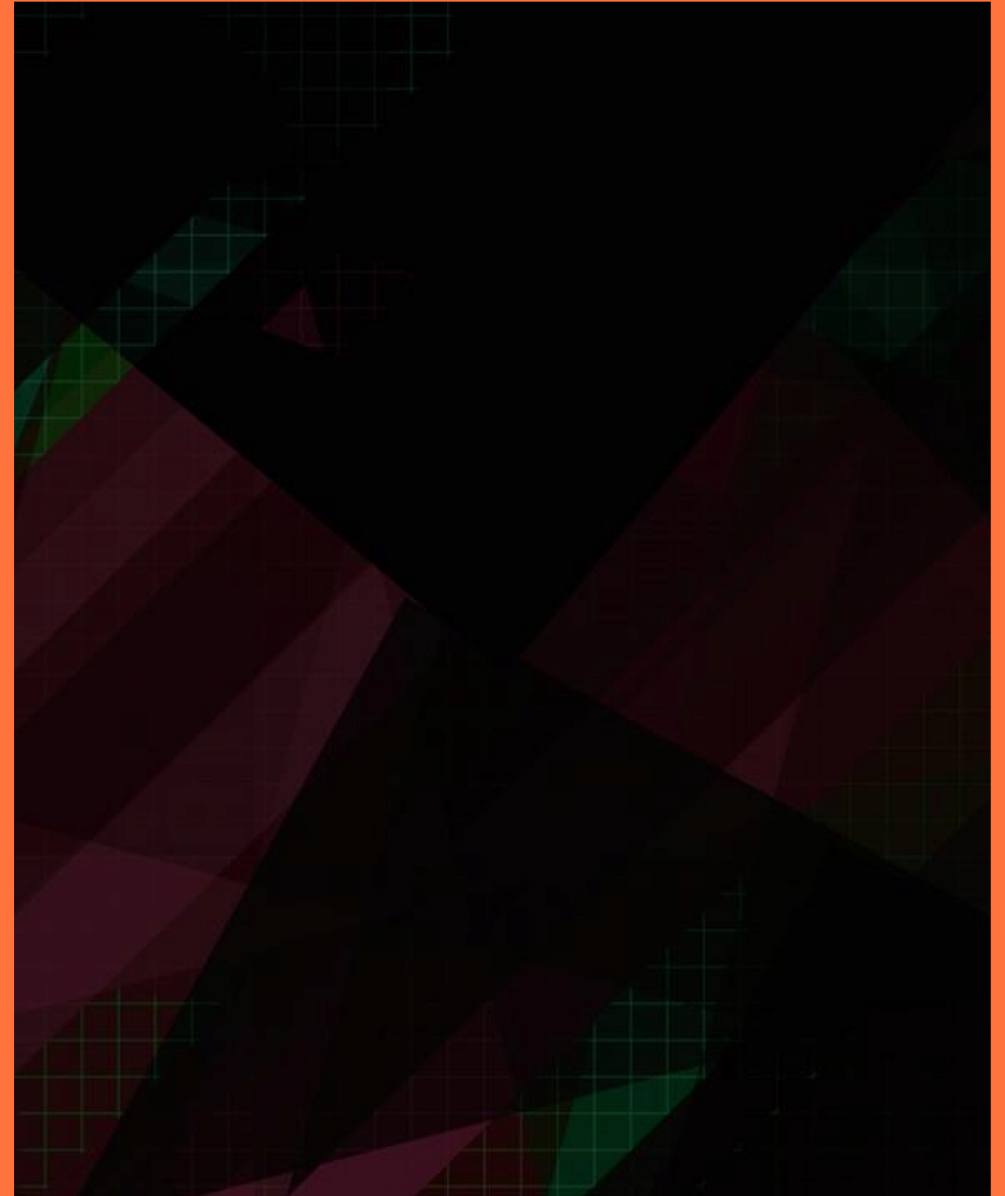


Practice IV. Work in groups. Functions to compute sectorial indices

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Objectives of Practice IV



Apply R programming to compute sectorial indices.



Work collaboratively in small groups.



Learn how to use both package-based and custom R functions.

Why Use R for This Analysis?



R IS OPEN-SOURCE,
REPRODUCIBLE, AND
COMMUNITY-SUPPORTED.



