
geom_point()
```



Why write functions?

To make repetitive code reusable

To make complex code understandable

To make useful code shareable

Function arguments

```
add_one <- function(x) {
    x <- x + 1
    x
}
add_one(1)
#> 2
Create function
```

```
function name
add_one <- function(x) {</pre>
   x < -x + 1
               function
                   body
→ add_one(1)
#> 2
```

```
add_one <- function(x) {
    x <- x + 1
    x output
}
add_one(1) — input
#> 2
```

Create a function called sim_data that doesn't take any arguments. In the function body, we'll return a tibble.

For x, have rnorm() return 50 random numbers.

For sex, use rep() to create 50 values of "male" and "female". Hint: You'll have to give rep() a character vector. for the first argument. The times argument is how many times rep() should repeat the first argument, so make sure you account for that.

For age() use the sample() function to sample 50 numbers from 25 to 50 with replacement.

Call sim_data()

```
1 sim_data <- function() {
2    tibble(
3          x = rnorm(50),
4          sex = rep(c("male", "female"), times = 25),
5          age = sample(25:50, size = 50, replace = TRUE)
6    )
7 }
8
9 sim_data()</pre>
```

```
# A tibble: 50 \times 3
       x sex
                 age
   <dbl> <chr> <int>
1 -1.74 male 46
2 - 0.301 female 46
3 - 0.894 male 38
4 0.484 female 49
5 0.0600 male
                 47
6 0.357 female
                 46
7 0.305 male
                 42
8 -1.38 female
                 34
9 -0.223 male
                 38
10 - 1.51 female
                 33
```

E-Values

The strength of unmeasured confounding required to explain away a value

Rate ratio: 3.9 = E-value: 7.3

Write a function to calculate an E-Value given an RR.

Call the function evalue and give it an argument called estimate. In the body of the function, calculate the E-Value using estimate + sqrt(estimate * (estimate - 1))

Call evalue() for a risk ratio of 3.9

```
1 evalue <- function(estimate) {
2   estimate + sqrt(estimate * (estimate - 1))
3 }</pre>
```

Invoking the function with ()

```
1 evalue(3.9)
```

```
[1] 7.263034
```

Control Flow

```
1 if (PREDICATE) {
2  true_result
3 }
```

Control Flow

```
1 if (PREDICATE) {
2   true_result
3 } else {
4   default_result
5 }
```

Control Flow

```
1 if (PREDICATE) {
2   true_result
3 } else if (ANOTHER_PREDICATE) {
4   true_result
5 } else {
6   default_result
7 }
```

If-else with vectors

```
ifelse(PREDICATE_VECTOR, true_result, false_result)
dplyr::case_when(
    PREDICATE_VECTOR ~ true_result,
    PREDICATE_VECTOR ~ next_true_result,
    .default = default_result

)
switch(
    x,
    value1 = result1,
    value2 = result2
)
```

Validation and stopping

if (is.numeric(x)) stop(), warn()

```
1 function(x) {
2   if (is.numeric(x)) stop("x must be a character")
3   # do something with a character
4 }
```

Use if () together with is.numeric() to make sure estimate is a number. Remember to use! for not.

If the estimate is less than 1, set estimate to be equal to 1 / estimate.

Call evalue() for a risk ratio of 3.9. Then try 0.80. Then try a character value.

```
1 evalue <- function(estimate) {
2   if (!is.numeric(estimate)) stop("`estimate` must be not if (estimate < 1) estimate <- 1 / estimate 
4   estimate + sqrt(estimate * (estimate - 1))
5 }</pre>
```

```
1 evalue(3.9)
[1] 7.263034

1 evalue(.80)
[1] 1.809017

1 evalue("3.9")
Error in evalue("3.9"): `estimate` must be numeric
```

Add a new argument called type. Set the default value to "rr"

Check if type is equal to "or". If it is, set the value of estimate to be sqrt(estimate)

Call evalue() for a risk ratio of 3.9. Then try it again with type = "or".

