Working with multiple plates

Mike Blazanin

Contents

Where are we so far?	1
A brief primer on lists in R	2
Importing data for multiple plates	3
Reading files for multiple block-shaped plates	3
Separated block-shaped plate files	3
Interleaved block-shaped plate files	4
Reading files for multiple wide-shaped plates	4
Transforming multiple plates to tidy-shaped	4
Merging designs with multiple plates	5
Plates with different designs	5
Plates with a shared design	6

Where are we so far?

- 1. Introduction: vignette("gc01_gcplyr")
- 2. Importing and transforming data: vignette("gc02_import_reshape")
- 3. Incorporating experimental designs: vignette("gc03_incorporate_designs")
- 4. Pre-processing and plotting your data: vignette("gc04_preprocess_plot")
- 5. Processing your data: vignette("gc05_process")
- 6. Analyzing your data: vignette("gc06_analyze")
- 7. Dealing with noise: vignette("gc07_noise")
- 8. Statistics, merging other data, and other resources: vignette("gc08_conclusion")
- 9. Working with multiple plates: vignette("gc09_multiple_plates")

So far, we've covered everything you can do with gcplyr with a single plate. In this vignette, I'll be briefly covering how gcplyr can also easily handle multiple plates of data at the same time. I'll be assuming that you've already read the analogous sections on working with a single plate at a time, so if you haven't, now is a good time to go back and do that!

If you haven't already, load the necessary packages.

```
library(gcplyr)
library(dplyr)
library(ggplot2)
```

A brief primer on lists in R

lists are a powerful data format in R, because they can contain any set of R objects. However, this can also make them a little more difficult to navigate than simpler formats like vectors or data.frames.

Let's make a simple example to explore how lists can be navigated.

```
mylist <- list("A", c(5, 6, 7), matrix(1:9, nrow = 3))
mylist
#> [[1]]
#> [1] "A"
#>
#>
#> [[2]]
#> [1] 5 6 7
#>
#> [[3]]
#> [,1] [,2] [,3]
#> [1,] 1 4 7
#> [2,] 2 5 8
#> [3,] 3 6 9
```

We've created a list with 3 elements:

- the first element is a vector with only one entry, "A"
- the second element is a vector with three entries
- the third element is a matrix with the numbers 1 through 9

If we want to extract the contents of any of these elements, we simply use double brackets [[

```
mylist[[1]]
#> [1] "A"
mylist[[2]]
#> [1] 5 6 7
mylist[[3]]
#> [,1] [,2] [,3]
#> [1,] 1 4 7
#> [2,] 2 5 8
#> [3,] 3 6 9
```

If we want to modify the contents of any of these elements, we can also use double brackets

When working with multiple plates, you'll often be dealing with lists of data.frames, rather than the single data.frames we've been working with before. With multiple plates, if you want to carry out an operation on only one of the plate data.frames, simply use double brackets to select it specifically. If you want to carry out an operation on all of the data.frames, gcplyr functions like trans_wide_to_tidy and merge_dfs will be able to handle the list automatically.

Importing data for multiple plates

If your data is block-shaped: use import_blockmeasures and start in the next section: Reading files for multiple block-shaped plates

If your data is wide-shaped: use read_wides and skip down to the Reading files for multiple wide-shaped plates section

Reading files for multiple block-shaped plates

If you're reading files for multiple block-shaped plates, your approach depends on how the files themselves are arranged: * Each plate's worth of block-shaped files is easily separable (e.g. saved in different folders) * The files from separate plates are interleaved such that they are alternating (as is common when multiple plates are read by a plate-reader robot in a single run)

Separated block-shaped plate files

In the first case, you will have to read each plate of files as an individual plate following the typical process with import_blockmeasures.

Let's generate some example files to see how this works. When working with real growth curve data, these files would be output by the plate reader.

```
filenames_sep <- make_example(vignette = 9, example = 1, dir = "./example_data")
#> Files have been written
head(filenames_sep)
#> [1] "./example_data/Plate1-0_00_00.csv" "./example_data/Plate1-0_15_00.csv"
#> [3] "./example_data/Plate1-0_30_00.csv" "./example_data/Plate1-0_45_00.csv"
#> [5] "./example_data/Plate1-1_00_00.csv" "./example_data/Plate1-1_15_00.csv"
```

As we can see, the names for the files have either "Plate1" or "Plate2" in them. We can use that to simply import them separately. If you would like to later work with them as a unit, you can save them into a list.

```
plates <- list(
  plate1 = import_blockmeasures(
    list.files(path = "./example_data/", pattern = "Plate1", full.names = TRUE),
    startrow = 4,
    metadata = list(Time = c(2, "C"))),
  plate2 = import_blockmeasures(
    list.files(path = "./example_data/", pattern = "Plate2", full.names = TRUE),
    startrow = 4,
    metadata = list(Time = c(2, "C"))))</pre>
```

Interleaved block-shaped plate files

In the second case, files from different plates are interleaved together. You can read and separate them with import_blockmeasures.

Let's generate some example files to see how this works. When working with real growth curve data, these files would be output by the plate reader.

```
filenames_mixed <- make_example(vignette = 9, example = 2, dir = "./example_data")

#> Files have been written

head(filenames_mixed)

#> [1] "./example_data/00_00_00.csv" "./example_data/00_00_01.csv"

#> [3] "./example_data/00_15_00.csv" "./example_data/00_15_01.csv"

#> [5] "./example_data/00_30_00.csv" "./example_data/00_30_01.csv"
```

In this scenario, there are two plates: the first plate was read every 15 minutes, and the second plate was read 1 second after each time the first plate was read. To read them and separate them out, we set the num_plates argument to the number of plates.

```
plates <- import_blockmeasures(
    filenames_mixed,
    startrow = 4,
    metadata = list(Time = c(2, "C")),
    num_plates = 2)</pre>
```

The output is a list, where the first element in the list is a wide-shaped data.frame for plate #1, and the second element is a wide-shaped data.frame for plate #2.

