

Working Environment

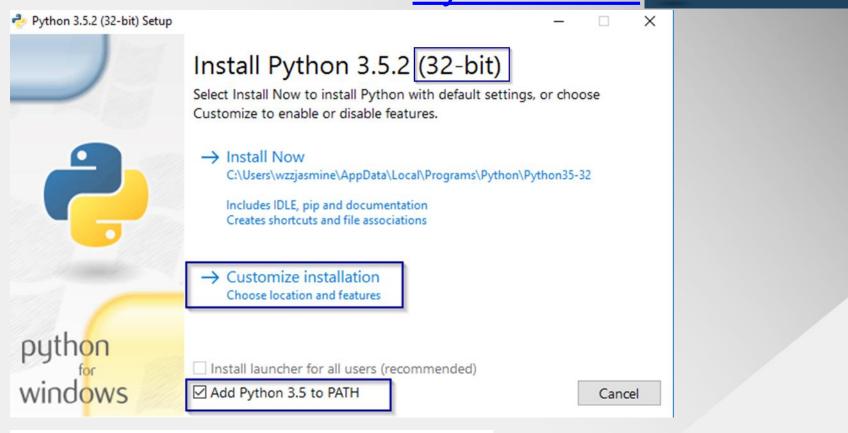
Customize install location

You will require write permissions for the selected location.

C:\Python35

Download and install Python3.5.2 python





Browse

Package to install

- Textblob
 - pip install textblob
 - python -m textblob.download_corpora
- Snownlp
 - pip install snownlp
- Jieba
 - pip install jieba
- Beautifulsoup
 - pip install beautifulsoup4
- PyNotebook
 - pip install jupyter

Introduction

- "Sentimental analysis, also called opinion mining, is the field of study that analyzes people's opinions, sentiments, evaluations, appraisals, attitudes, and emotions towards entities such as products, services, organizations, individuals, issues, events, topics, and their attributes." (Liu, 2012)
 - opinion mining
 - opinion extraction
 - sentiment mining
 - subjectivity analysis
 - affect analysis
 - emotion analysis
 - review mining
 - **–** ...



Why Sentimental Analysis

- We want to know others' opinion: surveys, opinion polls, and focus groups
- Explosive growth of social media: reviews, forum discussions, blogs, micro-blogs, Twitter

• We feel fine(video)



Posted by: John Smith

Date: September 10, 2011

(1) I bought a Canon G12 camera six months ago. (2) I simply love it. (3) The picture quality is amazing. (4) The battery life is also long. (5) However, my wife thinks it is too heavy for her.

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• Definition (Opinion): An opinion is a quintuple, (g, a, s, h, t), where g is the opinion (or sentiment) target, a is attribute or component of the target, s is the sentiment about the target, h is the opinion holder and t is the time when the opinion wasexpressed.

- Objective of sentiment analysis: Given an opinion document d, discover all opinion quintuples (g, a, s, h, t) in d
- Task 1: Extract all entities/aspects/holders in D, and categorize or group synonymous expressions into entity clusters (or categories)
- Task 2: Extract the times when opinions are given and standardize different time formats
- Task 3: Determine whether an opinion on an aspect is positive, negative or neutral, or assign a numeric sentiment rating to the aspect.
- Task 4: Produce all opinion quintuples (g, a, s, h, t) expressed in document d based on the results of the above tasks

Example (1)

Posted by: bigJohn Date: Sept. 15, 2011 (1) I bought a Samsung camera and my friends brought a Canon camera yesterday. (2) In the past week, we both used the cameras a lot. (3) The photos from my Samy are not that great, and the battery life is short too. (4) My friend was very happy with his camera and loves its picture quality. (5) I want a camera that can take good photos. (6) I am going to return it tomorrow.

