



Data wrangling, data quality

Web content acquisition and extraction

Pierre Senellart



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Outline

Crawling the Web

- Discovering new URLs

- Identifying duplicates

- Crawling architecture

Crawling complex content

Structured Web content extraction

Conclusion



Web Crawlers

- **crawlers, (Web) spiders, (Web) robots**: autonomous user agents that retrieve pages from the Web
- Basics of crawling:
 1. Start from a given URL or set of URLs
 2. Retrieve and process the corresponding page
 3. Discover new URLs (cf. next slide)
 4. Repeat on each found URL
- No real termination condition (virtual unlimited number of Web pages!)
- **Graph-browsing** problem
 - deep-first**: not very adapted, possibility of being lost in **robot traps**
 - breadth-first**
 - combination of both**: breadth-first with limited-depth deep-first on each discovered website



Sources of new URLs

- From HTML pages:
 - hyperlinks `...`
 - media `` `<embed src="...">`
`<object data="...">`
 - frames `<frame src="...">` `<iframe src="...">`
 - JavaScript links `window.open("...")`
 - etc.
- Other hyperlinked content (e.g., PDF files)
- Non-hyperlinked URLs that appear anywhere on the Web (in HTML text, text files, etc.): use regular expressions to extract them
- Referrer URLs
- Sitemaps [sitemaps.org, 2008]



Scope of a crawler

- Web-scale
 - The Web is infinite! Avoid robot traps by putting depth or page number **limits** on each Web server
 - Focus on **important** pages [Abiteboul et al., 2003]
- Web servers under a list of **DNS domains**: easy filtering of URLs
- A given topic: **focused crawling** techniques [Chakrabarti et al., 1999, Diligenti et al., 2000, Gouriten et al., 2014] based on classifiers of Web page content and predictors of the interest of a link.
- The national Web (cf. **public deposit**, national libraries): what is this? [Abiteboul et al., 2002]
- A given Web site: what is a Web site? [Senellart, 2005]



A word about hashing

Definition

A **hash function** is a deterministic mathematical function transforming objects (numbers, character strings, binary...) into fixed-size, seemingly random, numbers. The more random the transformation is, the better.

Example

Java hash function for the `String` class:

$$\sum_{i=0}^{n-1} s_i \times 31^{n-i-1} \bmod 2^{32}$$

where s_i is the (Unicode) code of character i of a string s .



Identification of duplicate Web pages

Problem

Identifying duplicates or near-duplicates on the Web to prevent multiple indexing

trivial duplicates: same resource at the same **canonized** URL:

`http://example.com:80/toto`

`http://example.com/titi/../toto`

exact duplicates: identification by **hashing**

near-duplicates: (timestamps, tip of the day, etc.) more complex!



Near-duplicate detection

Edit distance. Count the **minimum number of basic modifications** (additions or deletions of characters or words, etc.) to obtain a document from another one. Good measure of similarity, and can be computed in $O(mn)$ where m and n are the size of the documents. But: **does not scale** to a large collection of documents (unreasonable to compute for every pair!).

Shingles. Idea: two documents similar if they mostly share the same **succession of k -grams** (succession of tokens of length k).

Example

I like to watch the sun set with my friend.

My friend and I like to watch the sun set.

$S = \{\text{i like, like to, my friend, set with, sun set, the sun, to watch, watch the, with my}\}$

$T = \{\text{and i, friend and, i like, like to, my friend, sun set, the sun, to watch, watch the}\}$



Hashing shingles to detect duplicates [Broder et al., 1997]

- Similarity: **Jaccard coefficient** on the set of shingles:

$$J(S, T) = \frac{|S \cap T|}{|S \cup T|}$$

- Still **costly to compute!** But can be approximated as follows:
 1. Choose N **different hash functions**
 2. For each hash function h_i and each set of shingles $S_k = \{s_{k1} \dots s_{kn}\}$, store $\phi_{ik} = \min_j h_i(s_{kj})$
 3. Approximate $J(S_k, S_l)$ as the **proportion** of ϕ_{ik} and ϕ_{il} that are equal
- Possibly to repeat in a hierarchical way with **super-shingles** (we are only interested in **very** similar documents)



Crawling ethics

- Standard for robot exclusion: **robots.txt** at the root of a Web server [Koster, 1994].

```
User-agent: *
```

```
Allow: /searchhistory/
```

```
Disallow: /search
```

