

# Datenvisualisierung 4

## Multi-part plots and customisations

Daniela Palleschi

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### Lesungen

For further reading and practice on this topic, I suggest reading [Section 11.5 \(Communication: Themes\)](#) in Wickham et al. (2023), and [Chapter 4 \(Representing summary statistics\)](#) in Nordmann et al. (2022).

### Learning objectives

In this section we will learn to

- build multi-part plots
- adjust the position of geoms
- customise our plots for better data communicate

## Set-up

### Packages

Today, we're loading our relevant **tidyverse** packages directly: **dplyr** and **ggplot**. These are the only To aid us in loading in our data, we're also loading the **here** package, and the **janitor** package which is useful for tidying up our data (e.g., the `clean_names()` function). To customise our plots we're also using the **ggthemes** and **patchwork** packages. The former helps us produce plots that are colour-blind friendly, while the latter allows us to print multiple plots together. We also need a new package: **gghalves**.

```
pacman::p_load(tidyverse,
               here,
               janitor,
               ggthemes,
               patchwork,
               gghalves
               )
```

### Data

We're again working with our slightly altered version of the **english** dataset from the **languageR** package.

```
df_eng <- read_csv(
  here(
    "daten",
    "languageR_english.csv"
  )
) |>
clean_names() |>
rename(
  rt_lexdec = r_tlexdec,
  rt_naming = r_tnaming
)
```

# 1 Dodged density plots

We can produce density plots mapped along a categorical variable by using `geom_half_violin()` from the `gghalves` package.

```
df_eng %>%  
  ggplot() +  
  aes(x = age_subject, y = rt_lexdec) +  
  geom_half_violin(alpha = .8)
```

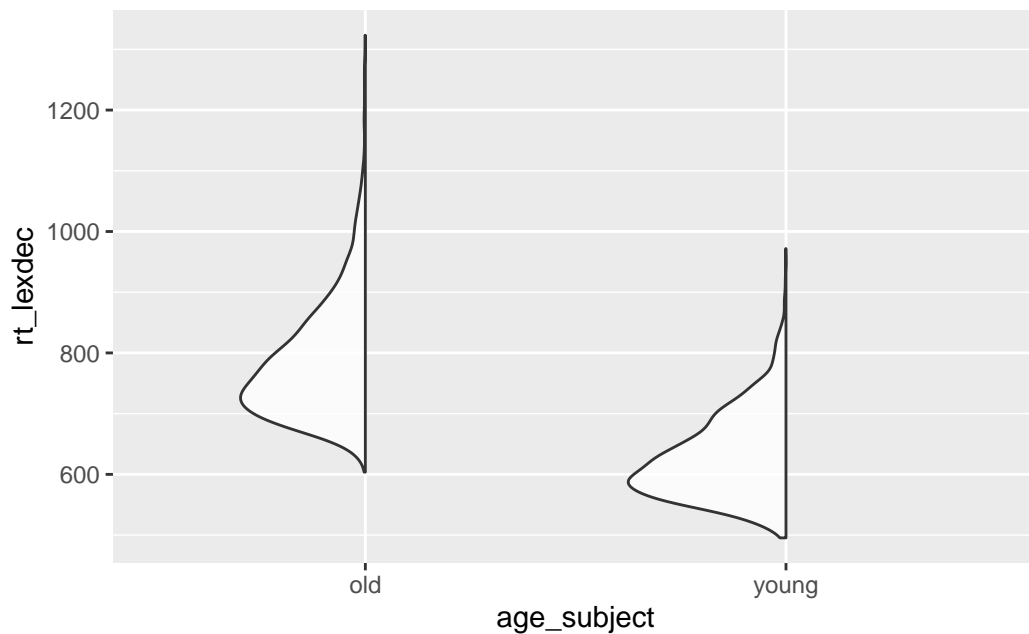


Abbildung 1: Dodged density plots with `gghalves::geom_half_violin()`

## 1.1 Adding a boxplot

We can also add another geom to add more information to the plot. Let's add a boxplot.

```
df_eng %>%  
  ggplot() +  
  aes(x = age_subject, y = rt_lexdec) +  
  geom_half_violin(alpha = .8) +  
  geom_boxplot()
```