IDEAL FOR WORKING WITH TIME
SERIES, SPATIAL DATA, AND
CLIMATE INDICATORS.



COMPATIBLE WITH CLIMATE-
SPECIFIC PACKAGES AND
GENERAL DATA ANALYSIS TOOLS.

Hands-on practice

1. Install and load packages

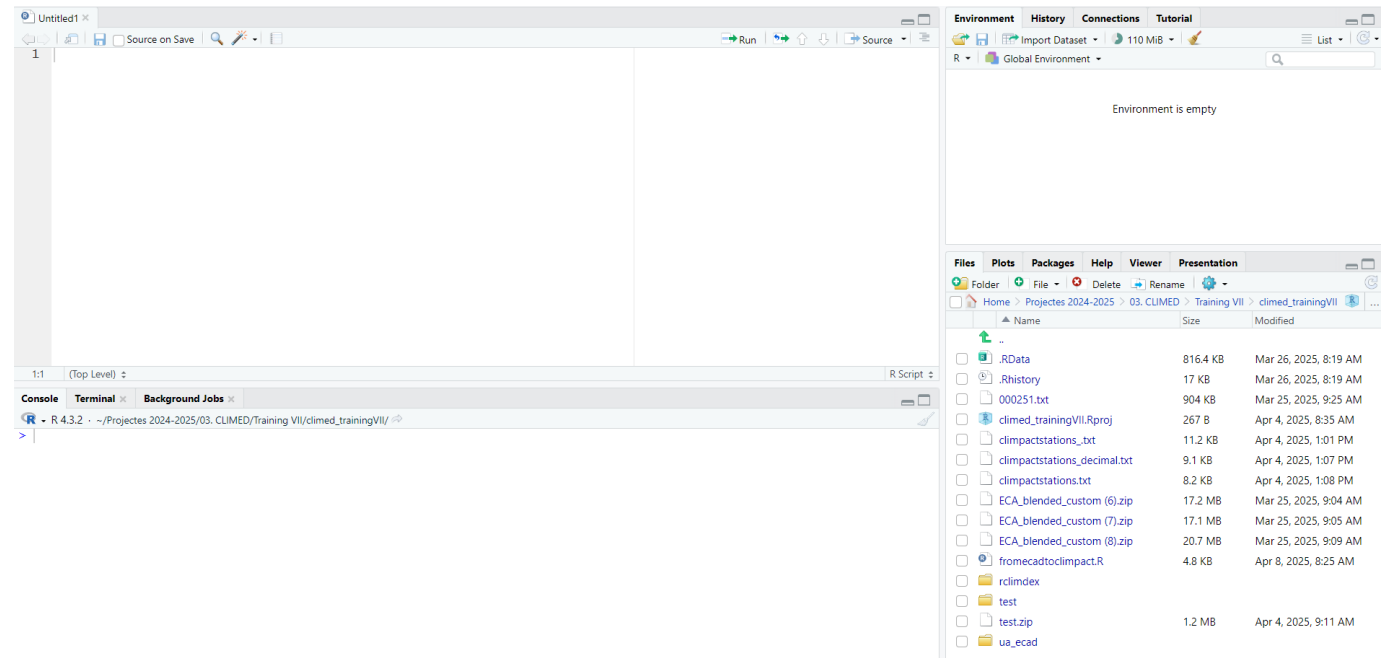
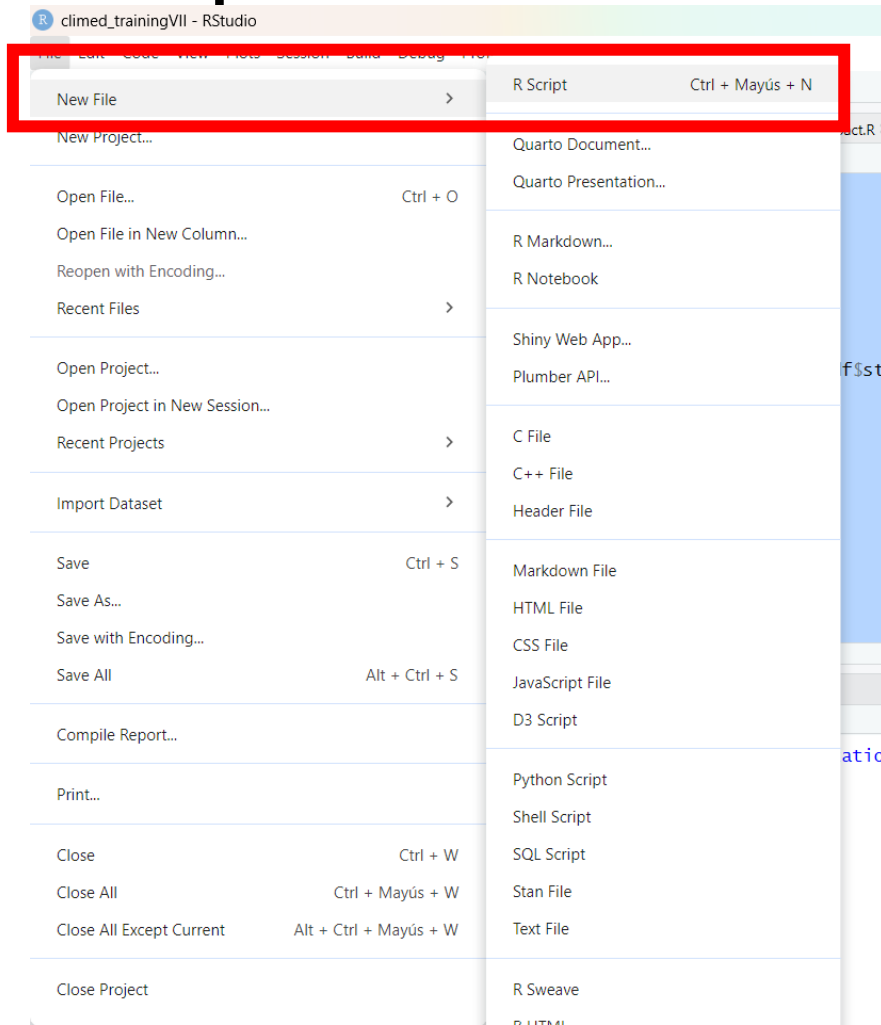
2. Load and prepare data