- Per-page exclusion.

```
<meta name="ROBOTS" content="NOINDEX,NOFOLLOW">
```

- Per-link exclusion.

```
<a href="toto.html" rel="nofollow">Toto</a>
```

- Avoid **Denial Of Service** (DOS), wait ≈ 1 s between two repeated requests to the same Web server



Legal aspects (France) – 1/2

- General principles:
 - to access or keep access to a “system for automated data processing” *in a fraudulent manner* is punished of two years of prison and 60,000 euros fine (Code pénal 323-1, modified by law 2015-912 on “Renseignement”)
 - to disrupt the functioning of a “system for automated data processing” is punished of five years of prison and 150,000 euros fine, extended to seven years and 300,000 euros when the system is a public one containing personal information (Code pénal 323-2, modified by law 2015-912 on “Renseignement”)
- A Web site hosted in a different country may invoke completely different legal principles, under a different jurisdiction
- Crawling content can be considered accessing and keeping access to a “system for automated data processing” (Cour d’appel de Paris, 5 February 2014, “Bluetouff case”)



Legal aspects (France) – 2/2

- robots.txt files are a de facto standard, and instructions in robots.txt files a receivable way to specify what can be crawled (Cour d'appel de Paris, 26 January 2011, Google vs SAIF)
- Frequent requests to a Web site can be considered as a way to disrupt the functioning of a “system for automated data processing” (Cour d'appel de Bordeaux, 15 November 2011, Cédric M. vs C-Discount), but only if it reaches abusive levels and can be shown to have cause disruption
- Web content is subject to “droit d'auteur” (Code de la propriété intellectuelle, Première partie, Livre 1er) and cannot generally be broadcast by third-parties; only transient copies are allowed (CJEU, 5 June 2014, PRCA vs NLA)
- Web content containing personal data is even more sensitive (GDPR): personal data should be collected for a specific purpose, kept updated, and protected

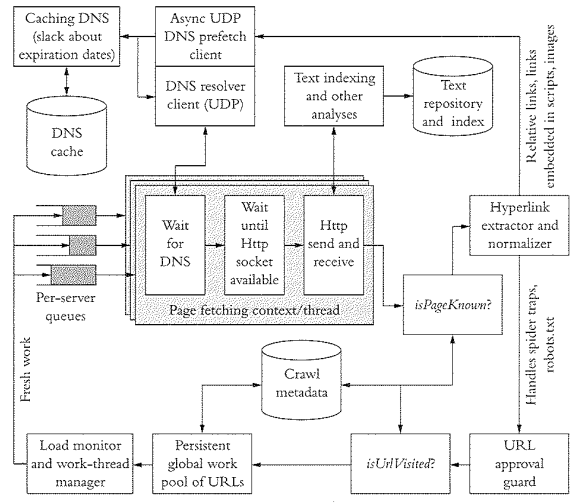


Parallel processing

Network delays, waits between requests:

- **Per-server queue** of URLs
- Parallel processing of requests to different hosts:
 - **multi-threaded** programming
 - **asynchronous** inputs and outputs (`select`, classes from `java.util.concurrent`): less overhead
- Use of **keep-alive** to reduce connexion overheads

General Architecture [Chakrabarti, 2003]





Refreshing URLs

- Content on the Web **changes**
- Different **change rates**:
 - online newspaper main page: every hour or so
 - published article: virtually no change
- **Continuous** crawling, and identification of change rates for **adaptive** crawling



Importance of Timely Crawling

- The Web is very **volatile**, with a typical half-life of URLs of a few years [Koehler, 2003]
- For many purposes (archiving, analytics), a crawl quality can be measured by its **temporal coherence** [Spaniol et al., 2009]
- Ideally, Web pages pointed to by a Web page should be crawled **at the same time**. Unrealistic in practice.
- Crawling **takes time** and **consumes resources**:
 - **Limited bandwidth**, limiting computing power on the crawling side
 - Because of crawling ethics, crawling a 5 million page site takes around **2 months**!
 - Limitations of social networking APIs **drastic**: on Twitter, using the Search API, at most 3 000 tweets per minute; using the Timeline API, at most 20 000 tweets per minute. . . (and 2 million per day)



