

Introduction to the Use of

Link Analysis

by Web Search Engines

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Outline

- Introduction to Information Retrieval (IR)
- Link Analysis
- HITS Algorithm
- PageRank Algorithm

Short History of IR

IR = search within doc. coll. for particular info. need (query)

- B. C. cave paintings
- 7-8th cent. A.D. Beowulf
- 12th cent. A.D. invention of paper, monks in scriptoriums
 - Gutenberg's printing press
 - Franklin's public libraries
 - Dewey's decimal system
 - Card catalog
 - Computer
 - Salton's SMART system
 - Berner-Lee's WWW

1940s-1950s 1960s

1989

1450

1700s

1872

the pre-1998 Web

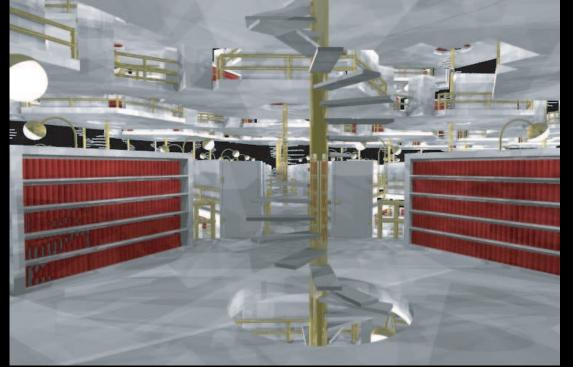
Yahoo

- hierarchies of sites
- organized by humans

Best Search Techniques

- word of mouth
- expert advice

Overall Feeling of Users



• Jorge Luis Borges' 1941 short story, *The Library of Babel*

When it was proclaimed that the Library contained all books, the first impression was one of extravagant happiness. All men felt themselves to be the masters of an intact and secret treasure. There was no personal or world problem whose eloquent solution did not exist in some hexagon.

... As was natural, this inordinate hope was followed by an excessive depression. The certitude that some shelf in some hexagon held precious books and that these precious books were inaccessible, seemed almost intolerable.

1998 ... enter Link Analysis

Change in User Attitudes about Web Search

Today

- "It's not my homepage, but it might as well be. I use it to ego-surf. I use it to read the news. Anytime I want to find out anything, I use it." - Matt Groening, creator and executive producer, The Simpsons
- "I can't imagine life without Google News. Thousands of sources from around the world ensure anyone with an Internet connection can stay informed. The diversity of viewpoints available is staggering." - Michael Powell, chair, Federal Communications Commission
- Google is my rapid-response research assistant. On the run-up to a deadline, I may use it to check the spelling of a foreign name, to acquire an image of a particular piece of military hardware, to find the exact quote of a public figure, check a stat, translate a phrase, or research the background of a particular corporation. It's the Swiss Army knife of information retrieval." - Garry Trudeau, cartoonist and creator, Doonesbury



IR before the Web = traditional IR IR on the Web = web IR



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- It's huge.
 - over 10 billion pages, average page size of 500KB
 - 20 times size of Library of Congress print collection
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- It's self-organized.
 - no standards, review process, formats
 - errors, falsehoods, link rot, and spammers!

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How is the Web different from other document collections?

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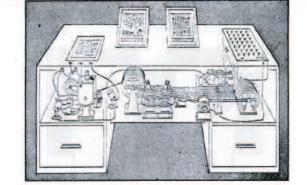
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A Herculean Task!

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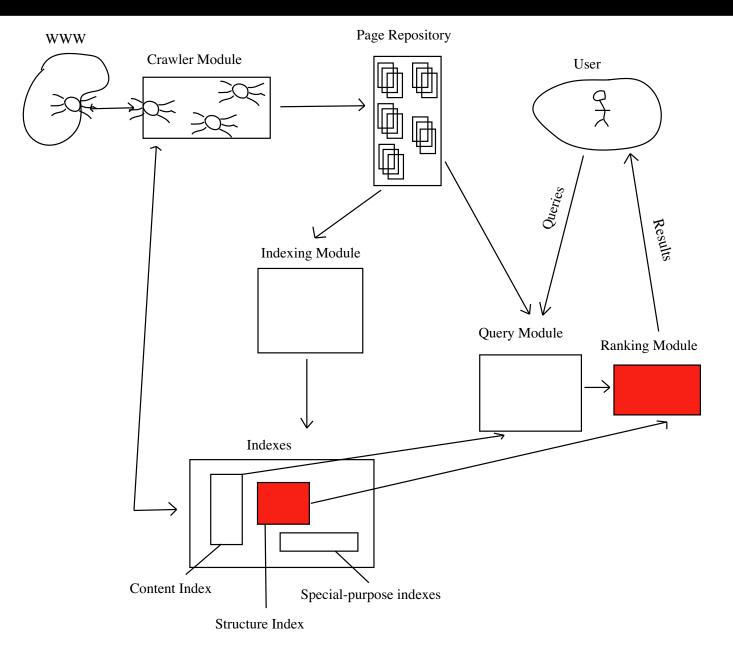
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- Ah, but it's *hyperlinked* !
 - Vannevar Bush's 1945 memex

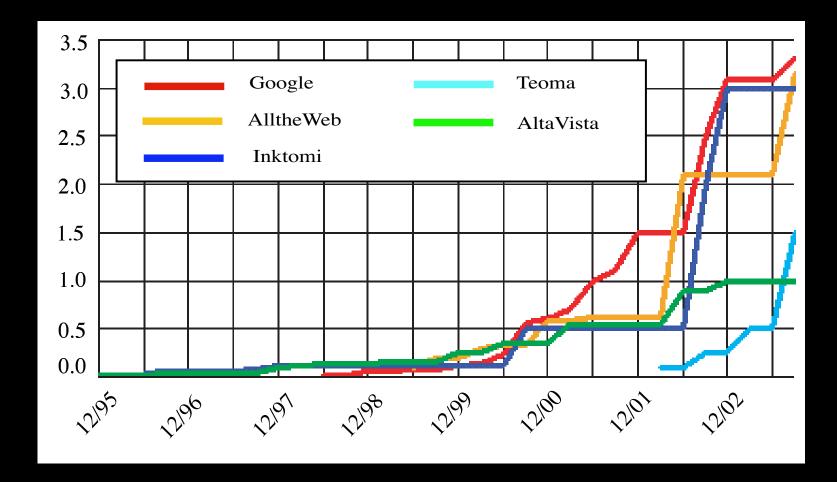


Memiex in the form of a desk would instantly bring files and material on any subject to the operator's fingertips. Slanting ranshueent viewing access magnify supermicrolilus filed by code numbers. At left is a mechanism which automatically holographic longhand notes, pictures and letters, then files them in the desk for future reference (LIFE 19(11), p. 123).

Elements of a Web Search Engine



Indexing Wars



Actual Index King =

Internet Archive - http://web.archive.org

Query Processing

Step 1: User enters query, i.e., aztec baby

Step 2: Inverted file consulted

- term 1 (aardvark) 3, 117, 3961
- term 10 (aztec) 3, 15, 19, 101, 673, 1199
- term 11 (baby) 3, 31, 56, 94, 673, 909, 11114, 253791 :
- term m (zymurgy) 1159223

Step 3: Relevant set identified, i.e. (3,673) Simple traditional engines stop here.

Modification to Inverted File

 add more features to inverted file by appending vector to each page identifier, i.e., [in title?, in descrip.?, # of occurrences]

Modified inverted file

- term 1 (aardvark) 3 [0,0,3], 117 [1,1,10], 3961 [0,1,4]
- term 10 (aztec) 3 [1, 1, 27], 15 [0,0,1], 19 [1,1,21], 101 [0,1,7], 673 [0, 0, 3], 1199 [0,0,3]
- term 11 (baby) 3 [1, 1, 10], 31 [0,0,2], 56 [0,1,3], 94 [1,1,11], 673 [1, 1, 14], 909 [0,0,2], 11114 [1,1,22], 253791 [0,1,6] .
- term m (zymurgy) 1159223 [1,1,9]
- IR score computed for each page in relevant set.
 EX: IR score (page 3) = (1 + 1 + 27) × (1 + 1 + 10) = 348 IR score (page 673) = (0 + 0 + 3) × (1 + 1 + 14) = 48 Early web engines stop here.
 Problem = Ranking by IR score is not good enough.

