



# Tidy data reshaping & summaries

Athanasia Monika Mowinckel

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LCBC  
LIFESPAN CHANGES  
*in brain and cognition*

# Part 2

## Tidy data reshaping & summaries



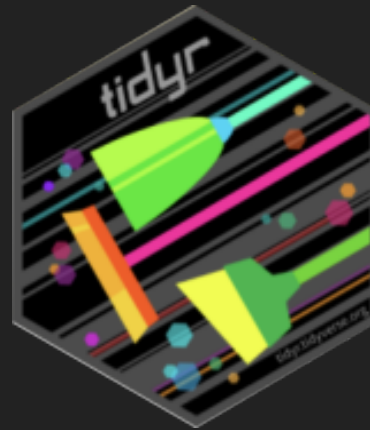
# Tidy data reshaping & summaries

- pivoting data with `tidyr` (~25 min)
- grouped summaries with `dplyr` (~25 min)
- working with nested data using `purrr` (~25 min)

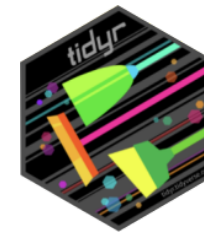
# tidyr

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pivoting / altering data shape



# tidyr



The goal of tidyr is to help you create tidy data.

Tidy data is data where:

- Every column is variable.
- Every row is an observation.
- Every cell is a single value.

Tidy data describes a standard way of storing data that is used wherever possible throughout the tidyverse. If you ensure that your data is tidy, you'll spend less time fighting with the tools and more time working on your analysis. Learn more about tidy data in `vignette("tidy-data")`.



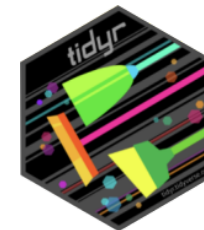
## Tall/long vs. wide data

- Tall (or long) data are considered "tidy", in that they adhere to the three tidy-data principles
- Wide data are not necessarily "messy", but have a shape less ideal for easy handling in the tidyverse

Example in longitudinal data design:

- wide data: each participant has a single row of data, with all longitudinal observations in separate columns
- tall data: a participant has as many rows as longitudinal time points, with measures in separate columns

