DATA COLLECTION

Petra Isenberg

Slides originaly by WESLEY WILLETT

VISUAL ANALYTICS

WHERE DOES DATA COME FROM?

We tend to think of data as a thing...

in a database...

somewhere...

WHY DO YOU NEED DATA?

(HINT: Usually, because you have a question you need to answer!)

DATA ANSWERS

ANALYSIS IS A CYCLE

GATHERING DATA,
APPLYING STATISTICAL TOOLS, AND
CONSTRUCTING GRAPHICS TO
ADDRESS QUESTIONS



INSPECT "ANSWERS" AND ASSESS <u>NEW</u> QUESTIONS

(SOMETIMES YOU'LL ALREADY START WITH DATA...)

"EXPLORATORY DATA ANALYSIS"



We already saw this...

(...BUT OFTEN YOU <u>START</u> WITH A QUESTION AND NEED TO COLLECT DATA TO FIT IT)

CHOOSING A QUESTION

"How has language evolved over time?"

"What will the weather be like next month?"

"Are the right people seeing my advertisements?"

"What is the current temperature?"

A PROBLEM OF SCALE

"How has language problems over time?"
TO FIND DATA

"What will the weathe be like next month?"

"Are the right people see g my advertisements?"

"What is the NOTAS mperature?" INTERESTING

HOW TO OBTAIN DATA?

COLLECT IT

- OBSERVATION
- SURVEYS
- LOGGING
- SENSORS
- CROWDSOURCING

FIND OR EXTRACT IT

- OPEN CORPUSES
- DATA RETAILERS
- APIS
- SCRAPING THE WEB

GENERATE IT

- SIMULATIONS

PROS/CONS

THIS LIST IS NOT EXHAUSTIVE

This lecture is intended to expose you to just a few useful data sources and collection methods.

COLLECTING DATA

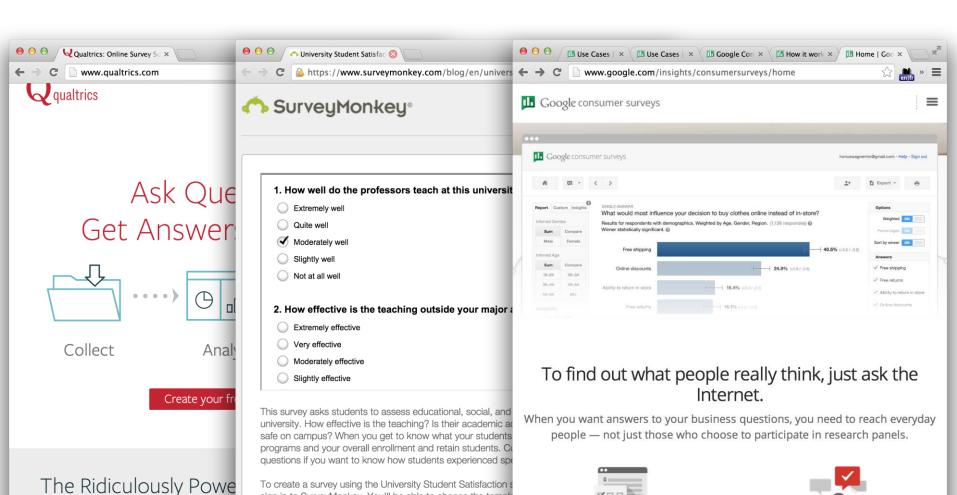
Choosing the best way to capture information you need.

SURVEYS

Paper surveys / In person interviews

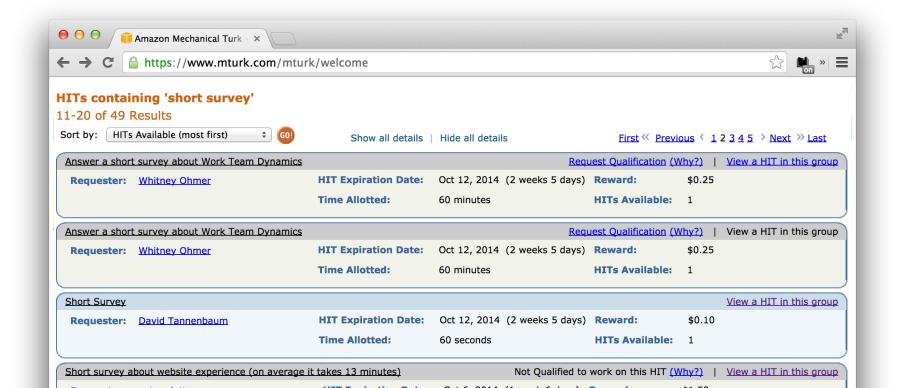
STILL ONE OF THE BEST WAYS TO GET DETAILED DATA OR DATA ABOUT SENSITIVE SUBJECTS

SURVEYS ONLINE



sign in to Curroy Mankoy, Vou'll be able to aboos the tomple

CROWDSOURCING DATA COLLECTION

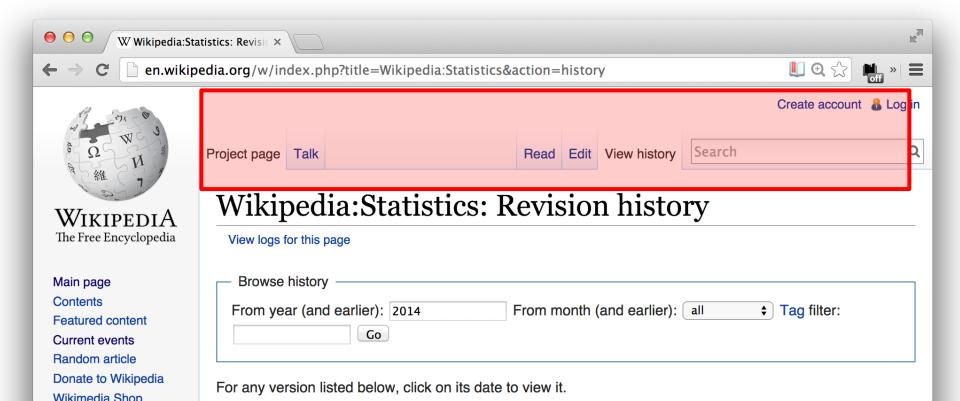


WEB LOGGING

Tracking Visits, Click-Throughs, and Traffic Patterns and other measures of User Activity.