CSC issues in Crawling and Indexing

- create parallel crawlers but avoid overlap
- ethical spidering
- how often to crawl pages, which pages to update
- best way to store huge inverted file
- how to efficiently update inverted file
- store the files across processors
- provide for parallel access
- create robust, failure-resistant system

Link Analysis

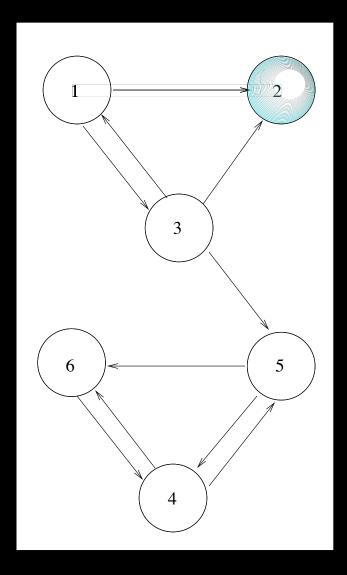
• uses *hyperlink* structure to focus the relevant set

• combine IR score with popularity or importance score

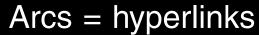


HITS - Kleinberg
$$\Rightarrow$$

The Web as a Graph



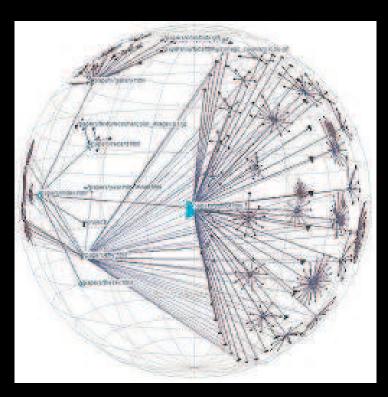
Nodes = webpages

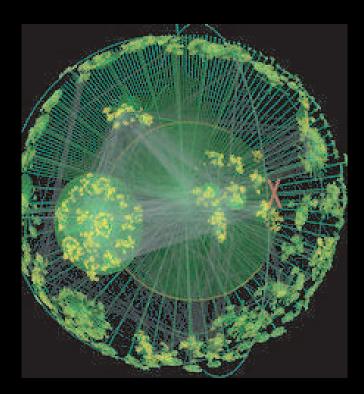


Web Graphs

CSC and MATH problems here:

- store adjacency matrix
- update adjacency matrix
- visualize web graph
- locate clusters in graph





How to Use Web Graph for Search

Hyperlink = **Recommendation**

- page with 20 recommendations (inlinks) must be more important than page with 2 inlinks.
- but status of recommender matters.
 EX: letters of recommendation: 1 letter from Trump vs. 20 from unknown people
- but what if recommender is generous with recommendations?
 EX: suppose Trump has written over 40,000 letters.
- each inlink should be weighted to account for status of recommender and # of outlinks from that recommender

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PAGERANK - importance/popularity score given to each page



Our Search: Google Technology

<u>Home</u>

All About Google

Help Central

Google Features

Services & Tools

Our Technology • Why Use Google <u>Benefits of Google</u>

Find on this site:



Google searches more sites more quickly, delivering the most relevant results.

Introduction

Google runs on a unique combination of advanced hardware and software. The speed you experience can be attributed in part to the efficiency of our search algorithm and partly to the thousands of low cost PC's we've networked together to create a superfast search engine.

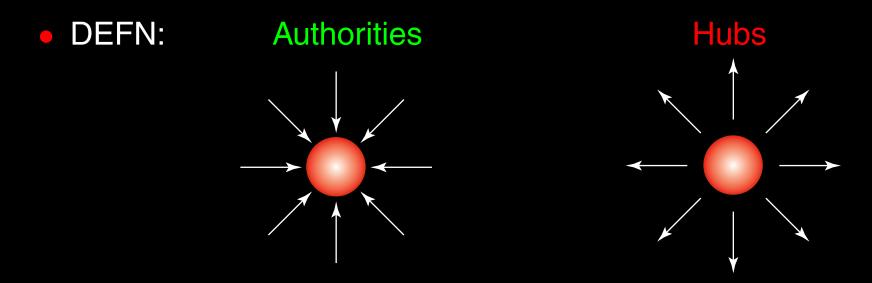
The heart of our software is PageRank[™], a system for ranking web pages developed by our founders <u>Larry Page</u> and <u>Sergey Brin</u> at Stanford University. And while we have dozens of engineers working to improve every aspect of Google on a daily basis, PageRank continues to provide the basis for all of our web search tools.

PageRank Explained

PageRank relies on the uniquely democratic nature of the web by using its

Other Way to Use Web Graph for Search

 give each page 2 scores (hub and authority scores) instead of just 1.



- pages can be both hubs and authorities (EX: ATL airport)
- Good hub pages point to good authority pages, and good authorities are pointed to by good hubs.
- HITS hub and authority score given to each page HITS - (Hypertext Induced Topic Search)



Determine Authority & Hub Scores

• a_i = authority score for P_i • h_i = hub score for P_i

HITS Algorithm

(J. Kleinberg 1998)

 $\mathbf{L}_{ij} = \begin{cases} \mathbf{1} & P_i \to P_j \\ \mathbf{0} & P_i \not\to P_j \end{cases}$

Hypertext Induced Topic Search

Successive Refinement

- Start with $h_i(0) = 1$ for all pages P_i
- Successively refine rankings

- For
$$k = 1, 2, ...$$

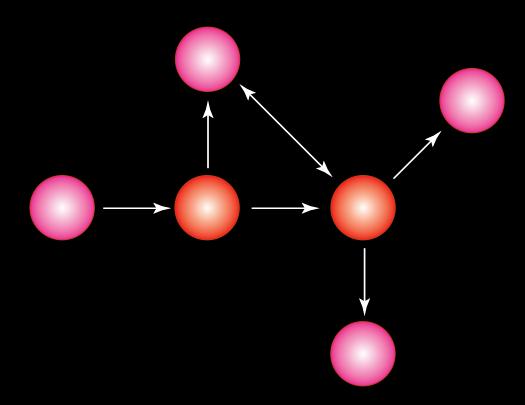
 $a_i(k) = \sum_{j:P_j \to P_i} h_j(k-1) \Rightarrow a_k = L^T h_{k-1}$
 $h_i(k) = \sum_{j:P_i \to P_j} a_j(k) \Rightarrow h_k = La_k$

 $- A = L^{T}L \quad a_{k} = Aa_{k-1} \rightarrow e\text{-vector}$ $- H = LL^{T} \quad h_{k} = Hh_{k-1} \rightarrow e\text{-vector}$