3. Apply a predefined function

4. Create a custom function

5. Visualization of the results

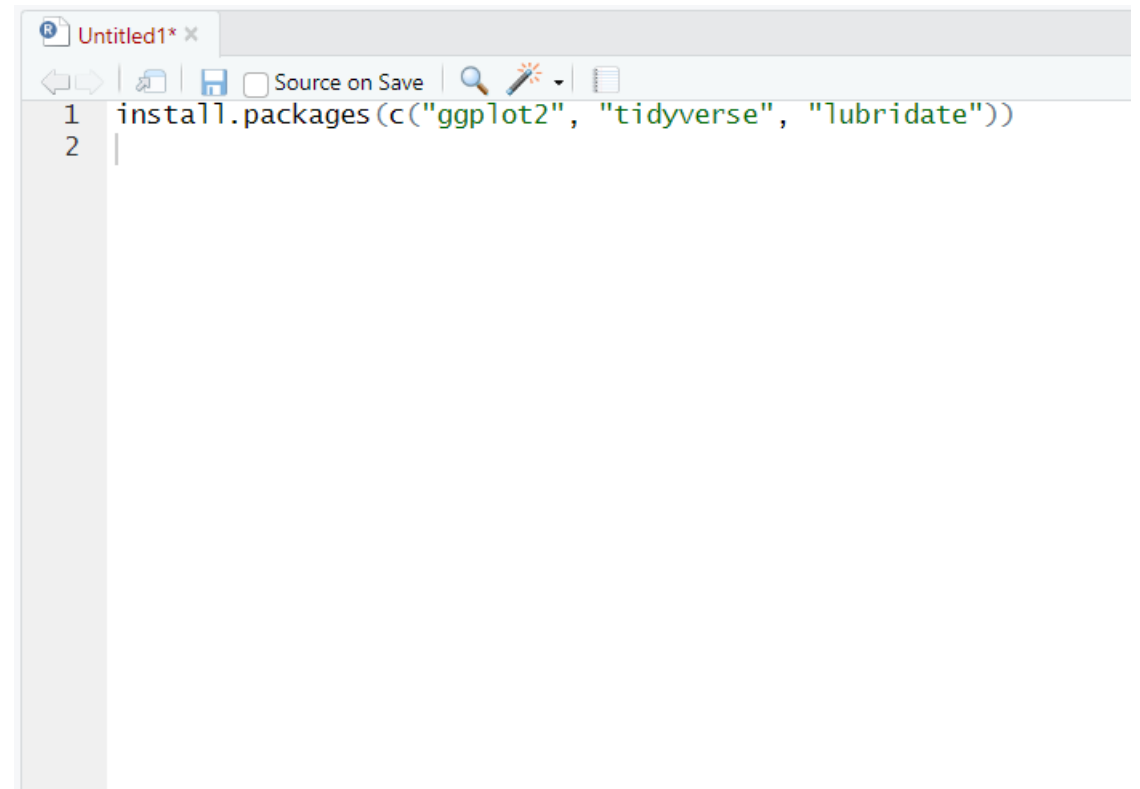
0. Prepare R-Studio



1. Install and load packages

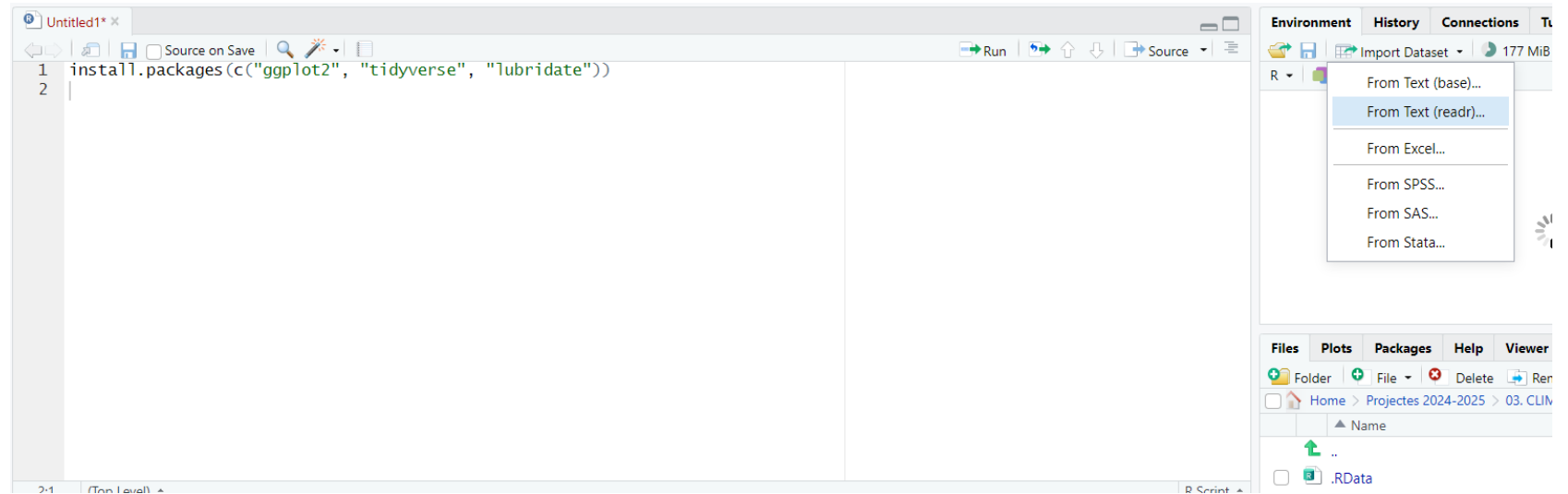
- # Install packages
- `install.packages(c("tidyverse", "lubridate"))`

