# Introduction to Information Retrieval

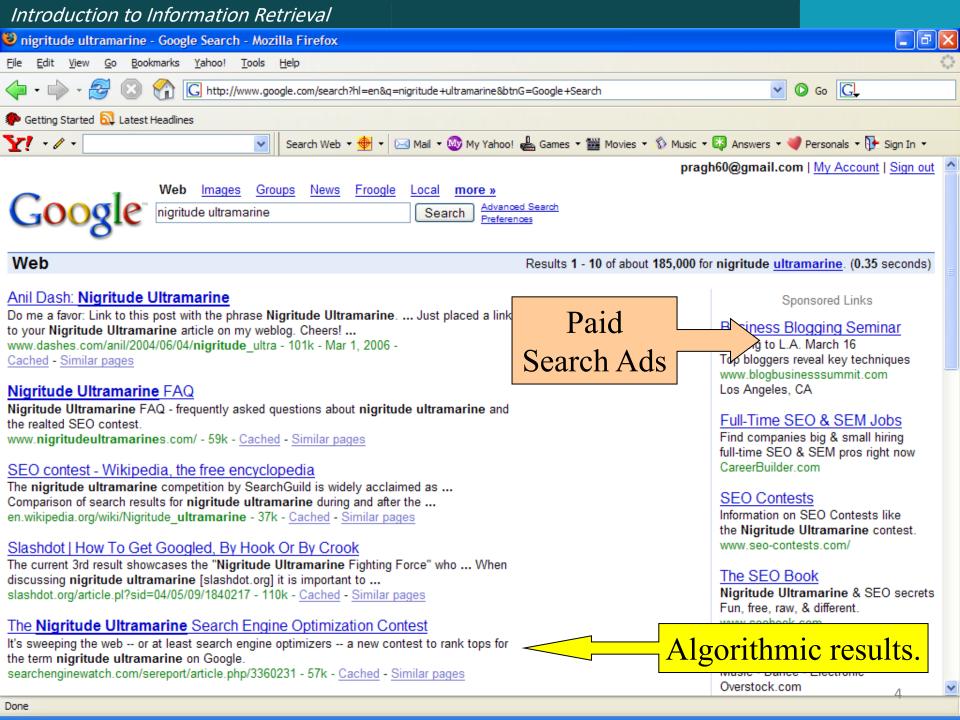
CS276
Information Retrieval and Web Search
Pandu Nayak and Prabhakar Raghavan
Lecture 15: Web search basics

## Brief (non-technical) history

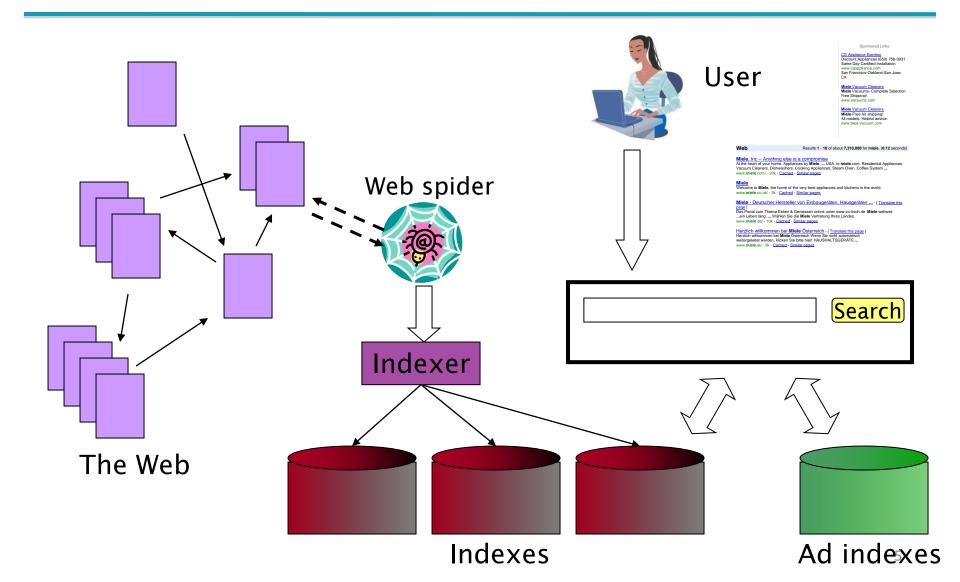
- Early keyword-based engines ca. 1995-1997
  - Altavista, Excite, Infoseek, Inktomi, Lycos
- Paid search ranking: Goto (morphed into Overture.com → Yahoo!)
  - Your search ranking depended on how much you paid
  - Auction for keywords: casino was expensive!

