

# Practical Web Scraping for Economists

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# Web Scraping Applications in Economics

- ▶ **Collect Pricing Data:** Analyze price fluctuations, market competition, or inflation trends. (Baye and Morgan, 2009; Boivin, Clark, and Vincent, 2012; Cavallo and Rigobon, 2016)
- ▶ **Monitor Job Listings:** Study labor market dynamics, unemployment trends, or wage changes. (Deming and Kahn, 2018; Kuhn and Shen, 2013; Kureková, Beblavý, and Thum-Thysen, 2015)
- ▶ **Gather Public Discourse Data:** Analyze public sentiment or policy content using Media/Twitter/etc. (Shamsi, 2024)
- ▶ **Compile Real Estate Information:** Study housing markets, pricing trends, and geographic economic disparities. (Bricongne, Meunier, and Pouget, 2023; Halket and Custozza, 2015)

# Web Scraping vs. APIs

- ▶ **Web Scraping:** Extracts content from web pages when no direct data access is available.
- ▶ **APIs:** Directly request data from servers; more reliable and efficient than scraping when available.
- ▶ **SERP APIs:** Retrieve search engine results (e.g., Google) without scraping, avoiding layout changes and CAPTCHAs.

# Ethical and Legal Considerations

- ▶ **Respect Terms of Service:** Many websites have specific guidelines for data usage. Scraping may violate these terms, so it's important to review them before starting.
- ▶ **Server Load:** Implement rate limiting and respect robots.txt files to avoid overwhelming servers. Use polite scraping practices by spacing requests.
- ▶ **Intellectual Property:** Some website content is protected by copyright, and scraping such data without permission can lead to legal issues.

**Disclaimer:** Please consult your department's IRB or ethics office before conducting web scraping. These materials reflect personal methods from past projects. There are certainly other, more technical resources and textbooks available.

# Types of Web Scraping: Static vs. Dynamic Pages

## ▶ **Static Pages:**

- ▶ Content is directly available in the page's source code.
- ▶ Easier since all data is loaded once the page is accessed.
- ▶ Tools like `requests` and `BeautifulSoup` are effective.

## ▶ **Dynamic Pages:**

- ▶ Data is loaded asynchronously using JavaScript after the initial page load.
- ▶ Requires simulating user interactions (e.g., clicks) or waiting for JavaScript execution.
- ▶ Tools like `Selenium` or `Scrapy+Splash` are needed to handle these pages.

**Key Challenge:** Dynamic pages require handling delays and JavaScript-rendered content, making scraping more complex.

# Static Web Scraping: Overview

**Static Web Scraping:** Extracting data from web pages where content is fixed once the page is loaded.

- ▶ All data is delivered in the initial HTML response.
- ▶ Simpler than dynamic scraping, but still requires planning.

# Steps in Static Web Scraping: Sending HTTP Requests

## Step 1: Sending an HTTP Request

- ▶ Send a request to the server using methods like GET or POST.
- ▶ **Tool:** Use requests library in Python.
- ▶ **Challenge:** Ensure valid response, avoid bot detection, handle CAPTCHAs.

## Example:

```
1 import requests
2 response = requests.get('http://example.com')
3 if response.status_code == 200:
4     html_content = response.content
```

# Steps in Static Web Scraping: Parsing the HTML

## Step 2: Parsing the HTML

- ▶ Identify and extract relevant data from the HTML.
- ▶ **Tool:** Use BeautifulSoup to navigate HTML.
- ▶ **Challenge:** Understanding complex or inconsistent HTML structures.

### Example:

```
1 from bs4 import BeautifulSoup
2 soup = BeautifulSoup(html_content, 'html.parser')
3 titles = soup.find_all('h1')
```



# Challenges in Static Scraping: Identifying HTML Elements

## Key Challenge: Identifying the Right HTML Elements

- ▶ Websites have extraneous information like ads or sidebars.
- ▶ **Approach:** Use browser dev tools to inspect HTML structure.
- ▶ Use **CSS selectors** or **XPath** for precise targeting.

## Example:

```
1     products = soup.select('div.product-name')
2     for product in products:
3         print(product.text)
```

# Challenges in Static Scraping: Paginated Content

## Key Challenge: Navigating Paginated Content

- ▶ Many websites split content across multiple pages.
- ▶ Identify URL patterns and loop through pages to scrape all data.

## Example:

```
1     base_url = 'http://example.com/page/'
2     for page_number in range(1, 10):
3         url = base_url + str(page_number)
4         response = requests.get(url)
5         soup = BeautifulSoup(response.content, 'html.
           parser')
```

# Challenges in Static Scraping: Rate Limiting and Captchas

## Key Challenge: Rate Limiting and Captchas

- ▶ Websites may throttle requests or use CAPTCHAs to block bots.
- ▶ **Approach:**
  - ▶ Use delays between requests to avoid overwhelming servers.
  - ▶ Set user-agent headers to mimic real browser behavior and reduce detection.
  - ▶ Employ proxies or rotate IPs to avoid being blocked.