- Google Analytics
- Open Web Analytics
- and many others...

EDITS & ACCESSS LOGS ON WIKIPEDIA



SENSORS

- Weather stations
- Personal activity trackers
- Cameras
- Mobile phones







HOW TO OBTAIN DATA?

COLLECT IT

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- CROWDSOURCING

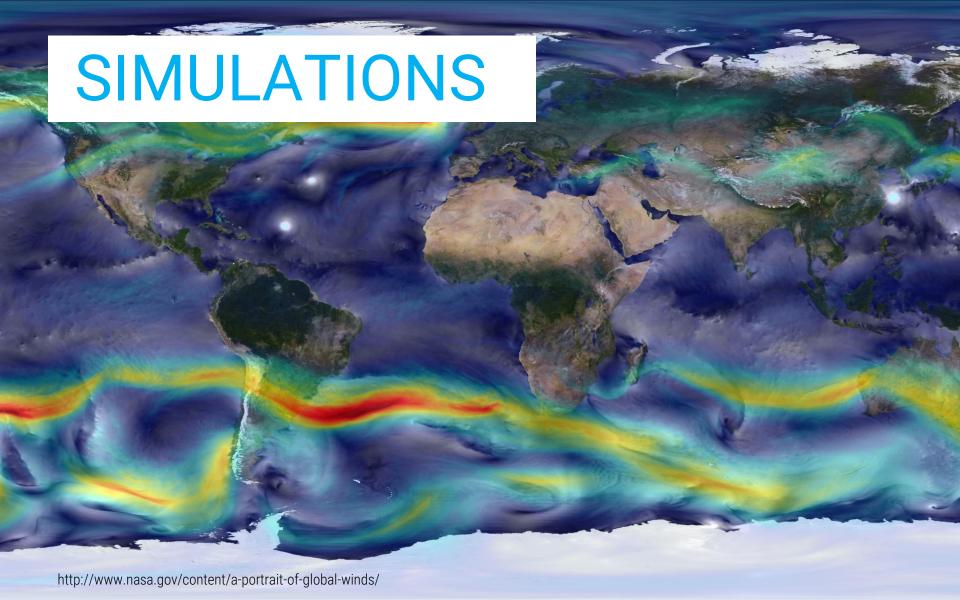
FIND OR EXTRACT IT

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GENERATE IT

- SIMULATIONS

GENERATING DATA











Is It Better to Rent or Buy?

By MIKE BOSTOCK, SHAN CARTER and ARCHIE TSE

The choice between buying a home and renting one is among the biggest financial decisions that many adults make. But the costs of buying are more varied and complicated than for renting, making it hard to tell which is a better deal. To help you answer this question, our calculator takes the most important costs associated with buying a house and computes the equivalent monthly rent. RELATED ARTICLE

Home Price

A very important factor, but not



If you can rent a similar home for less than ...

HOW TO OBTAIN DATA?

COLLECT IT

- OBSERVATION
- SURVEYS
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FIND OR EXTRACT IT

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GENERATE IT

- SIMULATIONS

FINDING AND EXTRACTING EXISTING DATA

LARGE OPEN CORPUSES

DBPEDIA



About: Iceland

An Entity of Type: place, from Named Graph:

http://dbpedia.org, within Data Space : dbpedia.org

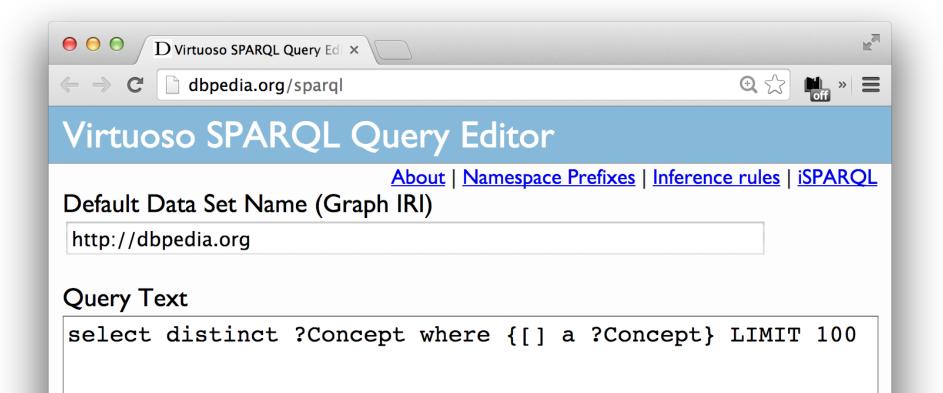


Iceland /aɪslənd/ (Icelandic: Ísland ['istlant]), sometimes referred to in full as the Republic of Iceland (Lýðveldið Ísland), is a Nordic island country marking the juncture between the North Atlantic and the Arctic Ocean, on the Mid-Atlantic Ridge. The country has a population of 325,671 and a total area of 103,000 km2 (40,000 sq mi), which makes it the most sparsely populated country in Europe.

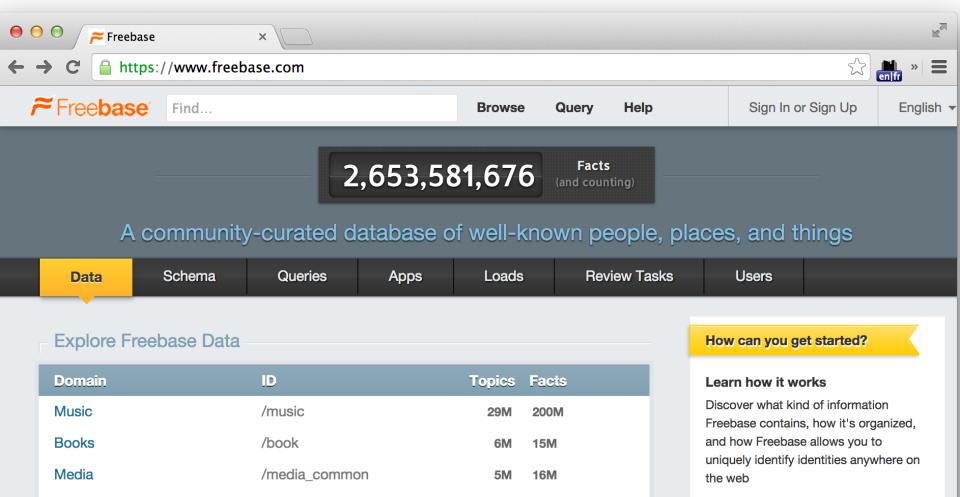
Property

Value

QUERYING DBPEDIA



FREEBASE



WIKIDATA



Main page

Community portal

Project chat

Create a new item

Item by title

Recent changes

Random item

Help

Donate

Print/export

Create a book Download as PDF

Printable version

Tools

What links here

Related changes

Special pages

Permanent link

Page information

Main Page Discussion





Wikidata is a free linked database that can be read and edited by both humans and machines.