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Outline

Crawling the Web

Crawling complex content

- Modern Web Sites

- CMS-based Web Content

- Social Networking Sites

- The Deep Web

Structured Web content extraction

Conclusion



Crawling Modern Web Sites

- Some modern Web sites only work when cookies are activated (**session cookies**), or when **JavaScript code** is interpreted
- Regular Web crawlers (**wget**, **Heritrix**, **Apache Nutch**) do not usually perform any cookie management and do not interpret JavaScript code
- Crawling of some Websites therefore require more **advanced tools**



Advanced crawling tools

Web scraping frameworks such as **scrapy** (Python) or **WWW::Mechanize** (Perl) simulate a Web browser interaction and cookie management (but no JS interpretation)

Headless browsers such as **htmlunit** simulate a Web browser, including simple JavaScript processing

Browser instrumentors such as **Selenium** allow full instrumentation of a regular Web browser (Chrome, Firefox, Internet Explorer)

Proxys such as **mitmproxy** capable of recording and replaying a complex set of HTTP requests

OXPath: a **full-fledged navigation and extraction language** for complex Web sites [Sellers et al., 2011]



Templated Web Site

- Many Web sites (especially, Web forums, blogs) use one of a few **content management systems** (CMS)
- Web sites that use the same CMS will be **similarly structured**, present a similar layout, etc.
- Information is **somewhat structured** in CMSs: publication date, author, tags, forums, threads, etc.
- **Some structure differences** may exist when Web sites use different versions, or different themes, of a CMS





Crawling CMS-Based Web Sites

- Traditional crawling approaches crawl Web sites **independently** of the nature of the sites and of their CMS
- When the CMS is known:
 - Potential for much more **efficient crawling strategies** (avoid pages with redundant information, uninformative pages, etc.)
 - Potential for **automatic extraction** of structured content
- Two ways of approaching the problem:
 - Have a **handcrafted knowledge base** of known CMSs, their characteristics, how to crawl and extract information [Faheem and Senellart, 2013b,a] (AAH)
 - **Automatically infer** the best way to crawl a given CMS [Faheem and Senellart, 2014] (ACE)
- Need to be **robust** w.r.t. template change

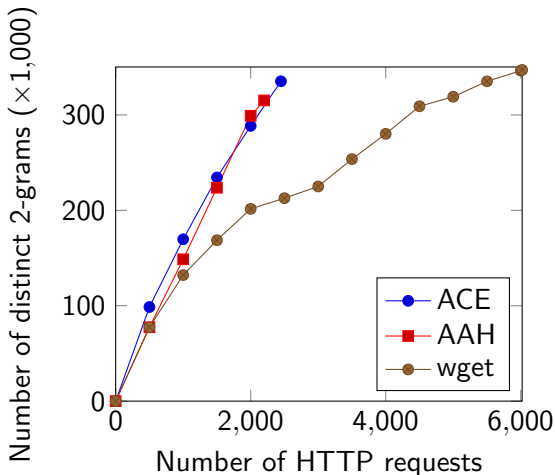


Detecting CMSs

- One main challenge in intelligent crawling and content extraction is to identify the CMS and then perform the **best crawling strategy** accordingly
- Detecting CMS using:
 1. URL patterns,
 2. HTTP metadata,
 3. textual content,
 4. XPath patterns, etc.
- These can be manually described (AAH), or automatically inferred (ACE)
- For instance the **vBulletin** Web forum content management system, that can be identified by searching for a reference to a `vbulletin_global.js` JavaScript script by using a simple `//script/@src` XPath expression.



Crawling <http://www.rockamring-blog.de/> [Faheem and Senellart, 2014]





Social data on the Web

Huge numbers of users of social networking sites (2019):

Facebook 2.4 billion

YouTube 2.0 billion

Instagram 1.0 billion

QZone 554 million

Weibo 486 million

Reddit 330 million

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Huge volume of shared data:

500 million tweets per day on Twitter (6,000 per second on average!). . .

. . . including statements by heads of states, revelations of political activists, etc.



