Introduction to Web Scraping in Python

Jan Knuf July 24, 2019

Why Should You Consider Learning How to Web Scrape?

- 1) Quite simple
- 2) Sounds fancy
- 3) Might make you think differently about available data sources
- 4) Access to highly-specific information (of tailored frequency)
- 5) Potentially saves a ton of time with tedious tasks

Web Scraping as a Tool for Economic Research

- Price data: online retailers, menus of restaurants, flight and train tickets
- User-produced data (social media, online auctions, blog posts, forum entries)
- Access to natural experiments that are otherwise not documented or very tedious to digitize
- Person-specific data on career achievements, media mentions, etc.

Some Examples

- Cavallo (2018) High-frequency price data from online retailers to measure price stickiness
- Card and DellaVigna (2013) Google scholar citation counts on 95% of all econ papers to identify historical developments of publishing in econ
 - **Zimmerman (2019)** Linking career information of top executives in Chile to administrative records
 - **Lowe (2019, WP)** Identifying treated MP's in a weekly natural experiment in the British parliament to capture exogenous variation in their visibility to party leaders

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Basics

The Web Scraping Process

Fetch Download the HTML-file from the server and store it locally

Parse Search and navigate through the content and structure of the HTML file

Extract Filter out the relevant bits of the website

(Manipulate) Interact with the website

Relevant Packages

Requests Fetch the HTML-file

Ixml, html5lib Parsing engine

BeautifulSoup Navigation, Searching, (Manipulation)

re Need regex very often

Extensions & Alternatives

Selenium:

- When interaction with the website is required (entry in search form, login, load more results,...)
- Replaces requests
- Simulates a browser (web driver)

Scrapy

- Entire web scraping framework (fetch, parse, manipulate)
- Documentation here
- Program a "spider" (or bot) to automatically fetch and extract a or multiple websites
- Combines **requests** and **BeautifulSoup** (fetch and parse)

Tutorial



Constraints

Constraints

Time: In most applications you will likely parse a large number of webpages, downloading each of them and search for the objects of interest.

Solutions: Pickle HTML-files, parallelization

Irregularities: (i) the web-developer changes the structure of the HTML-file or (ii) a content producer makes (human) errors when entering the data that lead to deviations from the expected regex patterns.

Solutions: Robust algorithm, exception rules, testing

Reproducibility: As if, some stranger has access to the data-folder of your project and can add, change or delete variables and observations.

Solutions: Store HTML-files locally and make them available with code

Structure or Regex?

- Often both is necessary and helpful
- Searching by *Regex* almost always produces False Positives, if we are searching through the/multiple entire website(s)
- Searching by structure is not always stable over time / websites
- It is crucial to make the algorithm robust:
 - 1. Define a set of possible (structure) locations
 - 2. Define a set of possible regex pattern
 - 3. Define case-specific extraction rules depending on where and which pattern was found
 - 4. Carefully implement state variables and set exception rules

The Issue of Reproducibility

- The raw data can be manipulated over time or disappear
- Even if HTML-source files are stable from initial scrape to review/post publication: might
 make reproducibilty even more tedious if the data-set has to be assembled again, before
 replication analysis can be conducted

Storing HTML-source files locally and supplying them together with other data, can help. Which makes the pickling-step even more useful.

Ideally, some not manipulable format including time stamp for the original HTML-file.

References

References

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