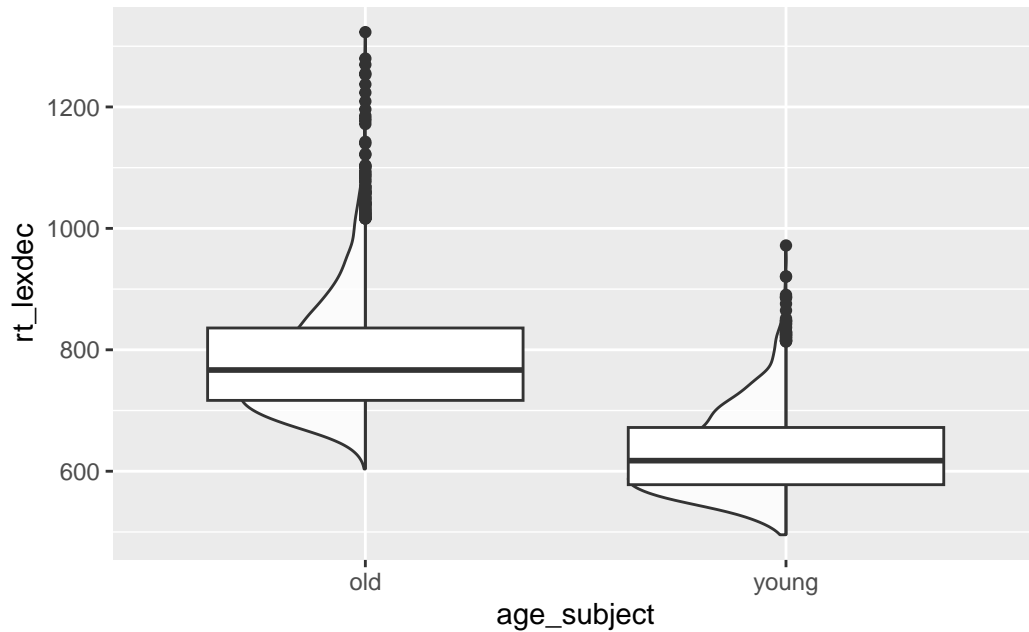


Abbildung 2: Boxplot on top of our dodged density plot

## 1.2 position\_nudge()

Maybe we want to move the boxplot so that it's not overtop of the density plots, and so that it's not quite as wide. We can do this by setting `position` to `position_nudge()`, and `width` to some value smaller than `.75`, which is the default width.

```
df_eng %>%
  ggplot() +
  aes(x = age_subject, y = rt_lexdec) +
  geom_half_violin(alpha = .8) +
  geom_boxplot(width = .3, # make less wide
              position = position_nudge(x=0.2)
              )
```

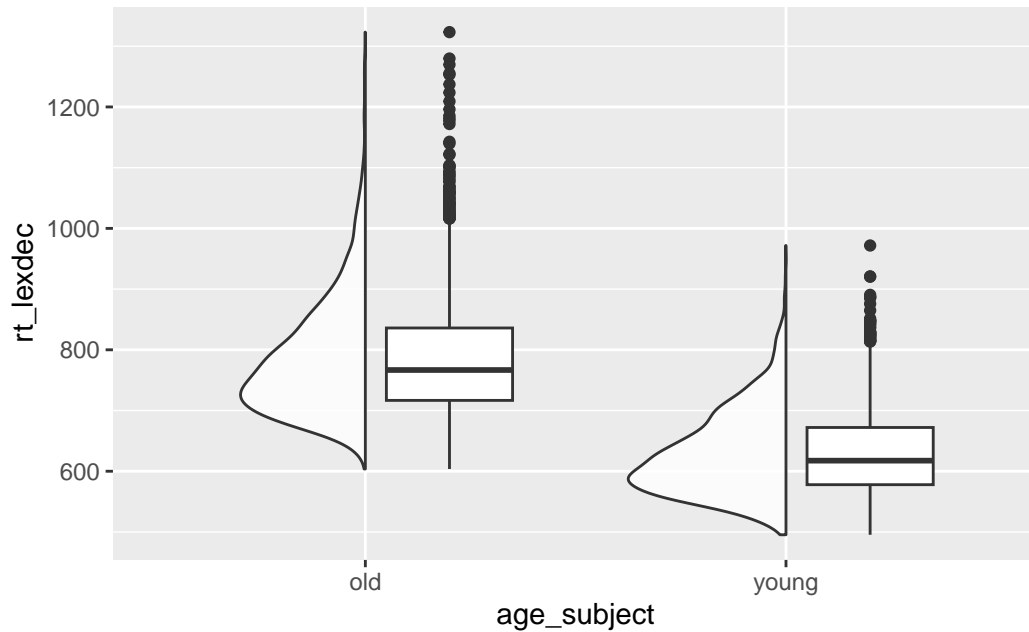


Abbildung 3: Boxplot on top of our dodged density plot

### 1.3 position\_jitter() for scatterplots

This is from a family of options that allow us to alter the position of geoms. For example, Abbildung 4 A and B both show the exact same data, but Abbildung 4 B includes `position = position_jitter(0.2)` to move overlapping points. This way we get a good idea of how many observations there were across reaction times (y-axis).

```
df_eng |>
  ggplot() +
  aes(x = age_subject, y = rt_lexdec) +
  geom_point() +
  labs(title = "geom_point()") +
df_eng |>
  ggplot() +
  aes(x = age_subject, y = rt_lexdec) +
  geom_point(position = position_jitter(0.2),
            alpha = 0.2) +
  labs(title = "geom_point(position = position_jitter(0.2))") +
  plot_annotation(tag_levels = "A")
```