HITS Neighborhood Graph

- 1. Find relevant set by consulting inverted file
- 2. Build neighborhood graph



3. Compute authority & hub scores for just the neighborhood

HITS Example

2

3

1

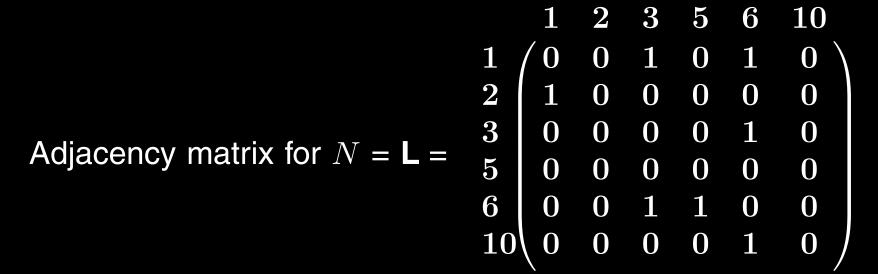
10

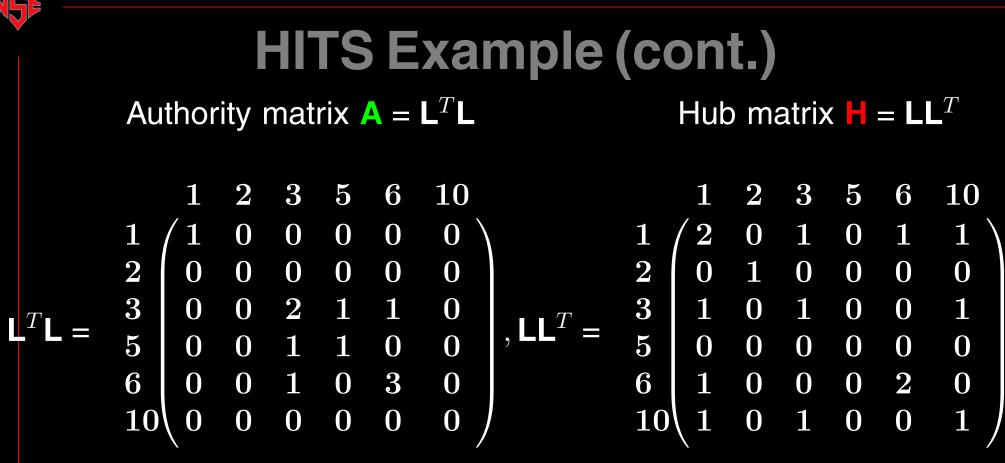
6

5

- 1. Relevant set = [1, 6]
- 2. Neighborhood graph N

3. Compute authority & hub scores.





Authority score vector a

$$\mathbf{a}^{T} = \left(\begin{array}{cccccc} 1 & 2 & 3 & 5 & 6 & 10 \\ \mathbf{a}^{T} = \left(\begin{array}{cccccc} 0 & 0 & .3660 & .1340 & .5 & 0 \end{array}\right)$$

Hub score vector h

$$\begin{array}{cccccccccc} 1 & 2 & 3 & 5 & 6 & 10 \\ \mathbf{h}^T = \left(.3660 & 0 & .2113 & 0 & .2113 & .2113 \right) \end{array}$$

HITS Convergence

- HITS with normalization step always converges.
- Rate of convergence depends on eigengap $\lambda_1 \lambda_2$.
- BUT λ_1 may be a repeated root \Rightarrow nonunique solutions. Different h_0 and a_0 can lead to different h_{∞} and a_{∞} .
- h_{∞} and a_{∞} can contain 0 values for some pages, which is undesirable in ranking context.

Pros & Cons

Advantages

- Returns satisfactory results
 - Client gets both authority & hub scores
- Some flexibility

Disadvantages

- Too much must happen while client is waiting; query-dependent
 - Custom built neighborhood graph needed for each query
 - Two eigenvector computations needed for each query
- Scores can be manipulated by creating artificial hubs

Modified HITS in Teoma

Other Approach to Ranking: PageRank

The PageRank Idea

(Sergey Brin & Lawrence Page 1998)

Ranking is preassigned

(An off-line calculation)

- Your page P has some rank r(P)
- Adjust r(P) higher or lower depending on ranks of pages that point to P
- Importance is not just number, but *quality* of in-links
 - role of outlinks relegated
 - much less sensitive to spamming

PageRank

The Definition

•
$$r(P) = \sum_{P \in \mathcal{B}_P} \frac{r(P)}{|P|}$$
 — $\mathcal{B}_P = \{ \text{all pages pointing to } P \}$
— $|P| = \text{number of out links from } P$

 $_{i}(P)$

Successive Refinement

- Start with $r_0(P_i) = 1/n$ for all pages P_1, P_2, \ldots, P_n
- Iteratively refine rankings for each page

$$- r_1(P_i) = \sum_{P \in \mathcal{B}_{P_i}} \frac{r_0(P)}{|P|}$$

$$- r_2(P_i) = \sum_{P \in \mathcal{B}_{P_i}} \frac{r_1(P)}{|P|}$$

$$\vdots$$

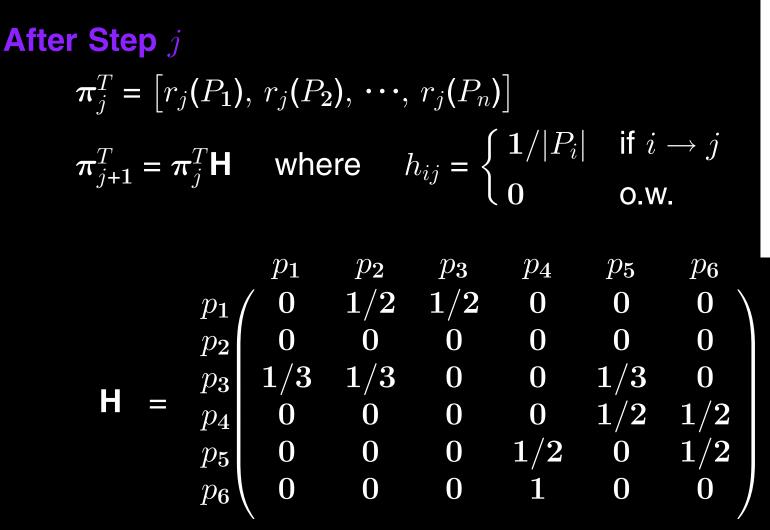
$$- r_{j+1}(P_i) = \sum_{P \in \mathcal{B}_{P_i}} \frac{r_1(P)}{|P|}$$

In Matrix Notation

After Step j $\pi_j^T = [r_j(P_1), r_j(P_2), \dots, r_j(P_n)]$ $\pi_{j+1}^T = \pi_j^T \mathbf{H}$ where $h_{ij} = \begin{cases} \mathbf{1}/|P_i| & \text{if } i \to j \\ \mathbf{0} & \text{o.w.} \end{cases}$

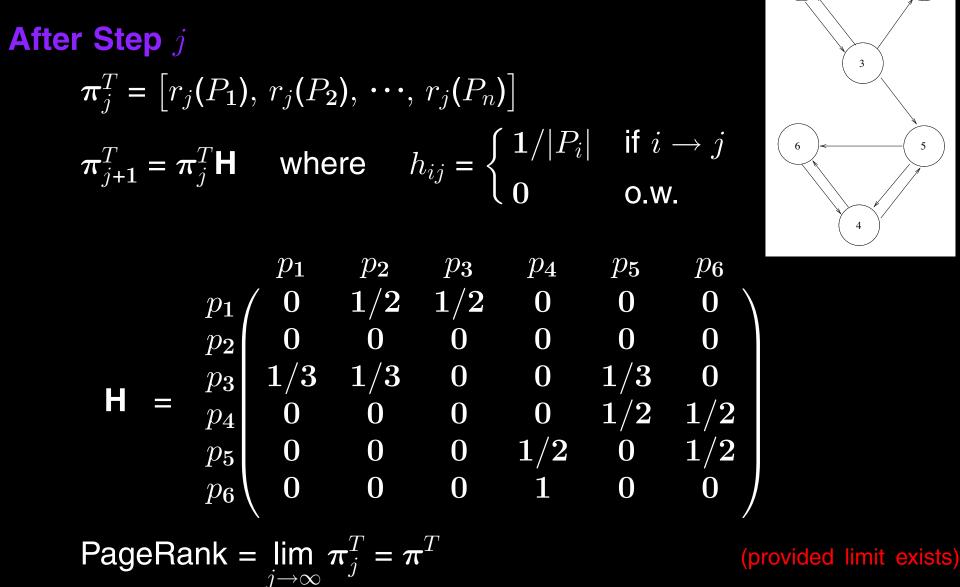
S

In Matrix Notation



In Matrix Notation

5



It's Almost a Markov Chain

H has row sums = 1 for ND nodes, row sums = 0 for D nodes

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• **H** has row sums = 1 for ND nodes, row sums = 0 for D nodes Stochasticity Fix: $\mathbf{S} = \mathbf{H} + \mathbf{av}^T$. (*a_i*=1 for *i*∈*D*, 0, o.w.)