```
1 evalue <- function(estimate, type = "rr") {
2   if (!is.numeric(estimate)) stop("`estimate` must be not
3   if (type == "or") estimate <- sqrt(estimate)
4   if (estimate < 1) estimate <- 1 / estimate
5   estimate + sqrt(estimate * (estimate - 1))
6 }</pre>
```

```
1 evalue(3.9)
[1] 7.263034

1 evalue(3.9, type = "or")
[1] 3.362342
```

Your Turn 7: Challenge!

Create a new function called transform_to_rr with arguments estimate and type.

Use the same code above to check if type == "or" and transform if so. Add another line that checks if type == "hr". If it does, transform the estimate using this formula: (1 - 0.5^sqrt(estimate)) / (1 - 0.5^sqrt(1 / estimate)).

Move the code that checks if estimate < 1 to transform_to_rr (below the OR and HR transformations)

Return estimate

In evalue(), change the default argument of type to be a character vector containing "rr", "or", and "hr".

Get and validate the value of type using match.arg(). Follow the pattern argument_name <-match.arg(argument_name)

Transform estimate using transform_to_rr(). Don't forget to pass it both estimate and type!

Your Turn 7: Challenge!

```
1 transform to rr <- function(estimate, type) {</pre>
      if (type == "or") estimate <- sqrt(estimate)</pre>
      if (type == "hr") {
      estimate <-
           (1 - 0.5^{\circ} \text{sqrt}(\text{estimate})) / (1 - 0.5^{\circ} \text{sqrt}(1 / \text{estimate}))
 6
      if (estimate < 1) estimate <- 1 / estimate
 8
      estimate
10 }
11
12 evalue <- function(estimate, type = c("rr", "or", "hr")) {</pre>
      # validate arguments
13
14
      if (!is.numeric(estimate)) stop("`estimate` must be numeric")
      type <- match.arg(type)</pre>
15
16
      # calculate evalue
17
18
      estimate <- transform to rr(estimate, type)</pre>
      estimate + sqrt(estimate * (estimate - 1))
19
20 }
```

Your Turn 7: Challenge!

```
1 evalue(3.9)
[1] 7.263034
         1 evalue(3.9, type = "or")
[1] 3.362342
         1 evalue(3.9, type = "hr")
[1] 4.474815
         1 evalue(3.9, type = "rd")
Error in match.arg(type): 'arg' should be one of "rr", "or", "hr"
```

Programming with the tidyverse

Pass the dots: . . .

```
1 select_gapminder <- function(...) {
2    gapminder |>
3        select(...)
4  }
5
6 select_gapminder(pop, year)
```

Pass the dots: . . .

```
# A tibble: 1,704 \times 2
       pop year
      <int> <int>
 1 8425333 1952
   9240934 1957
 3 10267083 1962
 4 11537966 1967
 5 13079460 1972
 6 14880372 1977
 7 12881816 1982
 8 13867957 1987
 9 16317921 1992
10 22227415 1997
```

Use ... to pass the arguments of your function, filter_summarize(), to filter().

In summarize, get the n and mean life expectancy for the data set

Check filter_summarize() with year ==
1952.

Try filter_summarize() again for 2002, but also filter countries that have "and" in the country name. Use str_detect() from the stringr package.

```
filter_summarize <- function(...) {
   gapminder |>
   filter(...) |>
   summarize(n = n(), mean_lifeExp = mean(lifeExp))
}
```

Writing functions with dplyr, ggplot2, and friends

```
plot_hist <- function(x) {
    ggplot(gapminder, aes(x = x)) + geom_histogram()
}

plot_hist(lifeExp)</pre>
```

```
Error in `geom_histogram()`:
! Problem while computing aesthetics.
i Error occurred in the 1st layer.
Caused by error:
! object 'lifeExp' not found
```

Programming with dplyr, ggplot2, and friends

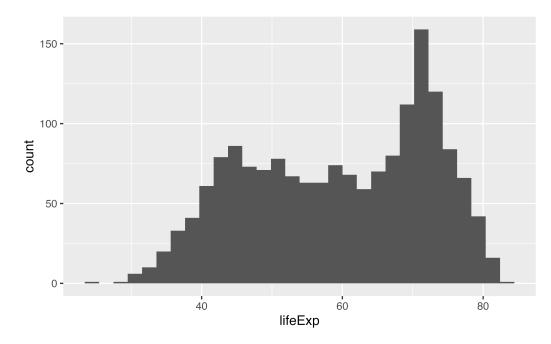
```
1 plot_hist <- function(x) {
2     ggplot(gapminder, aes(x = x)) + geom_histogram()
3 }
4     5 plot_hist("lifeExp")

Error in `geom_histogram()`:
! Problem while computing stat.
i Error occurred in the 1st layer.
Caused by error in `setup_params()`:
! `stat_bin()` requires a continuous x aesthetic.</pre>
* the x aesthetic is discrete.
```

Curly-curly: { variable }