Wikidata acts as central storage for the **etructured data** of its Wikimedia sister projects including Wikinedia



New to the wonderful world out to speed and feeling com

PROJECT GUTENBERG



All Books (sorted by popularity)



Sort Alphabetically



Sort by Release Date



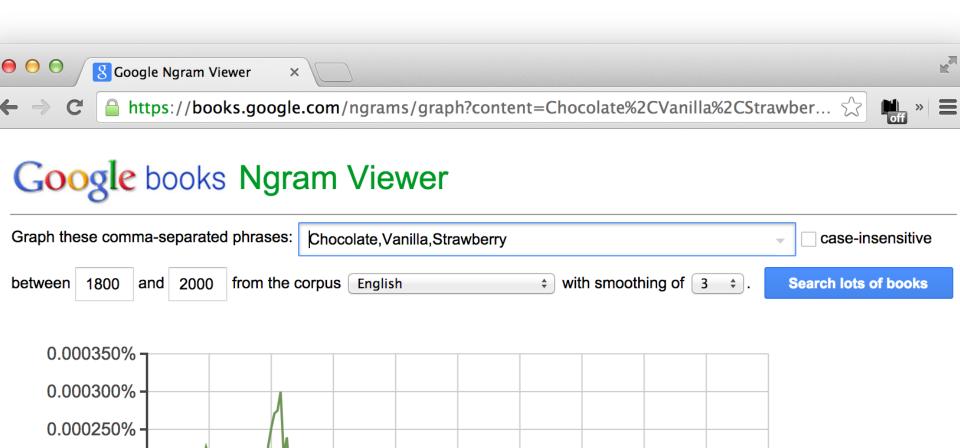
The Kama Sutra of Vatsyayana

Vatsyayana

13285 downloads

GOOGLE N-GRAMS

0.000200%

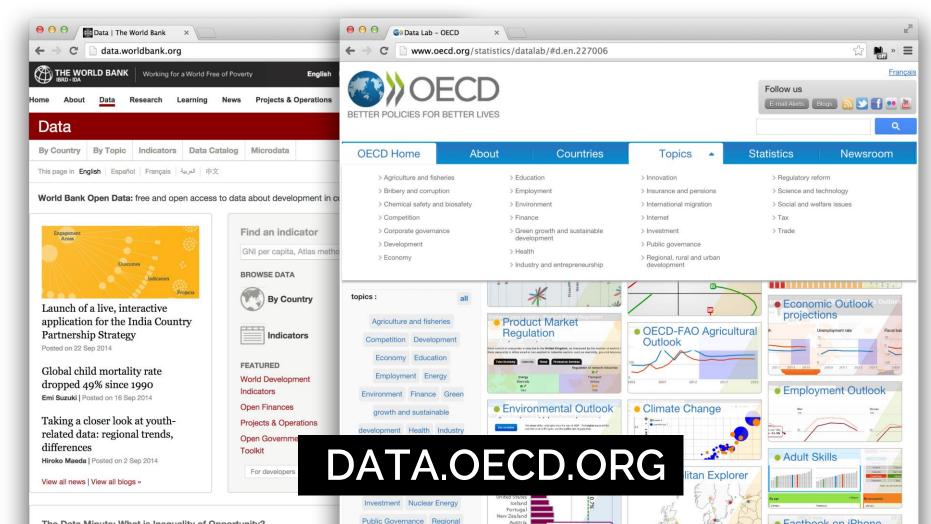


Chocolate

FINDING AND EXTRACTING EXISTING DATA

GOVERNMENT AND INTERNATIONAL DATA INITIATIVES

DATA.WORLDBANK.ORG

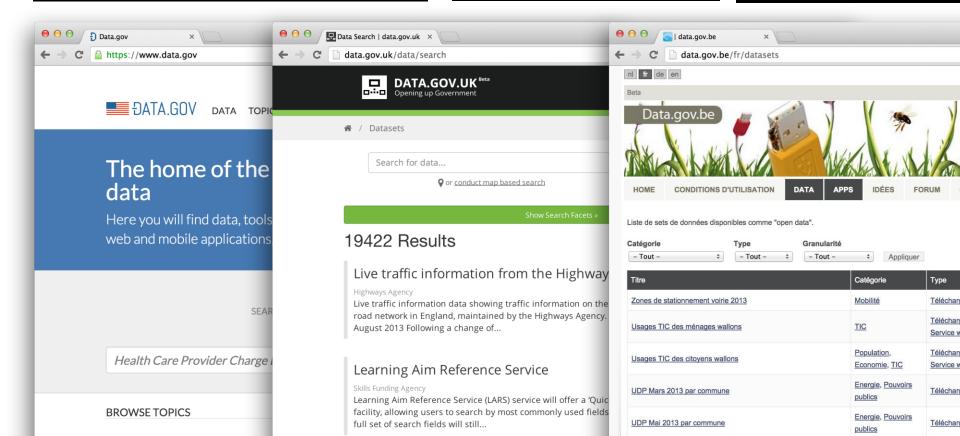


GOVERNMENT INITIATIVES

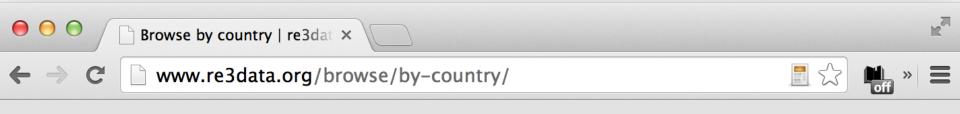
WWW.DATA.GOV (US)

