

# r\_notebook

## Advantages of R Notebooks!

---

Notebooks are a ***fantastic tool*** when it comes to writing R scripts for research. This is because they make the code ***easy to comprehend and inter-operable***, making them a good practice of **open science!**

Let's look at the same R code again, but now with the advantages of a notebook!

### 1. Demarcate code more clearly with markdown!

#### 1. 1 Load the libraries

```
suppressPackageStartupMessages({  
  library(dplyr)  
  library(ggplot2)  
})
```

#### 1. 2 Create the data set

```
# create fitness data  
source("R/create_synth_data.R")
```

Attaching package: 'lubridate'

The following objects are masked from 'package:base':

date, intersect, setdiff, union

```
df_fitness <- create_fitness_data(num_participants = 10, seed = 123)
```

### 2. Interact with the data as you code!

#### 2.1 Look at chunks of data within the notebook!

```
head(df_fitness)
```

	study_number	Datetime	week_id	is_compliant_day	avg_heart_rate
1	1	2021-01-01 12:00:00	1	TRUE	85.6
2	1	2021-01-01 12:01:00	1	TRUE	70.7
3	1	2021-01-01 12:02:00	1	TRUE	71.3
4	1	2021-01-01 12:03:00	1	TRUE	87.1
5	1	2021-01-01 12:04:00	1	TRUE	74.6
6	1	2021-01-01 12:05:00	1	TRUE	57.3

### 2. Make operations on data more visually accessible!

#### 2.1 Consider code chunk below with a single output

```
daily_avg_hr_per_study <- df_fitness |>
  mutate(day = as.Date(Datetime)) |>
  group_by(study_number, day) |>
  summarise(avg_daily_heart_rate = mean(avg_heart_rate, na.rm = TRUE), .groups = 'drop')

head(daily_avg_hr_per_study)
```

```
# A tibble: 6 × 3
  study_number day      avg_daily_heart_rate
    <int> <date>          <dbl>
1         1 2021-01-01          70.4
2         1 2021-01-02          69.2
3         1 2021-01-03          69.5
4         1 2021-01-04          67.6
5         1 2021-01-05          66.1
6         1 2021-01-06          65.3
```

## 2.1 Break it into parts!

### First step

```
mutated_data <- df_fitness |>
  mutate(day = as.Date(Datetime), .after = study_number)

head(mutated_data)
```

```
  study_number      day      Datetime week_id is_compliant_day
1           1 2021-01-01 2021-01-01 12:00:00         1          TRUE
2           1 2021-01-01 2021-01-01 12:01:00         1          TRUE
3           1 2021-01-01 2021-01-01 12:02:00         1          TRUE
4           1 2021-01-01 2021-01-01 12:03:00         1          TRUE
5           1 2021-01-01 2021-01-01 12:04:00         1          TRUE
6           1 2021-01-01 2021-01-01 12:05:00         1          TRUE

  avg_heart_rate
1           85.6
2           70.7
3           71.3
4           87.1
5           74.6
6           57.3
```

### Second step!

```
grouped_data <- mutated_data |>
  group_by(study_number, day)

head(grouped_data)
```

```
# A tibble: 6 × 6
# Groups:   study_number, day [1]
  study_number day      Datetime      week_id is_compliant_day
    <int> <date>      <dtm>          <int> <lgl>
1         1 2021-01-01 2021-01-01 12:00:00         1 TRUE
```

```

2          1 2021-01-01 2021-01-01 12:01:00      1 TRUE
3          1 2021-01-01 2021-01-01 12:02:00      1 TRUE
4          1 2021-01-01 2021-01-01 12:03:00      1 TRUE
5          1 2021-01-01 2021-01-01 12:04:00      1 TRUE
6          1 2021-01-01 2021-01-01 12:05:00      1 TRUE
# i 1 more variable: avg_heart_rate <dbl>

```

### Third and final step!

```

summarized_data <- grouped_data |>
  summarise(avg_daily_heart_rate = mean(avg_heart_rate, na.rm = TRUE), .groups = 'drop')

head(summarized_data)

```

```

# A tibble: 6 × 3
  study_number day      avg_daily_heart_rate
      <int> <date>          <dbl>
1         1 2021-01-01          70.4
2         1 2021-01-02          69.2
3         1 2021-01-03          69.5
4         1 2021-01-04          67.6
5         1 2021-01-05          66.1
6         1 2021-01-06          65.3

```

## 3. Using additional features!

There are **many more tools** that can be used to improve the readability of your code! A non-exhaustive list includes: **tables, images/figures, links etc.** Let's revisit the same code chunks with these features!