## Brief (non-technical) history

- 1998+: Link-based ranking pioneered by Google
  - Blew away all early engines save Inktomi
  - Great user experience in search of a business model
  - Meanwhile Goto/Overture's annual revenues were nearing \$1 billion
- Result: Google added paid search "ads" to the side, independent of search results
  - Yahoo followed suit, acquiring Overture (for paid placement) and Inktomi (for search)
- 2005+: Google gains search share, dominating in Europe and very strong in North America
  - 2009: Yahoo! and Microsoft propose combined paid search offering



#### Web search basics



#### **User Needs**

- Need [Brod02, RL04]
  - Informational want to learn about something (~40% / 65%)

Low hemoglobin

Navigational – want to go to that page (~25% / 15%)

United Airlines

- <u>Transactional</u> want to do something (web-mediated) (~35% / 20%)
  - Access a service

Seattle weather

Downloads

Mars surface images

Shop

Canon S410

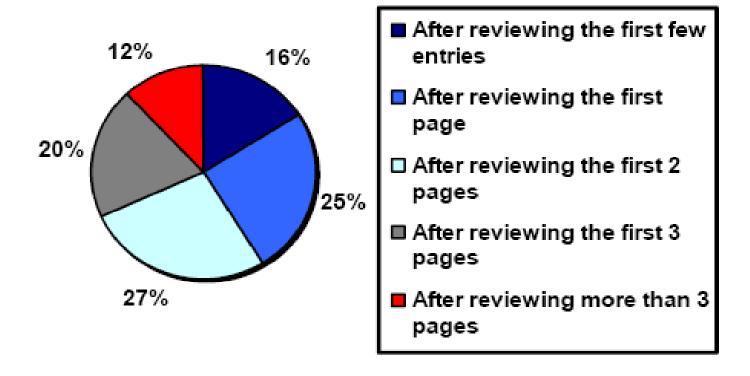
- Gray areas
  - Find a good hub

Car rental Brasil

Exploratory search "see what's there"

#### How far do people look for results?

"When you perform a search on a search engine and don't find what you are looking for, at what point do you typically either revise your search, or move on to another search engine? (Select one)"



(Source: <u>iprospect.com</u> WhitePaper\_2006\_SearchEngineUserBehavior.pdf)

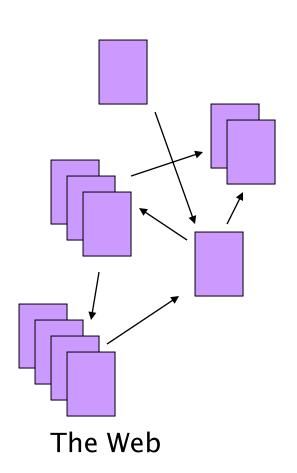
#### Users' empirical evaluation of results

- Quality of pages varies widely
  - Relevance is not enough
  - Other desirable qualities (non IR!!)
    - Content: Trustworthy, diverse, non-duplicated, well maintained
    - Web readability: display correctly & fast
    - No annoyances: pop-ups, etc.
- Precision vs. recall
  - On the web, recall seldom matters
- What matters
  - Precision at 1? Precision above the fold?
  - Comprehensiveness must be able to deal with obscure queries
    - Recall matters when the number of matches is very small
- User perceptions may be unscientific, but are significant over a large aggregate

#### Users' empirical evaluation of engines

- Relevance and validity of results
- UI Simple, no clutter, error tolerant
- Trust Results are objective
- Coverage of topics for polysemic queries
- Pre/Post process tools provided
  - Mitigate user errors (auto spell check, search assist,...)
  - Explicit: Search within results, more like this, refine ...
  - Anticipative: related searches
- Deal with idiosyncrasies
  - Web specific vocabulary
    - Impact on stemming, spell-check, etc.
  - Web addresses typed in the search box
- "The first, the last, the best and the worst ..."

