

# R Basics

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## Loading R packages

In your previous statistics course at Carleton, you likely loaded at least one add-on R package. In this course, we'll use a lot of tools found in the tidyverse of R packages. To load many of these packages at once, you can use the `library(<package_name>)` command. So to load the tidyverse we run:

```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr   1.1.4   v readr   2.1.5
v forcats 1.0.0   v stringr 1.5.1
v ggplot2 3.5.1   v tibble  3.2.1
v lubridate 1.9.4 v tidyr   1.3.1
v purrr   1.0.4
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()    masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

### Note

Above we see a lot of extra info printed when we load the tidyverse. These messages are just telling you what packages are now available to you and warning you that a few functions (e.g., `filter`) has been replaced by the tidyverse version. We'll see how to suppress these messages later.

## Creating and naming objects

All R statements where you create objects have the form:

```
object_name <- value
```

At first, we'll be creating a lot of data objects. For example, we can load a data set containing the ratings for each episode of The Office using the code

```
office_ratings <- readr::read_csv("https://raw.githubusercontent.com/rfordatascience/tidytuesday/master/data/2020/02/01/office_ratings.csv")
```

In this class you will be creating a lot of objects, so you'll need to come up with names for those objects. Trying to think of informative/meaningful names for objects is hard, but necessary work! Below are the fundamental rules for naming objects in R:

- names can't start with a number
- names are case-sensitive
- some common letters are used internally by R and should be avoided as variable names (c, q, t, C, D, F, T, I)
- There are reserved words that R won't let you use for variable names (for, in, while, if, else, repeat, break, next)
- R will let you use the name of a predefined function—but don't do it!

You can always check to see if the name you want to use is already taken via `exists()`:

For example `lm` exists

```
exists("lm")
```

```
[1] TRUE
```

but `carleton_college` doesn't.

```
exists("carleton_college")
```

```
[1] FALSE
```

There are also a lot of naming styles out there, and if you have coded in another language, you may have already developed a preference. Below is an illustration by Allison Horst



I generally following the [tidyverse style guide](#), so you'll see that I use only lowercase letters, numbers, and \_ (snake case).

## Overviews of data frames

Above, you loaded in a data set called `office_ratings`. Data sets are stored as a special data structure called a **data frame**. Data frames are the most-commonly used data structure for data analysis in R. For now, think of them like spreadsheets.

Once you have your data frame, you can get a quick overview of it using a few commands (below I use `data_set` as a generic placeholder for the data frame's name):

Command	Description
<code>head(data_set)</code>	print the first 6 rows
<code>tail(data_set)</code>	print the last 6 rows

Command	Description
<code>glimpse(data_set)</code>	a quick overview where columns run down the screen and the data values run across. This allows you to see every column in the data frame.
<code>str(data_set)</code>	a quick overview like <code>glimpse()</code> , but without some of the formatting
<code>summary(data_set)</code>	quick summary statistics for each column
<code>dim(data_set)</code>	the number of rows and columns
<code>nrow(data_set)</code>	the number of rows
<code>ncol(data_set)</code>	the number of columns

## Tibbles

A **tibble**, or a `tbl_df` is another version of a data frame which is used by default in a lot of the tidyverse packages that we'll use.

Tibbles are `data.frames` that are lazy and surly: they do less (i.e. they don't change variable names or types, and don't do partial matching) and complain more (e.g. when a variable does not exist). This forces you to confront problems earlier, typically leading to cleaner, more expressive code. Tibbles also have an enhanced `print()` method which makes them easier to use with large datasets containing complex objects.

### Check point

Run the above commands on the `office_ratings` data set. Compare and contrast the information returned by each command.

### Getting a spreadsheet

In RStudio, you can run the command `View(data_set)` to pull up a spreadsheet representation of a data frame. You can also click on the name of the data frame in the Environment pane. This can be a great way help you think about the data, and even has some interactive functions (e.g., filtering and searching); however, **never include `View(data_set)` in an .Rmd file!!**

### Review from intro stats

In intro stats we used the terms **cases** (or **observations**) and **variables** to describe the rows and columns of a data frame, respectively.

## Extracting pieces of data frames

Since data frames are the fundamental data structure for most analyses in R, it's important to know how to work with them. You already know how to get an overview of a data frame, but that isn't always very informative. Often, you want to extract pieces of a data frame, such as a specific column or row.

### Extracting rows

Data frames can be indexed by their row/column numbers. To extract elements of a data frame, the basic syntax is `data_set[row.index, column.index]`. So, to extract the 10th row of `office_ratings` we run

```
office_ratings[10, ]
```

```
# A tibble: 1 x 6
  season episode title      imdb_rating total_votes air_date
  <dbl>  <dbl> <chr>         <dbl>      <dbl> <date>
1     2     4 The Fire      8.4        2713 2005-10-11
```

Notice that to extract an entire row, we leave the column index position blank.

We can also extract multiple rows by creating a vector of row indices. For example, we can extract the first 5 rows via

