CC5212-1

Procesamiento Masivo de Datos Otoño 2019

Lecture 8

Information Retrieval: Ranking

Aidan Hogan aidhog@gmail.com

Apache Lucene



- Inverted Index
 - They built one so you don't have to!
 - Open Source in Java



Apache Lucene



```
🔎 Tasks 📮 Console 🛭
SearchWikiIndex [Java Application] C:\Program Files\Java\jre1.8.0 65\bin\javaw.exe (03-05-2017 12:45:22 a. m.)
Opening directory at lucene
Enter a keyword search phrase:
obama
Running query: obama
Parsed query: TITLE:obam^5.0 ABSTRACT:obam
Matching documents: 255
Showing top 10 results
        http://es.wikipedia.org/wiki/Obama Republican
                                                          Obama Republican
1
2
        http://es.wikipedia.org/wiki/Obama (Fukui)
                                                          Obama (Fukui)
        http://es.wikipedia.org/wiki/Republicanos por Obama
                                                                  Republicanos por Obama
3
        http://es.wikipedia.org/wiki/Engonga Obame
                                                          Engonga Obame
5
        http://es.wikipedia.org/wiki/Barack Obama
                                                          Barack Obama
        http://es.wikipedia.org/wiki/Michelle Obama
6
                                                          Michelle Obama
7
        http://es.wikipedia.org/wiki/Cartel %22Hope%22 de Obama Cartel "Hope" de Obama
        http://es.wikipedia.org/wiki/Transición_presidencial de Barack Obama
                                                                                   Transición presidencial de Barack Obama
8
9
        http://es.wikipedia.org/wiki/Por_qué_Obama_ganará_en_2008_y_en_2012
                                                                                   Por qué Obama ganará en 2008 y en 2012
        http://es.wikipedia.org/wiki/Ricardo Mangue Obama Nfubea
10
                                                                           Ricardo Mangue Obama Nfubea
```

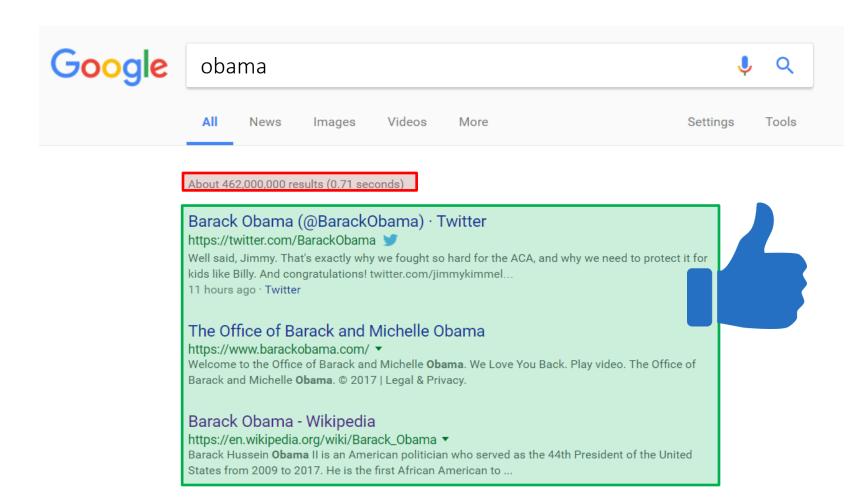




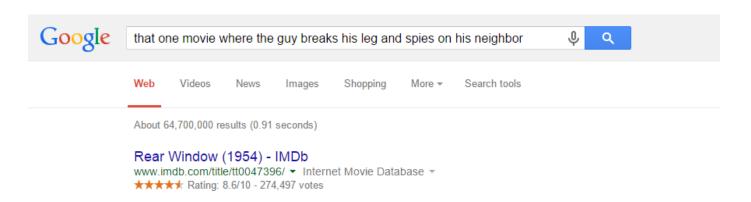
My God. It's full of win.

INFORMATION RETRIEVAL: RANKING

How Does Google Get Such Good Results?



How does Google Get Such Good Results?