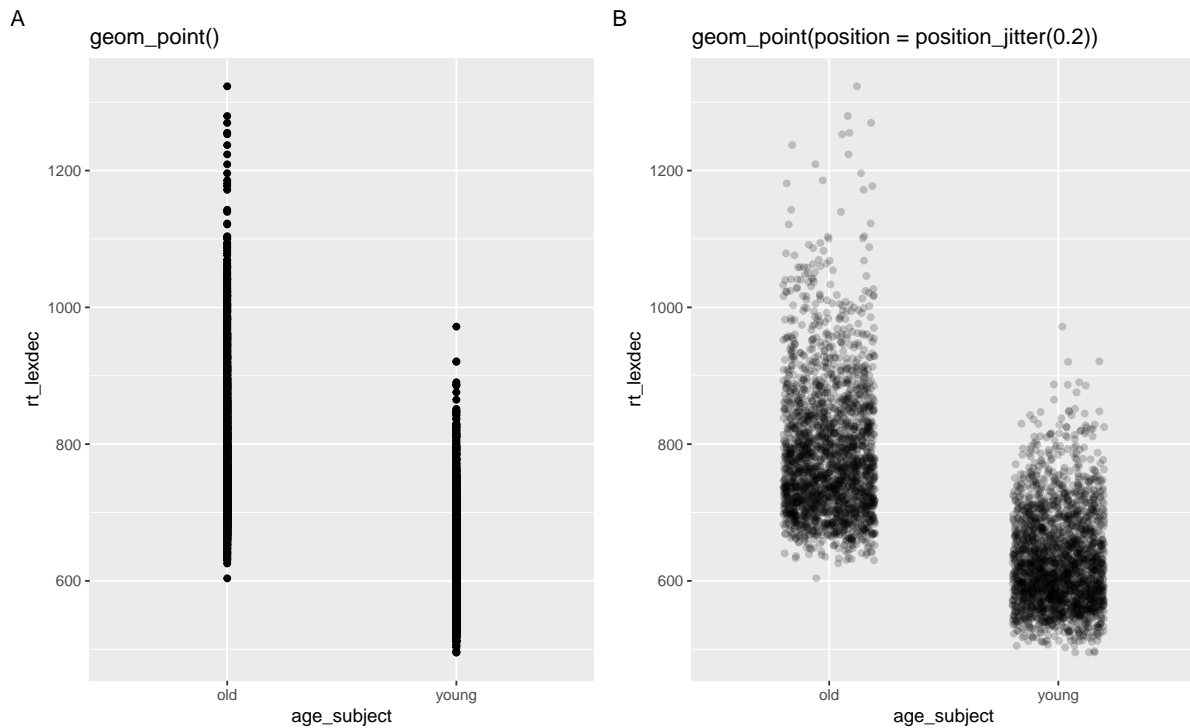


Abbildung 4: Plotting points along a categorical variable without (A) and with (B) position = position\_jitter(0.2). Plot B also includes alpha = 0.2

## 1.4 Combining all three

If we put all of these plots together, we get a [Abbildung 5](#).

```
fig_no_colour <-
  df_eng %>%
  ggplot() +
  aes(x = age_subject, y = rt_lexdec) +
  geom_point(position = position_jitter(0.2),
            alpha = 0.2) +
  geom_half_violin() +
  geom_boxplot(
    outlier.shape = NA,
    width = .3,
    position = position_nudge(x=0.2))
```

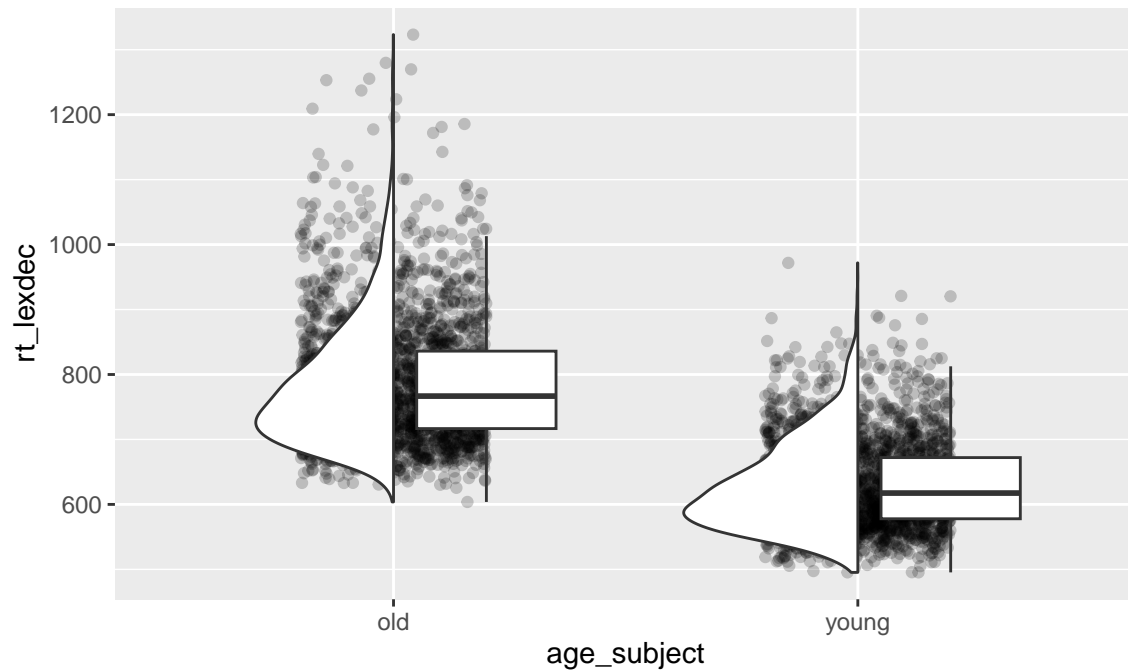


Abbildung 5: Density, boxplot, scatterplot

## 2 Positioning errorbar plots

In the second report, you produced errorbar plots, but the errorbars were overlapping.