Reading files for multiple wide-shaped plates

You can also read multiple wide-shaped datasets. Whether they're from a single file, or from multiple files, simply input the corresponding information and read_wides will return a list containing each wide-shaped data.frame

Transforming multiple plates to tidy-shaped

Once you have your data in a list of wide-shaped data.frames, it's easy to transform it to tidy-shaped. Assuming all the plates have the same columns, you can simply put the list into trans_wide_to_tidy.

The output is a list, where each element of the list is a tidy-shaped data.frame corresponding to one of the plates.

If you don't have any designs to merge, you can collapse this list into a single data.frame with merge_dfs:

```
tidy_plates_collapsed <- merge_dfs(tidy_plates, collapse = TRUE)
#> Joining, by = c("file", "Experiment", "Start_date", "Time", "Well", "Measurements")

print_df(head(tidy_plates_collapsed), col.names = TRUE)
#> file Experiment Start_date Time Well Measurements
#> widedata Experiment_1 2024-01-29 0 A1 0.002
#> widedata Experiment_1 2024-01-29 0 B1 0.002
#> widedata Experiment_1 2024-01-29 0 C1 0.002
#> widedata Experiment_1 2024-01-29 0 D1 0.002
#> widedata Experiment_1 2024-01-29 0 E1 0.002
#> widedata Experiment_1 2024-01-29 0 F1 0.002
#> widedata Experiment_1 2024-01-29 0 F1 0.002
```

Merging designs with multiple plates

Plates with different designs

If you have different designs for each of your plates, you'll have to merge them with each plate separately using brackets:

```
example_design1 <- make_design(</pre>
  pattern_split = ",", nrows = 8, ncols = 12,
  "Bacteria_strain" = make_designpattern(
    values = paste("Strain", 1:48),
    rows = 1:8, cols = 1:6, pattern = 1:48, byrow = TRUE),
  "Bacteria strain" = make designpattern(
    values = paste("Strain", 1:48),
    rows = 1:8, cols = 7:12, pattern = 1:48, byrow = TRUE))
example_design2 <- make_design(</pre>
  pattern split = ",", nrows = 8, ncols = 12,
  "Bacteria_strain" = make_designpattern(
    values = paste("Strain", 49:96),
    rows = 1:8, cols = 1:6, pattern = 1:48, byrow = TRUE),
  "Bacteria_strain" = make_designpattern(
    values = paste("Strain", 49:96),
    rows = 1:8, cols = 7:12, pattern = 1:48, byrow = TRUE))
tidy_plates[[1]] <- merge_dfs(tidy_plates[[1]], example_design1)</pre>
#> Joining, by = "Well"
tidy_plates[[2]] <- merge_dfs(tidy_plates[[2]], example_design2)</pre>
#> Joining, by = "Well"
```

Then afterwards, you can collapse your data together:

```
data_and_designs <- merge_dfs(tidy_plates, collapse = TRUE)

#> Joining, by = c("file", "Experiment", "Start_date", "Time", "Well", "Measurements",

#> "Bacteria_strain")

print_df(head(data_and_designs))

#> widedata Experiment_1 2024-01-29 0 A1 0.002 Strain 1

#> widedata Experiment_1 2024-01-29 0 B1 0.002 Strain 7

#> widedata Experiment_1 2024-01-29 0 C1 0.002 Strain 13

#> widedata Experiment_1 2024-01-29 0 D1 0.002 Strain 19

#> widedata Experiment_1 2024-01-29 0 E1 0.002 Strain 25

#> widedata Experiment_1 2024-01-29 0 F1 0.002 Strain 31
```

Plates with a shared design

If, instead, all your plates have the same design, you can merge the design with all of them at once by collapsing your list of data in the merge step. Don't worry about losing track of your plates, the file column that is automatically generated should differentiate the separate plates you read in.

```
example_design <- make_design(</pre>
  pattern_split = ",", nrows = 8, ncols = 12,
  "Bacteria_strain" = make_designpattern(
   values = paste("Strain", 1:48),
   rows = 1:8, cols = 1:6, pattern = 1:48, byrow = TRUE),
  "Bacteria strain" = make designpattern(
   values = paste("Strain", 1:48),
   rows = 1:8, cols = 7:12, pattern = 1:48, byrow = TRUE))
data_and_designs <- merge_dfs(tidy_plates, example_design, collapse = TRUE)
\# Joining, by = c("file", "Experiment", "Start_date", "Time", "Well", "Measurements",
#> "Bacteria_strain")
#> Joining, by = c("Well", "Bacteria_strain")
print_df(head(data_and_designs))
#> widedata Experiment_1 2024-01-29 0 A1 0.002 Strain 1
#> widedata Experiment_1 2024-01-29 0 B1 0.002 Strain 7
#> widedata Experiment_1 2024-01-29 0 C1 0.002 Strain 13
#> widedata Experiment_1 2024-01-29 0 D1 0.002 Strain 19
#> widedata Experiment_1 2024-01-29 0 E1 0.002 Strain 25
#> widedata Experiment_1 2024-01-29 0 F1 0.002 Strain 31
```