Example (2)

- Four opinion quintuples
 - (Samsung, picture_quality, negative, bigJohn, Sept-15-2011)
 - (Samsung, battery_life, negative, bigJohn, Sept-15-2011)
 - (Canon, GENERAL, positive, bigJohn's_friend, Sept-15-2011)
 - -(Canon, picture_quality, positive, bigJohn's_friend, Sept-15-2011)

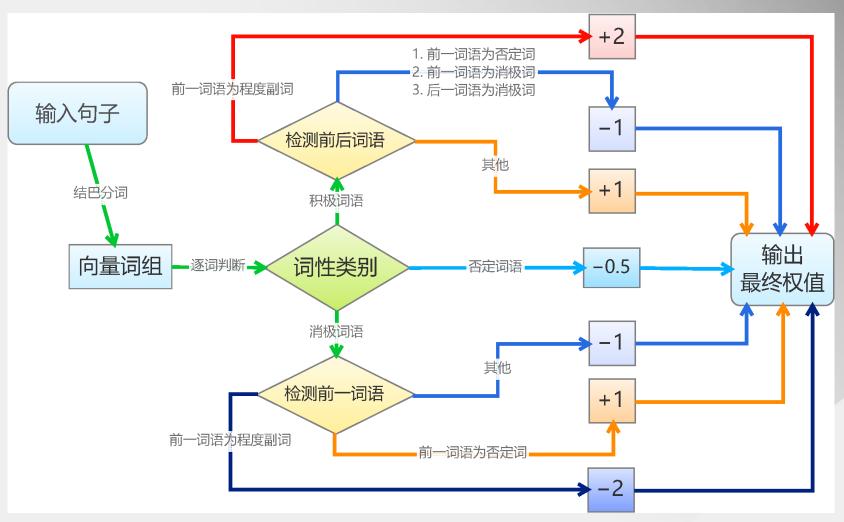
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Unsupervised Learning

- Dictionary based approach
 - Sentiment words dictionary
 - Amplifier words dictionary
 - Shifter words dictionary

Dictionary Based Sentiment Analysis



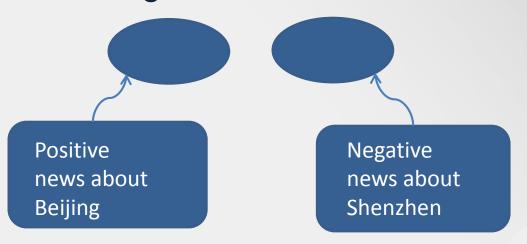
(Source: http://spaces.ac.cn/archives/3360/)

Mine Sentiment Dictionary

- Feature selection
- Word association

Feature Selection: Starter Edition

- Finding terms that best represent texts in each category.
- Starter edition: pick up the most popular term in each category.
- Problem: What if the most popular terms are the same across categories?



Feature Selection: Starter Edition

- Finding terms that best represent texts in each category.
- Starter edition: pick up the most popular term in each category.
- Problem: What if the most popular terms are the same across categories?



Feature Selection: Advanced Edition

 Picking up the most discriminant terms in each category.

 χ² calculates whether the occurrence of the term and occurrence of the category are independent.

Feature Selection: Using χ^2 to Realize the Advanced Edition

•
$$\chi^2(t,c) = \sum_{e_t \in \{0,1\}} \sum_{e_c \in \{0,1\}} \frac{(N_{e_t e_c} - E_{e_t e_c})^2}{E_{e_t e_c}}$$

- Et=1 (the document contains term t)
- Et=0 (the document does not contain term t)
- Ec=1 (the document is in class c)
- Et=0 (the document is not in class c)
- χ^2 measures how much expected counts E and observed counts N deviate from each other. It calculates the relative importance of terms for each category.

Feature Selection: Using χ^2 to Realize the Advanced Edition

•
$$\chi^2(t,c) = \sum_{e_t \in \{0,1\}} \sum_{e_c \in \{0,1\}} \frac{(N_{e_t e_c} - E_{e_t e_c})^2}{E_{e_t e_c}}$$

- Et=1 (the document contains term t)
- Et=0 (the document does not contain term t)
- Ec=1 (the document is in class c)
- Et=0 (the document is not in class c)
- χ^2 measures how much expected counts E and observe olympics its haze jate from each other. It calculates the relative importance of term: Positive category

Positive news about Beijing

news about Shenzhen

Exercise

•
$$\chi^2(t,c) = \sum_{e_t \in \{0,1\}} \sum_{e_c \in \{0,1\}} \frac{(N_{e_t e_c} - E_{e_t e_c})^2}{E_{e_t e_c}}$$

- Et=1 (the document contains term t)
- Et=0 (the document does not contain term t)
- Ec=1 (the document is in class c)
- Et=0 (the document is not in class c)
- A corpus of 801,948 news articles
- 27,701 articles are classified as "positive". Of them, 49 articles contain the word "Olympics".
- Of the rest 774,247 articles, 141 contain the word "Olympics".