DATA, GOV, UK

DATA.GOV.B



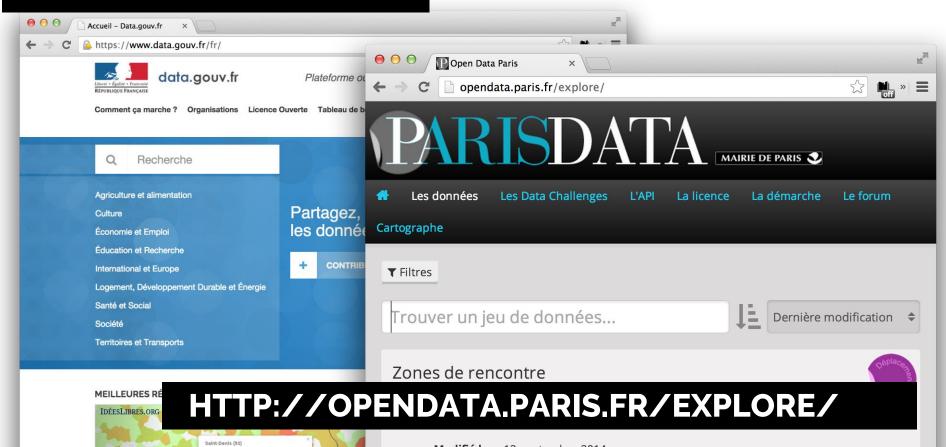
NEW DATA INITIATIVES JUST TO TRACK ALL THE DATA INITIATIVES





INITIATIVES IN FRANCE

HTTP://DATA.GOUV.FR



FINDING AND EXTRACTING EXISTING DATA

OTHER PUBLIC DATA REPOSITORIES

MORE REPOSITORIES OF PUBLIC DATA SETS

VISUALIZING.ORG

http://visualizing.org/data/browse

AMAZON PUBLIC DATA HOSTING

http://aws.amazon.com/publicdatasets/

GOOGLE PUBLIC DATA

http://www.google.com/publicdata/directory

KAGGLE

https://www.kaggle.com/

FINDING AND EXTRACTING EXISTING DATA

DATA RETAILERS

DATA RETAILERS

FACTUAL

http://www.factual.com/

Dawex

https://www.dawex.com/en/

Datamean

https://datmean.com/

Weather stations, ...

AND AGAIN, THERE ARE MANY, MANY MORE...

FINDING AND EXTRACTING EXISTING DATA

APIS





TWITTER

Streaming APIs (live data by users and by topics)

The "Firehose" (all of live twitter)

Complete Archives via "Gnip" and eventually (maybe) the US Library of Congress

HTTPS://DEV.TWITTER.COM

■ API reference index

The API reference index is a central list of all endpoints included on the Twitter Developer Platform across our different APIs.

Jump to...

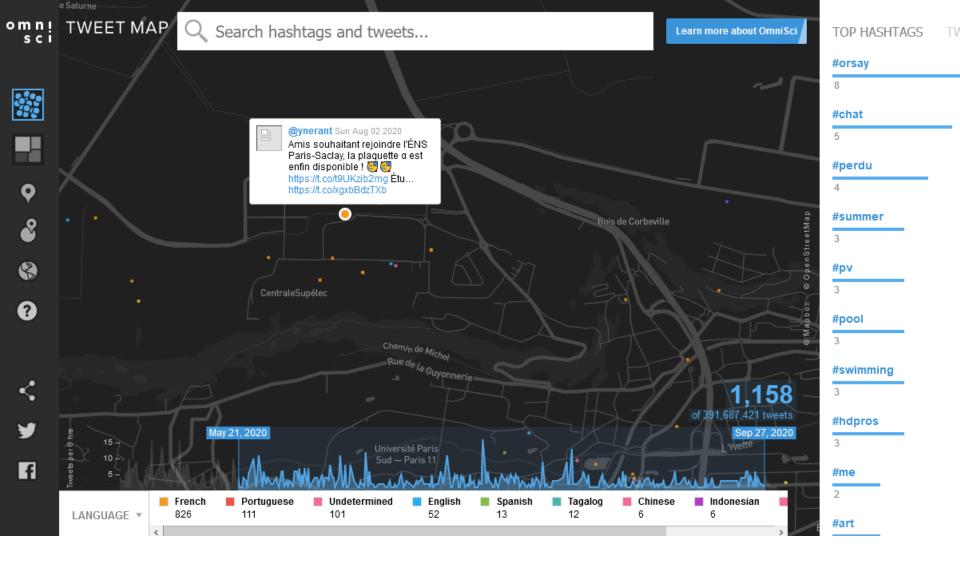
- Twitter API v2
- Twitter API v1.1
- Twitter Ads API
- Labs
- Platform-wide

Twitter API v2: Early Access

Tweets

Filtered stream

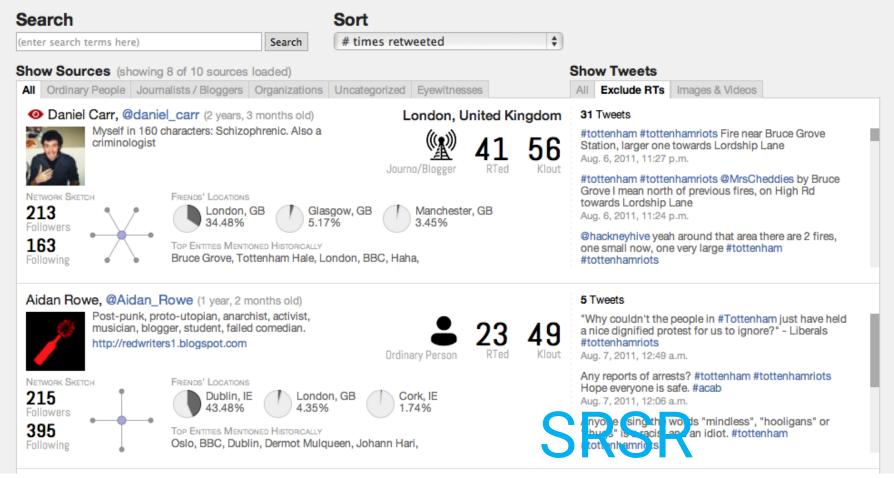
- GET /2/tweets/search/stream
- GET /2/tweets/search/stream/rules
- POST /2/tweets/search/stream/rules



https://www.omnisci.com/demos/tweetmap

Tottenham Riots

402 sources sharing 551 tweets matching "tottenhamriots" or "tottenham"



GOOGLE EARTH ENGINE

HTTPS://EARTHENGINE.GOOGLE.ORG/

MORE APIS (APPLICATION PROGRAMMING INTERFACES)

NEW YORK TIMES APIS

http://developer.nytimes.com/

(Archival news articles from 1851, books, movies, geographical, and political data)

OPEN STREET MAP

http://wiki.openstreetmap.org/wiki/API

(Detailed location and map data for the whole world)

AND THE LIST GOES ON!