Crawling Social Networks

- Theoretically possible to crawl social networking sites using a **regular Web crawler**
- Sometimes not possible:
`https://www.facebook.com/robots.txt`
- Often **very inefficient**, considering politeness constraints
- Better solution: Use provided social networking APIs
`https://developer.twitter.com/en/docs/api-reference-index`
`https://developers.facebook.com/docs/graph-api/reference/`
`https://developer.linkedin.com/apis`
`https://developers.google.com/youtube/`
- Also possible to buy access to the data, directly from the social network or from brokers such as `http://gnip.com/`



Social Networking APIs

- Most social networking Web sites (and some other kinds of Web sites) provide **APIs** to effectively access their content
- Usually a **RESTful** API, occasionally SOAP-based
- Usually require a **token** identifying the application using the API, sometimes a cryptographic signature as well
- May access the API as an authenticated user of the social network, or as an **external party**
- APIs seriously limit the **rate of requests**:
`https://developer.twitter.com/en/docs/tweets/search/api-reference/get-search-tweets`



REST

- Mode of interaction with a **Web service**
- Follow the KISS (**Keep it Simple, Stupid**) principle
- Each request to the service is a **simple HTTP GET method**
- Base URL is the **URL of the service**
- Parameters of the service are sent as **HTTP parameters** (in the URL)
- **HTTP response code** indicates success or failure
- Response contains **structured output**, usually as JSON or XML
- **No side effect**, each request independent of previous ones



The Case of Twitter

- Two main APIs:
 - **REST APIs**, including search, getting information about a user, a list, followers, etc.
`https://dev.twitter.com/docs/api/1.1`
 - **Streaming API**, providing real-time result
- **Very limited history** available
- Search can be on **keywords**, **language**, **geolocation** (for a small portion of tweets)



Cross-Network Crawling

- Often useful to combine results from **different social networks**
- Numerous libraries facilitating SN API accesses (twipy, Facebook4J, FourSquare VP C++ API. . .) **incompatible with each other**. . . Some efforts at generic APIs (OneAll, APIBlender [Gouriten and Senellart, 2012])
- **Example use case**: No API to get all check-ins from FourSquare, but a number of check-ins are available on Twitter; given results of Twitter Search/Streaming, use FourSquare API to get information about check-in locations.



The Deep Web

Definition (Deep Web, Hidden Web, Invisible Web)

All the content on the Web that is not directly accessible through **hyperlinks**. In particular: HTML forms, Web services.



Size estimate: 500 times more content than on the **surface Web!**
 [BrightPlanet, 2000]. Hundreds of thousands of deep Web
 databases [Chang et al., 2004]



Sources of the Deep Web

Example

- *Yellow Pages* and other directories;
- Library catalogs;
- Weather services;
- US Census Bureau data;
- etc.



Discovering Knowledge from the Deep Web [Nayak et al., 2012]

- Content of the deep Web hidden to classical Web search engines (they just follow links)
- But very valuable and high quality!
- Even services allowing access through the surface Web (e.g., e-commerce) have more semantics when accessed from the deep Web
- How to **benefit** from this information?
- How to **analyze**, **extract** and **model** this information?

Focus here: Automatic, unsupervised, methods, for a given domain of interest



Notes on the Extensional Approach

- Main issues:
 - Discovering services
 - Choosing appropriate data to submit forms
 - Use of data found in result pages to bootstrap the siphoning process
 - Ensure good coverage of the database
- Approach **favored by Google**, used in production [Madhavan et al., 2006]
- Not always feasible (huge load on Web servers)

Intensional Approach



discovery

Google Scholar **Advanced Scholar Search** [Advanced Search Tips](#) | [About Google Scholar](#)

Find articles	with all of the words	<input type="text"/>	10 results	<input type="button" value="Search Scholar"/>
	with the exact phrase	<input type="text"/>		
	with at least one of the words	<input type="text"/>		
	without the words	<input type="text"/>		
	where my words occur	anywhere in the article		
Author	Return articles written by	<input type="text"/>		
		e.g., "P.J. Hayes" or "McCarthy"		
Publication	Return articles published in	<input type="text"/>		
		e.g., J Biol Chem or Nature		
Date	Return articles published between	<input type="text"/> - <input type="text"/>		
		e.g., 1996		

probing

Form wrapped as
a Web service

analyzing

query



Google Scholar [Web](#) [Images](#) [Video](#) [Maps](#) [Books](#) [More](#) [Advanced Scholar Search](#)
[Help](#) [Feedback](#)

Scholar All articles - **Recent articles** Results 1 - 18 of about 91,000,000 for *data* [Advanced] [8.78 seconds]