### 2.1 `pivot_longer()` |> `summarise()`

Let's reproduce something similar using the `english` dataset. First, we'll use `pivot_longer()` to lengthen our data, then we'll create a summary of reaction times for the lexical decision task and naming task per age group.

```
sum_eng <-
  df_eng |>
  pivot_longer(
    cols = c(rt_lexdec, rt_naming),
    names_to = "task",
    values_to = "rt"
  ) |>
  summarise(
    mean = mean(rt, na.rm = T),
    sd = sd(rt, na.rm = T),
```

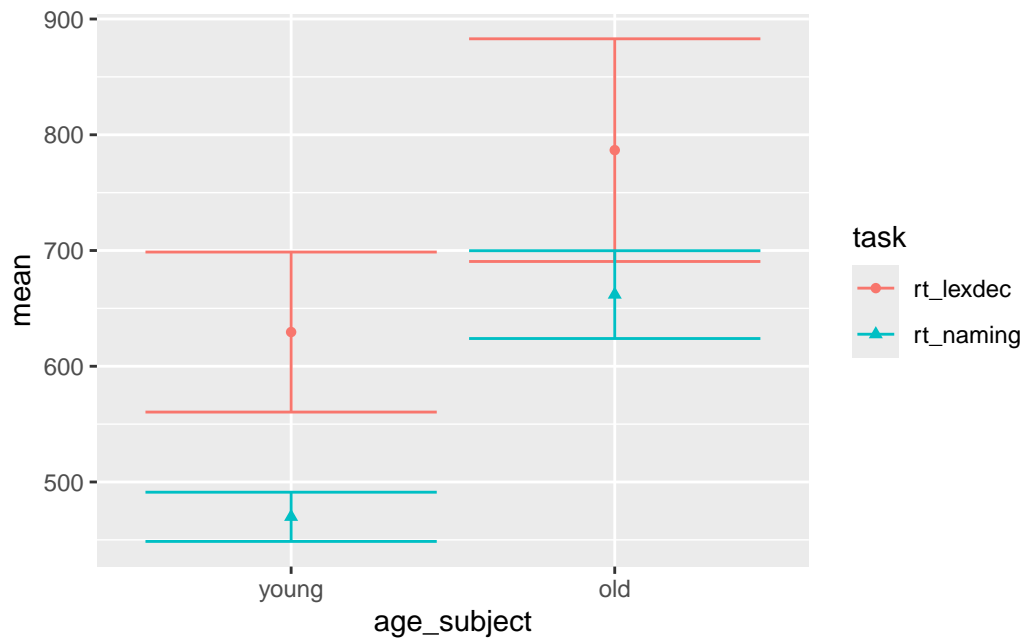


Abbildung 6: Overlapping errorbars

```
.by = c(age_subject, task)
) |>
mutate(age_subject = factor(age_subject, levels = c("young", "old")))
```

## 2.2 Overlapping errorbars

If we create an errorbar plot of this data, we get [Abbildung 7](#).

```
sum_eng |>
  ggplot() +
  aes(x = age_subject, y = mean, colour = task, shape = task) +
  geom_point() +
  geom_errorbar(aes(ymin = mean-sd, ymax = mean+sd))
```



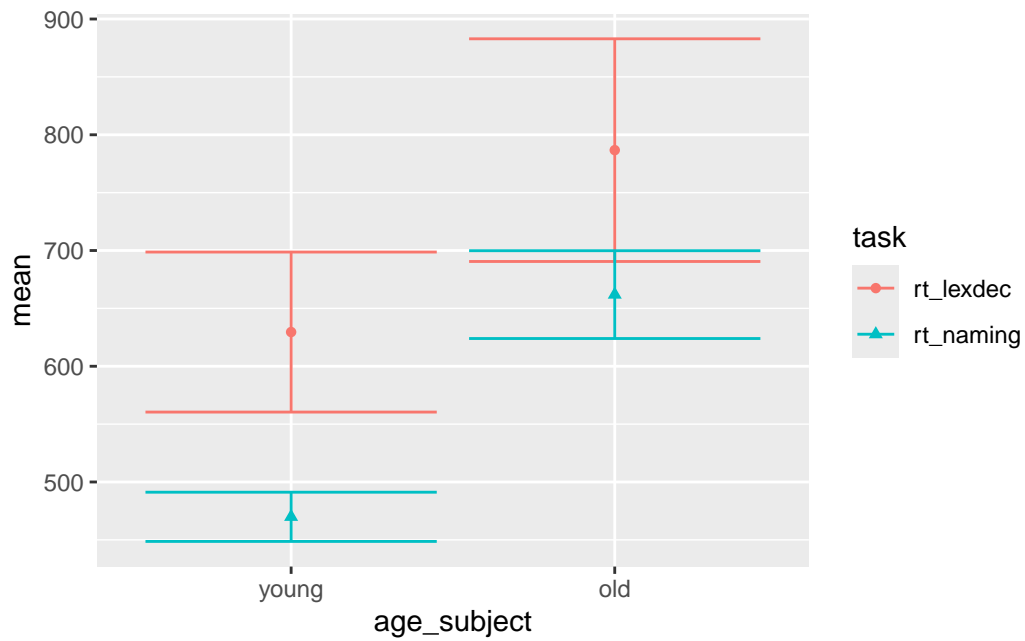


Abbildung 7: Overlapping errorbar plot

### 2.3 position\_dodge()

We can add `position = position_dodge(0.2)` to force the errorbars to not overlap. We'll also adjust their width so they're not so wide (any value lower than 0.75).

```
sum_eng |>
  ggplot() +
  aes(x = age_subject, y = mean, colour = task, shape = task) +
  geom_point() +
  geom_errorbar(aes(ymin = mean-sd, ymax = mean+sd),
                position = position_dodge(0.2),
                width = 0.2)
```