#### The Web document collection



- No design/co-ordination
- Distributed content creation, linking, democratization of publishing
- Content includes truth, lies, obsolete information, contradictions ...
- Unstructured (text, html, ...), semistructured (XML, annotated photos), structured (Databases)...
- Scale much larger than previous text collections ... but corporate records are catching up
- Growth slowed down from initial "volume doubling every few months" but still expanding
- Content can be dynamically generated

## **SPAM**(SEARCH ENGINE OPTIMIZATION)

#### The trouble with paid search ads ...

- It costs money. What's the alternative?
- Search Engine Optimization:
  - "Tuning" your web page to rank highly in the algorithmic search results for select keywords
  - Alternative to paying for placement
  - Thus, intrinsically a marketing function
- Performed by companies, webmasters and consultants ("Search engine optimizers") for their clients
- Some perfectly legitimate, some very shady

#### Search engine optimization (Spam)

#### Motives

- Commercial, political, religious, lobbies
- Promotion funded by advertising budget

#### Operators

- Contractors (Search Engine Optimizers) for lobbies, companies
- Web masters
- Hosting services

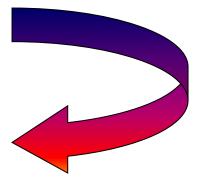
#### Forums

- E.g., Web master world ( <u>www.webmasterworld.com</u> )
  - Search engine specific tricks
  - Discussions about academic papers ©

## Simplest forms

- First generation engines relied heavily on tf/idf
  - The top-ranked pages for the query maui resort were the ones containing the most maui's and resort's
- SEOs responded with dense repetitions of chosen terms
  - e.g., maui resort maui resort maui resort
  - Often, the repetitions would be in the same color as the background of the web page
    - Repeated terms got indexed by crawlers
    - But not visible to humans on browsers

Pure word density cannot be trusted as an IR signal



## Variants of keyword stuffing

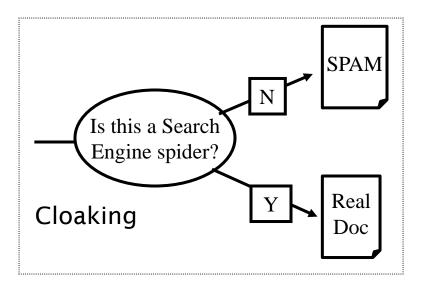
- Misleading meta-tags, excessive repetition
- Hidden text with colors, style sheet tricks, etc.

#### Meta-Tags =

"... London hotels, hotel, holiday inn, hilton, discount, booking, reservation, sex, mp3, britney spears, viagra, ..."

## Cloaking

- Serve fake content to search engine spider
- DNS cloaking: Switch IP address. Impersonate



### More spam techniques

#### Doorway pages

 Pages optimized for a single keyword that re-direct to the real target page

#### Link spamming

- Mutual admiration societies, hidden links, awards more on these later
- Domain flooding: numerous domains that point or redirect to a target page

#### Robots

- Fake query stream rank checking programs
  - "Curve-fit" ranking programs of search engines
- Millions of submissions via Add-Url

## The war against spam

- Quality signals Prefer authoritative pages based on:
  - Votes from authors (linkage signals)
  - Votes from users (usage signals)
- Policing of URL submissions
  - Anti robot test
- Limits on meta-keywords
- Robust link analysis
  - Ignore statistically implausible linkage (or text)
  - Use link analysis to detect spammers (guilt by association)

- Spam recognition by machine learning
  - Training set based on known spam
- Family friendly filters
  - Linguistic analysis, general classification techniques, etc.
  - For images: flesh tone detectors, source text analysis, etc.
- Editorial intervention
  - Blacklists
  - Top queries audited
  - Complaints addressed
  - Suspect pattern detection

#### More on spam

- Web search engines have policies on SEO practices they tolerate/block
  - http://help.yahoo.com/help/us/ysearch/index.html
  - http://www.google.com/intl/en/webmasters/
- Adversarial IR: the unending (technical) battle between SEO's and web search engines
- Research <a href="http://airweb.cse.lehigh.edu/">http://airweb.cse.lehigh.edu/</a>

#### **DUPLICATE DETECTION**

#### Duplicate documents

- The web is full of duplicated content
- Strict duplicate detection = exact match
  - Not as common
- But many, many cases of near duplicates
  - E.g., last-modified date the only difference between two copies of a page