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PDK2 2-deficient mice lack male fertility owing to inability to initiate Y-chromosome
28 February 2004; available via Internet at [http://www.pdk2.org.uk/](#) [Accessed 10 Feb 2004].
-> Both genetic and biochemical data point toward a physiological role for this complex in the cellular insulin-signaling activity in YSD, insulin resistance <<
[Cited in 122] - Related articles - Scholar - Web Search - Web Search

Bayesian data analysis and model selection
28 February 2004; available via Internet at [http://www.bayesian.org.uk/](#) [Accessed 10 Feb 2004].
-> Both genetic and biochemical data point toward a physiological role for this complex in the cellular insulin-signaling activity in YSD, insulin resistance <<
[Cited in 122] - Related articles - Scholar - Web Search - Web Search

Data mining: the art of machine learning and techniques with Java implementations - [webinfo.ec.nyu.edu](#)
28 February 2004; available via Internet at [http://www.webinfo.ec.nyu.edu/](#) [Accessed 10 Feb 2004].
-> Both genetic and biochemical data point toward a physiological role for this complex in the cellular insulin-signaling activity in YSD, insulin resistance <<
[Cited in 122] - Related articles - Scholar - Web Search - Web Search



Notes on the Intensional Approach

- More **ambitious** [Chang et al., 2005, Senellart et al., 2008]
- Main issues:
 - Discovering services
 - Understanding the structure and semantics of a form
 - Understanding the structure and semantics of result pages
 - Semantic analysis of the service as a whole
 - Query rewriting using the services
- No significant load imposed on Web servers



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Languages for extraction

- Based on serialization: regular expressions
- Based on DOM:

DOM navigation expresses local navigation in the DOM, from a node to its parent, its children, its attribute, etc. Standard API [W3C] but variations.

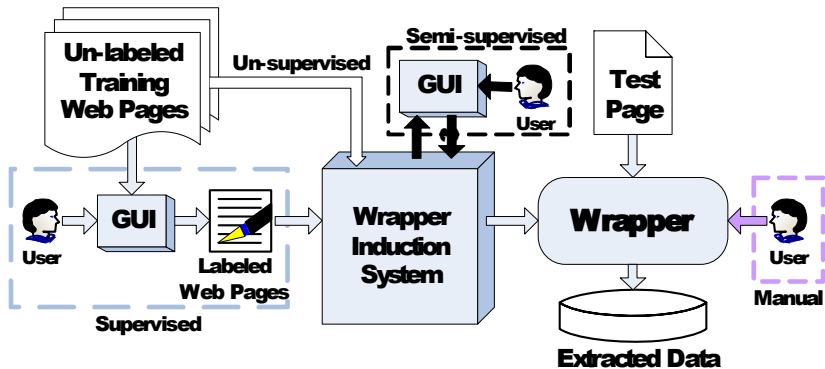
searching elements by tag names, identifiers, names, class names

CSS selectors

XPath



Wrapper induction [Chang et al., 2006]





Supervised, semi-supervised, and domain-based techniques

- Many academic approaches and systems
- No ready-to-use free software for supervised and semi-supervised extraction (as far as I know)



Unsupervised techniques

- Exploiting data redundance within a page [Liu et al., 2004] or across pages [Crescenzi et al., 2001, Arasu and Garcia-Molina, 2003]
- RoadRunner: freely downloadable and existing demos at <http://www.dia.uniroma3.it/db/roadRunner/>



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Conclusion

What you should remember

- Crawling as a **graph-browsing** problem.
- **Shingling** for identifying duplicates.
- Numerous **engineering issues** in building a Web-scale crawler.
- Crawling modern Web content is **not as easy** as launching a traditional Web crawler
- Often critical to **focus the crawl** towards content of interest
- Ideally: a traditional large-scale crawler that knows **when to delegate** to more specialized crawling mechanisms (tools querying social networking APIs, deep Web crawlers, JS-aware crawlers, etc.)
- Huge variety of tools, techniques, suitable for different needs



References

Free software

wget simple yet effective Web spider

Heritrix Web-scale highly configurable Web crawler, used by the Internet Archive

Beautiful Soup Python module for parsing real-world Web pages

Scrapy rich Python module for Web crawling and content extraction

Selenium browser instrumentor, with API in several languages

To go further

- A good textbook [Chakrabarti, 2003]
- Main references:
 - HTML 4.01 recommendation [W3C, 1999]
 - HTTP/1.1 RFC [IETF, 1999]

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