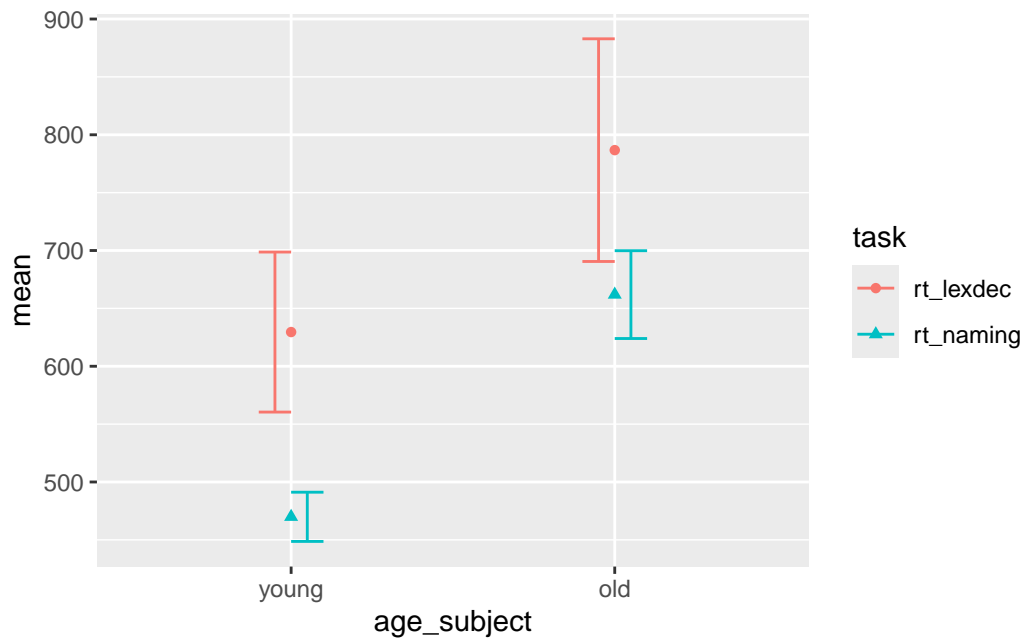


Abbildung 8: Overlapping errorbar plot

## 2.4 dodging all relevant geoms

But now we've left the points behind. We need to also dodge the points, so we add `position_dodge()` to `geom_point()`, making sure to use the same value as we did with `geom_errorbar()`.

```
sum_eng |>
  ggplot() +
  aes(x = age_subject, y = mean, colour = task, shape = task) +
  geom_point(position = position_dodge(0.2)) +
  geom_errorbar(aes(ymin = mean-sd, ymax = mean+sd),
               position = position_dodge(0.2),
               width = 0.2)
```