- # Load librarys
- `library(tidyverse)`
- `library(lubridate)`

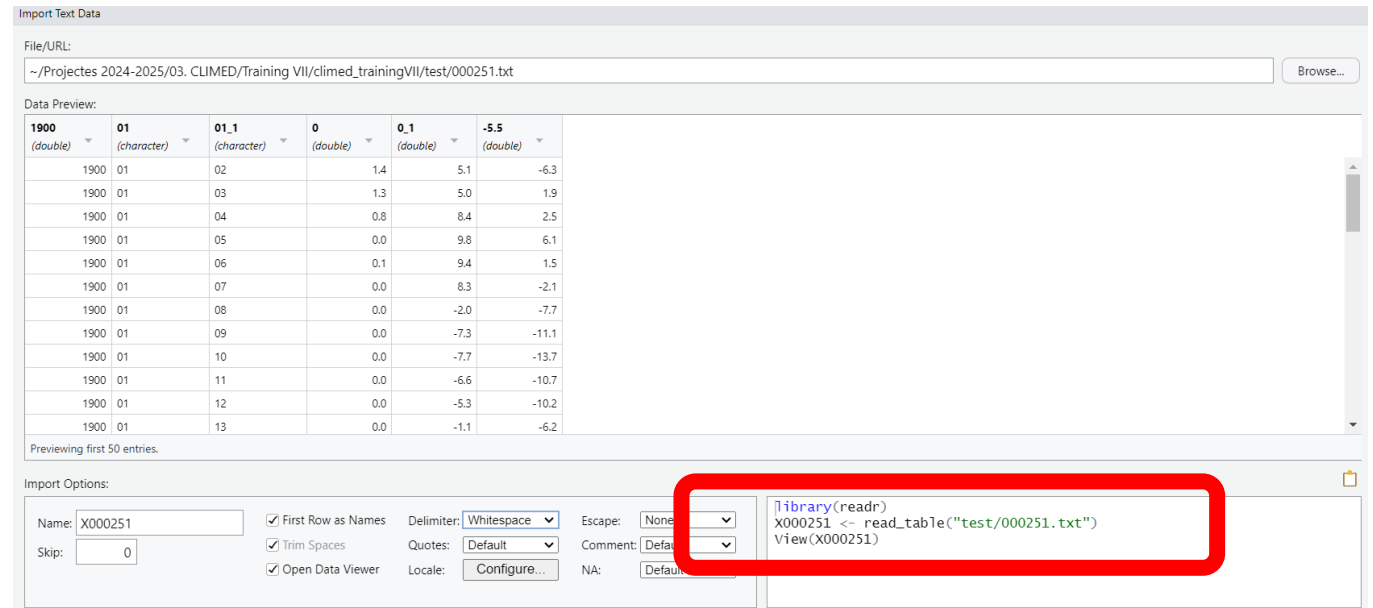
A screenshot of an R script editor window titled "Untitled1* x". The window has a toolbar with icons for navigation, saving, and search. Below the toolbar, the script contains two lines of code: line 1 is `install.packages(c("ggplot2", "tidyverse", "lubridate"))` and line 2 is a blank line with a cursor.

```
1 install.packages(c("ggplot2", "tidyverse", "lubridate"))
2 |
```

2. Load Data



```
library(ggplot2)
library(tidyverse)
library(lubridate)
library(readr)
climate_data <- read_table("test/000251.txt")
View(climate_data)
```

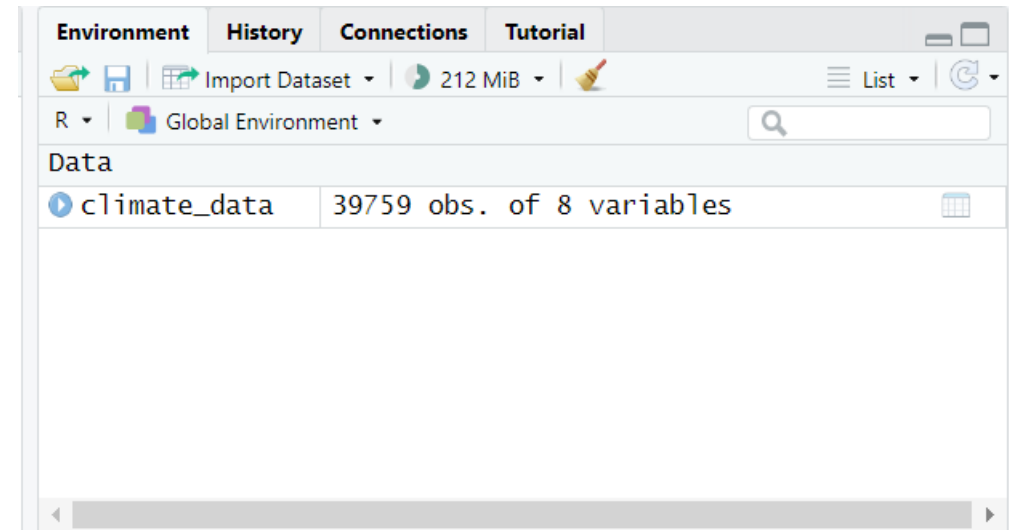


```
4 library(readr)
5 climate_data <- read_table("test/000251.txt")
6 View(climate_data)
```