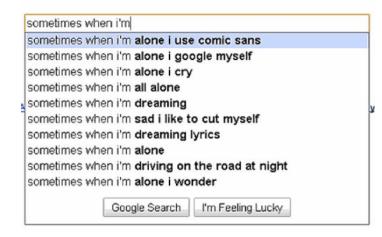




About 107,000 results (0.36 seconds)

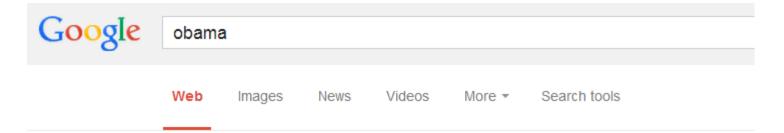






Two aspects of ranking: Relevance vs. Importance

Two Sides to Ranking: Relevance



About 16,700,000 results (0.23 seconds)

Broccoli - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Broccoli *

Broccoli is an edible green plant in the cabbage family, whose large flowering head is used as a vegetable. The word **broccoli** comes from the Italian plural of ... Cauliflower - Romanesco broccoli - Broccoli (disambiguation) - Broccolini

Broccoli - The World's Healthiest Foods

www.whfoods.com/genpage.php?tname=foodspice&dbid=9 *

Broccoli can provide you with some special cholesterol-lowering benefits if you will cook by steaming. The fiber-related components in **broccoli** do a better job ...

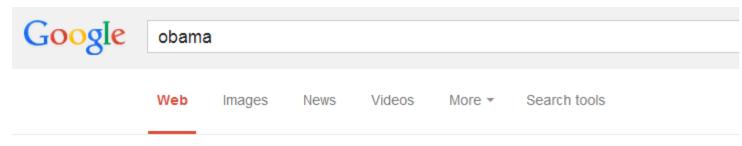
ews for broccoli

Mistakes We All Make Wan Spaghetti, Steak And

Auffington Post - 2 days ago

But in her new book Brassicas: Cooking the World's Healthiest Vegetables, she sa plunking **broccoli**, cauliflower or brussels sprouts into ...

Two Sides to Ranking: Importance



About 48,100,000 results (0.26 seconds)

Mount Obama - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Mount_Obama ▼

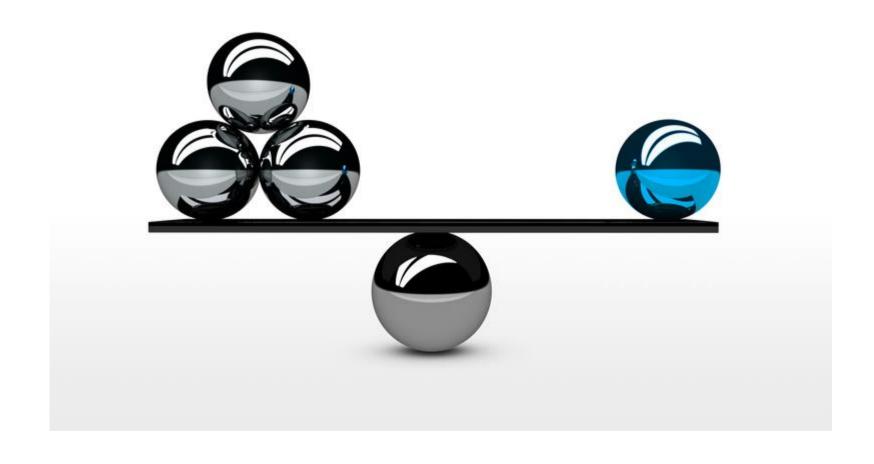
Mount Obama (known as **Boggy Peak** until August 4, 2009) is the highest point in the nation of Antigua and Barbuda and on the island of Antigua. It lies in the far ...

