

## **Importing Data in R Directly From the Internet Using APIs: Exploring Various Packages**

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**Abstract**

This handout complements the presentation on importing data in R directly from the internet using web APIs. It introduces the basic ideas behind HTTP-based APIs, shows how typical response formats such as JSON can be parsed in R, and outlines a reproducible workflow using `httr2`, `jsonlite`, and `tidyverse`. A worked example with the GitHub REST API illustrates the full pipeline from request construction to data wrangling and interpretation. The handout concludes with common pitfalls, best practices, and a short template exercise for further practice.

*Keywords:* R, APIs, data import, `httr2`, `jsonlite`, `tidyverse`

## Importing Data in R Directly From the Internet Using APIs: Exploring Various Packages

### 1 Introduction

Modern business analytics rarely relies only on static CSV or Excel files that are downloaded once and then analysed in isolation. Instead, many real-world projects require **up-to-date information** from online platforms, financial services, or open data portals. Web APIs (Application Programming Interfaces) are the standard way to access these data sources programmatically.

In the course *Data Analysis for Decision-Making (WS 2025/26)*, this topic focuses on how to use R to import data directly from the internet via web APIs. The goal of this handout is to provide a compact but complete overview that supports the presentation:

- What web APIs are and why they matter.
- How JSON responses can be turned into usable R objects.
- Which R packages are most useful for API workflows.
- A small, realistic example using the GitHub API.
- Challenges and best practices when working with APIs.
- A short template for an exercise you can adapt to other APIs.

### 2 Web APIs and Response Formats

#### 2.1 What is a web API?

A **web API** is an interface that allows one piece of software (for example, an R script) to request data from another system over the internet. Most modern APIs are **HTTP-based** and use standard methods such as:

- GET – retrieve data (read-only),
- POST – send or create data,
- PUT / PATCH / DELETE – update or delete data

APIs expose **endpoints**, which are specific URLs that provide a certain kind of information. Examples include:

- `https://api.github.com/repos/{owner}/{repo}`,
- `https://api.exchangerate.host/latest`,
- `https://api.coindesk.com/v1/bpi/currentprice.json`.

Each endpoint has documented parameters (for example, which currency to use or how many results to return) and a clear specification of what the response will look like.

## 2.2 JSON as the dominant format

Most modern APIs respond with data in **JSON (JavaScript Object Notation)**.

Conceptually, JSON is a nested structure of:

- objects (key–value pairs),
- arrays (ordered lists),
- primitive values (numbers, strings, logicals, or null).

From R's perspective, JSON is naturally represented as nested lists. A simplified example response might look like:

```
{
  "time": "2025-11-19T07:00:00Z",
  "value": 123.45,
  "meta": {
    "source": "example-api",
    "units": "index points"
  }
}
```

In an R workflow, we typically:

1. Send a request and receive the raw HTTP response.
2. Extract the body as text or as JSON.

3. Parse the JSON into R objects using **jsonlite**.
4. Convert the nested lists into a tidy tibble or data frame.

Other formats such as XML or CSV still appear in some APIs, but JSON has become the default in many modern services.

### 3 R Packages and Basic Workflow

#### 3.1 Core packages

For this topic, a small set of R packages covers most needs:

- **httr2** – a modern HTTP client for building, sending, and inspecting requests and responses.
- **jsonlite** – a robust JSON parser that converts JSON into lists or data frames.
- **dplyr**, **tidyr**, **purrr** – tidyverse tools for wrangling and transforming nested data structures.
- **ggplot2** – for visualising the results of the analysis.

These packages can be combined into a simple but powerful workflow.

#### 3.2 A generic API workflow in R

The basic pattern for working with an HTTP API in R looks like this:

The code itself is usually short. The main intellectual work lies in understanding the **structure** of the JSON and designing a tidy representation that is suitable for analysis.

### 4 Worked Example: GitHub Repository Metadata

To connect the presentation and this handout, we use the same simple example: querying the GitHub REST API for information about a public repository. GitHub exposes rich metadata about repositories, commits, issues, and more.

#### 4.1 Goal of the example

Retrieve basic metadata about the `tidyverse/ggplot2` repository and summarise it in a small tibble that can be used in further analysis.

### 4.1.1 Step 1 – Inspect the endpoint

The GitHub API documentation tells us that repository metadata is available at:

```
https://api.github.com/repos/{owner}/{repo}
```

For the repository tidyverse/ggplot2, the URL becomes:

```
https://api.github.com/repos/tidyverse/ggplot2
```

### 4.1.2 Step 2 – Request and parse in R

### 4.1.3 Step 3 – Light transformation

The tibble can be enriched with a derived variable and then simplified:

Even though this is a small example, it demonstrates the complete chain from endpoint to tidy tibble, which is the core skill required for more complex APIs.

## 5 Challenges and Best Practices

Working with APIs is powerful but also comes with recurring challenges. The most important ones for students in this course are summarised below.

### 5.1 Authentication and rate limits

Many APIs require **API keys** or OAuth tokens. Even public APIs often impose **rate limits**, which restrict how many requests can be made per minute or hour. Best practice is to:

- read the documentation carefully,
- store keys in environment variables (for example, via `.Renviron`),
- avoid hard-coding secrets in scripts or sharing them on GitHub.

### 5.2 Nested and evolving JSON structures

Deeply nested JSON can be hard to convert into a tidy format. It is helpful to:

- start by inspecting the raw structure with `str()` or `glimpse()`,

- use `purrr::map_*()` and `tidyr::unnest_*()` when lists of lists appear,
- keep wrangling code modular and well commented.

APIs sometimes change their structure over time, so code should be written in a way that makes such assumptions explicit and easy to update.

### 5.3 Error handling and reproducibility

Requests can fail due to network problems, invalid parameters, or server-side issues. The `httr2` package provides helpers such as `resp_status()` and `resp_status_desc()` to inspect responses and react accordingly.

For reproducibility, it is good practice to:

- log which endpoints and parameters were used,
- cache important responses (for example, as `.rds` files),
- document the date on which the data were retrieved.

## 6 Conclusion

Importing data in R directly from the internet using APIs is a key skill for data-driven decision-making. Once the basic ideas of HTTP requests and JSON parsing are understood, the combination of **httr2**, **jsonlite**, and the tidyverse allows analysts to build workflows that are:

- automated rather than manual,
- transparent and reproducible,
- flexible enough to incorporate different APIs and domains.

The small GitHub example in this handout is only a starting point. The same principles extend to more advanced use cases such as authenticated endpoints, scheduled data collection, interactive dashboards, or integration with Shiny applications.

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