```
plot_hist <- function(x) {
    ggplot(gapminder, aes(x = {{ x }})) + geom_histogram(
}

plot_hist(lifeExp)</pre>
```



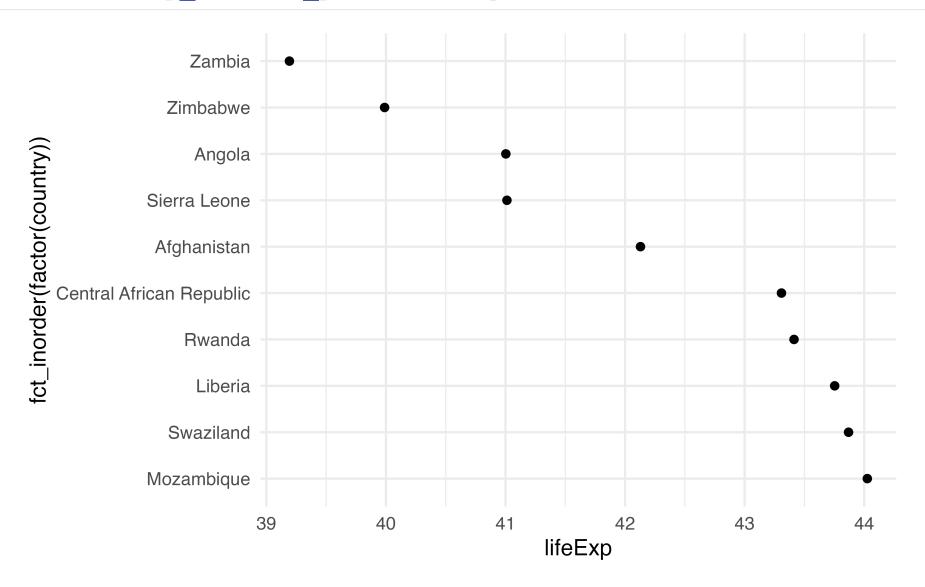
Filter gapminder by year using the value of .year (notice the period before hand!). You do NOT need curly-curly for this. (Why is that?)

Arrange it by the variable. This time, do wrap it in curly-curly!

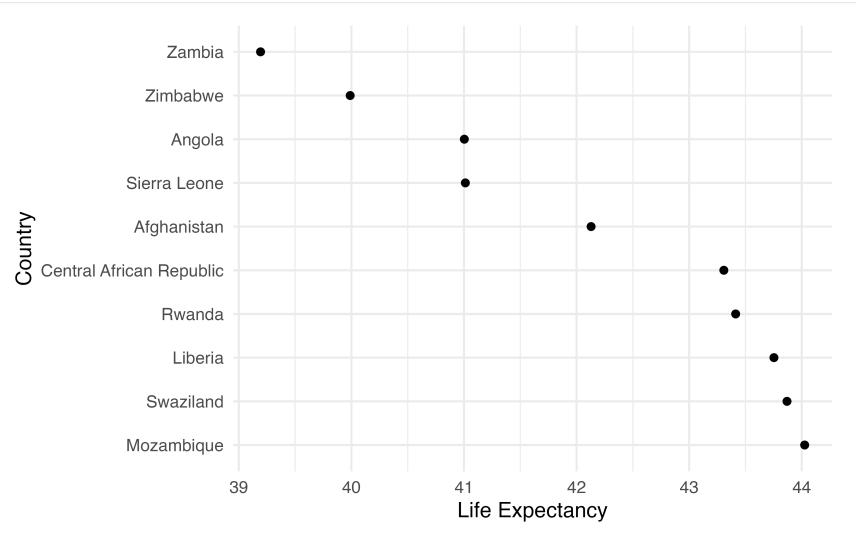
Make a scatter plot. Use variable for x. For y, we'll use country, but to keep it in the order we arranged it by, we'll turn it into a factor. Wrap the the factor() call with fct_inorder(). Check the help page if you want to know more about what this is doing.

```
1 top_scatter_plot <- function(variable, .year) {
2    gapminder |>
3        filter(year == .year) |>
4        arrange(desc({{ variable }})) |>
5        # take the 10 lowest values
6        tail(10) |>
7        ggplot(aes(x = {{ variable }}, y = fct_inorder(factors)
8        geom_point() +
9        theme_minimal()
10 }
```

1 top_scatter_plot(lifeExp, 2002)



```
1 top_scatter_plot(lifeExp, 2002) +
2 labs(x = "Life Expectancy", y = "Country")
```



Resources

R for Data Science, 2nd ed.: A comprehensive but friendly introduction to the Tidyverse. Free online.

Advanced R, 2nd ed.: Detailed guide to how R works and how to make your code better. Free online.