Images for mount obama

Report images



Relevance vs. Importance: A Balancing Act



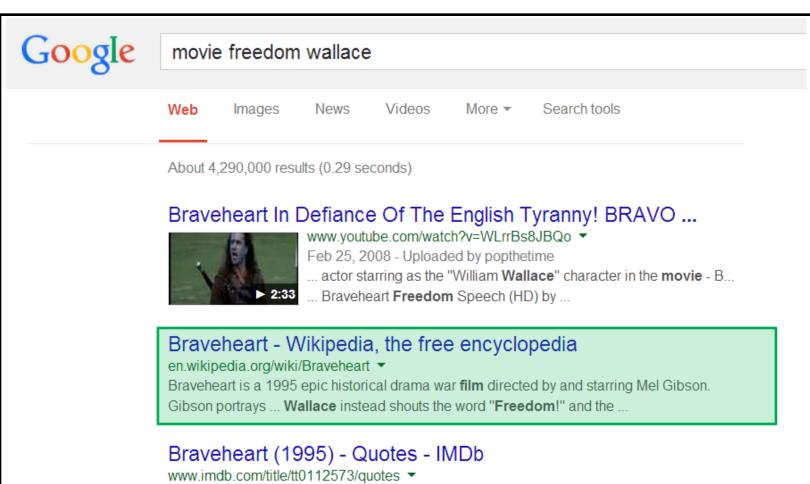
RANKING:

RELEVANCE

Example Query

Which of these three keyword terms is most "important"?