```
office_ratings[1:5, ]
```

```
# A tibble: 5 x 6
  season episode title      imdb_rating total_votes air_date
  <dbl>  <dbl> <chr>         <dbl>      <dbl> <date>
1     1     1 Pilot       7.6        3706 2005-03-24
2     1     2 Diversity Day 8.3        3566 2005-03-29
3     1     3 Health Care   7.9        2983 2005-04-05
4     1     4 The Alliance   8.1        2886 2005-04-12
5     1     5 Basketball   8.4        3179 2005-04-19
```

Here, `1:5` create a sequence of integers from 1 to 5.

We could also specify arbitrary row index values by combining the values into a vector. For example, we could extract the 1st, 13th, 64th, and 128th rows via

```
office_ratings[c(1, 13, 64, 128), ]
```

```
# A tibble: 4 x 6
```

	season	episode	title	imdb_rating	total_votes	air_date
	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<date>
1	1	1	Pilot	7.6	3706	2005-03-24
2	2	7	The Client	8.6	2631	2005-11-08
3	4	13	Job Fair	7.9	1977	2008-05-08
4	7	11	Classy Christmas	8.9	2138	2010-12-09

## Extracting columns

Similar to extracting rows, we can use a numeric index to extract the columns of a data frame. For example, to extract the 3rd column, we can run

```
office_ratings[,3]
```

```
# A tibble: 188 x 1
```

```
title
<chr>
1 Pilot
2 Diversity Day
3 Health Care
4 The Alliance
5 Basketball
6 Hot Girl
7 The Dundies
8 Sexual Harassment
9 Office Olympics
10 The Fire
# i 178 more rows
```

Alternatively, we can pass in the column name in quotes instead of the column number

```
office_ratings[, "title"]
```

```
# A tibble: 188 x 1
```

```
title
<chr>
```

```
1 Pilot
2 Diversity Day
3 Health Care
4 The Alliance
5 Basketball
6 Hot Girl
7 The Dundies
8 Sexual Harassment
9 Office Olympics
10 The Fire
# i 178 more rows
```

Notice that the extracted column is still formatted as a data frame (or tibble). If you want to extract the contents of the column and just have a vector of titles, you have a few options.

- You could use double brackets with the column number:

```
office_ratings[[3]]
```

- You could use double brackets with the column name in quotes:

```
office_ratings[["title"]]
```

- You could use the \$ extractor with the column name (not in quotes):

```
office_ratings$title
```

☒ **Check point**

1. Extract the 35th row of `office_ratings`.
2. Extract rows 35, 36, 37, and 38 of `office_ratings`.
3. Extract the `imdb_rating` column from office ratings using the column index number.
4. Extract the `imdb_rating` column from office ratings using the column name.

## Lists

It turns out that data frames are special cases of **lists**, a more general data structure. In a data frame, each column is an element of the data list and each column must be of the same length. In general, lists can be comprised of elements of vastly different lengths and data types.

As an example, let's construct a list of the faculty in the MAST department and what is being taught this winter.

```
stat_faculty <- c("Kelling", "Loy", "Luby", "Poppick", "St. Clair", "Wadsworth")
stat_courses <- c(120, 220, 230, 250, 285, 330)
math_faculty <- c("Brooke", "Davis", "Egge", "Gomez-Gonzales", "Haunsperger", "Johnson",
                  "Meyer", "Montee", "Shrestha", "Terry", "Thompson", "Turnage-Butterbaugh")
math_courses <- c(101, 106, 111, 120, 210, 211, 232, 236, 240, 241, 251, 321, 333, 395)

mast <- list(stat_faculty = stat_faculty, stat_courses = stat_courses,
             math_faculty = math_faculty, math_courses = math_courses)
```

## Overview of a list

You can get an overview of a list a few ways:

- `glimpse(list_name)` and `str(list_name)` list the elements of the list and the first few entries of each element.

```
glimpse(mast)
```

List of 4

```
$ stat_faculty: chr [1:6] "Kelling" "Loy" "Luby" "Poppick" ...
$ stat_courses: num [1:6] 120 220 230 250 285 330
$ math_faculty: chr [1:12] "Brooke" "Davis" "Egge" "Gomez-Gonzales" ...
$ math_courses: num [1:14] 101 106 111 120 210 211 232 236 240 241 ...
```

- `length(list_name)` will tell you how many elements are in the list

```
length(mast)
```

```
[1] 4
```



## Extracting elements of a list

Since data frames are lists, you've already seen how to extract elements of a list. For example, to extract the `stat_faculty` you could run

```
mast[[1]]
```

```
[1] "Kelling" "Loy"     "Luby"    "Poppick" "St. Clair" "Wadsworth"
```

or

```
mast[["stat_faculty"]]
```

```
[1] "Kelling" "Loy"     "Luby"    "Poppick" "St. Clair" "Wadsworth"
```

### Note

If you had only used a single bracket above, the returned object would still be a list, which is typically not what we would want.