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$$\mathbf{S} = \begin{bmatrix} 0 & 1/2 & 1/2 & 0 & 0 & 0 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/3 & 1/3 & 0 & 0 & 1/3 & 0 \\ 0 & 0 & 0 & 0 & 1/2 & 1/2 \\ 0 & 0 & 0 & 1/2 & 0 & 1/2 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix}, \text{where } \mathbf{a} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \mathbf{v}^{T} = 1/6 \mathbf{e}^{T}$$

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• Each π_i^T is a probability distribution vector

 $\left(\sum_{i}r_{j}(P_{i})=1\right)$

• $\pi_{j+1}^T = \pi_j^T \mathbf{S}$ is random walk on the graph defined by links • $\pi^T = \lim_{j \to \infty} \pi_j^T$ = stationary probability distribution

Web Surfer Randomly Clicks On Links

(Back button not a link)

Long-run proportion of time on page P_i is π_i

Problems

Web Surfer Randomly Clicks On Links (Back button not a link)

Long-run proportion of time on page P_i is π_i

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Dead end page (nothing to click on) $(\pi^T \text{ not well defined})$ Could get trapped into a cycle $(P_i \rightarrow P_j \rightarrow P_i)$ (No convergence)

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Convergence

Markov chain must be irreducible and aperiodic

Web Surfer Randomly Clicks On Links (Back button not a link) Long-run proportion of time on page P_i is π_i **Problems** Dead end page (nothing to click on) $(\pi^T \text{ not well defined})$ Could get trapped into a cycle $(P_i \rightarrow P_i \rightarrow P_i)$ (No convergence) Convergence Markov chain must be irreducible and aperiodic DEFN: a chain is *irreducible* if every page is reachable from every other page. DEFN: every *reducible* chain can be permuted to the form $\begin{vmatrix} \mathbf{X} & \mathbf{Y} \\ \mathbf{0} & \mathbf{Z} \end{vmatrix}$.

Bored Surfer Enters Random URL

Irreducibility Fix: $\mathbf{G} = \alpha \mathbf{S} + (1 - \alpha) \mathbf{E}$ $e_{ij} = 1/n$ $\alpha \approx .85$ $\mathbf{G} = \alpha \mathbf{H} + \alpha \mathbf{a} \mathbf{v}^T + (1 - \alpha) \mathbf{E}$ (trivially irreducible)

• π^T is now guaranteed to exist and be unique and power method is guaranteed to converge to π^T .

Bored Surfer Enters Random URL Irreducibility Fix: $\mathbf{G} = \alpha \mathbf{S} + (1 - \alpha) \mathbf{E}$ $e_{ij} = 1/n$ $\alpha \approx .85$ $\mathbf{G} = \alpha \mathbf{H} + \alpha \mathbf{a} \mathbf{v}^T + (1 - \alpha) \mathbf{E}$ (trivially irreducible)

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• Different $\mathbf{E} = \mathbf{e}\mathbf{v}^T$ and α allow customization & speedup, yet rank-one update maintained; $\mathbf{G} = \alpha \mathbf{H} + (\alpha \mathbf{a} + (1 - \alpha) \mathbf{e})\mathbf{v}^T$

$$\mathbf{\hat{s}} = \alpha \mathbf{\hat{S}} + (1 - \alpha)\mathbf{\hat{E}} = \begin{bmatrix} 1/60 & 7/15 & 7/15 & 1/60 & 1/60 & 1/60 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 19/60 & 19/60 & 1/60 & 1/60 & 19/60 & 1/60 \\ 1/60 & 1/60 & 1/60 & 1/60 & 7/15 & 7/15 \\ 1/60 & 1/60 & 1/60 & 1/60 & 7/15 & 1/60 & 7/15 \\ 1/60 & 1/60 & 1/60 & 11/12 & 1/60 & 1/60 \end{bmatrix}$$

Computing π^{T}

A Big Problem

Solve
$$\pi^T = \pi^T \mathbf{G}$$

$$\pi^T(\mathbf{I}-\mathbf{G})=\mathbf{0}$$

(stationary distribution vector)

(too big for direct solves)

¹² CLEVE'S **THE WORLD'S LARGEST** CORNER **MATRIX COMPUTATION**

Google's PageRank is an eigenvector of a matrix of order 2.7 billion.

One of the reasons why Google is such an effective search engine is the PageRank[™] algorithm, developed by Google's founders, Larry Page and Sergey Brin, when they were graduate students at Stanford University. PageRank is determined entirely by the link structure of the Web. It is recomputed about once a month and does not involve any of the actual content of Web pages or of any individual query. Then, for any particular query, Google finds the pages on the Web that match that query and lists those pages in the order of their PageRank.

Imagine surfing the Web, going from page to page by randomly choosing an outgoing link from one page to get to the next. This can lead to dead ends at pages with no outgoing links, or cycles around cliques of interconnected pages. So, a certain fraction of the time, simply choose a random page from anywhere on the Web. This theoretical random walk of the Web is a *Markov chain* or *Markov process*. The limiting probability that a dedicated random surfer visits any particular page is its PageRank. A page has high rank if it has links to and from other pages with high rank.

Let W be the set of Web pages that can reached by following a chain of hyperlinks starting from a page at Google and let n be the number of pages in W. The set W actually varies with time, but in May 2002, n was about 2.7 billion. Let G be the n-by-n connectivity matrix of

BY CLEVE MOLER

It tells us that the largest eigenvalue of *A* is equal to one and that the corresponding eigenvector, which satisfies the equation

x = Ax,

exists and is unique to within a scaling factor. When this scaling factor is chosen so that

 $\sum_i x_i = 1$

then *x* is the state vector of the Markov chain. The elements of *x* are Google's PageRank.

If the matrix were small enough to fit in MATLAB, one way to compute the eigenvector *x* would be to start with a good approximate solution, such as the PageRanks from the previous month, and simply repeat the assignment statement

x = Ax

until successive vectors agree to within specified tolerance. This is known as the power method and is about the only possible approach for very large *n*. I'm not sure how Google actually computes PageRank, but one step of the power method would require one pass over a database of Web pages, updating weighted reference counts generated by the hyperlinks between pages.

Computing π^{T}

A Big Problem

Solve $\pi^T = \pi^T \mathbf{G}$ (stationary distribution vector) $\pi^T (\mathbf{I} - \mathbf{G}) = 0$ (too big for direct solves) Start with $\pi_0^T = \mathbf{e}/n$ and iterate $\pi_{i+1}^T = \pi_i^T \mathbf{G}$ (power method)

Power Method to compute PageRank

 $\boldsymbol{\pi}_{\mathbf{0}}^{T} = \mathbf{e}^{T}/n$

until convergence, do

$$oldsymbol{\pi}_{j+1}^T = oldsymbol{\pi}_j^T$$
 G

(dense computation)

end

S

Power Method to compute PageRank

 $\boldsymbol{\pi}_{\mathbf{0}}^{T} = \mathbf{e}^{T}/n$

SP

until convergence, do

X $\pi_{j+1}^T = \pi_j^T \mathbf{G}$ (dense computation)

•
$$\boldsymbol{\pi}_{j+1}^T = \alpha \; \boldsymbol{\pi}_j^T \; \mathbf{S} + (\mathbf{1} - \alpha) \; \boldsymbol{\pi}_j^T \; \mathbf{e} \; \mathbf{v}^T$$

(sparser computation)

end

Power Method to compute PageRank

 $\boldsymbol{\pi}_{\mathbf{0}}^{T} = \mathbf{e}^{T}/n$

SP

until convergence, do

 $\begin{array}{l} \mathsf{X} \quad \pi_{j+1}^T = \pi_j^T \; \mathsf{G} & (\text{dense computation}) \\ \\ \mathsf{X} \quad \pi_{j+1}^T = \alpha \; \pi_j^T \; \mathsf{S} + (1 - \alpha) \; \pi_j^T \; \mathsf{e} \; \mathsf{v}^T & (\text{sparser computation}) \\ \\ \bullet \quad \pi_{j+1}^T = \alpha \; \pi_j^T \; \mathsf{H} + (\alpha \; \pi_j^T \; \mathsf{a} + (1 - \alpha)) \; \mathsf{v}^T & (\text{even less computation}) \\ \\ \\ \text{end} \end{array}$

• **H** is very, very sparse with about 3-10 nonzeros per row.