Exercise

	e _{positive} =1	e _{positive} =0
e _{olympics} =1	N ₁₁ =49 E ₁₁ =6.6	N ₁₀ =141 E ₁₀ =183.4
e _{olympics} =0	N ₀₁ =27652 E ₀₁ =27694.4	N ₀₀ =774106 E ₀₀ =774063.6

$$\chi^{2}(t,c) = \sum_{e_{t} \in \{0,1\}} \sum_{e_{c} \in \{0,1\}} \frac{(N_{e_{t}e_{c}} - E_{e_{t}e_{c}})^{2}}{E_{e_{t}e_{c}}} = 284$$

Word Association

- Select benchmark words
- Calculate PMI of the target word and benchmark words
- Estimate sentiment orientation of the target word based on above calculation

Word Association

PMI: pointwise mutural information

$$PMI(term_1, term_2) = \log_2 \left(\frac{\Pr(term_1 \land term_2)}{\Pr(term_1)\Pr(term_2)} \right)$$

- Pr(term1^term2) is the actual co-occurrence probability of term1 and term2
- Pr(term1)Pr(term2) is the co-occurrence probability of the two terms if they are statistically independent.
- PMI measures the degree of statistical dependence between two terms.

Word Association

Estimate sentiment orientation

$$SO(phrase) = \log_2 \left(\frac{hits(phrase NEAR "excellent")hits("poor")}{hits(phrase NEAR "poor")hits("excellent")} \right)$$

Exercise

 Use Google/Baidu to mine sentiment orientation of 'Olympics'

Supervised Learning

- Naive Bayes (NB) Classification
 - Retrieving a training set (usually manually),
 with texts already assigned to a known list of categories
 - Based on the training set, determining the contribution of each term for each category
 - Based on the terms in texts, assigning new texts into categories.

NB Classification: Example

Doc Type	Doc ID	Terms in doc	in $c = China$?
Training Set	1	China Beijing China Tokyo	yes
	2	China China Shanghai	yes
	3	China Macao Japan	yes
	4	Tokyo Japan China	no
Testing Set	5	China China Tokyo	?
		Japan	

NB Classification:

```
 \hat{p}(China|c)=5/10 \qquad \hat{p}(Japan|c)=1/10 \qquad \hat{p}(Tokyo|c)=1/10   \hat{p}(China|\bar{c})=1/3 \qquad \hat{p}(Japan|\bar{c})=1/3 \qquad \hat{p}(Tokyo|\bar{c})=1/3
```

•
$$\hat{p}(China|\bar{c})=1/3$$
 $\hat{p}(Japan|\bar{c})=1/3$ $\hat{p}(Tokyo|\bar{c})=1/3$

- $\hat{p}(c|doc_5)=(5/10)^3*(1/10)*(1/10)=0.00125$
- $\hat{p}(\bar{c}|doc_5)=(1/3)^3*(1/3)*(1/3)=0.00412$

(Revised from Manning, Raghavan, & Schütze, 2008)

Use NB Classification to do Sentiment Analysis (1)

Doc Type	Doc ID	Terms in doc	Classification
Training Set	1	I love this car	Positive
	2	This view is amazing	Positive
	3	I feel great this morning	Positive
	4	I am so excited about the concert	Positive
	5	He is my best friend	Positive
	6	I do not like this car	Negative
	7	This view is horrible	Negative
	8	I feel tired this morning	Negative
	9	I am not looking forward to the concert	Negative
	10	He is my enemy	Negative

Use NB Classification to do Sentiment Analysis (2)

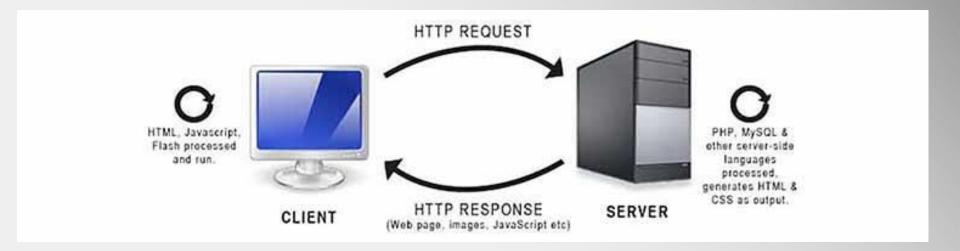
Doc Type	Doc ID	Terms in doc	NB Classification	Human coding
Testing Set	1	feel happy this morning	N	Р
	2	Oh I love my friend	Р	Р
	3	not like that man	N	N
	4	house not great	N	N
	5	your song annoying	N	N