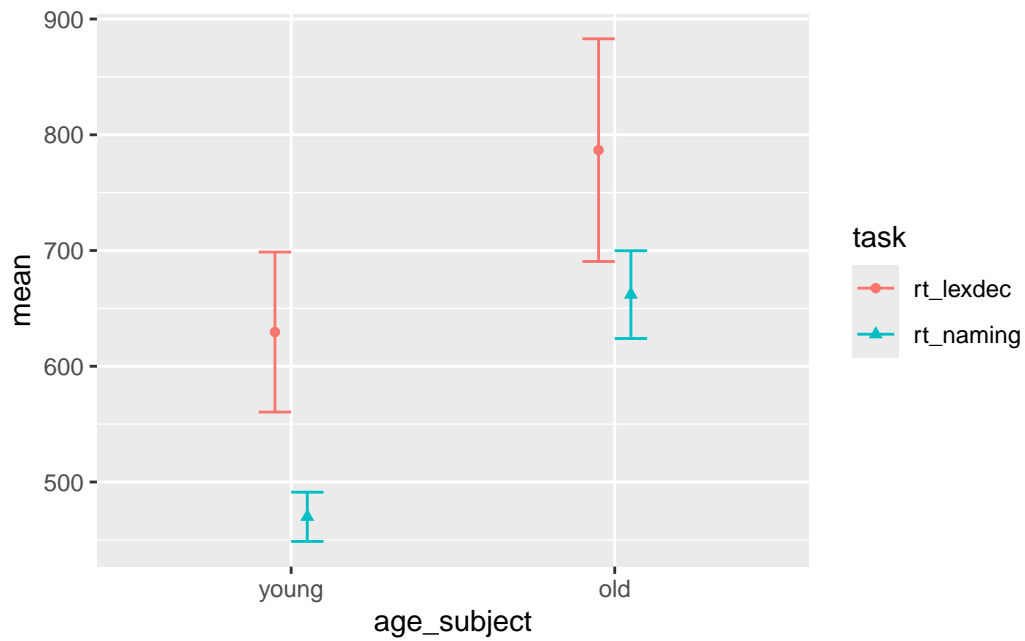


Abbildung 9: Overlapping errorbar plot

### 3 Customisations

What customisations do you see in the plots in [Abbildung 10](#)?

```
fig_dens_colour <-
  df_eng %>%
  ggplot(aes(x = age_subject, y = rt_lexdec, )) +
  geom_point(
    color = "grey",
    position = position_jitter(0.2),
    alpha = 0.2) +
  geom_half_violin(
    aes(fill = age_subject)) +
  geom_boxplot(
    outlier.shape = NA,
    aes(color = age_subject),
    width = .3,
    position = position_nudge(x=0.2)) +
  labs(title = "Distribution of reaction times",
    x = "Age group",
    y = "LDT reaction time (ms)",
```

```

    fill = "Age group") +
  scale_color_colorblind() +
  scale_fill_colorblind() +
  theme_minimal() +
  theme(legend.position = "none")

fig_point_colour <-
  df_eng %>%
  ggplot(aes(x = age_subject, y = rt_lexdec, )) +
  geom_point(
    aes(color = age_subject),
    position = position_jitter(0.2),
    alpha = 0.2) +
  geom_half_violin() +
  geom_boxplot(
    outlier.shape = NA,
    # aes(color = age_subject),
    width = .3,
    position = position_nudge(x=0.2)) +
  labs(title = "Distribution of reaction times",
    x = "Age group",
    y = "LDT reaction time (ms)",
    fill = "Age group") +
  scale_color_colorblind() +
  scale_fill_colorblind() +
  theme_minimal() +
  theme(legend.position = "none")

fig_default <-
  sum_eng %>%
  ggplot(aes(x = age_subject, y = mean,
    colour = task, shape = task)) +
  geom_point() +
  geom_errorbar(aes(ymin=mean-sd,ymax=mean+sd))

fig_custom <-
  sum_eng %>%
  mutate(task = fct_recode(task,
    "LDT" = "rt_lexdec",
    "Naming" = "rt_naming"),
  age_subject = fct_recode(age_subject,
    "Young" = "young",

```

```

      "Old" = "old")) |>
ggplot(aes(x = age_subject, y = mean,
          colour = task, shape = task)) +
geom_point(position = position_dodge(0.3),
          size = 3) +
geom_errorbar(aes(ymin=mean-sd,ymax=mean+sd),
             position = position_dodge(0.3),
             width = .3) +
geom_line(aes(group = task,
             linetype = task),
          position = position_dodge(0.3)) +
theme_minimal() +
labs(
  title = "Reaction times per group and task",
  x = "Age group",
  y = "Reaction time (ms)",
  colour = "Task",
  shape = "Task",
  linetype = "Task"
) +
theme(axis.title = element_text(size = 12,
                                face = "bold"),
      plot.title = element_text(size = 14),
      legend.title = element_text(face = "bold"))

```

### 3.1 Default themes

Firstly, `theme_minimal()` was added to each plot to customise the general look. There are a variety of custom themes to try, like `theme_bw()` or `theme_classic()`. Try them out.

### 3.2 `theme()`

We can also control individual components of theme by adding customisations with `theme()`. For example we see in [Abbildung 10 A](#) the axis titles are bolded. This was achieved by adding `theme(axis.title = element_text(face = "bold"))`, where `axis.title =` indicates we want to make a change to the axis titles, `element_text()` indicates it's their text that we want to change, and `face = "bold"` indicates we want to make the text bold. The same was done for `legend.title =` to make the legend title bold.