• \Rightarrow one vector-matrix mult. is $O(nnz(\mathbf{H})) \approx O(n)$.

Convergence

Can prove $\lambda_2(\mathbf{G}) \leq \boldsymbol{lpha}$

(\Rightarrow asymptotic rate of convergence of PageRank method is rate at which $lpha^k
ightarrow 0$)

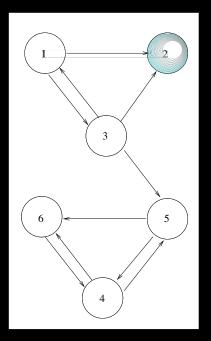
Google

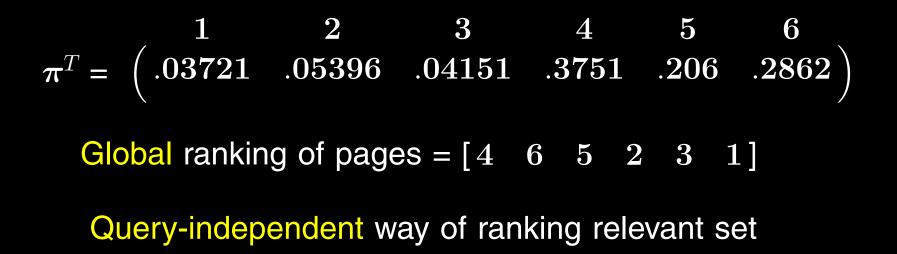
– uses α = .85

(5/6, 1/6 interpretation)

- report 50-100 iterations til convergence
- still takes days to converge

PageRank Example







PageRank Issues

Spamming

• Link Farms

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What's News-

& Business and Finance &

EWS CORP. and Liberty are no longer working together on a joint offer to take control of Hughes, with News Corp. proceeding on its own and Liberty considering an independent bid. The move threatens to cloud the process of finding a new owner for the GM unit.

(Article on Page A3)

* * * The SEC signaled it may file civil charges against Morgan Stanley, alleging it doled out IPO shares based partly on investors' commitments to buy more stock.

(Article on Page C1)

Ahold's problems deepened as U.S. authorities opened inquiries into accounting at the Dutch company's U.S. Foodservice unit. Fleming said the SEC upgraded to a formal investigation an inquiry into the food wholesaler's trade practices with suppliers.

(Articles on Page A2) * * *

• Consumer confidence fell to its lowest level since 1993, hurt by energy costs, the terrorism threat and a stagnant job market.

(Article on Page A3)

*

The industrials rebounded on

World-Wide

BUSH IS PREPARING to present Congress a huge bill for Iraq costs.

The total could run to \$95 billion depending on the length of the possible war and occupation. As horsetrading began at the U.N. to win support for a war resolution, the president again made clear he intends to act with or without the world body's imprimatur. Arms inspectors said Baghdad provided new data, including a report of a possible biological bomb. Gen. Franks assumed command of the war-operations center in Qatar. Allied warplanes are aggressively taking out missile sites that could threaten the allied troop buildup. (Column 4 and Pages A4 and A6)

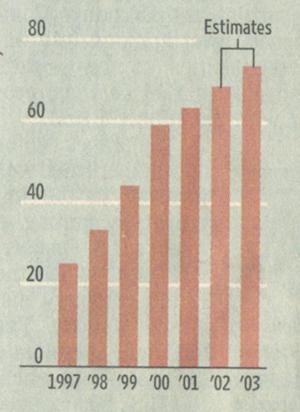
Turkey's parliament debated legislation to let the U.S. deploy 62,000 to open a northern front. Kurdish soldiers lined roads in a show of force as U.S. officials traveled into Iraq's north for an opposition conference.

* * * Powell said North Korea hasn't restarted a reactor and plutonium-processing facility at Yongbyon, hinting such forbearance might constitute an overture. But saber rattling continued a day after a missile test timed for the inauguration in Seoul. Pyongyang accused U.S. spy planes of violating its airspace and told its army to prepare for U.S. attack. (Page A14)

The FBI came under withering bipartisan criticism in a Senate Judiciary report in which Sen. Specter

Web Master

As the Web spreads... Total Internet users, by household, in millions



Sources: Forrester Research; Nielsen NetRatings

Bush to Seek up to \$95 Billion To Cover Costs of War on Iraq

By GREG JAFFE And JOHN D. MCKINNON

WASHINGTON-The Bush administration is preparing supplemental spending requests totaling as much as \$95 billion for a war with Iraq, its aftermath and new expenses to fight terrorism, officials said.

The total could be as low as \$60 billion because Pentagon budget planners don't know how long a military conflict will last, whether U.S. allies will contribute more than token sums to the effort and what damage Saddam Hussein might do

Google's U.S. presence expands

Top search engines, in millions Top shopping-referral sites, of unique visitors¹

in millions of referrals²

Google		Google			
	39.4	12.61			
Yahoo Search		DealTime			
	38.6	2.50			
MSN Search		BizRa	te		
A BARTER	36.8	1.93			
AOL Search		Overtu	ure		
	22.0	1.0	04		
Ask Jeeves		Epinions		² Number of people	
13.3	¹ Including visitors	0.7	8	the sites send to major online	
Overture 6.4	from home and work, in January 2003	CNET	6	stores, including only visitors from home, for Q4 2002	

to his own country to retaliate against conquering forces.

Budget planners also are awaiting the outcome of an intense internal debate over whether to include \$13 billion in the requests to Congress that the Pentagon says it needs to fund the broader war on terrorism, as well as for stepped up homeland security. The White House Office of Management and Budget argues that the money might not be necessary. President Bush, Defense Secretary Donald Rumsfeld and budget director Mitchell Daniels Jr. met yesterday to discuss the matter but didn't reach a final agreement. Mr. Rumsfeld plans to continue pressing his

Cat and Mouse

* *

As Google Becomes Web's Gatekeeper, Sites Fight to Get In

WSJ.com

Search Engine Punishes Firms That Try to Game System; Outlawing the 'Link Farms'

Exoticleatherwear Gets Cut Off

By MICHAEL TOTTY And MYLENE MANGALINDAN

Joy Holman sells provocative leather clothing on the Web. She wants what nearly everyone doing business online wants: more exposure on Google.

So from the time she launched exoticleatherwear.com last May, she tried all sorts of tricks to get her site to show up among the first listings when a user of Google Inc.'s popular search engine typed in "women's leatherwear" or "leather apparel." She buried hidden words in her Web pages intended to fool

Google's computers. She signed up with a service that promised to have hundreds of sites link to her online store-thereby

boosting a crucial measure in Google's system of ranking sites.

The techniques



Web Sites Fight for Prime Real Estate on Google Hom

Continued From First Page advertising that tried to capitalize on Google's formula for ranking sites. In effect, SearchKing was offering its clients a chance to boost their own Google rankings by buying ads on more-popular sites. SearchKing filed suit against the search company in federal court in Oklahoma, claiming that Google "purposefully devalued" SearchKing and its customers, damaging its reputation and hurting its advertising sales.

Google won't comment on the case. In court filings, the company said SearchKing "engaged in behavior that would lower the quality of Google search results" and alter the company's ranking system.