Accuracy: 80%

(Revised Wang Chengjun's blog)

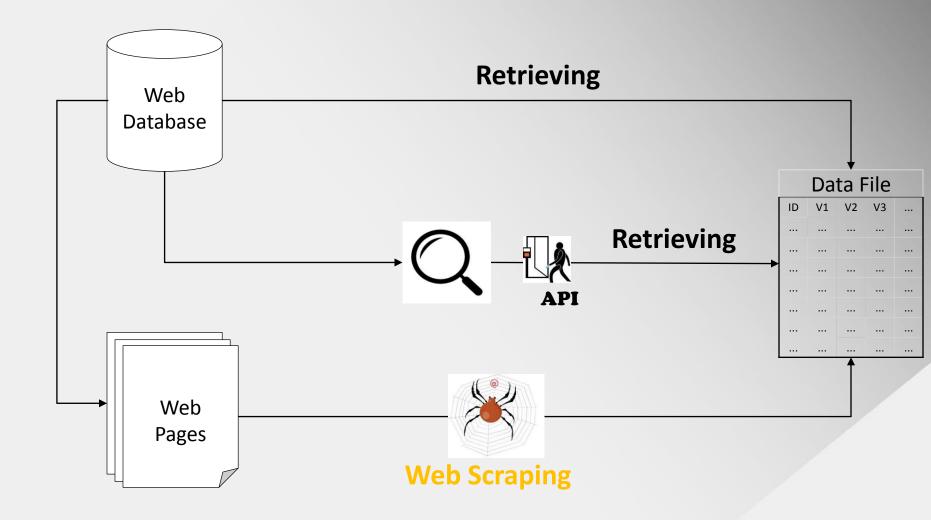


Access the Internet: The Human Way



(The client-server model)

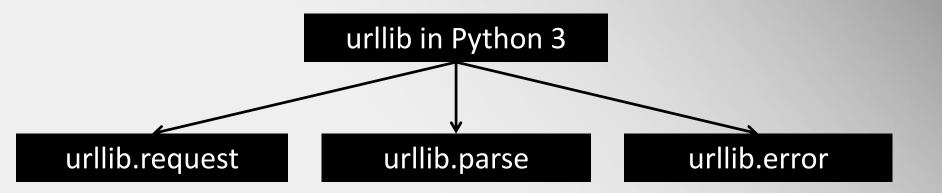
Access Internet: Alternatives



What does Web Scraper do?

- Retrieving HTML data from a domain name
- Parsing that data for target information
- Storing the target information
- Optionally, moving to another page to repeat the process

Your First Web Scrapper



http://pythonscraping.com/pages/page1.html

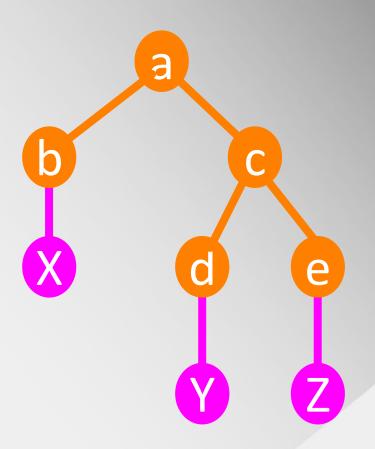
Introduction to HTML (1)

- HTML is a markup language for describing web documents (web pages).
 - HTML stands for Hyper Text Markup Language
 - A markup language is a set of markup tags
 - HTML documents are described by HTML tags
 - Each HTML tag describes different document content

(http://www.w3schools.com/html/html_intro.asp)

Introduction to HTML (2)