# tidyr

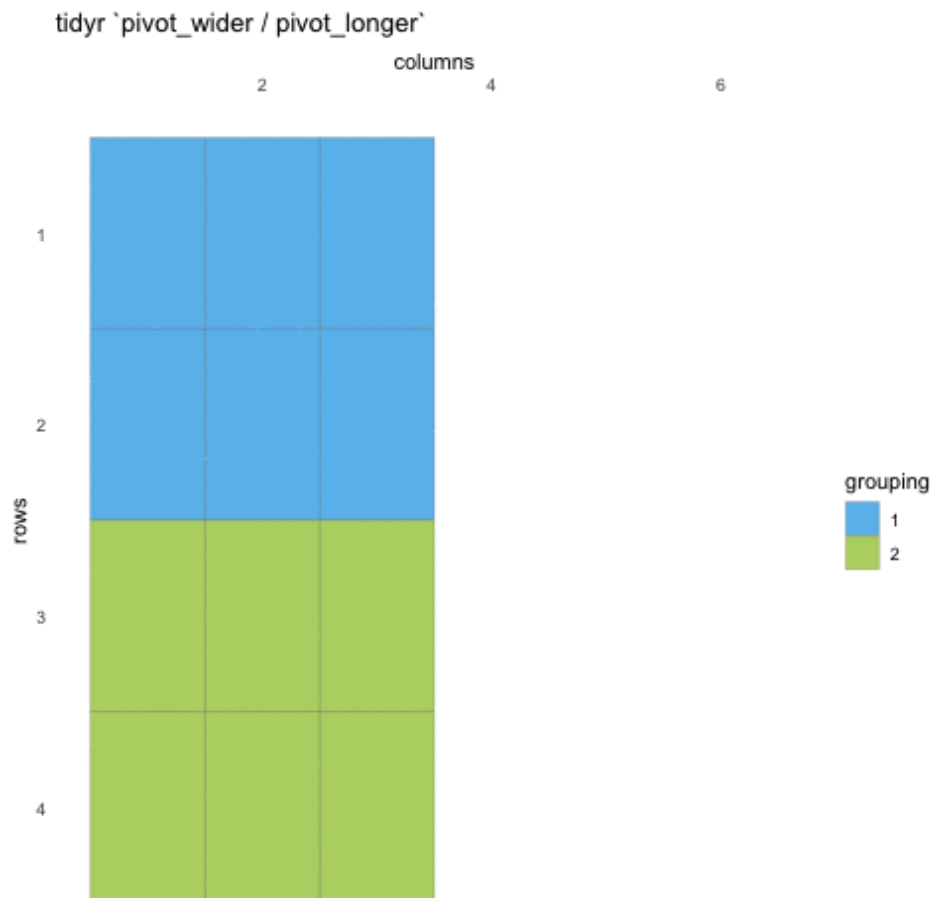


## pivoting

`pivot_longer()` - wide to long

`pivot_wider()` - long to wide

Transforms data shape





# Pivoting longer

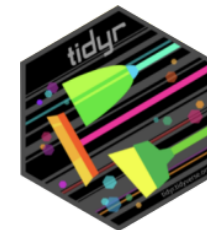
takes tidy-select column arguments, so it is easy to grab all the columns you are after.

```
penguins %>%  
  pivot_longer(contains("_"))
```

```
## # A tibble: 1,376 x 6  
##   species island    sex    year name          value  
##   <fct>   <fct>    <fct> <int> <chr>         <dbl>  
## 1 Adelie  Torgersen male    2007 bill_length_mm    39.1  
## 2 Adelie  Torgersen male    2007 bill_depth_mm    18.7  
## 3 Adelie  Torgersen male    2007 flipper_length_mm  181  
## 4 Adelie  Torgersen male    2007 body_mass_g      3750  
## 5 Adelie  Torgersen female  2007 bill_length_mm    39.5  
## 6 Adelie  Torgersen female  2007 bill_depth_mm    17.4  
## 7 Adelie  Torgersen female  2007 flipper_length_mm  186  
## 8 Adelie  Torgersen female  2007 body_mass_g      3800  
## 9 Adelie  Torgersen female  2007 bill_length_mm    40.3  
## 10 Adelie Torgersen female  2007 bill_depth_mm     18  
## # ... with 1,366 more rows
```