### 3.1 Process Overview



### 3.2 Variables overview

No.	Variable	Meaning
1.	mutated_data	Add a <b>"day"</b> column to the fitness data
2.	grouped_data	Group <b>mutated_data</b> according to the the <b>day</b> column
3.	summarized_data	Summarize the <b>average heart rate</b> data

### 3.3 Additional resources

- [Mutate function: stack overflow](#)
- [Grouping and summarizing data: medium article](#)

### 3.4 In line code segments

```
# Calculate mean and standard deviation
# Filter the data based on the condition and calculate mean and standard deviation for avg_heart_rate
filtered_data <- df_fitness |>
  filter(study_number == 2)

# Calculate mean and standard deviation for avg_heart_rate
mean_value <- mean(filtered_data$avg_heart_rate, na.rm = TRUE)
std_value <- sd(filtered_data$avg_heart_rate, na.rm = TRUE)
```

The **mean** and **standard deviation** of **heart rates** for **study 2** are 64.9245487 and 10.4119959 respectively.

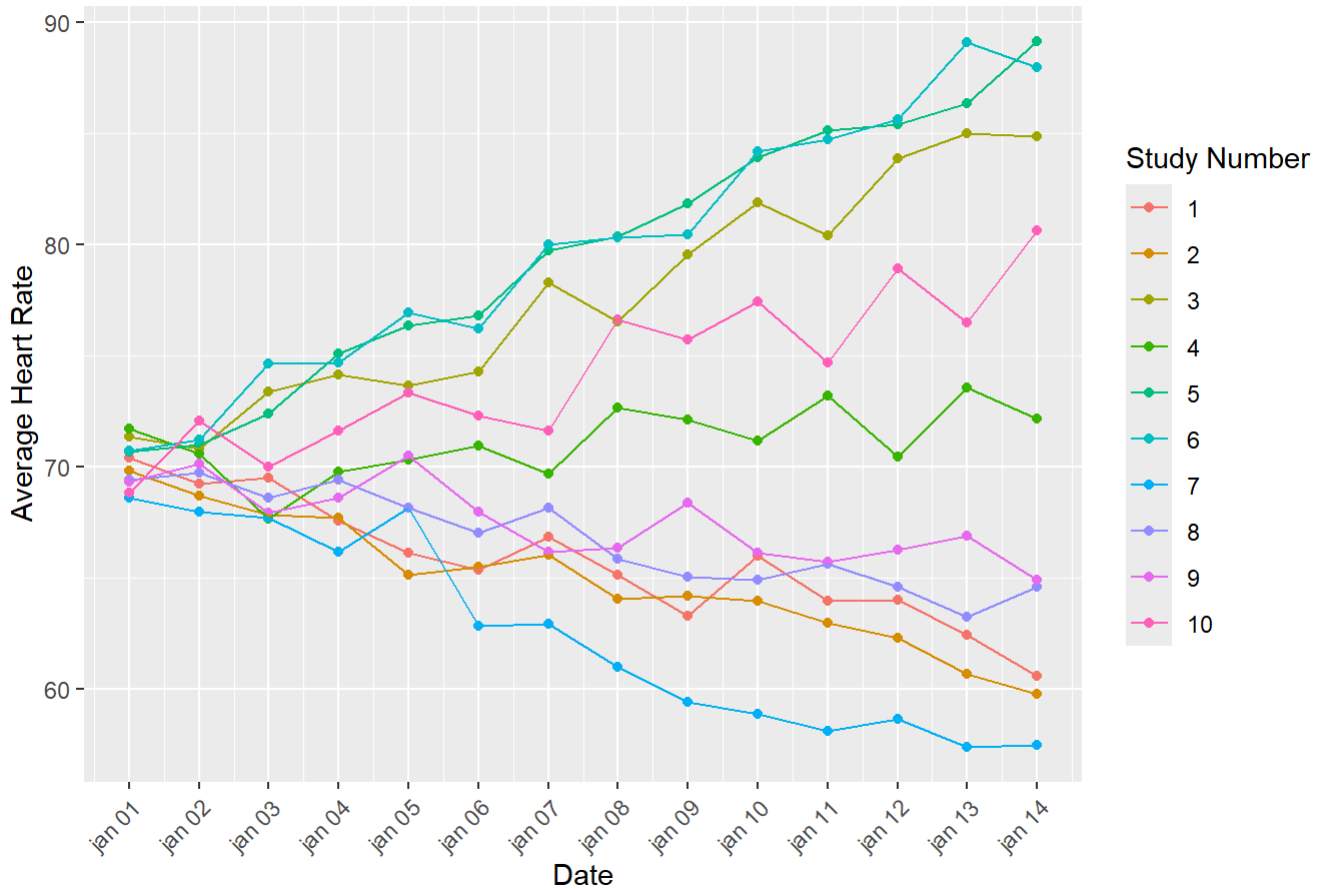
## 4. Visualize results (more easily)!

