

# Module 1: Tabular Data

Working with larger-than-RAM data using duckdbfs

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```
# Set CRAN mirror to avoid prompts
library(duckdbfs)
library(dplyr)
```

Attaching package: 'dplyr'

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

filter, lag

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

intersect, setdiff, setequal, union

```
library(ggplot2)
```

Warning: package 'ggplot2' was built under R version 4.5.2

```
# Remote S3 path to EXIOBASE 3 (Source Cooperative)
```

```
duckdbfs::duckdb_secrets(  
  key = "",  
  secret = "",  
  endpoint = "s3.amazonaws.com",  
  region = "us-west-2"  
)
```

```
[1] 1
```

```
s3_url <- "s3://us-west-2.opendata.source.coop/youssef-harby/exiobase-3/4588235/parquet/**"
```

```
# Open the dataset lazily  
exio <- open_dataset(s3_url)
```

```
# View the schema (column names and types) without reading data  
glimpse(exio)
```

```

Rows: ??
Columns: 8
Database: DuckDB 1.4.4 [root@Darwin 24.5.0:R 4.5.1/:memory:]
$ stressor <chr> "Value Added", "Value Added", "Value Added", "Value Added", "~
$ region <chr> "AT", "AT", "AT", "AT", "AT", "AT", "AT", "AT", "AT", "AT", "~
$ sector <chr> "Cultivation of wheat", "Cultivation of cereal grains nec", "~
$ value <dbl> 183.1118891, 402.2305799, 830.2127384, 101.9705426, 31.763189~
$ unit <chr> "M.EUR", "M.EUR", "M.EUR", "M.EUR", "M.EUR", "M.EUR", "M.EUR"~
$ year <dbl> 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1~
$ format <chr> "ixi", "ixi", "ixi", "ixi", "ixi", "ixi", "ixi", "ixi", "ixi"~
$ matrix <chr> "F_impacts", "F_impacts", "F_impacts", "F_impacts", "F_impact~

# Get all CO2-related stressors from F_satellite matrix
co2 <- exio |>
  filter(matrix == "F_satellite") |>
  filter(grepl("CO2|carbon dioxide", stressor, ignore.case = TRUE)) |>
  collect() # Collect data upfront to avoid connection issues

co2

# A tibble: 973,538 x 8
  stressor          region sector          value unit  year format matrix
  <chr>          <chr> <chr>          <dbl> <chr> <dbl> <chr> <chr>
1 CO2 - combustion - air AT      Cultivation o~ 2.27e8 kg    1995 ixi F_sat~
2 CO2 - combustion - air AT      Cultivation o~ 2.16e8 kg    1995 ixi F_sat~
3 CO2 - combustion - air AT      Cultivation o~ 1.01e8 kg    1995 ixi F_sat~
4 CO2 - combustion - air AT      Cultivation o~ 4.98e7 kg    1995 ixi F_sat~
5 CO2 - combustion - air AT      Cultivation o~ 1.10e7 kg    1995 ixi F_sat~
6 CO2 - combustion - air AT      Cultivation o~ 1.13e4 kg    1995 ixi F_sat~
7 CO2 - combustion - air AT      Cultivation o~ 1.74e6 kg    1995 ixi F_sat~
8 CO2 - combustion - air AT      Cattle farming 6.79e7 kg    1995 ixi F_sat~
9 CO2 - combustion - air AT      Pigs farming  4.87e7 kg    1995 ixi F_sat~
10 CO2 - combustion - air AT      Poultry farmi~ 4.62e7 kg    1995 ixi F_sat~
# i 973,528 more rows

identify which regions are the top 5 co2 emitters

co2_top5 <- co2 |>
  filter(year == 2022) |>
  group_by(region) |>
  summarize(total_co2 = sum(value, na.rm = TRUE)) |>
  arrange(desc(total_co2)) |>
  head(5)

co2_top5

# A tibble: 5 x 2
  region total_co2

```

	<chr>	<dbl>
1	CN	2.24e13
2	US	7.43e12
3	IN	5.31e12
4	WA	4.12e12
5	WM	4.11e12

filter the co2 data down to just these top countries, and plot thier total co2 emissions by year