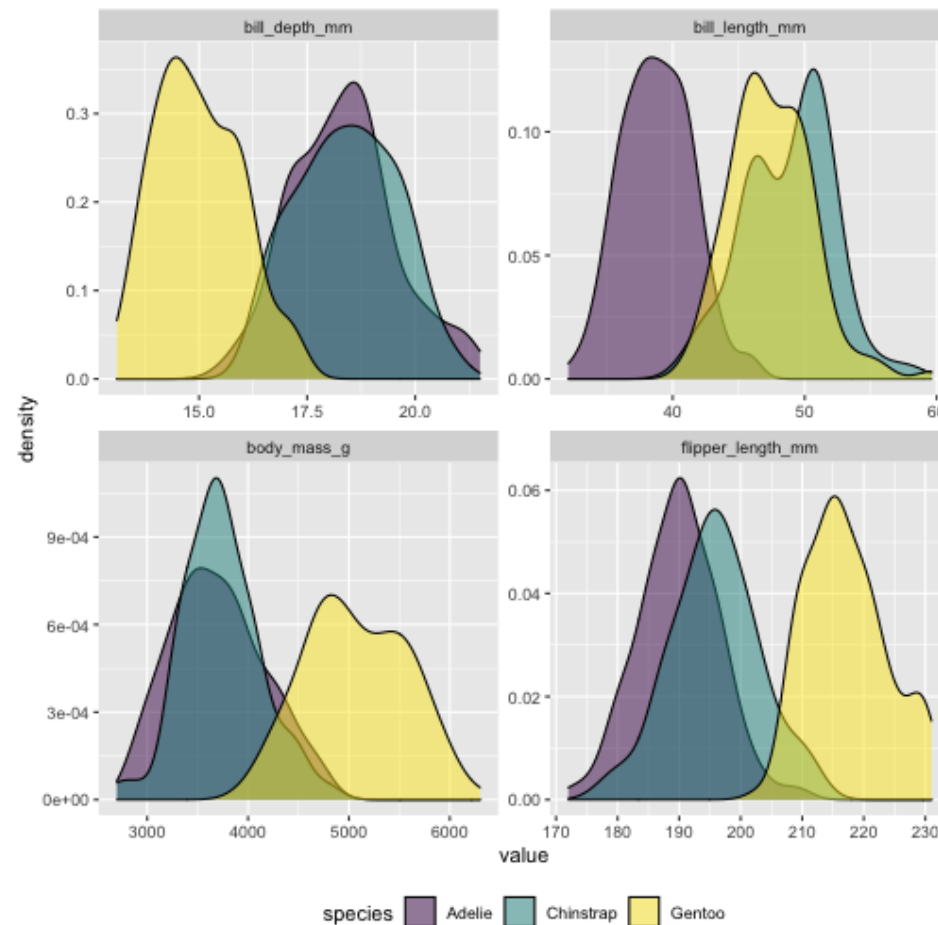


# Why pivot longer?

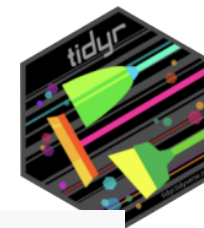


Can be convenient for easy sub-plots with ggplot

```
penguins %>%  
  pivot_longer(contains("_")) %>%  
  
  ggplot(aes(x = value, fill = species)) +  
  geom_density() +  
  facet_wrap(~ name, scales = "free") +  
  scale_fill_viridis_d(alpha = .5) +  
  theme(legend.position = "bottom")
```



# pivoting wider



```
penguins_long <- penguins %>%
  mutate(id = row_number()) %>%
  pivot_longer(contains("_"),
               names_to = c("body_part",
                           "measure",
                           "unit"),
               names_sep = "_")

penguins_long %>%
  pivot_wider(names_from = c("body_part", "measure", "unit"), # pivot these columns
              values_from = "value", # take the values from here
              names_sep = "_") # separate names_from with this character
```

```
## # A tibble: 344 x 9
##   species island sex    year    id bill_length_mm bill_depth_mm
##   <fct>    <fct> <fct> <int> <int>      <dbl>          <dbl>
## 1 Adelie  Torge... male   2007     1        39.1           18.7
## 2 Adelie  Torge... fema... 2007     2        39.5           17.4
## 3 Adelie  Torge... fema... 2007     3        40.3            18
## 4 Adelie  Torge... <NA>    2007     4         NA           NA
## 5 Adelie  Torge... fema... 2007     5        36.7           19.3
## 6 Adelie  Torge... male   2007     6        39.3           20.6
## 7 Adelie  Torge... fema... 2007     7        38.9           17.8
## 8 Adelie  Torge... male   2007     8        39.2           19.6
```