With notebooks, you can view multiple visualizations in with different cells!

```
daily_avg_hr_per_study <- df_fitness |>
  mutate(day = as.Date(Datetime)) |>
  group_by(study_number, day) |>
  summarise(avg_daily_heart_rate = mean(avg_heart_rate, na.rm = TRUE), .groups = 'drop')

ggplot(daily_avg_hr_per_study, aes(x = day, y = avg_daily_heart_rate, color = as.factor(study_number))) +
  geom_point() +
  geom_line() +
  labs(
    title = "Daily Average Heart Rate for Each Study Number",
    x = "Date",
    y = "Average Heart Rate",
    color = "Study Number"
  ) +
  scale_x_date(date_breaks = "1 day", date_labels = "%b %d") +
  scale_color_hue() +
  #theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

## Daily Average Heart Rate for Each Study Number



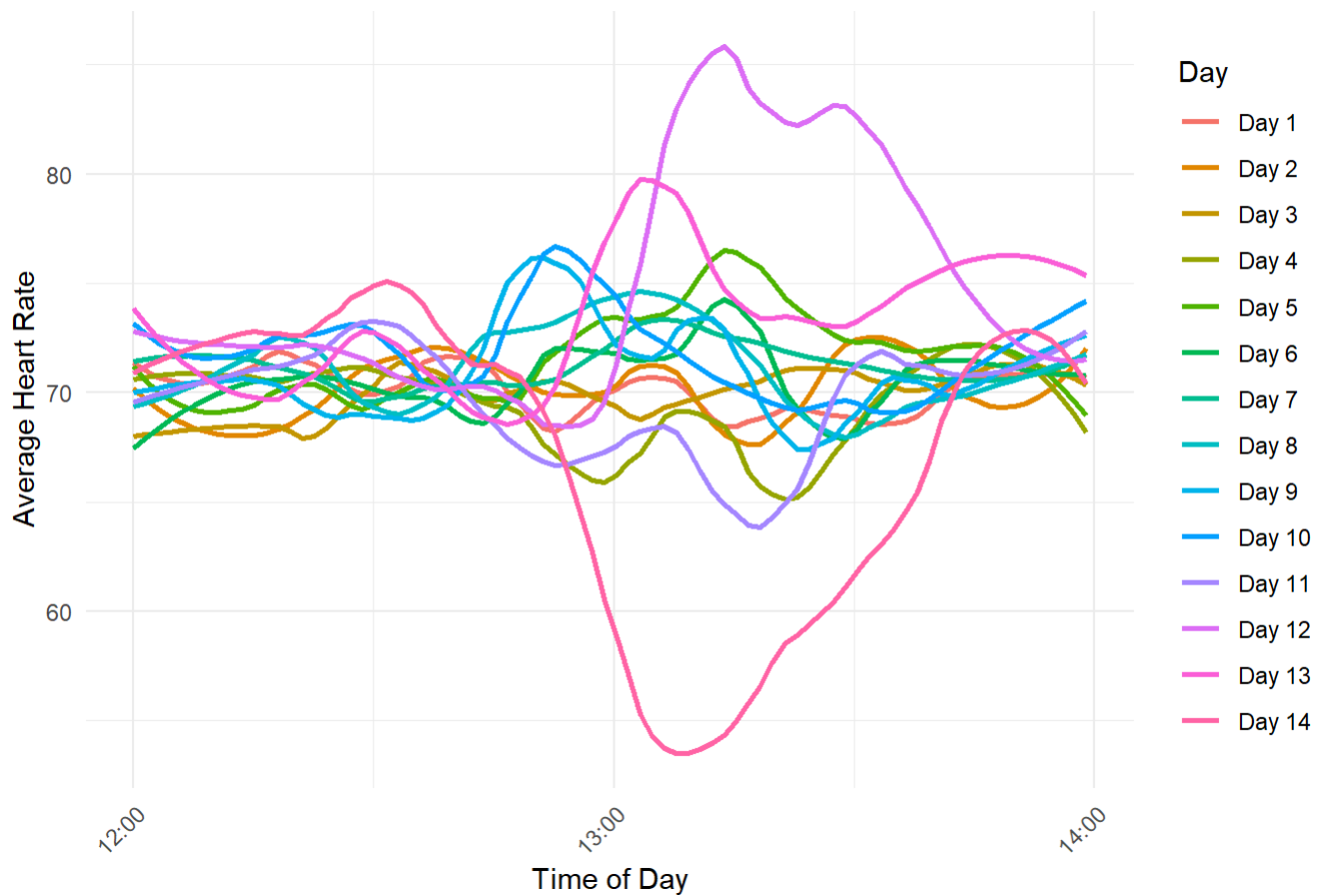
```
hourly_avg_hr <- df_fitness |>
  mutate(
    time_of_day = format(Datetime, "%H:%M"),
    day = as.Date(Datetime),
    day_label = paste("Day", as.integer(day - min(day) + 1)) # Assigns Day 1, Day 2, etc.
  ) |>
  group_by(day, time_of_day, day_label) |>
  summarise(avg_hr_across_studies = mean(avg_heart_rate, na.rm = TRUE), .groups = 'drop')