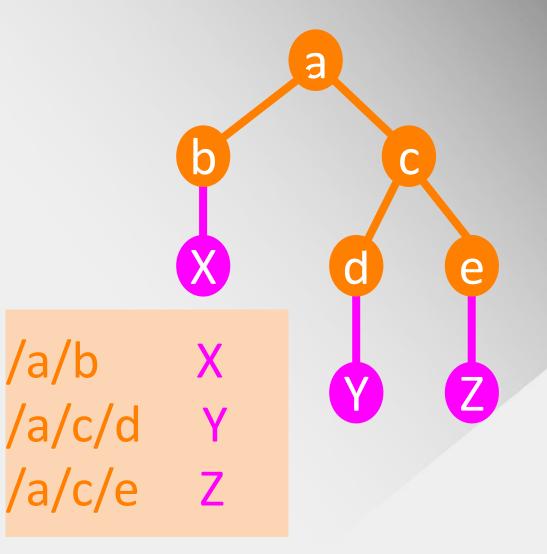
HTML as Tree



Introduction to HTML (3)

HTML as Path

```
<a>>
 <b>X</b>
  <c>
   <d>Y</d>
   <e>Z</e>
</c>
```



Your First Web Scrapper (1)

http://pythonscraping.com/pages/page1.html

An Interesting Title

Lorem ipsum dolor sit amet, consectetur adipisicing elit, sed do eiusmod tempor incididunt ut labore et dolore magna aliqua. Ut enim ad minim veniam, quis nostrud exercitation ullamco laboris nisi ut aliquip ex ea commodo consequat. Duis aute irure dolor in reprehenderit in voluptate velit esse cillum dolore eu fugiat nulla pariatur. Excepteur sint occaecat cupidatat non proident, sunt in culpa qui officia deserunt mollit anim id est laborum.

```
from urllib.request import urlopen
html=urlopen("http://pythonscraping.com/pages/page1.html")
html.read()
```

Your First Web Scrapper (2)

```
<html>
        <head>
                <title>A Useful Page</title>
        </head>
        <body>
                <h1>An Interesting Title</h1>
                <div>Lorem ipsum dolor sit amet, consectetur
            adipisicing elit, sed do eiusmod tempor incididunt ut labore
            et dolore magna aliqua. Ut enim ad minim veniam, quis
            nostrud exercitation ullamco laboris nisi ut aliquip ex ea
            commodo consequat. Duis aute irure dolor in reprehenderit
            in voluptate velit esse cillum dolore eu fugiat nulla pariatur.
            Excepteur sint occaecat cupidatat non proident, sunt in
            culpa qui officia deserunt mollit anim id est laborum.</div>
        </body>
</html>
```

Your First Web Scrapper (2)

```
<html>
        <head>
                 <title>A Useful Page</title>
        </head>
        <body>
                                                           html
                 <h1>An Interesting Title</h1>
                 <div>Lorem ipsum dolor sit amet, consecretur
             adipisicing elit, sed do eiusmod tempor incididunt ut labore
             et dolore magna aliqua. Ut enim head im veniar nostrud exercitation ullamco lab head ut aliquip
             commodo consequat. Duis aute ir re dolor in rep ehanderit
             in voluptate velit esse cillum dolo e eu fugiat nulla pariatur.
             Excepteur sint occaecat cupident non proid
             culpa qui officia deserunt me title midest
        </body>
</html>
                                                             text
                                                                            text
                                            text
```

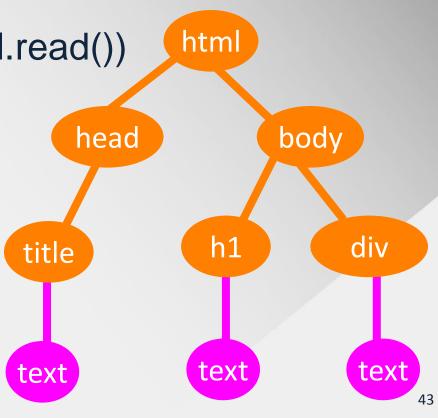
Your First Web Scrapper (3)

- from bs4 import BeautifulSoup
- html=urlopen("http://pythonscraping.com/pages/ page1.html")

bsobj=BeautifulSoup(html.read())

print (bsobj.html.body.h1)

- print (bsobj.body.h1)
- print (bsobj.html.h1)



Save Webpages (1)