FINDING AND EXTRACTING EXISTING DATA

SCRAPING THE WEB

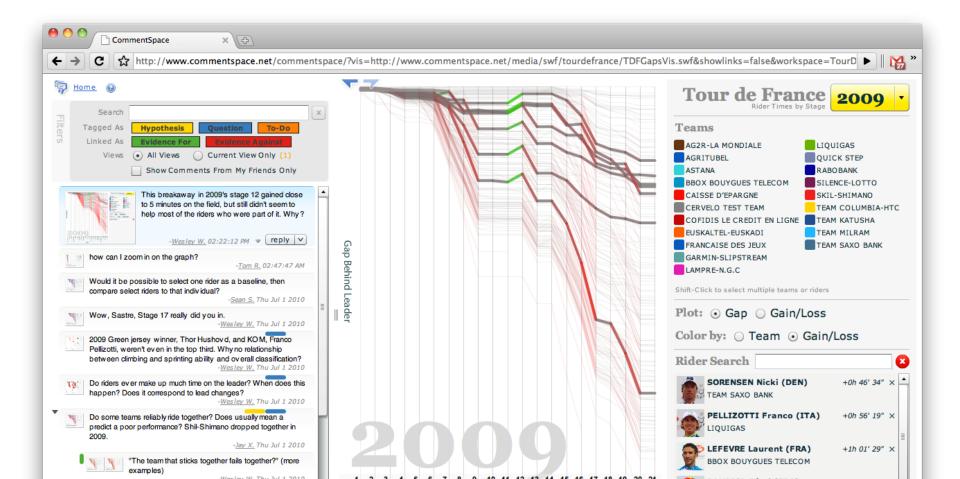
WHY SCRAPE?

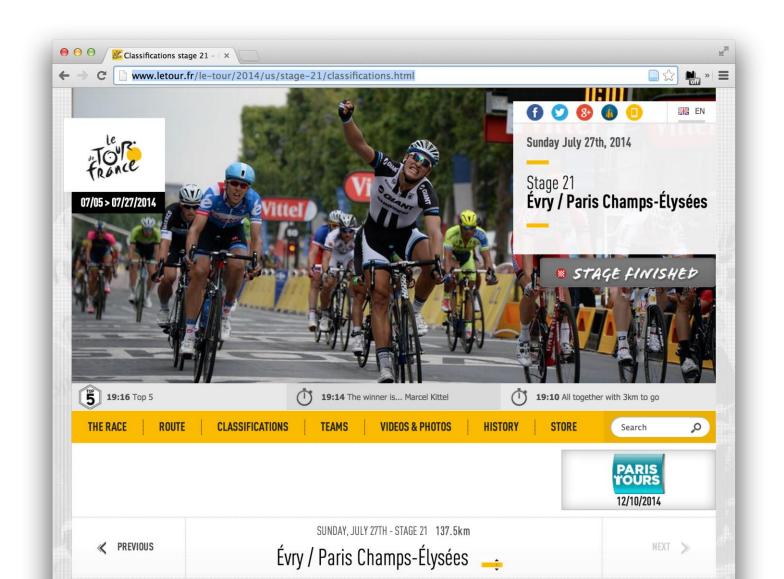
No API exists for the data you want (can't access the right data, wrong format, etc.)

<u>Simplicity</u> – Usually don't need to authenticate, no ratelimiting, etc.

Want to capture <u>context of pages</u> or relationship between them.

FOR EXAMPLE...







Combative

LCL

Overall individual time classification

Total distance covered: **3660.5 KM**

Individual

RANK	RID	ER	RIDER NO.	ТЕАМ	TIMES	GAP
1.		NIBALI Vincenzo	41	ASTANA PRO TEAM	89h 59' 06"	
2.		PÉRAUD Jean-Christophe	81	AG2R LA MONDIALE	90h 06' 43"	+ 07' 37"
3.		PINOT Thibaut	127	FDJ.FR	90h 07' 21"	+ 08' 15"
4.	\$	VALVERDE BELMONTE Alejandro	11	MOVISTAR TEAM	90h 08' 46''	+ 09' 40"
5.		VAN GARDEREN Tejay	141	BMC RACING TEAM	90h 10' 30"	+ 11' 24"
6.		BARDET Romain	82	AG2R LA MONDIALE	90h 10' 32"	+ 11' 26"
7.		KONIG Leopold	201	TEAM NETAPP-ENDURA	90h 13' 38"	+ 14' 32"
8.	6	ZUBELDIA AGIRRE Haimar	169	TREK FACTORY RACING	90h 17' 03''	+ 17' 57"
9.		TEN DAM Laurens	67	BELKIN PRO CYCLING	90h 17' 17''	+ 18' 11"
10.		MOLLEMA Bauke	61	BELKIN PRO CYCLING	90h 20' 21"	+ 21' 15"
11.		ROLLAND Pierre	151	TEAM EUROPCAR	90h 22' 13"	+ 23' 07"
12.		SCHLECK Frank	161	TREK FACTORY RACING	90h 24' 54"	+ 25' 48"
13.		VAN DEN BROECK Jurgen	131	LOTTO-BELISOL	90h 33' 07"	+ 34' 01"
14.		TROFIMOV Yury	29	TEAM KATUSHA	90h 35' 47''	+ 36' 41"
15.		KRUIJSWIJK Steven	64	BELKIN PRO CYCLING	90h 37' 21"	+ 38' 15"
16.		FEILLU Brice	211	BRETAGNE - SECHE ENVIRONNEMENT	90h 43' 05''	+ 43' 59"
17.		HORNER Christopher	114	LAMPRE - MERIDA	90h 43' 37''	+ 44' 31"
18.	5	NIEVE ITURRALDE Mikel	5	TEAM SKY	90h 45' 37''	+ 46' 31"
19.		GADRET John	13	MOVISTAR TEAM	90h 46' 36''	+ 47' 30"

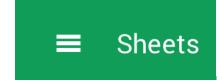
SOMETIMES YOU DON'T NEED A SCRAPER!

A few tips and tricks...