ggplot(hourly_avg_hr, aes(x = as.POSIXct(time_of_day, format = "%H:%M"), y = avg_hr_across_studies)) +
  geom_smooth(se = FALSE, span = 0.3) + # Using geom_smooth with a smaller span for smoothing
  labs(
    title = "Variation of Average Heart Rate Over the Day (Smoothed)",
    x = "Time of Day",
    y = "Average Heart Rate",
    color = "Day"
  ) +
  scale_x_datetime(date_breaks = "1 hour", date_labels = "%H:%M") + # Breaks every hour
  scale_color_hue() +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

`geom\_smooth()` using method = 'loess' and formula = 'y ~ x'

Warning: Removed 10 rows containing non-finite outside the scale range  
(`stat\_smooth()`).

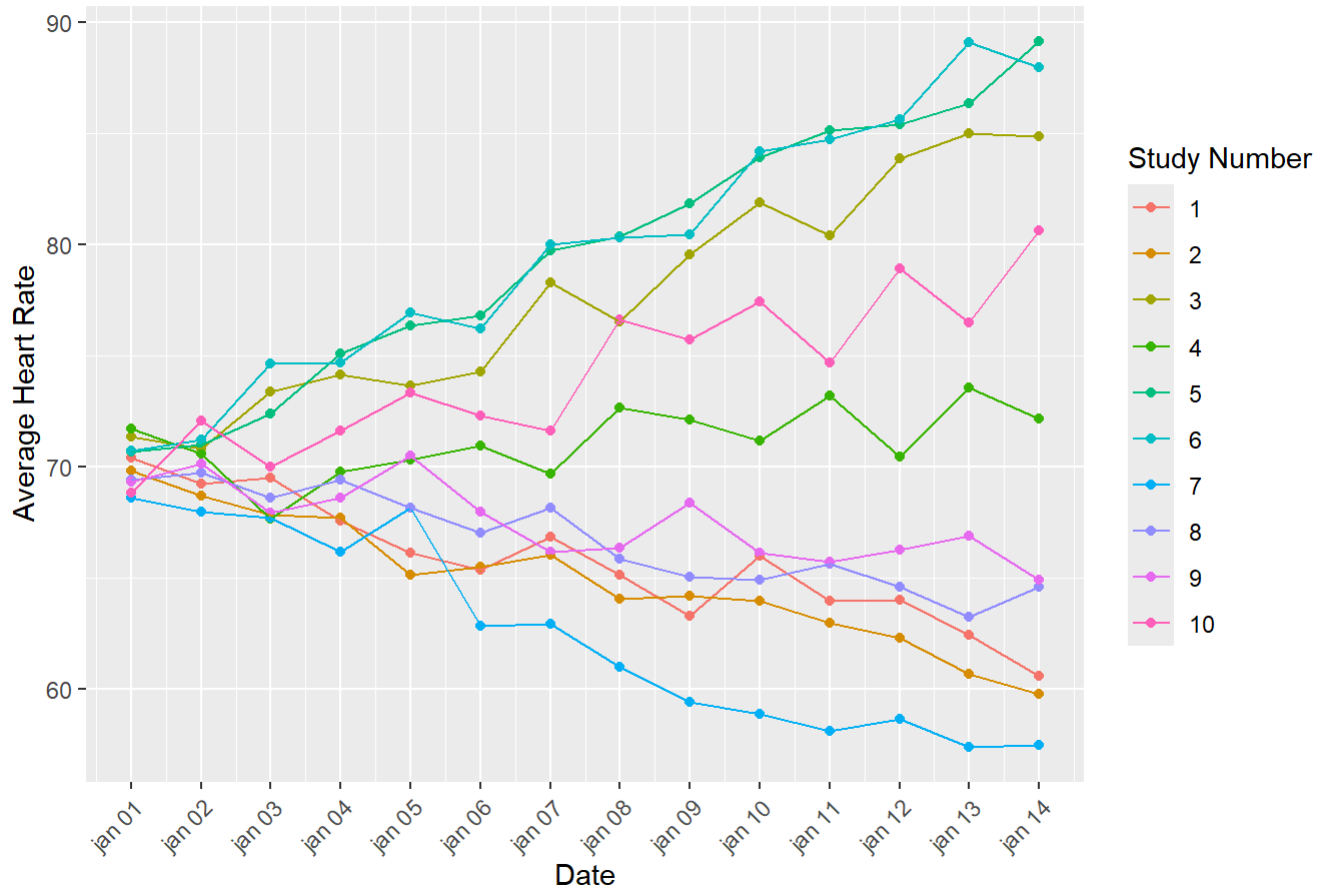
Variation of Average Heart Rate Over the Day (Smoothed)



```
daily_avg_hr_per_study <- df_fitness |>
  mutate(day = as.Date(Datetime)) |>
  group_by(study_number, day) |>
  summarise(avg_daily_heart_rate = mean(avg_heart_rate, na.rm = TRUE), .groups = 'drop')

ggplot(daily_avg_hr_per_study, aes(x = day, y = avg_daily_heart_rate, color = as.factor(study_number))) +
  geom_point() +
  geom_line() +
  labs(
    title = "Daily Average Heart Rate for Each Study Number",
    x = "Date",
    y = "Average Heart Rate",
    color = "Study Number"
  ) +
  scale_x_date(date_breaks = "1 day", date_labels = "%b %d") +
  scale_color_hue() +
  #theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

Daily Average Heart Rate for Each Study Number



## 5. Export notebooks to Slides/ PDFs!

This makes viewing your notebook on different devices possible without an IDE!