Google, a closely held company founded by Stanford University graduate students Sergey Brin and Larry Page, says Web companies that want to rank high should concentrate on improving their Web pages rather than gaming its system. "When people try to take scoring into their own hands, that turns into a worse experience for users," says Matt Cutts, a Google software engineer.

Coding Trickery

Efforts to outfox the search engines have been around since search engines first became popular in the early 1990s. Early tricks included stuffing thousands of widely used search terms in hidden coding, called "metatags." The coding fools a search engine into identifying a site with popular words and phrases that may not actually appear on the site.

Another gimmick was hiding words or terms against a same-color background. The hidden coding deceived search engines that relied heavily on the number of times a word or phrase appeared in ranking a site. But Google's system, based on links, wasn't fooled.

Mr. Brin, 29, one of Google's two founders and now its president of technology, boasted to a San Francisco searchengine conference in 2000 that Google wasn't worried about having its results clogged with irrelevant results because its search methods couldn't be manipulated.

That didn't stop search optimizers from finding other ways to outfox the system. Attempts to manipulate Google's results even became a sport, called Goole benching. Desplotters would the to creating Web sites that were nothing more than collections of links to the clients' site, called "link farms." Since Google ranks a site largely by how many links or "votes" it gets, the link farms could boost a site's popularity.

In a similar technique, called a link exchange, a group of unrelated sites would agree to all link to each other, thereby fooling Google into thinking the sites have a multitude of votes. Many sites also found they could buy links to themselves to boost their rankings.

Ms. Holman, the leatherwear retailer, discovered the consequences of trying to fool Google. The 42-year-old hospital laboratory technician, who learned computer skills by troubleshooting her hospital's

'The big search engines determine the laws of how commerce runs,' says Mr. Massa.

equipment, operates her online apparel store as a side business that she hopes can someday replace her day job.

When she launched her Exotic Leather Wear store from her home in Mesa, Ariz., she quickly learned the importance of appearing near the top of search-engine results, especially on Google. She boned up on search techniques, visiting online discussion groups dedicated to search engines and reading what material she could find on the Web.

At first, Ms. Holman limited herself to modest changes, such as loading her page with hidden metatag coding that would help steer a search toward her site when a user entered words such as "haltertops" or "leather miniskirts." Since Google doesn't give much weight to metatags in determining its rankings, the efforts had little effect on her search results.

She then received an e-mail advertisement from AutomatedLinks.com, a Wirral, England, company that promised to send traffic "through the roof" by linking more than 2,000 Web sites to hers. Aside from attracting customers, the links were designed to improve her site's caserch engine rankings by taking In theory, when Google encounters the AutomatedLinks code, it treats it as a legitimate referral to the other sites and counts them in toting up the sites' popularity.

Shortly after Ms. Holman signed up with AutomatedLinks in July, she read on an online discussion group that Google objected to such link arrangements. She says she immediately stripped the code from her Web pages. For a while her site gradually worked its way up in Google search results, and business steadily improved because links to her site still remained on the sites of other AutomatedLinks customers. Then, sometime in November, her site was suddenly no longer appearing among the top results. Her orders plunged as much as 80%.

Ms. Holman, who e-mailed Google and AutomatedLinks, says she has been unable to get answers. But in the last few months, other AutomatedLinks customers say they have seen their sites apparently penalized by Google. Graham McLeay, who runs a small chauffeur service north of London, saw revenue cut in half during the two months he believes his site was penalized by Google.

The high-stakes fight between Google and the optimizers can leave some Website owners confused. "I don't know how people are supposed to judge what is right and wrong," says Mr. McLeay.

AutomatedLinks didn't respond to requests for comment. Google declined to comment on the case. But Mr. Cutts, the Google engineer, warns that the rules are clear and that it's better to follow them rather than try to get a problem fixed after a site has been penalized. "We want to return the most relevant pages we can," Mr. Cutts says. "The best way for a site owner to do that is follow our guidelines."

Crackdown

Google has been stepping up its enforcement since 2001. It warned Webmasters that using trickery could get their sites kicked out of the Google index and it provided a list of forbidden activities, including hiding text and "link schemes," such as the link farms. Google also warned against "cloaking" – showing a search engine a page that's designed to score well while giving visitors a different, more attractive page – or creating multiple Web addresses that take visitors to a single site.

To stay one step ahead of the Web

homa City-based SearchKing, an online directory for hundreds of small, specialty Web sites. SearchKing also sells advertising links designed both to deliver traffic to an advertiser and boost its rankings in Google and other search results.

Bob Massa, SearchKing's chief executive, last August launched the PR Ad Network as a way to capitalize on Google's page-ranking system, known as Page-Rank. PageRank rates Web sites on a scale of one to 10 based on their popularity, and the rankings can be viewed by Web users if they install special Google software. PR Ad Network sells ads that are priced according to a site's Page-Rank, with higher-ranked sites commanding higher prices. When a site buys an advertising link on a highly ranked site, the ad buyer could see its ratings improve because of the greater weight Google gives to that link.

Shortly after publicizing the ad network, Mr. Massa discovered that his site suddenly dropped in Google's rankings. What's more, sites that participated in the separate SearchKing directory also had their Google rankings lowered. He filed a lawsuit in Oklahoma City federal court, claiming Google was punishing him for trying to profit from the company's page-ranking system.

A Google spokesman won't comment on the case. In its court filings, Google said it demoted pages on the SearchKing site because of SearchKing's attempts to manipulate search results. The company has asked for the suit to be dismissed, arguing that the PageRank represents its opinion of the value of a Web site and as such is protected by the First Amendment.

"The big search engines determine the laws of how commerce runs," says Mr. Massa, who is persisting with the lawsuit even though the sites have had their page rankings partly restored. "Someone needs to demand accountability."

Google is taking steps that many say could satisfy businesses trying to boost their rankings. Google has long sold sponsored links that show up on the top of many search-results pages, separate from the main listings. Last year, the company expanded its paid-listings program, so that there are now more slots where sites can pay for a prominent place in the results. Many sites now are turning to advertising instead of tactics to ontimize their rankings.

Home Depot E Amid First Qu

By CHAD TERHUNE

ATLANTA-Home Depot Inc. repor fiscal fourth-quarter earnings decli 3.4% on disappointing sales.

Speaking to investors and induanalysts, the company's chairman chief executive, Bob Nardelli, s Home Depot is prepared to win b dissatisfied customers and answe competitive challenge from its chief val with remodeled stores, increased ventory and improved customer serve

The nation's largest home-impr ment retailer said net income for the q ter ended Feb. 2 decreased to \$686 mill or 30 cents a share, from \$710 millior 30 cents a share, a year earlier. Sales 2% to \$13.21 billion from \$13.49 billion, first quarterly sales decline in the cor ny's 24-year history. Home Depot n the latest quarter was a week shorter a year earlier. Using comparable 13-w periods, the company said quarterly s increased 5% and net income rose 8.

Same-store sales, or sales at st open at least a year, declined 6% in quarter. Home Depot said stronger s last month offset a disastrous Decen and helped the retailer avoid its ea estimate that same-store sales could as much as 10%. In 4 p.m. New Stock Exchange composite trading, H Depot shares rose 66 cents to \$22.84

Fiat Patriarch Is Set to Becom

By ALESSANDRA GALLONI

ROME—Umberto Agnelli is due t named Fiat SpA chairman on Friday, ping into the driver's seat as the Italian glomerate works on an 11th-hour refin ing of its unprofitable car unit.