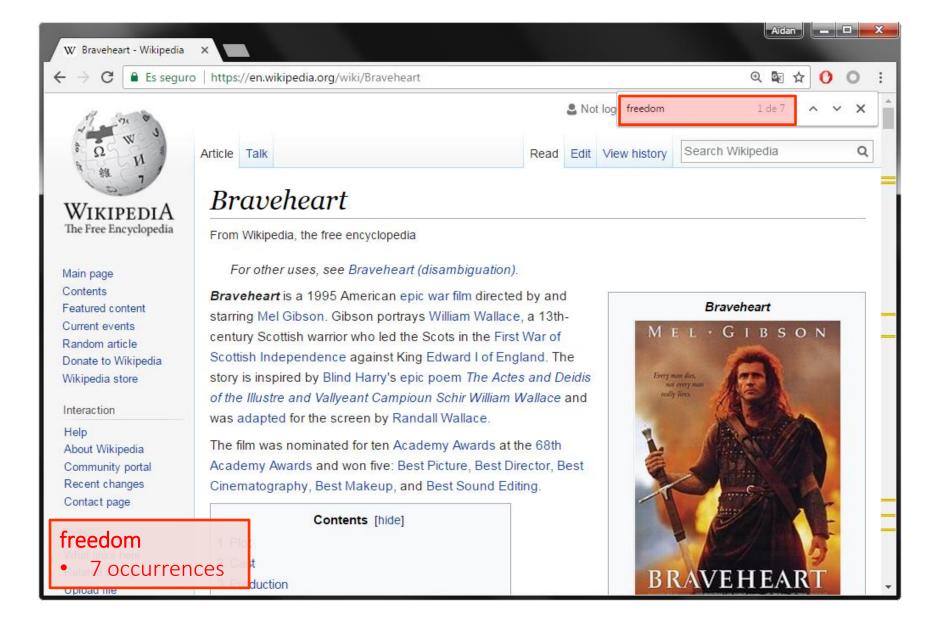




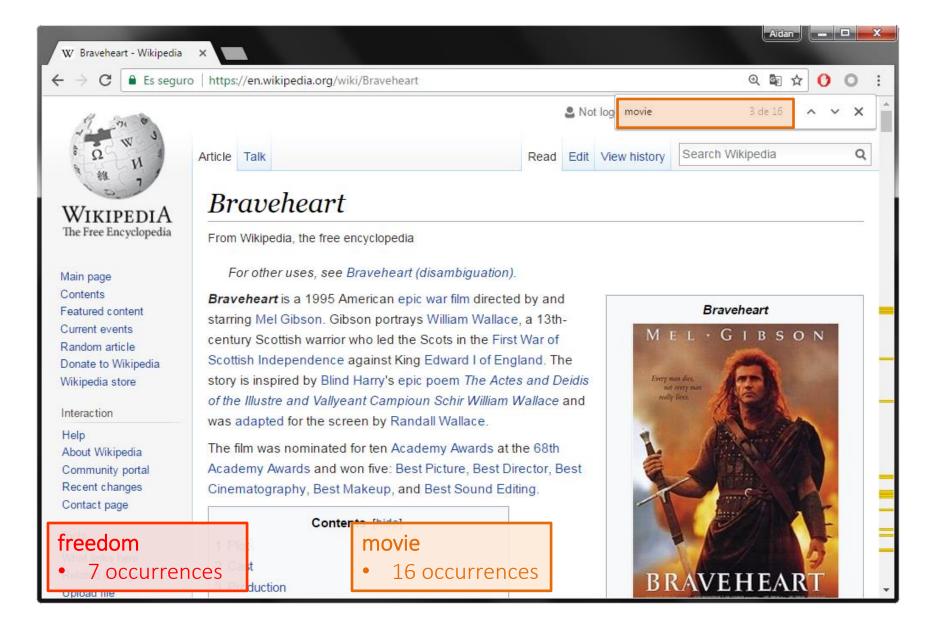
... (1995) Quotes on IMDb: Memorable quotes and exchanges from movies, TV series and

more...... William Wallace: It's all for nothing if you don't have freedom.