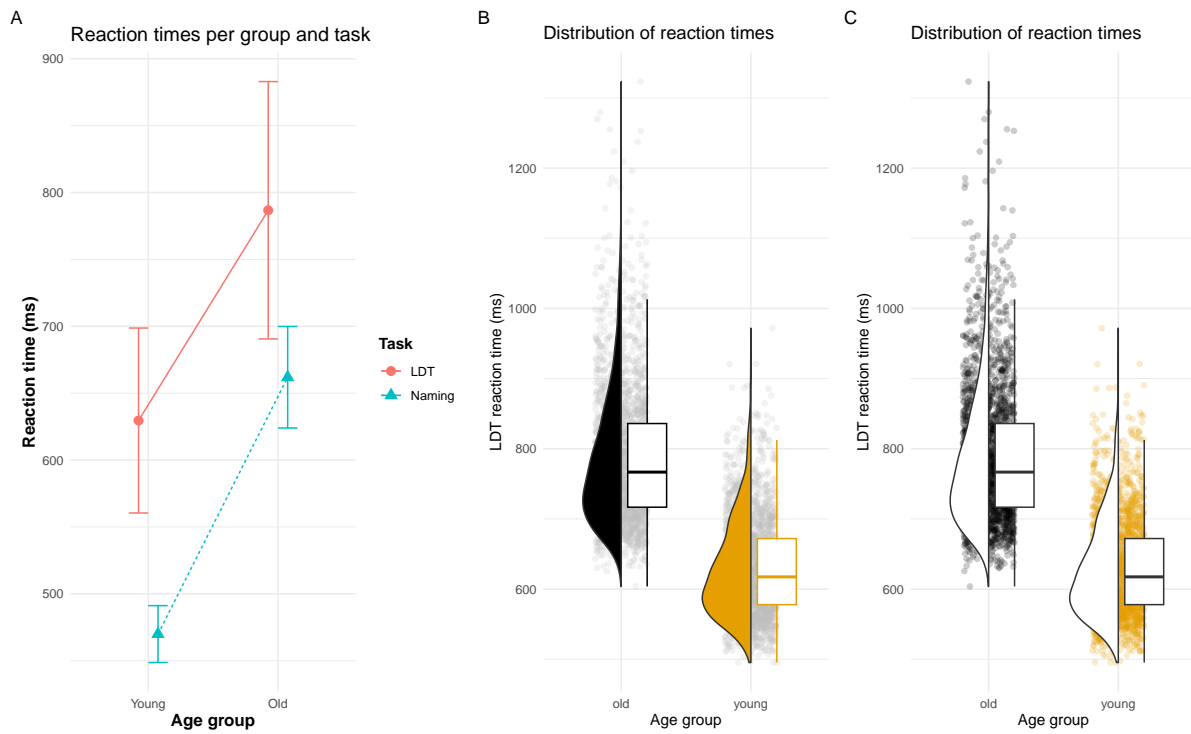


Abbildung 10: Customised plots to facilitation data communication.

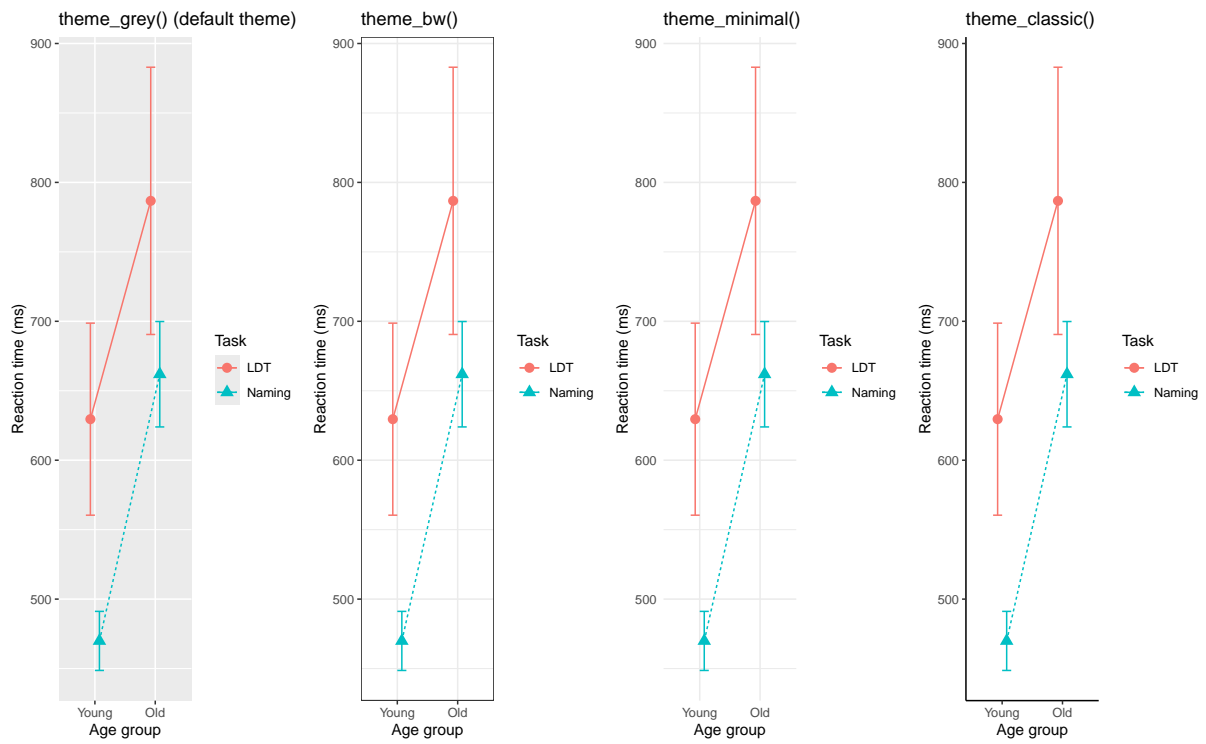


Abbildung 11: Preset themes

```
fig_no_colour + theme_minimal() +
  theme(
    axis.title = element_text(face = "italic")
  )
```