#### Duplicate/Near-Duplicate Detection

- Duplication: Exact match can be detected with fingerprints
- Near-Duplication: Approximate match
  - Overview
    - Compute syntactic similarity with an edit-distance measure
    - Use similarity threshold to detect near-duplicates
      - E.g., Similarity > 80% => Documents are "near duplicates"
      - Not transitive though sometimes used transitively

## **Computing Similarity**

- Features:
  - Segments of a document (natural or artificial breakpoints)
  - Shingles (Word N-Grams)
  - a rose is a rose is a rose →

```
a_rose_is_a

rose_is_a_rose

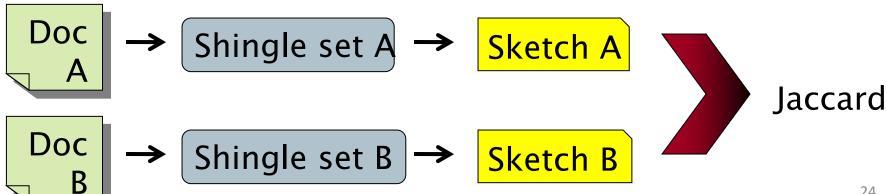
is_a_rose_is

a_rose_is_a
```

- Similarity Measure between two docs (= sets of shingles)
  - Jaccard coefficient: Size\_of\_Intersection / Size\_of\_Union

## Shingles + Set Intersection

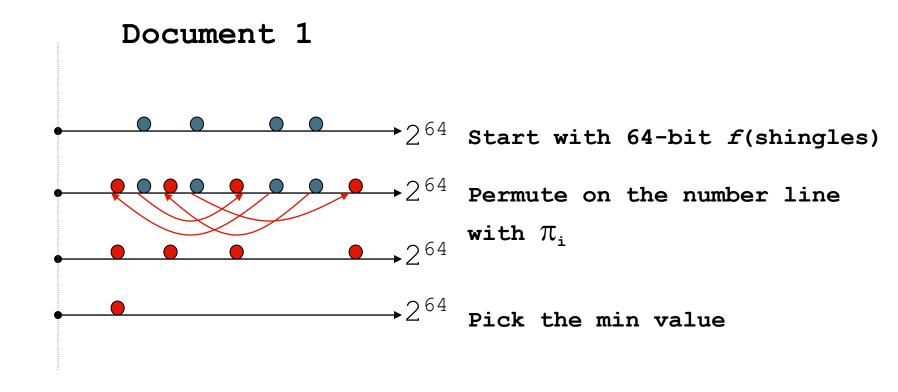
- Computing <u>exact</u> set intersection of shingles between <u>all</u> pairs of documents is expensive/intractable
  - Approximate using a cleverly chosen subset of shingles from each (a sketch)
- Estimate (size\_of\_intersection / size\_of\_union) based on a short sketch



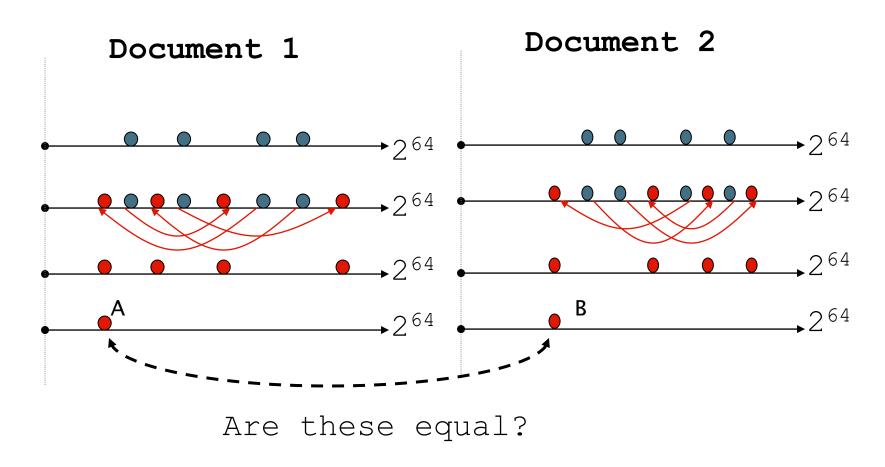
#### Sketch of a document

- Create a "sketch vector" (of size ~200) for each document
  - Documents that share ≥ t (say 80%) corresponding vector elements are near duplicates
  - For doc D, sketch<sub>D</sub>[ i] is as follows:
    - Let f map all shingles in the universe to 0..2<sup>m</sup>-1 (e.g., f = fingerprinting)
    - Let  $\pi_i$  be a *random permutation* on  $0..2^m-1$
    - Pick MIN  $\{\pi_i(f(s))\}$  over all shingles s in D