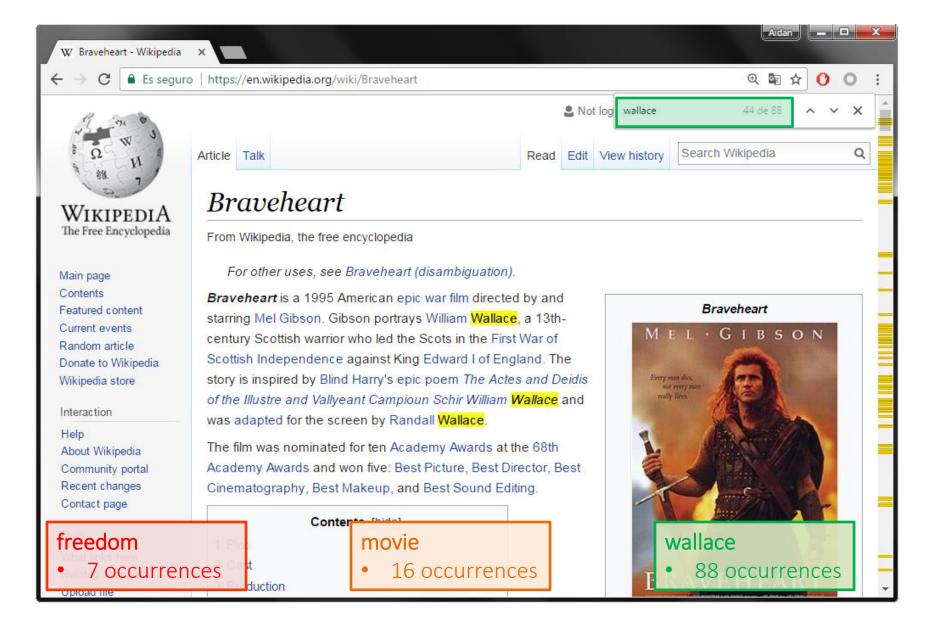
Matches in a Document



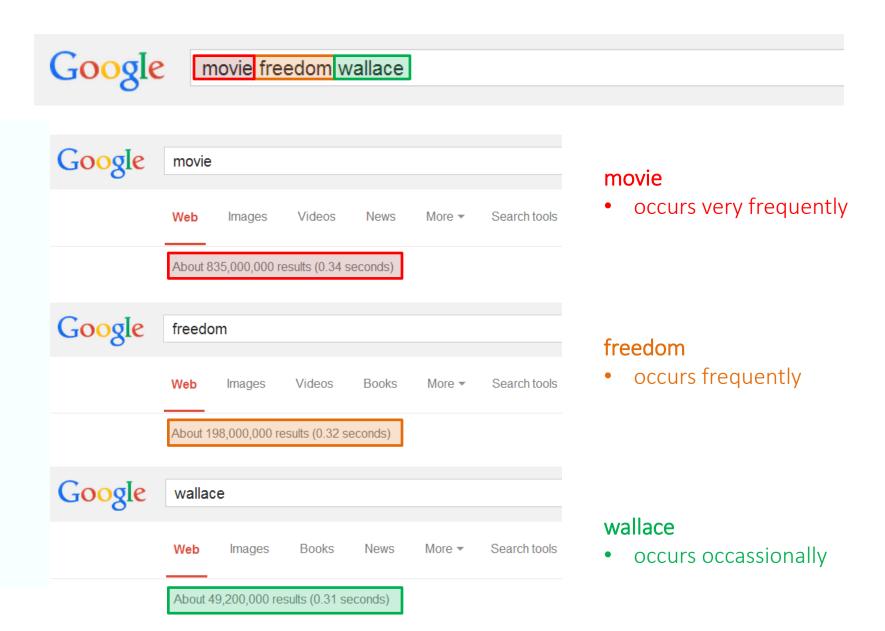
Matches in a Document



Matches in a Document



Usefulness of Words



Estimating Relevance

- Rare words more important than common words
 - wallace (49M) more important than freedom (198M)
 more important than movie (835M)

- Words occurring more frequently in a document indicate higher relevance
 - wallace (88) more matches than movie (16) more matches than freedom (7)

- TF: Term Frequency
 - Measures occurrences of a term in a document
 - tf(t,d) ... various options
 - Raw count of occurrences

$$tf(t,d) = count(t,d)$$

Logarithmically scaled

$$tf(t,d) = \log(count(t,d) + 1)$$

Normalised by document length

$$\operatorname{tf}(t,d) = \frac{\operatorname{count}(t,d)}{\sum_{t' \in d} \operatorname{count}(t',d)}$$

$$tf(t,d) = \frac{count(t,d)}{\max_{t' \in d} count(t',d)}$$

A combination / something else ☺

- IDF: Inverse Document Frequency
 - How common a term is across all documents
 - $-\operatorname{idf}(t,D)$...
 - Logarithmically scaled document occurrences

$$idf(t,D) = \log\left(\frac{|D|+1}{|\{d \in D : t \in d\}|+1}\right)$$

Note: The more rare, the larger the value

 TF—IDF: Combine Term Frequency and Inverse Document Frequency:

$$tf\text{-}idf(t,d) = tf(t,d) \times idf(t,D)$$

- Score for a query
 - Let query $q = (t_1, \ldots, t_n)$
 - Score for a query: $\operatorname{score}(q,d) = \sum_{t \in q} \operatorname{tf-idf}(t,d)$

(There are other possibilities)



Term Frequency

$$tf(t,d) = count(t,d)$$

Inverse Document Frequency

$$idf(t,D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$

\overline{t}	tf(t,d)
movie	16
freedom	7
wallace	43



Term Frequency

$$tf(t,d) = count(t,d)$$

$$idf(t, D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$

\overline{t}	tf(t,d)	$ \{d \in D : t \in d\} $
movie	16	835,000,000
freedom	7	198,000,000
wallace	43	49,200,000



Term Frequency

$$tf(t,d) = count(t,d)$$

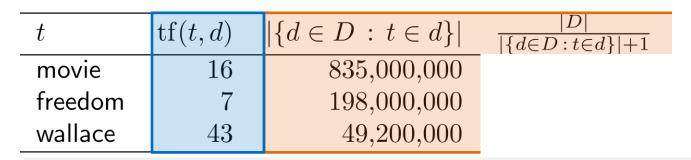
the

Web

Inverse Document Frequency

$$idf(t, D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$





Images

News

Books

More ▼

Search tools

|D| = 11,410,000,000

About 11,410,000,000 results (0.27 seconds)