PULLING DATA TABLES FROM THE WEB



IMPORTHTML



Imports data from a table or list within an HTML page.

Demographics of India

From Wikipedia, the free encyclopedia

The demographics of India are

This article is about the people from India. For other uses, see Indian (disambiguation).

inclusive of the second most populous country in the world, **Population** with over 1.21 billion people Growth rate (2011 census), more than a sixth Birth rate est.) of the world's population. Already containing 17.5% of the **Death rate** est.) world's population, India is Life projected to be the world's most expectancy populous country by 2025, male surpassing China, its population female reaching 1.6 billion by 2050.^{[4][5]} Fertility rate Its population growth rate is Infant 1.41%, ranking 102nd in the mortality world in 2010.^[6] Indian rate population reached the billion Age structure mark in 2000.

Demographics of India 1,236,344,631 (July 2014 est.)^[1] (2nd) 1.51% (2009 est.) (93rd) 20.22 births/1,000 population (2013 7.4 deaths/1,000 population (2013 68.89 years (2009 est.) 67.46 years (2009 est.) 72.61 years (2009 est.) 2.44 children born/woman (SRS 2011) 44 deaths/1,000 live births (2011 est.)

Population distribution in India by states Area [19] Urban [21] Rural [21] State / Density **Sex Ratio** % [18] Rank Type Population Literacy Males **Females** [20] **Union Territory** (km²) (/km²) **Population Population** 1 Uttar Pradesh 34,539,582 State 199,812,341 16.50 240,928 828 104,480,510 95,331,831 912 67.68 131,658,339 2 Maharashtra 9.28 307,713 58,243,056 54,131,277 82.34 55,777,647 41,100,980 State 121,455,333 929 3 Bihar 103,804,637 1,102 54,278,157 49,821,295 74,316,709 8,681,800 State 8.60 94,163 918 61.80 4 West Bengal 91,276,115 1,030 46,809,027 44,467,088 57,748,946 22,427,251 State 7.54 88.752 950 76.26 5 Madhya Pradesh 72,626,809 6.00 37,612,306 931 69.32 44,380,878 15,967,145 State 308,245 236 35,014,503 34,921,681 27,483,998 6 Tamil Nadu 72,147,030 5.96 130,058 36,137,975 996 80.09 State 555 36,009,055 7 Rajasthan 68,548,437 5.66 342,239 201 35,550,997 32,997,440 928 43,292,813 13,214,375 State 66.11 8 Karnataka 61,095,297 5.05 191,791 30,966,657 30,128,640 973 75.36 34,889,033 17,961,529 State 9 Guiarat 4.99 196.024 28.948.432 919 78.03 31.740.767 18.930.250 State 60.439.692 31.491.260 % .00 123 -В Arial 10 =ImportHtml("http://en.wikipedia.org/wiki/Demographics_of_India", "table",4) f_{\times} Α В С D Ε F State / Area [19] Union Territory Population % [18] (km²) Rank Type 2 Uttar Pradesh State 199,812,341 16.5 240,928 3 Maharashtra State 121,455,333 9.28 307,713 8.6 94,163 4 Bihar State 103,804,637

91,276,115

72,626,809

72,147,030

7.54

5.96

6

88,752

308,245

130,058

5

6

7

West Bengal

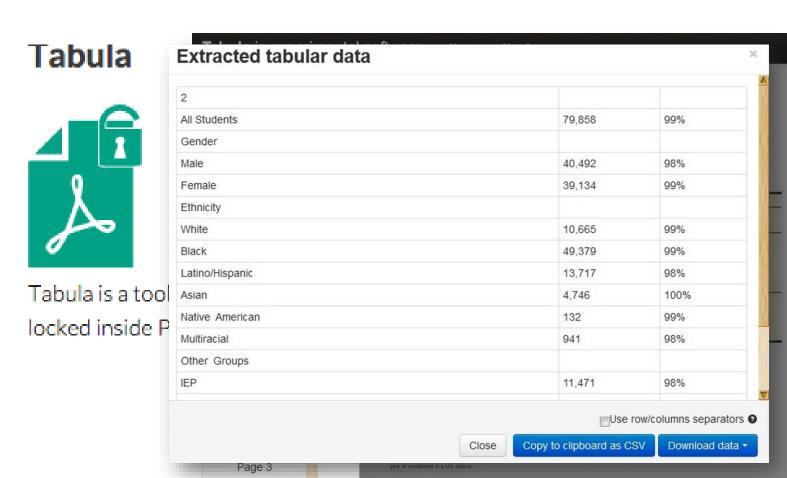
Tamil Nadu

Madhya Pradesh State

State

State

PARSING PDFS



BUILDING A WEB SCRAPER

FETCHING DATA + PARSING DATA

YOU SHOULD **SEPARATE** THESE PROCESSES **WHENEVER POSSIBLE!**

FETCHING DATA

DON'T DO EVERYTHING AT ONCE

Download complete pages and save them locally <u>before</u> you process them.

DEALING WITH PAGINATION

If results or records are spread across multiple pages, you may need to parse the page to find the link to the next page.

PARSING DATA

SERIOUSLY, DON'T DO EVERYTHING AT ONCE!

Processing data from local files means you don't have to get it right the first time.

USE YOUR BROWSER'S DEVELOPER TOOLS

All modern web browsers have built-in tools that let you inspect web pages.

BE CAREFUL - YOU CAN GET YOURSELF BLOCKED

Many sites will try to slow or block heavy access (both to prevent scraping and DoS attacks)

To get around this... You can introduce delays in your scraper or scrape from multiple locations.

A FEW MORE NOTES ABOUT DATA MANAGEMENT

FORMATS AND BEST-PRACTICES

DATA FORMATS

STRUCTURED vs. UNSTRUCTURED

STRUCTURED DATA is more like what you'd find in a traditional spreadsheet or database.

<u>UNSTRUCTURED DATA</u> can include **raw text**, **streaming data**, even **images** or **video**.