Mr. Agnelli, the 68-year-old broth Fiat patriarch Gianni Agnelli, who last month, was widely expected to over from current chairman, H Fresco, later this year. But Mr. Fr who has served as chairman since



PageRank Issues

Spamming

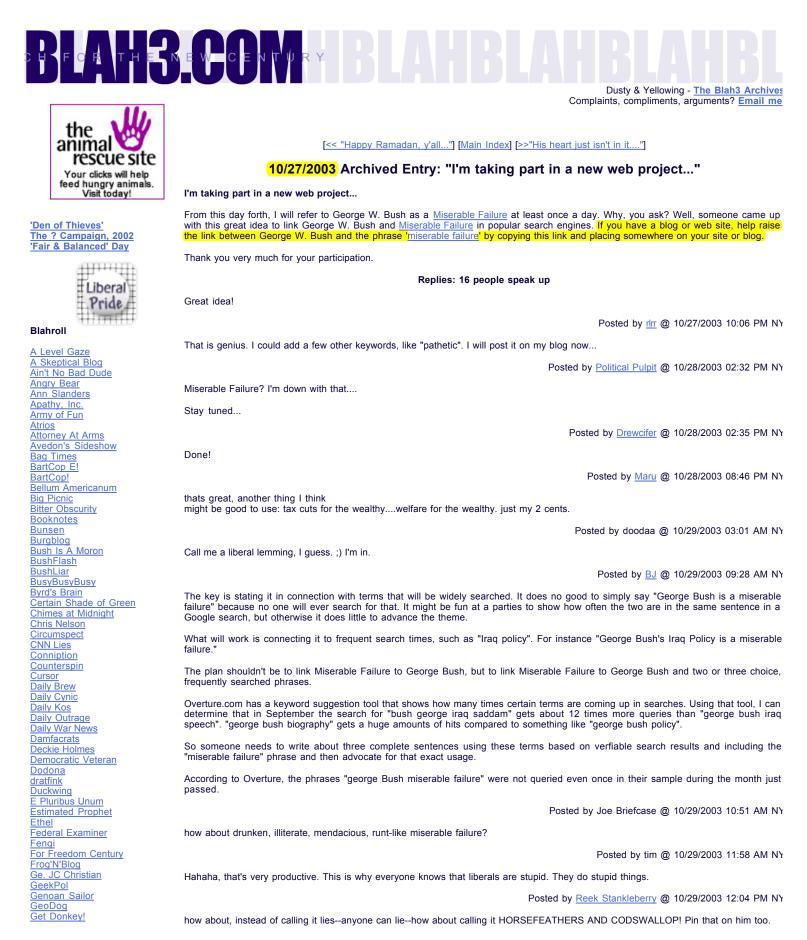
- Link Farms
- Google Bombs

Go

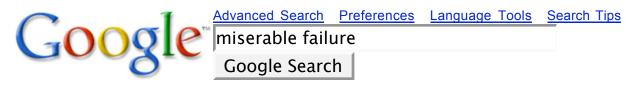
SEARCH

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		EDITION	WATCH/LISTEN TO BBC NEWS		
News Front Page	Last Updated: Sunday, 7 December,	2003, 15:04 GMT			
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- -	'Miserable failure' lir	nks to Bush			
Africa	George W Bush has been Google bombed.		SEE ALSO:		
Americas			WMD spoof is internet hit 04 Jul 03 West Midlands		
Asia-Pacific	Web users entering the words "miserable failure" into the popular		Google hit by link bombers		
Europe	search engine are directed to the biography of the president on the		13 Mar 02 Science/Nature		
Middle East	White House website.		RELATED INTERNET LINKS:		
South Asia	The trick is possible because Google		White House		
UK Business	searches more than just the contents of web pages - it also		Google bombing		
Health	counts how often a site is linked to,	AP	The BBC is not responsible for the content of external internet sites		
Science/Nature	and with what words.	Bush has been the target of similar pranks before			
Technology	Thus, members of an online commun		TOP AMERICAS STORIES NOW US army battles to keep soldiers		
Entertainment	searches - called "Google bombing" - one.	by linking their sites to a chosen	Report backs US Catholic bishops		
Have Your Say	Weblogger Adam Mathes is credited w when he used it to link the phrase "t	Weblogger Adam Mathes is credited with inventing the practice in 2001, when he used it to link the phrase "talentless hack" to a friend's website.			
Country Profiles In Depth	The search engine can be manipulated by a fairly small group of users, one report suggested.				
Programmes	Newsday newspaper says as few as 32 web pages with the words "miserable failure" link to the Bush				
RELATED SITES	biography.	name in the address bar, make sure that it is spelled			
BBB SPORT	The Bush administration has been	correctly (IRAQ)			
BBC WEATHER	on the receiving end of pointed Google bombs before.	Prank website			
BBC ON THIS DAY	In the run-up to the Iraq war interne	et users manipulated Google so the			
LANGUAGES	In the run-up to the Iraq war, internet users manipulated Google so the phrase "weapons of mass destruction" led to a joke page saying "These				
ESPAÑOL	Weapons of Mass Destruction cannot	be displayed."			
BRASIL CARIBBEAN	The site suggests "clicking the regime change button", or "If you are George Bush and typed the country's name in the address bar, make sure that it is spelled correctly (IRAQ)".				
	醛 E-mail this to a friend	Printable version			
	LINKS TO MORE AMERICAS STORIES				
	Select	Go			
	E-mail services Desktop ticker	' MODIIES/PDAS			
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http://www.blah3.com/graymatter/archives/00000654.html



Web Images Groups Directory News

Searched the web for **miserable failure**. Results **1** - **10** of about **257,000**. Search took **0.08** seconds. Tip: In most browsers you can just hit the return key instead of clicking on the search button.

Michael Moore.com

Wednesday, January 14th, 2004 I'll Be Voting For Wesley Clark / Good-Bye Mr. Bush — by Michael Moore. Many of you have written ... Description: Official site of the gadfly of corporations, creator of the film Roger and Me and the television show... Category: <u>Arts > Celebrities > M > Moore, Michael</u> www.michaelmoore.com/ - 43k - Cached - Similar pages

Biography of President George W. Bush

Home > President > Biography President George W. Bush En Español. George W. Bush is the 43rd President of the United States. He ... Description: Biography of the president from the official White House web site. Category: <u>Kids and Teens > School Time > ... > Bush, George Walker</u> www.whitehouse.gov/president/gwbbio.html - 29k - <u>Cached</u> - <u>Similar pages</u>

Biography of Jimmy Carter

Home > History & Tours > Past Presidents > Jimmy Carter. Jimmy Carter. Jimmy Carter aspired to make Government "competent and compassionate ... Description: Short biography from the official White House site. Category: <u>Society > History > ... > Presidents > Carter, James Earl</u> www.whitehouse.gov/history/presidents/jc39.html - 36k - <u>Cached</u> - <u>Similar pages</u>

Senator Hillary Rodham Clinton: Online Office Welcome Page

Dear Friend,. Thank you for visiting my on-line office! I appreciate your interest in the issues before the United States Senate. ... Description: Official US Senate web site of Senator Hillary Rodham Clinton (D - NY). Category: <u>Society > History > ... > First Ladies > Clinton, Hillary</u> clinton.senate.gov/ - 9k - <u>Cached</u> - <u>Similar pages</u>

BBC NEWS | Americas | 'Miserable failure' links to Bush

'**Miserable failure**' links to Bush. ... Prank website. Newsday newspaper says as few as 32 web pages with the words "**miserable failure**" link to the Bush biography. ... news.bbc.co.uk/2/hi/americas/3298443.stm - 31k - <u>Cached</u> - <u>Similar pages</u>

Atlantic Unbound | Politics & Prose | 2003.09.24

... Atlantic Unbound | September 24, 2003 Politics & Prose | by Jack Beatty "A **Miserable Failure**" Will Bush be re-elected? Only if voters ... www.theatlantic.com/unbound/polipro/pp2003-09-24.htm - 22k - <u>Cached</u> - <u>Similar pages</u>

miserable failure | Hillary Clinton | Hildebeest

... Miserable Failure. Quotes for the History Books. ... You may also want to check out the Miserable Failure Project. and the cuckolded dyke Project. and the ... miserable-failure.blogspot.com/ - 60k - <u>Cached</u> - <u>Similar pages</u>

Dick Gephardt for President - Welcome

... to preserve some large part of the Bush tax cut. I think retaining



PageRank Issues

Spamming

- Link Farms
- Google Bombs

Updating

• The Google Dance



<u>What's Google Dance?</u> - <u>Google Web API</u> - <u>Google Humor</u> -<u>Google Crawler</u> <u>Keyword Importance</u> - <u>Webmaster Tools</u> - <u>About Us</u> -Contact Us

Below you will find the Google Dance results for the search keyword **pagerank**. If you notice that there are any differences in results between the different Google data centers then Google is in the middle of spidering the internet. It's that simple!