Term Frequency

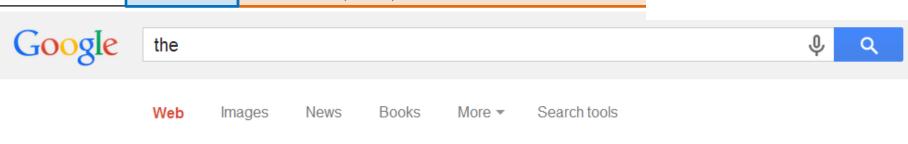
$$tf(t,d) = count(t,d)$$

Inverse Document Frequency

$$idf(t,D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$

\overline{t}	tf(t,d)	$ \{d \in D : t \in d\} $	$\frac{ D }{ \{d \in D : t \in d\} +1}$
movie	16	835,000,000	13.66
freedom	7	198,000,000	57.62
wallace	43	49,200,000	231.91



|D| = 11,410,000,000

About 11,410,000,000 results (0.27 seconds)



Term Frequency

$$tf(t,d) = count(t,d)$$

Inverse Document Frequency

$$idf(t,D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$

\overline{t}	tf(t,d)	$ \{d \in D : t \in d\} $	$\frac{ D }{ \{d \in D : t \in d\} +1}$	idf(t, D)
movie	16	835,000,000	13.66	3.77
freedom	7	198,000,000	57.62	5.84
wallace	43	49,200,000	231.91	7.85



Term Frequency

$$tf(t,d) = count(t,d)$$

Inverse Document Frequency

$$idf(t,D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$

\overline{t}	tf(t,d)	$ \{d\in D:t\in d\} $	$\frac{ D }{ \{d \in D : t \in d\} +1}$	idf(t, D)	tf-idf(t,d)
movie	16	835,000,000	13.66	3.77	60.36
freedom	7	198,000,000	57.62	5.84	40.94
wallace	43	49,200,000	231.91	7.85	337.87



Term Frequency

$$tf(t,d) = count(t,d)$$

Inverse Document Frequency

$$idf(t, D) = \log_2\left(\frac{|D|}{|\{d \in D : t \in d\}|+1}\right)$$

$$tf-idf(t,d) = tf(t,d) \times idf(t,D)$$

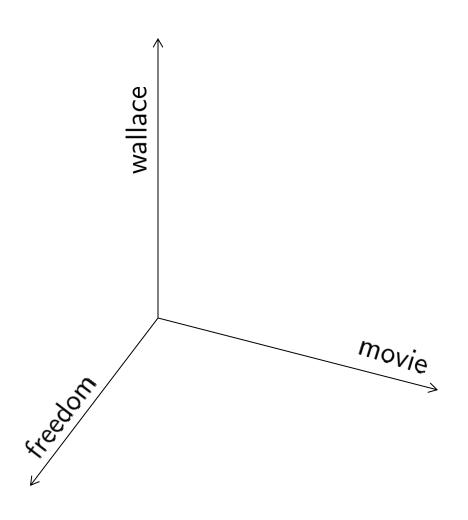
\overline{t}	tf(t,d)	$ \{d \in D : t \in d\} $	$\frac{ D }{ \{d \in D : t \in d\} +1}$	idf(t, D)	tf-idf(t,d)
movie	16	835,000,000	13.66	3.77	60.36
freedom	7	198,000,000	57.62	5.84	40.94
wallace	43	49,200,000	231.91	7.85	337.87

$$score(q, d) = \sum_{t \in q} tf-idf(t, d)$$

 $score((movie, freedom, wallace), http://en.wikipedia.org/Braveheart) \approx 439.17$

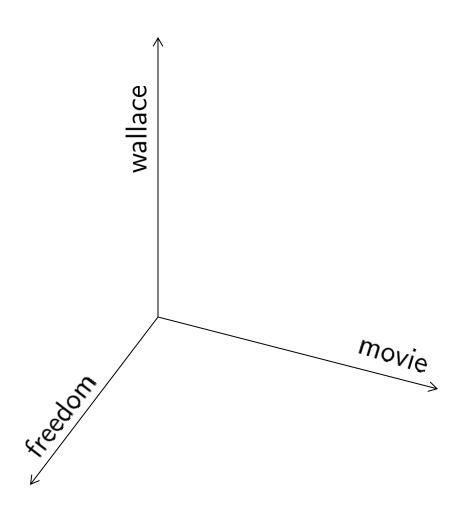
\overline{t}	tf(t,d)
movie	16
freedom	7
wallace	43