```
mast[1]
```

```
$stat_faculty
```

```
[1] "Kelling" "Loy"     "Luby"    "Poppick" "St. Clair" "Wadsworth"
```

### ☒ Check point

Extract the statistics courses offered this term.

## Vectors

The columns of the `office_ratings` data frame and the elements of the `mast` list were comprised of **(atomic) vectors**. Unlike lists, all elements within a vector share the same type. For example, all names in the `stat_faculty` vector were character strings and all ratings in the `imdb_rating` column were numeric. We'll deal with a variety of types of vectors in this course, including:

- numeric
- character (text)
- logical (TRUE/FALSE)

## Extracting elements of a vector

Just like with lists (and therefore data frames), we use brackets to extract elements from a vector. As an example, let's work with the title column from `office_ratings`.

```
title <- office_ratings$title # vector of titles
```

To extract the 111th title, we run

```
title[111]
```

```
[1] "New Leads"
```

or two extract the 100th through 111th titles, we run

```
title[100:111]
```

```
[1] "Double Date"      "Murder"           "Shareholder Meeting"  
[4] "Scott's Tots"     "Secret Santa"      "The Banker"  
[7] "Sabre"           "Manager and Salesman" "The Delivery: Part 1"  
[10] "The Delivery: Part 2" "St. Patrick's Day" "New Leads"
```

## Negative indices

Sometimes, we want to “kick out” elements of our vector. To do this, we can use a negative index value. For example,

```
title[-1]
```

returns all but the first title—that is, it kicks out the first title. To kick out multiple elements, we need to negate a vector of indices. For example, below we kick out the first 10 titles

```
title[-c(1:10)]
```

And now we kick out the 5th, 50th, and 150th titles

```
title[-c(5, 50, 150)]
```

This idea can be adapted to lists and data frames. For example, to kick out the first row of `office_ratings`, we run

```
office_ratings[-1,]
```

```
# A tibble: 187 x 6
```

	season	episode	title	imdb_rating	total_votes	air_date
	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<date>
1	1	2	Diversity Day	8.3	3566	2005-03-29
2	1	3	Health Care	7.9	2983	2005-04-05
3	1	4	The Alliance	8.1	2886	2005-04-12
4	1	5	Basketball	8.4	3179	2005-04-19
5	1	6	Hot Girl	7.8	2852	2005-04-26
6	2	1	The Dundies	8.7	3213	2005-09-20
7	2	2	Sexual Harassment	8.2	2736	2005-09-27
8	2	3	Office Olympics	8.4	2742	2005-10-04
9	2	4	The Fire	8.4	2713	2005-10-11
10	2	5	Halloween	8.2	2561	2005-10-18

```
# i 177 more rows
```

or to kick out the math courses from the mast list we run

```
mast[-4]
```

```
$stat_faculty
```

```
[1] "Kelling" "Loy" "Luby" "Poppick" "St. Clair" "Wadsworth"
```

```
$stat_courses
```

```
[1] 120 220 230 250 285 330
```

```
$math_faculty
```

```
[1] "Brooke" "Davis" "Egge"
[4] "Gomez-Gonzales" "Haunsperger" "Johnson"
[7] "Meyer" "Montee" "Shrestha"
[10] "Terry" "Thompson" "Turnage-Butterbaugh"
```

## Logical indices

It's great to be able to extract (or omit) elements using indices, but sometimes we don't know what index value we should use. For example, if you wanted to extract all of the 300-level statistics courses from the `stat_courses` vector, you would need to manually determine that positions 2:5 meet that requirement. That's a lot of work! A better alternative is to allow

R to find the elements meeting that requirement using **logical operators**. Below is a table summarizing common logical operators in R.

Comparison	Meaning
<	less than
>	greater than
<=	less than or equal to
>=	greater than or equal to
==	is equal to
!=	not equal to

In order to extract the 300-level statistics courses, we'll take two steps:

1. We'll determine whether each course is numbered at least 300,
2. then we'll use that sequence of TRUEs/FALSEs to extract the course.

So, first we use the logical operator `>=` to compare `stat_courses` and 300. This returns TRUE if the element meets the specification and FALSE otherwise.

```
stat_courses >= 300
```

```
[1] FALSE FALSE FALSE FALSE FALSE TRUE
```

Now, we can use this vector as our index. Only the TRUE elements will be extracted:

```
stat_courses[stat_courses >= 300]
```

```
[1] 330
```

The same idea can be used with data frames and lists, just remember how to format the brackets and indices!

☒ **Check point**

1. Extract all statistics courses below 250 from `stat_courses`.
2. Extract all math courses except for 240 (probability) from `math_courses`.
3. Extract all rows from season 3 of The Office.