3. Apply predefined function

- 3.1 Prepare data
 - # Extract year and TX
 - `climate_data$year <- climate_data[[1]]`
Column 1 = year
 - `climate_data$tx <- climate_data[[5]]`
Column 5 = TX

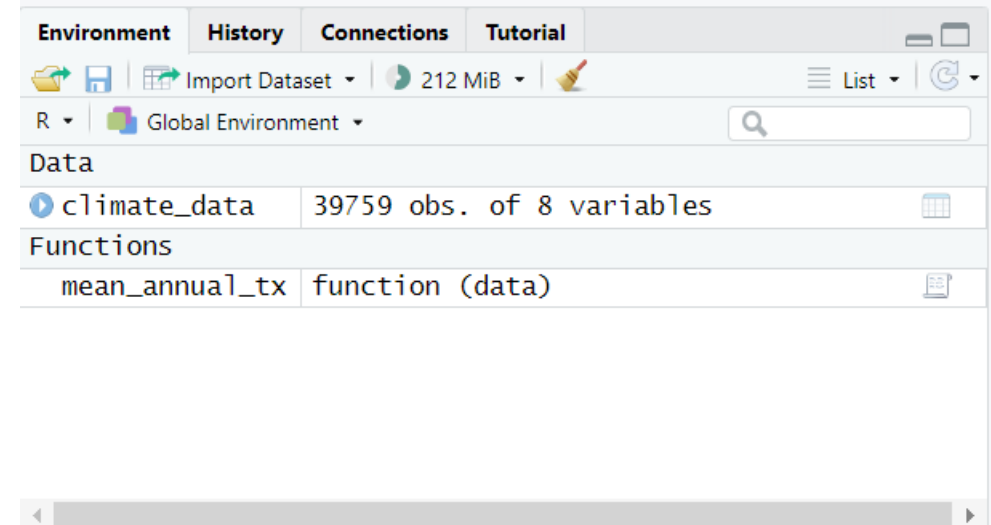
```
9 View(climate_data)
10
11
12 climate_data$year <- climate_data[[1]]
13 climate_data$tx <- climate_data[[5]]
14 |
```



3. Apply predefined function

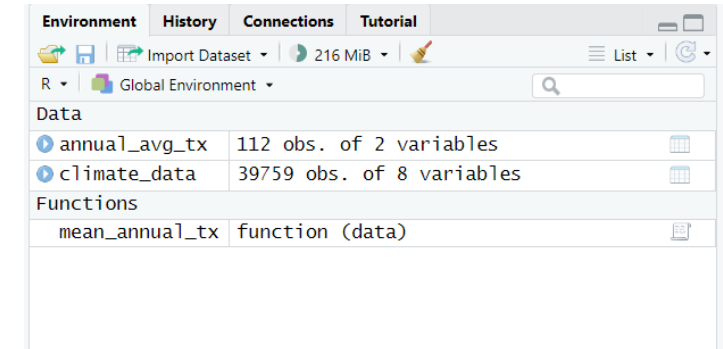
- 3.2 Create function for yearly mean and apply
- `mean_annual_tx <- function(data) {`
- `data %>%`
- `group_by(year) %>%`
- `summarise(mean_tx = mean(tx, na.rm = TRUE))`
- `}`
- #Apply function
- `annual_avg_tx <- mean_annual_tx(climate_data)`
- # See result
- `print(annual_avg_tx)`

```
16 mean_annual_tx <- function(data) {  
17   data %>%  
18     group_by(year) %>%  
19     summarise(mean_tx = mean(tx, na.rm = TRUE))  
20 }
```



```
22 # Apply  
23 annual_avg_tx <- mean_annual_tx(climate_data)  
24 |  
25 # See  
26 print(annual_avg_tx)  
24:1 (Top Level) ↓
```

```
R 4.3.2 · ~/Projectes 2024-2025/03. CLIMED/Training VII/climed_trainingVII/ ↗  
# A tibble: 112 × 2  
  year mean_tx  
  <dbl> <dbl>  
1 1900 15.6  
2 1901 15.8  
3 1902 14.8  
4 1903 15.7  
5 1904 15.0  
6 1905 15.6  
7 1906 15.8  
8 1907 14.2  
9 1908 14.7  
0 1909 16.0
```