## Example:

```
1 import time
2 headers = {'User-Agent': 'Mozilla/5.0'}
3 base_url = 'http://example.com/page/'
4 for page_number in range(1, 10):
5     url = base_url + str(page_number)
6     response = requests.get(url, headers=headers)
7     ...
8     time.sleep(2) # Wait 2 seconds between
                  requests
```

# Challenges in Static Scraping: Extracting Tables or Lists

## Key Challenge: Extracting Data from Tables or Lists

- ▶ Tables and lists are commonly used for structured data.
- ▶ Use BeautifulSoup to extract rows and cells from tables.
- ▶ For well-structured tables, use `pandas.read_html()` for direct extraction.

## Example:

```
1 import pandas as pd
2 tables = pd.read_html('http://example.com')
3 df = tables[0] # First table on the page
```

# Data Cleaning and Preprocessing

## Cleaning and Preprocessing the Data

- ▶ Extracted data is often messy, with extra characters or tags.
- ▶ Use string methods or regular expressions to clean the data.
- ▶ **Challenge:** Inconsistent formatting across pages.

### Example:

```
1 import re
2 clean_text = re.sub(r'\s+', ' ', text) # Replace
    multiple spaces
```

# Best Practices for Static Web Scraping

## Best Practices:

- ▶ Respect `robots.txt` and website terms of service.
- ▶ Store data in structured formats (CSV, JSON, SQL).
- ▶ Make code modular and scalable for reuse across websites.
- ▶ Regularly test and debug scraping scripts to handle site changes.

# Dynamic Web Scraping: Overview

**Dynamic Web Scraping:** Involves extracting data from websites where content is loaded dynamically via JavaScript after the initial page load.

- ▶ Essential for websites using JavaScript frameworks (React, Angular, Vue).
- ▶ Required for websites using infinite scrolling or needing user interaction.

# Steps in Dynamic Web Scraping: Simulating a Browser Environment

## Step 1: Simulating a Browser Environment

- ▶ Dynamic content requires executing JavaScript, so fetching the HTML using requests won't suffice.
- ▶ **Tool:** Use Selenium for browser automation to open a real browser and interact with the page like a user.

### Example:

```
1 from selenium import webdriver
2 driver = webdriver.Chrome()
3 driver.get('http://example.com')
```



# Steps in Dynamic Web Scraping: Waiting for JavaScript to Load

## Step 2: Waiting for JavaScript to Load

- ▶ Dynamic pages take time to load certain elements.
- ▶ **Tool:** Use Selenium's `WebDriverWait` to pause until specific elements are loaded.

### Example:

```
1      from selenium.webdriver.support.ui import
        WebDriverWait
2      from selenium.webdriver.support import
        expected_conditions as EC
3
4      element = WebDriverWait(driver , 10).until(
5          EC.presence_of_element_located((By.ID , "content
6          ")))
```

# Steps in Dynamic Web Scraping: Interacting with the Webpage

## Step 3: Interacting with the Webpage

- ▶ Simulate user interactions like clicking buttons or scrolling.
- ▶ **Tool:** Selenium can simulate these actions to load more data or reveal hidden content.

### Example:

```
1      # Clicking a button
2      button = driver.find_element(By.ID, 'load-more')
3      button.click()
4
5      # Scrolling to the bottom of the page
6      driver.execute_script("window.scrollTo(0,-document.
                             body.scrollHeight);")
```

# Steps in Dynamic Web Scraping: Extracting Data from Rendered Content

## Step 4: Extracting Data from Rendered Content

- ▶ After interactions, extract data by accessing the fully rendered page.
- ▶ **Tool:** Use Selenium with BeautifulSoup to parse the content and extract data.

### Example:

```
1 from bs4 import BeautifulSoup
2 soup = BeautifulSoup(driver.page_source, 'html.
   parser')
3 data = soup.find_all('div', class_='item')
```

# Steps in Dynamic Web Scraping: Dealing with Infinite Scrolling

## Step 5: Dealing with Infinite Scrolling

- ▶ Some sites load more content dynamically as you scroll.
- ▶ **Approach:** Use `Selenium.execute_script()` to scroll to the bottom and capture newly loaded content.

### Example:

```
1 while True:
2     driver.execute_script("window.scrollTo(0, -
        document.body.scrollHeight);")
3     time.sleep(3) # Wait for new content to load
```

# Challenges in Dynamic Web Scraping: Handling JavaScript-Rendered Content

## Key Challenge: Handling JavaScript-Rendered Content

- ▶ JavaScript renders content after the initial page load, so traditional scraping methods don't work.
- ▶ **Approach:** Use Selenium to render the page fully.
- ▶ **Challenges:**
  - ▶ Frameworks like React or Angular may have delayed element loading.
  - ▶ Selenium can be resource-intensive and slow for large-scale scraping.