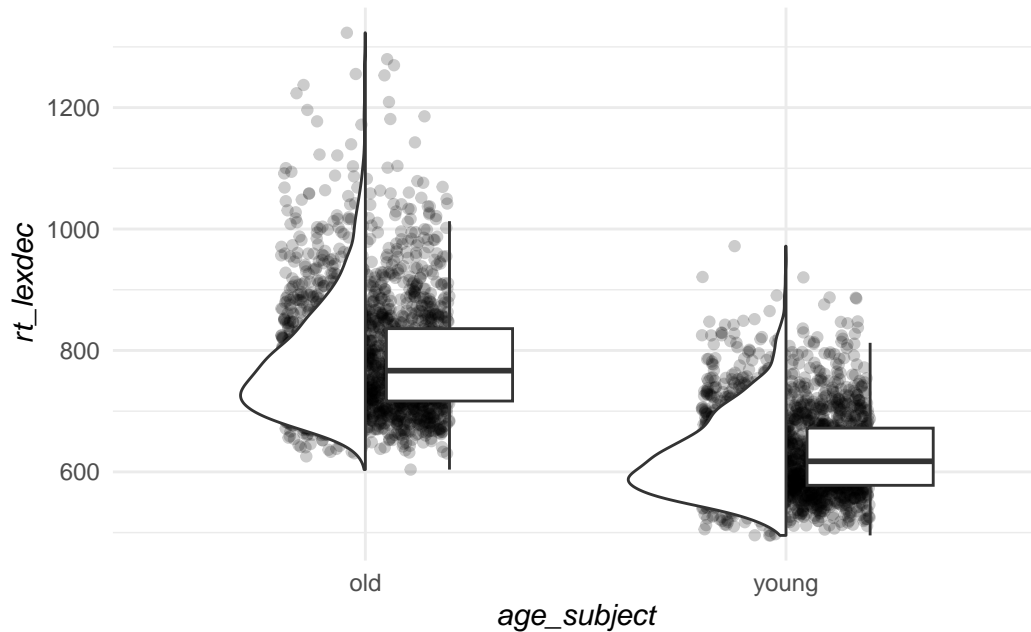


Abbildung 12: Using theme()

## Heutige Ziele

Heute haben wir gelernt, wie man...

- build multi-part plots
- adjust the position of geoms
- customise our plots for better data communicate

## 4 Aufgaben

1. Multi-part plot. Produce [Abbildung 9](#) and [Abbildung 5](#) for `rt_naming` (instead of `rt_lexdec`). Print the plots side-by-side using `patchwork`.



2. Labels. Use `labs()` to add labels for the title, x- and y-axes, and for any aesthetics you used (shape, colour, etc.) that result in a legend. This should end with your legend title also having a custom name.
3. Customisations. Add customisations to the two plots by choosing a default theme, followed by `theme()` with adjustments for the axis titles, legend title, and plot title. You can change `face`, `size`, `family` (i.e., font). You can type `?theme` in the Console or try Googling to get some ideas. If you aren't feeling creative, just try to replicate one of the customisations you see in Abbildung 10

## Session Info

Hergestellt mit R version 4.4.0 (2024-04-24) (Puppy Cup) und RStudioversion 2023.3.0.386 (Cherry Blossom).

```
print(sessionInfo(), locale = F)
```

```
R version 4.4.0 (2024-04-24)
Platform: aarch64-apple-darwin20
Running under: macOS Ventura 13.2.1
```

```
Matrix products: default
```

```
BLAS: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
```

```
LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib;
```

```
attached base packages:
```

```
[1] stats      graphics  grDevices datasets  utils      methods    base
```

```
other attached packages:
```

```
[1] magick_2.8.3    ggthemes_5.1.0
[5] janitor_2.2.0   here_1.0.1      lubridate_1.9.3 forcats_1.0.0
[9] stringr_1.5.1   dplyr_1.1.4     purrr_1.0.2     readr_2.1.5
[13] tidyr_1.3.1     tibble_3.2.1    ggplot2_3.5.1   tidyverse_2.0.0
```

```
loaded via a namespace (and not attached):
```

```
[1] utf8_1.2.4      generics_0.1.3  renv_1.0.7      stringi_1.8.3
[5] hms_1.1.3       digest_0.6.35   magrittr_2.0.3   evaluate_0.23
[9] grid_4.4.0      timechange_0.3.0 fastmap_1.1.1    rprojroot_2.0.4
[13] jsonlite_1.8.8  tinytex_0.50    fansi_1.0.6      scales_1.3.0
[17] cli_3.6.2       crayon_1.5.2    rlang_1.1.3      bit64_4.0.5
[21] munsell_0.5.1   withr_3.0.0     yaml_2.3.8       parallel_4.4.0
```

|      |                   |                   |                  |                  |
|------|-------------------|-------------------|------------------|------------------|
| [25] | tools_4.4.0       | tzdb_0.4.0        | colorspace_2.1-0 | pacman_0.5.1     |
| [29] | vctrs_0.6.5       | R6_2.5.1          | lifecycle_1.0.4  | snakecase_0.11.1 |
| [33] | bit_4.0.5         | vroom_1.6.5       | pkgconfig_2.0.3  | pillar_1.9.0     |
| [37] | gtable_0.3.5      | glue_1.7.0        | Rcpp_1.0.12      | xfun_0.43        |
| [41] | tidyselect_1.2.1  | rstudioapi_0.16.0 | knitr_1.46       | farver_2.1.1     |
| [45] | htmltools_0.5.8.1 | labeling_0.4.3    | rmarkdown_2.26   | compiler_4.4.0   |

## Literaturverzeichnis

- Nordmann, E., McAleer, P., Toivo, W., Paterson, H., & DeBruine, L. M. (2022). Data Visualization Using R for Researchers Who Do Not Use R. *Advances in Methods and Practices in Psychological Science*, 5(2), 251524592210746. <https://doi.org/10.1177/25152459221074654>
- Wickham, H., Çetinkaya-Rundel, M., & Grolemund, G. (2023). *R for Data Science* (2. Aufl.).