## Computing Sketch[i] for Doc1

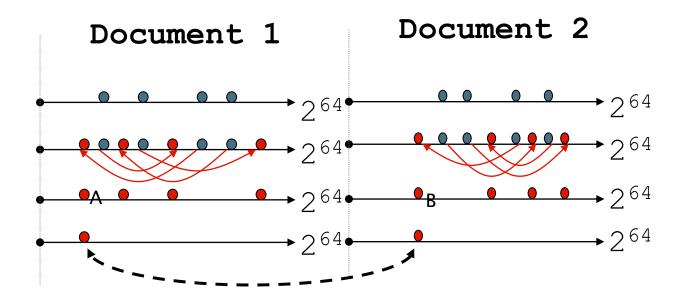


#### Test if Doc1.Sketch[i] = Doc2.Sketch[i]



Test for 200 random permutations:  $\pi_1$ ,  $\pi_2$ ,...  $\pi_{200}$ 

#### However...



A = B iff the shingle with the MIN value in the union of Doc1 and Doc2 is common to both (i.e., lies in the intersection)

Claim: This happens with probability

Size\_of\_intersection / Size\_of\_union

## Set Similarity of sets C<sub>i</sub>, C<sub>j</sub>

$$Jaccard(C_{i}, C_{j}) = \frac{\left|C_{i} \cap C_{j}\right|}{\left|C_{i} \cup C_{j}\right|}$$

- View sets as columns of a matrix A; one row for each element in the universe. a<sub>ij</sub> = 1 indicates presence of item i in set j
- Example

$$C_1$$
  $C_2$ 

1 0  
1 1 Jaccard(
$$C_1, C_2$$
) = 2/5 = 0.4  
0 0

#### **Key Observation**

For columns C<sub>i</sub>, C<sub>i</sub>, four types of rows

- Overload notation: A = # of rows of type A
- Claim

$$Jaccard(C_{i}, C_{j}) = \frac{A}{A + B + C}$$

## "Min" Hashing

- Randomly permute rows
- Hash h(C<sub>i</sub>) = index of first row with 1 in column
   C<sub>i</sub>
- Surprising Property

$$P\left(h(C_i) = h(C_j)\right) = Jaccard\left(C_i, C_j\right)$$

- Why?
  - Both are A/(A+B+C)
  - Look down columns C<sub>i</sub>, C<sub>i</sub> until first non-Type-D row
  - $h(C_i) = h(C_i) \longleftrightarrow$  type A row

#### Min-Hash sketches

- Pick P random row permutations
- MinHash sketch

Sketch<sub>D</sub> = list of P indexes of first rows with 1 in column C

- Similarity of signatures
  - Let sim[sketch(C<sub>i</sub>),sketch(C<sub>j</sub>)] = fraction of permutations where MinHash values agree
  - Observe E[sim(sketch(C<sub>i</sub>),sketch(C<sub>j</sub>))] = Jaccard(C<sub>i</sub>,C<sub>j</sub>)

## Example

$$\begin{array}{c|cccc} \mathbf{C_1} & \mathbf{C_2} & \mathbf{C_3} \\ \mathbf{R_1} & 1 & 0 & 1 \\ \mathbf{R_2} & 0 & 1 & 1 \\ \mathbf{R_3} & 1 & 0 & 0 \\ \mathbf{R_4} & 1 & 0 & 1 \\ \mathbf{R_5} & 0 & 1 & 0 \\ \end{array}$$

#### **Signatures**

Perm 1 = (12345) 
$$\begin{bmatrix} 1 & S_2 & S_3 \\ 1 & 2 & 1 \end{bmatrix}$$
  
Perm 2 = (54321)  $\begin{bmatrix} 4 & 5 & 4 \\ 4 & 5 & 4 \end{bmatrix}$   
Perm 3 = (34512)  $\begin{bmatrix} 3 & 5 & 4 \\ 3 & 5 & 4 \end{bmatrix}$ 

#### **Similarities**

|         | 1-2  | 1-3  | 2-3  |
|---------|------|------|------|
| Col-Col | 0.00 | 0.50 | 0.25 |
| Sig-Sig | 0.00 | 0.67 | 0.00 |