# Challenges in Dynamic Web Scraping: Timing and Delays

## Key Challenge: Timing and Delays

- ▶ Dynamic elements load at different times, making proper timing critical.
- ▶ **Approach:**
  - ▶ Use explicit waits with `WebDriverWait` for specific elements.
  - ▶ Set general wait times using implicit waits to balance speed and completeness.

## Example:

```
1 driver.implicitly_wait(10) # Sets a 10-second max  
   wait for all elements
```

# Challenges in Dynamic Web Scraping: Simulating User Behavior

## Key Challenge: Simulating User Behavior

- ▶ Websites may require interaction, such as clicking buttons or handling pop-ups, to reveal data.
- ▶ **Approach:** Use Selenium to simulate user interactions like clicks and form submissions.
- ▶ **Example:**

```
1         # Handle a pop-up modal
2         modal = driver.find_element(By.CLASS_NAME,
3                                     'close-modal')
4         modal.click()
```

# Challenges in Dynamic Web Scraping: Handling CAPTCHAs and Anti-Scraping Measures

## Key Challenge: CAPTCHAs and Anti-Scraping Measures

- ▶ Many dynamic websites use CAPTCHAs and rate-limiting to prevent scraping.
- ▶ **Approach:**
  - ▶ Use third-party services for CAPTCHA solving or avoid triggering CAPTCHAs.
  - ▶ Rotate proxies and use User-Agent spoofing to simulate a real user.

## Example:

```
1 from selenium.webdriver.chrome.options import
   Options
2 options = Options()
3 options.headless = True
4 options.add_argument('user-agent=Mozilla/5.0')
5 driver = webdriver.Chrome(options=options)
```



# Closing the Browser in Selenium

**Use of** `driver.quit()` **or** `driver.close()`:

- ▶ It is important to close the browser after completing the task to free up system resources.
- ▶ Without closing, the browser will continue running in the background, consuming memory and CPU.
- ▶ `driver.quit()` ends the entire browser session, while `driver.close()` closes the current window.

**Example:**

```
1      driver.quit()  # Close the browser and end the  
           session
```

# Best Practices for Dynamic Web Scraping

## Best Practices:

- ▶ **Use APIs:** If available, APIs provide structured data directly.
- ▶ **Optimize Scraping Speed:** Use headless browsers and disable unnecessary features like images.
- ▶ **Use Proxies and Rotate User Agents:** Avoid being blocked by rotating IPs and user agents.
- ▶ **Error Handling:** Implement error handling and logging to prevent script crashes.
- ▶ **Respect Website Policies:** Follow the website's `robots.txt` and terms of service.

# Scheduled Web Scraping

## Scheduled scraping is crucial for researchers to:

- ▶ **Price Monitoring:** Track price changes and market trends on e-commerce platforms.
- ▶ **Labor Market Studies:** Analyze wage dynamics and employment trends from job boards.
- ▶ **Housing Market Analysis:** Monitor property prices and rent trends from real estate sites.
- ▶ **Consumer Sentiment:** Scrape social media and reviews to study behavior and firm reputation.
- ▶ **Macro-Economic Indicators:** Collect real-time economic data like GDP and unemployment rates.

# Scheduling and Automating Web Scraping Tasks

There are several methods to schedule and automate web scraping efficiently:

- ▶ **Cron Jobs (Linux/macOS) or Task Scheduler (Windows):** Run scripts at specific intervals (daily, weekly) using system tools like Cron or Task Scheduler.
- ▶ **Python Libraries (e.g., `schedule`, `APScheduler`):** Schedule tasks directly in Python code. `schedule` runs tasks at intervals, while `APScheduler` provides advanced time-based scheduling.
- ▶ **Cloud Services (e.g., `AWS Lambda`, `Google Cloud Functions`):** Run scraping scripts in the cloud without managing servers. Trigger tasks based on time or events.
- ▶ **Others:** Docker containers, CI/CD pipelines, or third-party automation platforms like Zapier or Integromat.

# Questions?

Thanks for your attention!

If you have any questions, concerns, or just want to chat about  
web scraping (or coffee),  
feel free to reach out!

**[m.shamsi@lse.ac.uk](mailto:m.shamsi@lse.ac.uk)**

(No bots, please!)

# Further Reading on Web Scraping

- ▶ Boegershausen et al., 2022
- ▶ Śpiewanowski, Talavera, and Vi, 2022
- ▶ Jarmin, 2019
- ▶ Edelman, 2012

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