```
library(ggplot2)

co2_top5_filtered <- co2 |>
  filter(region %in% co2_top5$region) |>
  group_by(region, year) |>
  summarize(total_co2 = sum(value, na.rm = TRUE), .groups = "drop")
co2_top5_filtered
```

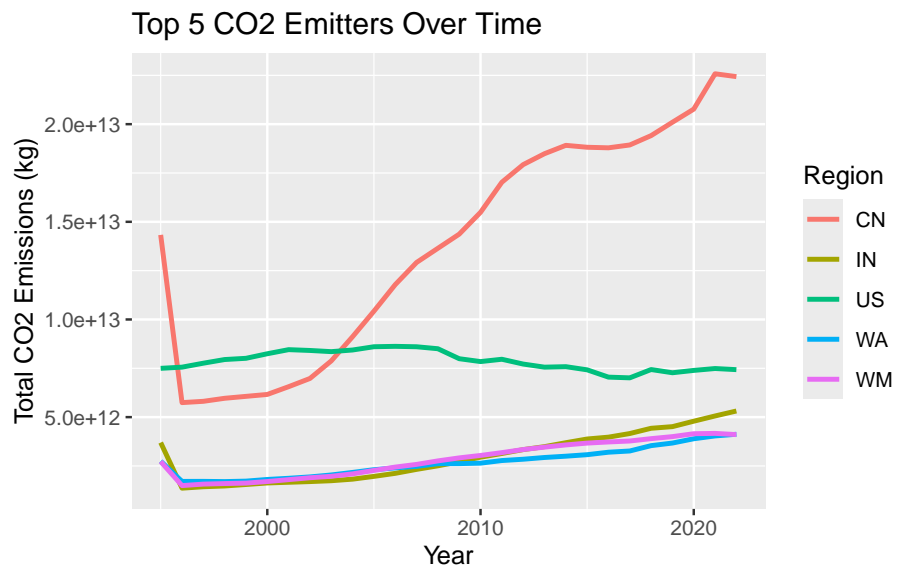
```
# A tibble: 140 x 3
  region year total_co2
  <chr>   <dbl>     <dbl>
1 CN     1995  1.43e13
2 CN     1996  5.74e12
3 CN     1997  5.81e12
4 CN     1998  5.96e12
5 CN     1999  6.06e12
6 CN     2000  6.16e12
7 CN     2001  6.56e12
8 CN     2002  6.98e12
9 CN     2003  7.88e12
10 CN    2004  9.12e12
```

# i 130 more rows

```
library(ggplot2)
p <- ggplot(co2_top5_filtered, aes(x = year, y = total_co2, color = region)) +
  geom_line(linewidth = 1) +
  labs(title = "Top 5 CO2 Emitters Over Time",
       x = "Year",
       y = "Total CO2 Emissions (kg)",
       color = "Region")
ggsave("co2_top5_plot.png", plot=p)
```

Saving 5.5 x 3.5 in image

p

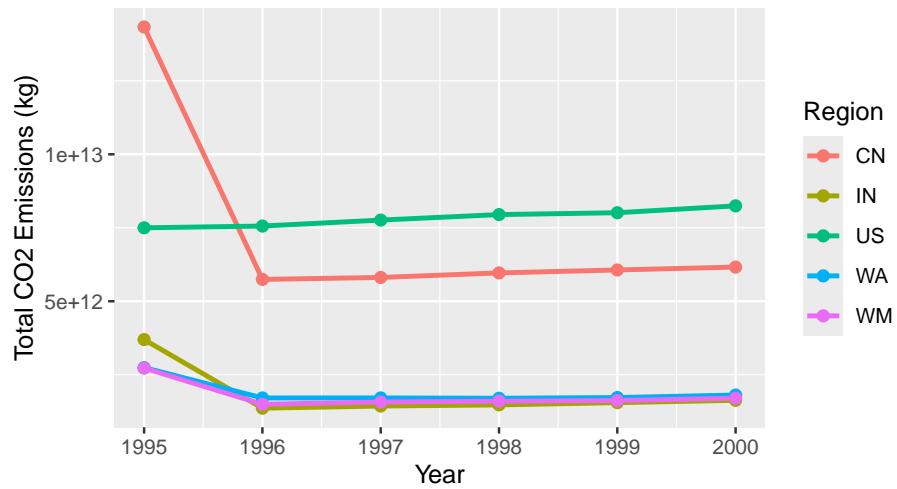


take a look at the early years of the data to see if we can identify the sudden drop pattern in the top 5 emitters

```
# Plot early years in detail
early_years <- co2 |>
  filter(region %in% co2_top5$region, year >= 1995, year <= 2000) |>
  group_by(region, year) |>
  summarize(total_co2 = sum(value, na.rm = TRUE), .groups = "drop")
ggplot(early_years, aes(x = year, y = total_co2, color = region)) +
  geom_line(linewidth = 1) +
  geom_point(size = 2) +
  labs(
    title = "CO2 Emissions: Early Years Detail (1995-2000)",
    subtitle = "Showing the sudden drop pattern",
    x = "Year",
    y = "Total CO2 Emissions (kg)",
    color = "Region"
  )
```

### CO2 Emissions: Early Years Detail (1995–2000)

Showing the sudden drop pattern



```
ggsave("co2_early_years_plot.png", plot = last_plot())
```

Saving 5.5 x 3.5 in image