SEMI-STRUCTURED DATA is more organized, but doesn't follow a fixed schema (e.g. DBPEDIA data)

CSV

(Comma-Separated Value)

- 1 firstName, lastName, age, streetAddress, city, state
- John, Smith, 25, 21 2nd Street, New York, NY, 10021, 2

firstName	lastName	age	streetAddress	city	state	postalCode	homePhoneNumber	faxPhoneNumber	gender
John	Smith	25	21 2nd Street	New York	NY	10021	212 555-1239	646 555-4567	male

We will mostly use CSV in this course

CSV BEST PRACTICES

- Remove unnecessary rows or cells
- empty cells, comments, write NA for missing values

Row	author keyword	author frequency	expert keyword	expert frequency	
1	data partitioning	1	data and data management	64	
2	visual knowledge discov	13	knowledge discovery	29	
3	feature selection	1	features and attributes	38	
4	guided visualization	1	interaction	152	
5	regression	2	machine learning & statistics	55	
6	model building	1	machine learning & statistics	55	
7	decision support systems	1	analysis process	113	
8	model validation and ana	1	machine learning & statistics	55	
9	program analysis	1	analysis process	113	
10	multi-variate statistics	1	multidimensional / multivariat	83	
11	visual analytics	86	visual analytics	86	
12	cultural heritage	2	applications	103	
13	wall paintings	1	art and aesthetics	10	
14	degradation	1	applications	103	
15	nonnegative matrix facto	1	matrices	10	
16	interactive clustering	3	clustering	50	

CSV BEST PRACTICES

Splits cells if you can

ous (capstone,

If needed create a second file

	First page	Last page	keynote, VAST challenge, panel, poster,)	Abstract	Author Names
	457	457	M		Donna J. Cox
	6	13, 460	С	The use of critical po	James Helman;Lambertus Hesselink
1	14	27, 461	C		Gordon V. Bancroft; Fergus Merritt; Todd Plessel; Paul G. Kelaita; R. Kevin McCabe; Al Globus
1	28	35, 462	C	The VIS-5D system	William L. Hibbard;David A. Santek
1	36	44, 462	С	The author presents	
1	45	50, 462	C	Some ideas and tech	Gregory M. Nielson;Bernd Hamann
1	51	58, 463	C	The use of qualitative	Yaser Yacoob
1	59	66	С	Visualizing the third	Del Lamb;Amit Bandopadhay
1	67	73	С	The animation of two	Anthony J. Maeder
1	74	82, 464	С		James V. Miller;David E. Breen;Michael J. Wozny
1	83	92, 465	C		Ping-Kang Hsiung;Robert H. Thibadeau;Christopher B. Cox;Robert H. P. Dunn;Michael Wu;Pau
1	93	96, 467	С	The authors describe	Richard A. Becker;Stephen G. Eick;Eileen O. Miller;Allan R. Wilks
1	97	106, 46	C		Andrew J. Hanson;Pheng-Ann Heng;B. C. Kaplan
Į	107	113	С	The authors describe	Bowen Alpern;Larry Carter;Ted Selker

CSV BEST PRACTICES

Give meaningful unique column names

d	Α	В	С	D
1	ExistingFieldName	UserFriendlyFieldName		
2	AccMngDpt	Department		
3	AccMngName	Account Manager		
4	CusAccMnglD	Account Manager ID		
5	CusAddress	Customer Address		
6	CusCoulD	Customer Country		
7	CusID	Customer ID		
8	CusName	Customer		
9	DelAddress	Delivery Address		
10	DelDate	Delivery Date		
11	DelDesc	Delivery Description		
12	DelID	Delivery ID		
13	DelTime	Delivery Time		
14				
15				
16				
17				

Make column name casing consistent

For pandas, snake_case is recommended → with it you can do:

df.column_name or
df['column_name']

XML

(eXtensible Markup Language)

```
<person>
 <firstName>John</firstName>
 <lastName>Smith
 <age>25</age>
 <address>
   <streetAddress>21 2nd Street/streetAddress>
   <city>New York</city>
   <state>NY</state>
   <postalCode>10021</postalCode>
 </address>
 <phoneNumbers>
   <phoneNumber type="home">212 555-1234</phoneNumber>
   <phoneNumber type="fax">646 555-4567</phoneNumber>
 </phoneNumbers>
 <gender>
```

firstName	lastName	age	streetAddress	city	state	postalCode	homePhoneNumber	faxPhoneNumber	gender
John	Smith	25	21 2nd Street	New York	NY	10021	212 555-1239	646 555-4567	male

JSON

(<u>JavaScript Object Notation</u>)

```
"firstName": "John",
"lastName": "Smith",
"age": 25,
"address": {
 "streetAddress": "21 2nd Street",
 "city": "New York",
 "state": "NY",
  "postalCode": "10021"
"phoneNumber": [
    "type": "home",
    "number": "212 555-1239"
   "type": "fax",
    "number": "646 555-4567"
```

firstName	lastName	age	streetAddress	city	state	postalCode	homePhoneNumber	faxPhoneNumber	gender
John	Smith	25	21 2nd Street	New York	NY	10021	212 555-1239	646 555-4567	male

YAML

(YAML Ain't Markup Language)

```
firstName: John
lastName: Smith
age: 25
address:
      streetAddress: 21 2nd Street
      city: New York
      state: NY
      postalCode: 10021
phoneNumber:
          type: home
          number: 212 555-1234
          type: fax
```

firstName	lastName	age	streetAddress	city	state	postalCode	homePhoneNumber	faxPhoneNumber	gender
John	Smith	25	21 2nd Street	New York	NY	10021	212 555-1239	646 555-4567	male

HANDLING DATA

STORING DATA

- Always keep <u>backups</u>
- Password protect or encrypt any data with personal or sensitive information

PROVENANCE

- Keep track of <u>where/when</u> data was collected
- Record any data processing steps so you (or others) can repeat them if necessary

IP, COPYRIGHT, AND (RE)SHARING DATA

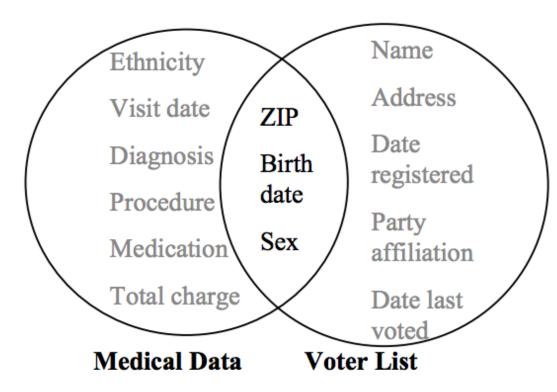
- Be sure you know who <u>owns</u> the data.
- Think early on about whether or not you'll need to <u>publish</u> or <u>(re)share</u> data.
- Be careful you <u>aren't violating copyright</u>, especially when scraping.