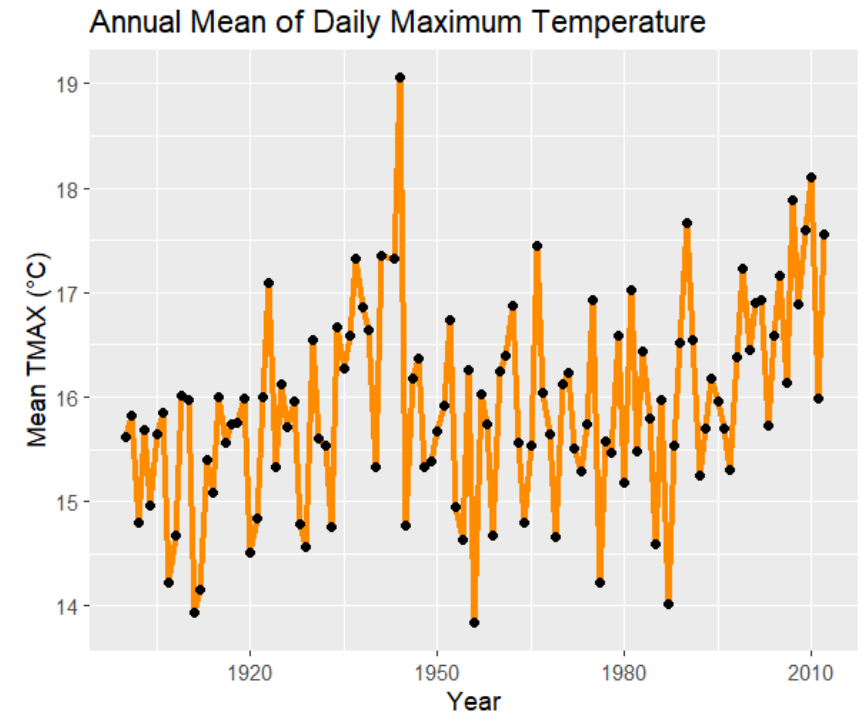


3. Apply predefined function

- 3.2 Visualize
- library(ggplot2)
- ggplot(annual_avg_tx, aes(x = year, y = mean_tx)) +
- geom_line(color = "darkorange", size = 1.2) +
- geom_point() +
- labs(title = "Annual Mean of Daily Maximum Temperature",
x = "Year", y = "Mean TMAX (°C)")

```
30 ggplot(annual_avg_tx, aes(x = year, y = mean_tx)) +  
31   geom_line(color = "darkorange", size = 1.2) +  
32   geom_point() +  
33   labs(title = "Annual Mean of Daily Maximum Temperature",  
34         x = "Year", y = "Mean TMAX (°C)")  
35
```

24:1 (Top Level) ↕



4. Sectorial functions

- 1. Agriculture: Dry Spells

```
climate_data$prec <- climate_data[[4]]
```

```
count_dry_spells <- function(prec, threshold = 1, min_length = 5) {  
  rle_obj <- rle(prec < threshold) # TRUE si precipitación < 1 mm  
  sum(rle_obj$values & rle_obj$lengths >= min_length)  
}
```

```
dry_spells <- count_dry_spells(climate_data$prec)
```

```
# Mostrar resultado
```

```
print(paste("Number of dry spells:", dry_spells))
```

4. Sectorial functions

- 1. Agriculture: Dry Spells

```
climate_data$tmax <- climate_data[[5]]  
  
count_hot_days <- function(temp, threshold = 35) {  
  sum(temp > threshold, na.rm = TRUE)  
}  
  
hot_days <- count_hot_days(climate_data$tmax)  
  
print(paste("Number of hot days:", hot_days))
```

4. Custom functions

- ```
rain_mm <- function(rain_mm) { if (rain_mm > 50) {
 return(5) } else if (rain_mm >= 31 && rain_mm <= 50) {
 return(4) } else if (rain_mm >= 21 && rain_mm <= 30) {
 return(3) } else if (rain_mm >= 11 && rain_mm <= 20) {
 return(2) } else if (rain_mm >= 0 && rain_mm <= 10) {
 return(1) } else { return(0) }}
```
- ```
climate_data$prec <- climate_data[[4]]
```
- ```
mm<-sapply(climate_data$prec ,FUN=cp_rain_mm)
```

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