Write to a file

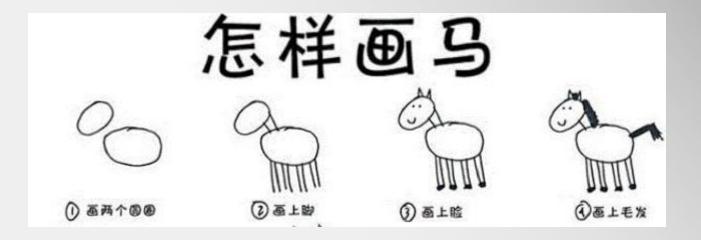
```
fileName="D://example.txt"
p=open(fileName,"w") #open for writing, truncating the file first
print("hello",file=p)
print("world",file=p)
p.close()
p=open(fileName, "a") #open for writing, appending to the end of
the file if it exists
print("hello world again",file=p)
p.close()
```

Save Webpages (2)

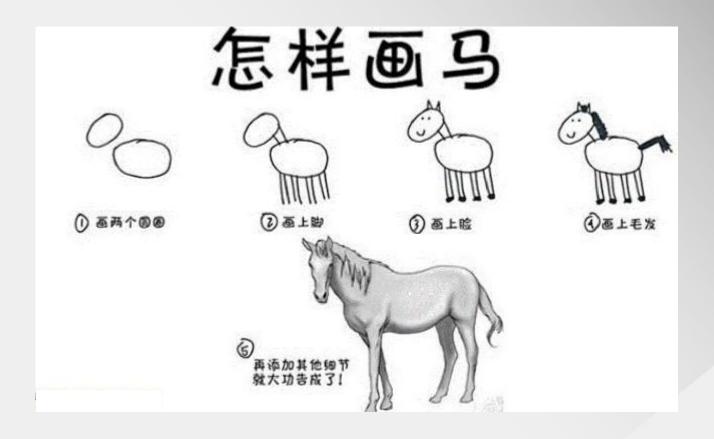
Crawl a webpage and write it to a file



Drawing Horse



Drawing Horse



Comparative Opinion

- Coke tastes better than Pepsi.
- Coke tastes the best.

Implicit Opinion

- I bought the mattress a week ago, and a valley has formed.
- The battery life of Nokia phones is longer than Samsung phones.

Standing Point

 The housing price has gone down, which is bad for the economy.

Sarcastic Sentences

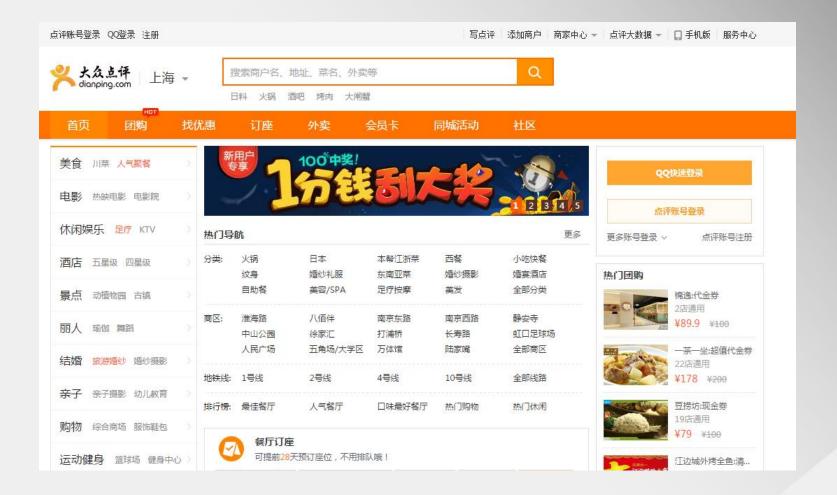
 What a great car! It stopped working in two days.



References(1)

• Liu, B. (2012). Sentiment analysis and opinion mining.





上海>>火锅>>page2

http://www.dianping.com/search/category/1/10/g110p2

City:

Beijing Shanghai

•••

Topic:

Restaurants

Movie

Hotel

•••

Cuisine:

Hotpot

Chuan

Jiangzhe

• • •

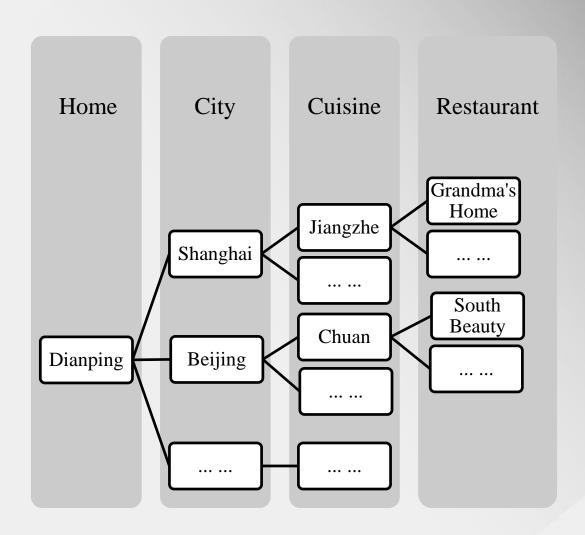
Page:

1

2

3

•••



Preparation:

- Generate a list of cities (or just one city)
- Generate a list of topics (or just one topic)
- Generate a list of cuisines (or just one cuisine)

Crawl:

- Combine city, topic, cuisine, and generate the url of index page
- Extract shops url on the index page
- crawl shop webpage one by one

Sampling Problem

- 北京>>美食>>素菜
 - http://www.dianping.com/search/category/2/10/g109
 - 171 shops, 12 pages, 100% displayed
- 北京>>美食>>江浙菜
 - http://www.dianping.com/search/category/2/10/g101
 - ??? shops, 50 pages, ?% displayed
- 北京>>美食>>粤菜
 - http://www.dianping.com/search/category/2/10/g103
 - ??? shops, 50 pages, ?% displayed
- 北京>>美食>>川菜
 - http://www.dianping.com/search/category/2/10/g102
 - ??? shops, 50 pages, ?% displayed
- 北京>>美食>>北京菜
 - http://www.dianping.com/search/category/2/10/g311
 - ??? shops, 50 pages, ?% displayed

ID-based System

 Shop URL: http://www.dianping.com/shop/XXX

a unique numeric value—shop ID

automatically generated by the web database

Draw a Random Sample (1)

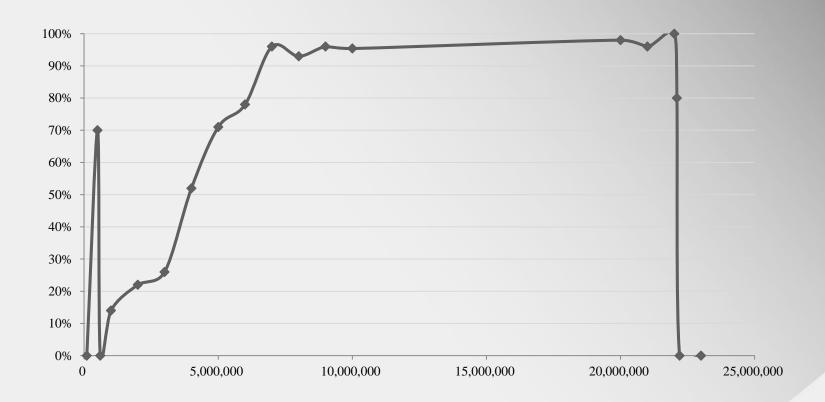
 1, Detect ID ranges: the likely minimal and maximal number of digits of IDs

			Sampling	Hit
No. of digits	Population	Sample size	fraction	
1	10	10	100%	NA
2	100	100	100%	NA
3	1,000	1,000	100%	NA
4	10,000	10,000	100%	NA
5	100,000	10,000	10%	NA
6	1,000,000	10,000	1%	776
7	10,000,000	10,000	0.1%	6,074
8	100,000,000	10,000	0.01%	1,273
9	1,000,000,000	10,000	0.001%	NA
10	10,000,000,000	10,000	0.0001%	NA

Draw a Random Sample (2)

- 2, Draw a small sample to calculate occupancy rate of the study population.
 - Occupancy Rate (OR) measures the proportion of potential user IDs being valid bloggers.
 - OR will be instrumental for projecting the population, designing sampling strategies for future studies, weighting resultant samples, and other practical purposes.

Draw a Random Sample (3)



Draw a Random Sample (4)

 3, Draw a substantive sample based on the occupancy rate.

References(2)

Ryan, M. (2015) Web scraping with Python Zhu, J. J. H., Mo, Q., Wang, F., & Lu, H. (2011). A random digit search (RDS) method for sampling of blogs and other usergenerated content. Social Science Computer Review, 29 (3), 327-339.