PRIVACY AND ANONYMIZING DATA

- Any information that could be used to <u>identify</u> individuals is sensitive!
- There may be **legal repercussions** for releasing it.
- In some cases you might need to <u>anonymize</u> data before sharing.

JUST REMOVING NAMES IS OFTEN NOT ENOUGH!

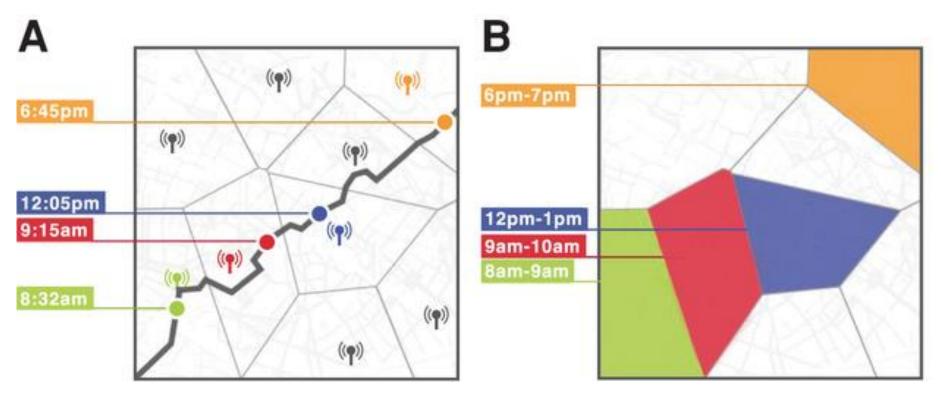
OTHER INFORMATION CAN STILL BE UNIQUE



[L. Sweeney. 2002]

k-ANONYMITY: A MODEL FOR PROTECTING PRIVACY

LOCATION DATA IS ESPECIALLY SENSITIVE



[de Montjoye et al. 2013]

Unique in the Crowd: The privacy bounds of human mobility

REGULATIONS (ACADEMIA AND RESEARCH)

Institutional Review and Ethics Boards may need to approve experiments or data collection <u>before it</u> <u>happens.</u>

Studies involving people may need informed consent.

REGULATIONS (INDUSTRY)

Some governments have placed limits on **how long** user data can be kept.

Some kinds of tracking (e.g., cookies) may now require <u>opt-in</u> or <u>notifications</u>. (However this varies by country).

SOCIAL EXPERIMENTS

SYNC

Senator asks FTC to investigate

Senator social networks

1888 FTC to investigate Exper conta Facebook's mood study Adam D After the social network altered the news feeds of nearly 700,000 users ^aCore Daf CA 9414 Without telling them, Sen. Mark R. Warner wants to know if there should be Edited oversight on these types of experiments. Emo^r cont in laboratory experimened in laboratory expe network, collected over a 20-y period (moods (e.g., depression, happiness) can be to networks [Fowler JH, Christakis NA (2008) BMJ 337:athough the results are controversial. In an experiment with people who use Facebook, we test whether emotional contagion occurs outside of in-person interaction between individuals by reducing the amount of emotional content in the News Feed. When positive expressions were reduced, people produced fewer positive posts and more negative posts; when negative expressions were restories, and activities undertaken by Inc. duced, the opposite pattern occurred. These results indicate that primary manner by which people see content that emotions expressed by others on Facebook influence our own Which content is shown or omitted in the News Feed emotions, constituting experimental evidence for massive-scale termined via a ranking algorithm that Facebook continually contagion via social networks. This work also suggests that, in develops and tests in the interest of showing viewers the content

they will find most relevant and engaging. One such test is

contrast to prevailing assumptions, in-person interaction and non-

"EXPERIMENTING ON **HUMAN BEINGS"**



We Experiment On Human Beings!

July 28th, 2014 by Christian Rudder





I'm the first to admit it: we might be popular, we might create a lot of great relationships we might blab blab blab But OkCupid doesn't

GDPR

General Data Protection Regulation

- → The world's strongest data protection rules
- → Define how organization can handle information about people (customers etc.)

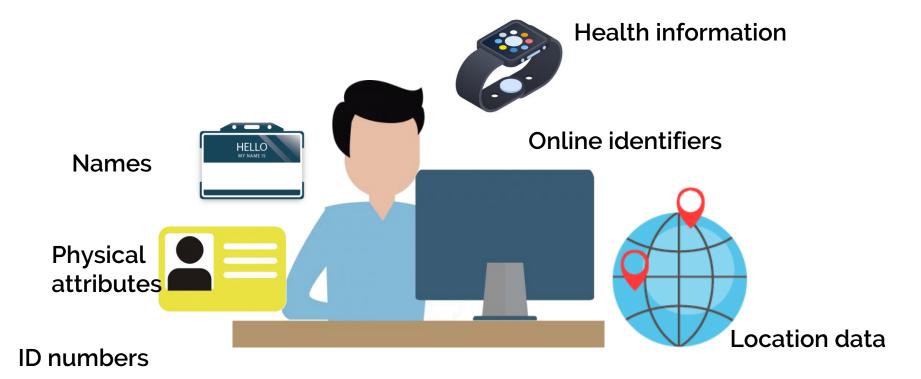
GDPR & RESEARCH

Collection of personal data

= data from which people can be identified

(data that is pseudonymous is still personal data)

PERSONAL DATA



=Any information relating to an identified or identifiable natural person ('data subject'); an identifiable natural person is one who can be identified, directly or indirectly, in particular by reference to an identifier such as a name, an identification number, location data, an online identifier or to one or more factors specific to the physical, physiological, genetic, mental, economic, cultural or social identity of that natural person; (Art 4, 1)

DATA PROCESSING

lawful, fair and transparent

get ethics approval

only process the minimal amount of necessary personal data; anonymize where possible

IN SUMMARY: THERE ARE LOTS OF TOOLS AT YOUR DISPOSAL!

COLLECT IT

- OBSERVATION
- SURVEYS
- LOGGING
- SENSORS
- CROWDSOURCING

FIND OR EXTRACT IT

- OPEN CORPUSES
- DATA RETAILERS
- APIS
- SCRAPING THE WEB

GENERATE IT

- SIMULATIONS

BEFORE NEXT CLASS

INSTALL:



OpenRefine (formerly Google Refine) http://openrefine.org/