# Go to RStudio

live demo

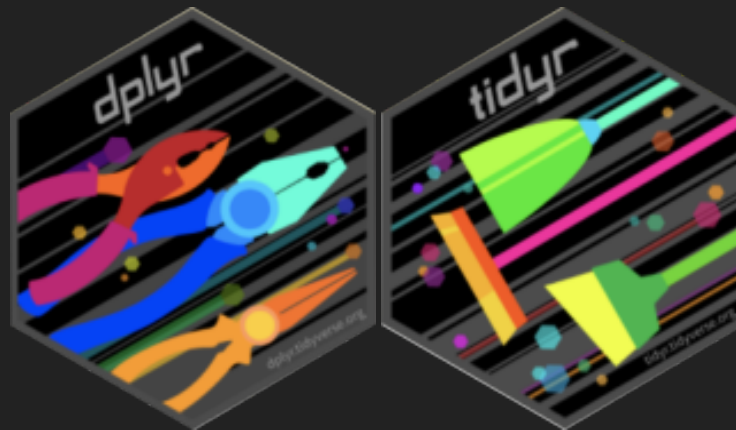
Go to subsetting exercises

```
learnr::run_tutorial("005-pivoting",  
  "tidyquintro")
```

08:00

# dplyr + tidyr

## data summaries





# dplyr - comparison to base-R

## tidy

```
penguins %>%  
  summarise(mean(bill_length_mm, na.rm = TRUE))
```

## base

```
mean(penguins$bill_length_mm, na.rm = TRUE)
```

<https://dplyr.tidyverse.org/articles/base.html>

# Go to RStudio

live demo

Go to subsetting exercises

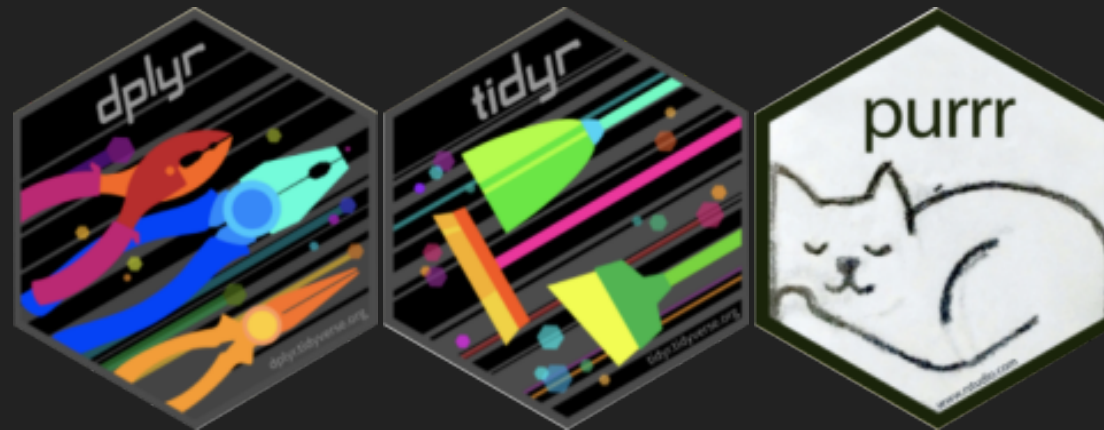
```
learnr::run_tutorial("006-summarising",  
  "tidyquintro")
```

08:00



# dplyr + tidyr + purrr

Working with nested data - avoiding loops





# comparison to base-R

## tidy

```
penguins %>%  
  nest_by(species, island) %>%  
  mutate(lm_model = list(  
    lm(bill_length_mm ~ bill_depth_mm, data = data)  
  ))
```

## base

```
penguins$groups <- interaction(penguins$species, penguins$island)  
  
models <- list()  
for(i in 1:length(unique(penguins$groups))){  
  tmp <- penguins[penguins$groups == groups[i],]  
  models[[i]] <- lm(bill_length_mm ~ bill_depth_mm, data = data)  
}  
  
# or  
lapply(unique(penguins$groups), function(x) lm(bill_length_mm ~ bill_depth_mm,  
  data = penguins[penguins$groups == x,]))
```

# Go to RStudio

live demo

Go to subsetting exercises

```
learnr::run_tutorial("006-nesting",  
  "tidyquintro")
```

08:00

# End of part 2

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