			Next Page>>
GOOGLE DANICETOOL	pagerank		•
	Lets Dance!		
Thought of a great nan	ne for your site?	w	
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What's Google Dance? - Goog	le Web API - <u>Google Humor</u> - <u>Google (</u>	<u>Crawler</u>	
<u>Keyword Importance</u> - <u>We</u>	ebmaster Tools - <u>About Us</u> - <u>Contact</u>	<u>Us</u>	
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PageRank Issues

Spamming

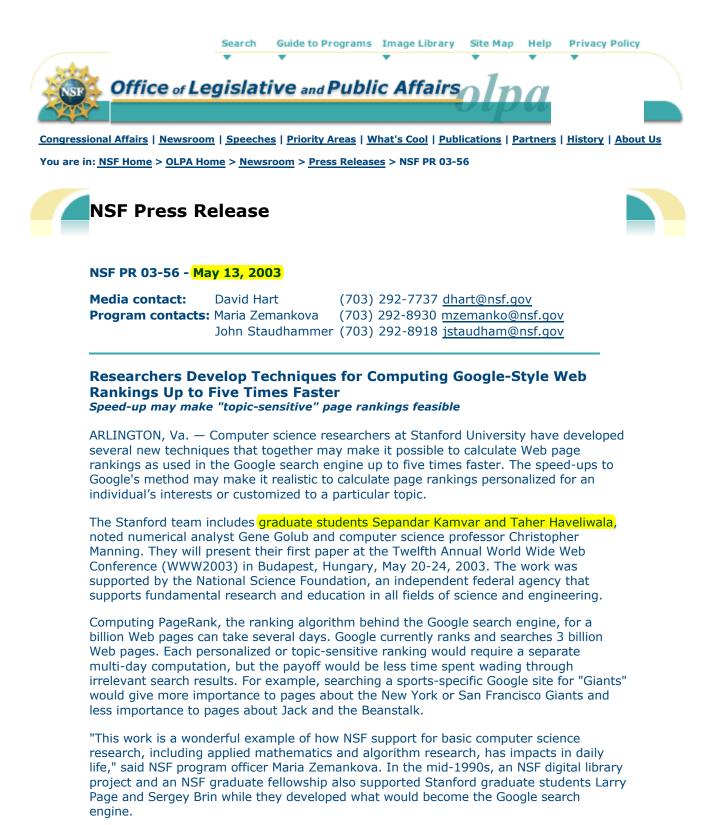
- Link Farms
- Google Bombs

Updating

• The Google Dance

Speed Improvements

• Enhancing Power Method



To speed up PageRank, the Stanford team developed a trio of techniques in numerical linear algebra. First, in the WWW2003 paper, they describe so-called "extrapolation" methods, which make some assumptions about the Web's link structure that aren't true, but permit a quick and easy computation of PageRank. Because the assumptions aren't true, the PageRank isn't exactly correct, but it's close and can be refined using the



PageRank Issues

Spamming

- Link Farms
- Google Bombs

Updating

• The Google Dance

Speed Improvements

- Enhancing Power Method
- Personalized PageRank





A stealth start-up out of Stanford University is hoping to raise the heat on one of the toughest problems in Web search--and possibly out-Google Google in the process.

Kaltix was formed in recent months by three members of Stanford's PageRank team--a research group created to advance the mathematical algorithm developed by Google co-founder and Stanford alum Larry Page that cemented Google's fame.

PageRank has helped steer people to Web sites like no other search technology before it, harnessing the link structure of the Web to determine the most popular pages. Now, Kaltix hopes to improve upon PageRank, with an attempt to speed up the underlying PageRank computations.

That, in turn, could lay the groundwork for a breakthrough in a cutting-edge area of Web search development known as "personalization," which aims to sort search results based on the specific needs and interests of individuals, instead of the consensus approach pioneered by Google.

"Kaltix is a 'stealth mode' start-up...(leveraging) research done at Stanford University as well as several new technologies developed at Kaltix to provide large-scale personalized and context-sensitive search." a



Kaltix representative said, declining to comment further.

Kaltix has disclosed few specifics about its plans or technology. But the company's general statements appear to place it in a sweet spot for innovation that's being pursued by all of the major search providers. Now that Web search has become a moneymaker for portals such as Yahoo and Microsoft's MSN. technologists from all the industry players are back in the labs developing formulas to personalize search.

Web companies outside the search industry have long made attempts to create personalization features, but most of these attempts have fallen short of expectations. Amazon.com, for example, regularly serves up book titles related to a visitor's previous purchases, which may no longer be relevant. A personalization feature offered through TiVo, a maker of video recording devices, was criticized when reports circulated that the device would recommend gay-themed television programs to viewers based on just a few program selections

Despite these flawed attempts, developers continue to have faith that personalization technology can be created that will ultimately unleash marketing and revenue opportunities.

If search developers are successful in building such technology, they could help millions of people better

integrated, low-overhead traffic monitor shows detailed information on traffic

Who/What: White paper

Intel® Xeon™ Processor MP Quick Reference Guide

Key findings:

Designed specifically for 4-way, multi-processor (MP)-based servers, available up to 2.80 GHz and 2 MB iL3 cache

Who/What: **Reference Guide** (pdf)

Free Office Pro 2003 **Evaluation Kit**

Key findings: Test-drive of all the elements of the latest Microsoft Office System; evaluate before you upgrade

Who/What: 30-day trial copy, plus a resource guide CD

ZDNet Power Centers

Webcasts & Video



Google Press Release

Google Acquires Kaltix Corp.

<u>Home</u>

All About Google

Press Center

Press Kit

Press Releases In the News Image Gallery User Testimonials Google Zeitgeist

Company

Overview Fact Sheet Management Milestones Fun Facts Investor Info Awards

Technology

Overview Life of a Query

Business Overview Case Studies



MOUNTAIN VIEW, Calif. - Sept. 30, 2003 - Google Inc. today announced it acquired Kaltix Corp., a Palo Alto, Calif.-based search technology start-up. Financial terms of the deal were not disclosed.

"Google and Kaltix share a common commitment to developing innovative search technologies that make finding information faster, easier and more relevant," said Larry Page, co-founder and president of Products at Google. "Kaltix is working on a number of compelling search technologies, and Google is the ideal vehicle for the continued development of these advancements."

Kaltix Corp. was formed in June 2003 and focuses on developing personalized and context-sensitive search technologies that make it faster and easier for people to find information on the web.

About Google

Google's innovative search technologies connect millions of people around the world with information every day. Founded in 1998 by Stanford Ph.D. students Larry Page and Sergey Brin, Google today is a top web property in all major global markets. Google's targeted advertising program, which is the largest and fastest growing in the industry, provides businesses of all sizes with measurable results, while enhancing the overall web experience for users. Google is headquartered in Silicon Valley with offices throughout North America, Europe, and Asia. For more information, visit www.google.com.

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Nathan Tyler Google Inc. +1 650-623-4311 nate@google.com



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Conclusions

- Link Analysis has drastically improved web search!
- There are many exciting open problems for CSC and MATH majors to solve.
- Often the challenge lies not in the modeling or theory, but in the massive scale of the problem.
- The continual battle between search engines and search engine optimizers means that methods must constantly adapt and innovate.
- There is huge financial potential for industrious entrepreneurs!