## Example

#### Permutation $\pi$ Input matrix (Shingles x Documents)

#### Signature matrix M

| 2 | 4 | 3 |
|---|---|---|
| 3 | 2 | 4 |
| 7 | 1 | 7 |
| 6 | 3 | 2 |
| 1 | 6 | 6 |
| 5 | 7 | 1 |
| 4 | 5 | 5 |

| 1 | 0 | 1 | 0 |
|---|---|---|---|
| 1 | 0 | 0 | 1 |
| 0 | 1 | О | 1 |
| 0 | 1 | О | 1 |
| 0 | 1 | 0 | 1 |
| 1 | 0 | 1 | 0 |
| 1 | 0 | 1 | 0 |



| 2 | 1 | 2 | 1 |
|---|---|---|---|
| 2 | 1 | 4 | 1 |
| 1 | 2 | 1 | 2 |

#### Similarities:

Col/C Sig/S

|     |      | 2-4  |   | 3-4 |
|-----|------|------|---|-----|
| Col | 0.75 | 0.75 | 0 | 0   |
| Sig | 0.67 | 1.00 | 0 | 0   |

#### All signature pairs

- Now we have an extremely efficient method for estimating a Jaccard coefficient for a single pair of documents.
- But we still have to estimate  $N^2$  coefficients where N is the number of web pages.
  - Still slow
- One solution: locality sensitive hashing (LSH)
- Another solution: sorting (Henzinger 2006)

#### More resources

IIR Chapter 19

## **SIZE OF THE WEB**

### What is the size of the web?

#### Issues

- The web is really infinite
  - Dynamic content, e.g., calendars
  - Soft 404: www.yahoo.com/<anything> is a valid page
- Static web contains syntactic duplication, mostly due to mirroring (~30%)
- Some servers are seldom connected
- Who cares?
  - Media, and consequently the user
  - Engine design
  - Engine crawl policy. Impact on recall.

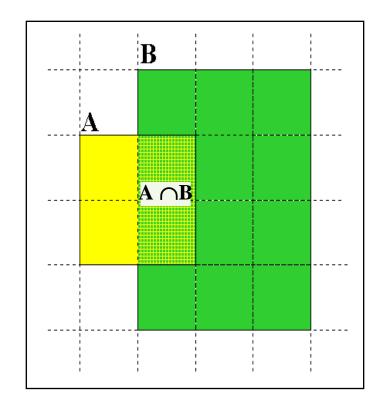
# What can we attempt to measure?

- •The relative sizes of search engines
  - The notion of a page being indexed is still reasonably well defined.
  - Already there are problems
    - Document extension: e.g., engines index pages not yet crawled, by indexing anchortext.
    - Document restriction: All engines restrict what is indexed (first n words, only relevant words, etc.)

#### New definition?

- The statically indexable web is whatever search engines index.
- IQ is whatever the IQ tests measure.
- Different engines have different preferences
- max url depth, max count/host, anti-spam rules, priority rules, etc.
- Different engines index different things under the same URL:
- frames, meta-keywords, document restrictions, document extensions, ...

# Relative Size from Overlap Given two engines A and B



Sample URLs randomly from A

Check if contained in B and vice versa

$$A \cap B = (1/2) * Size A$$
 $A \cap B = (1/6) * Size B$ 

$$(1/2) * Size A = (1/6) * Size B$$

$$\therefore Size A / Size B = (1/6) / (1/2) = 1/3$$

## Sampling URLs

- Ideal strategy: Generate a random URL and check for containment in each index.
- Problem: Random URLs are hard to find! Enough to generate a random URL contained in a given Engine.
- Approach 1: Generate a random URL contained in a given engine
  - Suffices for the estimation of relative size
- Approach 2: Random walks / IP addresses
  - In theory: might give us a true estimate of the size of the web (as opposed to just relative sizes of indexes)

### Statistical methods

- Approach 1
  - Random queries
  - Random searches
- Approach 2
  - Random IP addresses
  - Random walks

## Random URLs from random queries

- Generate <u>random query</u>: how?
  - Lexicon: 400,000+ words from a web crawl

Not an English dictionary

- Conjunctive Queries: w<sub>1</sub> and w<sub>2</sub>
   e.g., vocalists AND rsi
- Get 100 result URLs from engine A
- Choose a random URL as the candidate to check for presence in engine B
- This distribution induces a probability weight W(p) for each page.

## **Query Based Checking**

- Strong Query to check whether an engine B has a document D:
  - Download D. Get list of words.
  - Use 8 low frequency words as AND query to B
  - Check if D is present in result set.
- Problems:
  - Near duplicates
  - Frames
  - Redirects
  - Engine time-outs
  - Is 8-word query good enough?

- Statistically sound under the induced weight.
- Biases induced by random query
  - Query Bias: Favors content-rich pages in the language(s) of the lexicon
  - Ranking Bias: Solution: Use conjunctive queries & fetch all
  - Checking Bias: Duplicates, impoverished pages omitted
  - Document or query restriction bias: engine might not deal properly with 8 words conjunctive query
  - Malicious Bias: Sabotage by engine
  - Operational Problems: Time-outs, failures, engine inconsistencies, index modification.

#### Random searches

- Choose random searches extracted from a local log [Lawrence & Giles 97] or build "random searches" [Notess]
  - Use only queries with small result sets.
  - Count normalized URLs in result sets.
  - Use ratio statistics

- Advantage
  - Might be a better reflection of the human perception of coverage
- Issues
  - Samples are correlated with source of log
  - Duplicates
  - Technical statistical problems (must have non-zero results, ratio average not statistically sound)

#### Random searches

- 575 & 1050 queries from the NEC RI employee logs
- 6 Engines in 1998, 11 in 1999
- Implementation:
  - Restricted to queries with < 600 results in total</li>
  - Counted URLs from each engine after verifying query match
  - Computed size ratio & overlap for individual queries
  - Estimated index size ratio & overlap by averaging over all queries

#### Queries from Lawrence and Giles study

- adaptive access control
- neighborhood preservation topographic
- hamiltonian structures
- right linear grammar
- pulse width modulation neural
- unbalanced prior probabilities
- ranked assignment method
- internet explorer favourites importing
- karvel thornber
- zili liu

- softmax activation function
- bose multidimensional system theory
- gamma mlp
- dvi2pdf
- john oliensis
- rieke spikes exploring neural
- video watermarking
- counterpropagation network
- fat shattering dimension
- abelson amorphous computing

#### Random IP addresses

- Generate random IP addresses
- Find a web server at the given address
  - If there's one
- Collect all pages from server
  - From this, choose a page at random

Sec. 19.5

#### Random IP addresses

- HTTP requests to random IP addresses
  - Ignored: empty or authorization required or excluded
  - [Lawr99] Estimated 2.8 million IP addresses running crawlable web servers (16 million total) from observing 2500 servers.
  - OCLC using IP sampling found 8.7 M hosts in 2001
    - Netcraft [Netc02] accessed 37.2 million hosts in July 2002
- [Lawr99] exhaustively crawled 2500 servers and extrapolated
  - Estimated size of the web to be 800 million pages
  - Estimated use of metadata descriptors:
    - Meta tags (keywords, description) in 34% of home pages, Dublin core metadata in 0.3%

- Advantages
  - Clean statistics
  - Independent of crawling strategies
- Disadvantages
  - Doesn't deal with duplication
  - Many hosts might share one IP, or not accept requests
  - No guarantee all pages are linked to root page.
    - E.g.: employee pages
  - Power law for # pages/hosts generates bias towards sites with few pages.
    - But bias can be accurately quantified IF underlying distribution understood
  - Potentially influenced by spamming (multiple IP's for same server to avoid IP block)

#### Random walks

- View the Web as a directed graph
- Build a random walk on this graph
  - Includes various "jump" rules back to visited sites
    - Does not get stuck in spider traps!
    - Can follow all links!
  - Converges to a stationary distribution
    - Must assume graph is finite and independent of the walk.
    - Conditions are not satisfied (cookie crumbs, flooding)
    - Time to convergence not really known
  - Sample from stationary distribution of walk
  - Use the "strong query" method to check coverage by SE

- Advantages
  - "Statistically clean" method, at least in theory!
  - Could work even for infinite web (assuming convergence) under certain metrics.
- Disadvantages
  - List of seeds is a problem.
  - Practical approximation might not be valid.
  - Non-uniform distribution
    - Subject to link spamming

### Conclusions

- No sampling solution is perfect.
- Lots of new ideas ...
- ....but the problem is getting harder
- Quantitative studies are fascinating and a good research problem