$$l = \sqrt{\sum_{t \in q} \mathrm{tf}(t, d)^2}$$



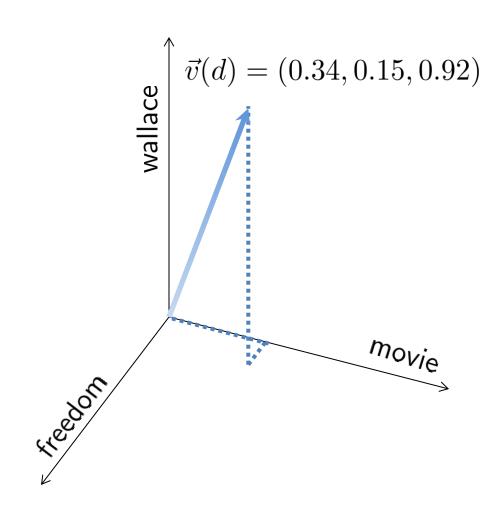
\overline{t}	tf(t,d)	$tf(t,d)^2$
movie	16	256
freedom	7	49
wallace	43	1,894

$$l = \sqrt{\sum_{t \in q} \operatorname{tf}(t, d)^2}$$



\overline{t}	tf(t,d)	$tf(t,d)^2$	$\frac{\operatorname{tf}(t,d)}{t}$
movie	16	$\phantom{00000000000000000000000000000000000$	$\frac{\iota}{0.34}$
freedom	7	49	0.15
wallace	43	1,894	0.92

$$l = \sqrt{\sum_{t \in q} \operatorname{tf}(t, d)^2}$$



Cosine Similarity

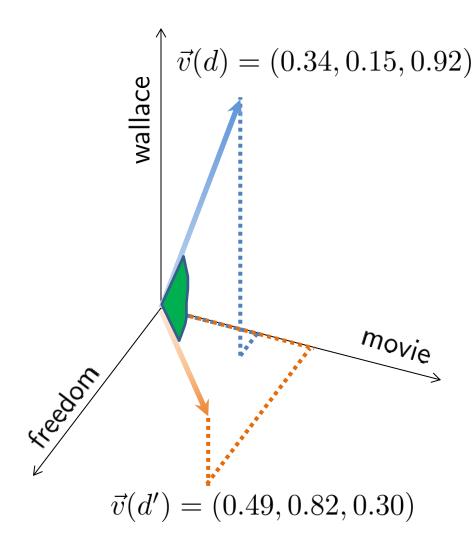
$$sim(d, d') = \vec{v}(d) \cdot \vec{v}(d')$$

t	$\vec{v}(d)$	$\vec{v}(d')$	×
movie	0.34	0.49	0.17
freedom	0.15	0.82	0.12
wallace	0.93	0.30	0.28

$$sim(d, d') \approx 0.57$$

Note:

$$\mathbf{a} \cdot \mathbf{b} = |\mathbf{a}| |\mathbf{b}| \cos(\angle(\mathbf{a}, \mathbf{b}))$$
$$|\vec{v}(d)| = |\vec{v}(d')| = 1$$



Hence the similarity is the cosine of the angle between the vectors

 TF—IDF: Combine Term Frequency and Inverse Document Frequency:

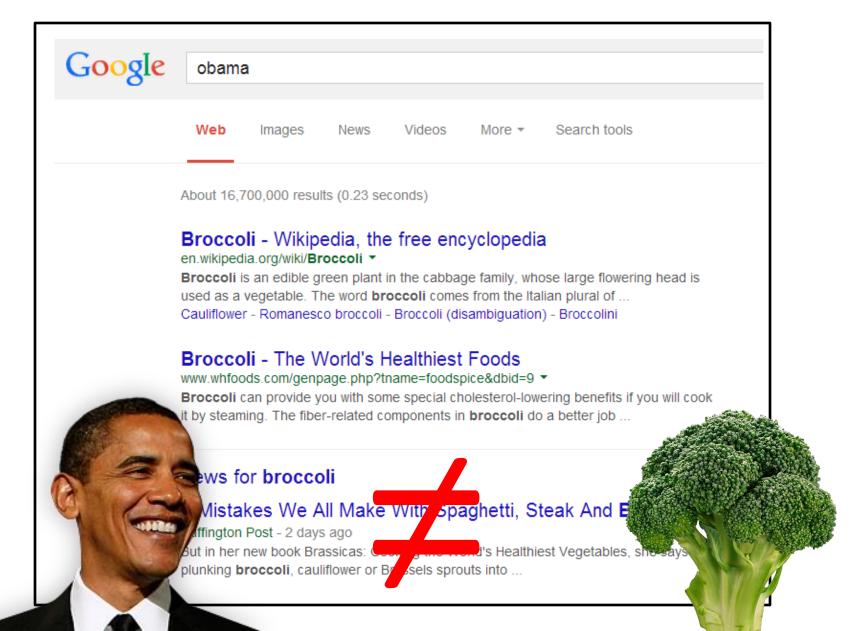
$$tf\text{-}idf(t,d) = tf(t,d) \times idf(t,D)$$

- Score for a query
 - Let query $q = (t_1, \ldots, t_n)$
 - Score for a query: $score(q, d) = \sum_{t \in q} tf\text{-}idf(t, d)$

(There are other possibilities)

... we could also use cosine similarity between query and document using TF-IDF weights

Two Sides to Ranking: Relevance



Field-Based Boosting

Not all text is equal: titles, headers, etc.

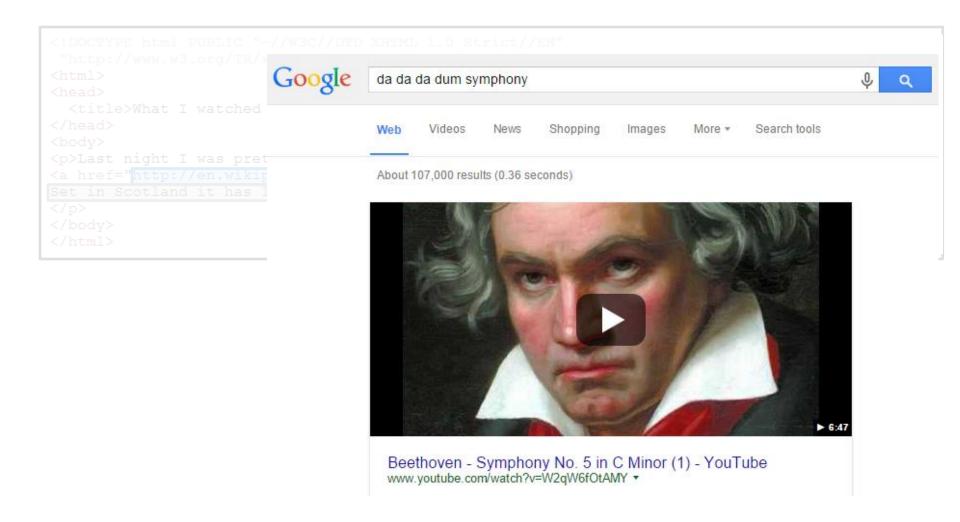


Anchor Text

See how the Web views/tags a page

Anchor Text

See how the Web views/tags a page



Lucene uses relevance scoring

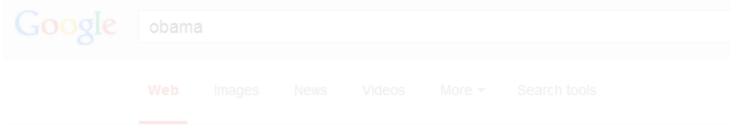


```
🔎 Tasks 📮 Console 🔀
SearchWikiIndex [Java Application] C:\Program Files\Java\jre1.8.0_65\bin\javaw.exe (03-05-2017 12:45:22 a. m.)
Opening directory at lucene
Enter a keyword search phrase:
obama
Running query: obama
Parsed query: TITLE:obam^5.0 ABSTRACT:obam
Matching documents: 255
Showing top 10 results
1
        http://es.wikipedia.org/wiki/Obama Republican
                                                          Obama Republican
2
        http://es.wikipedia.org/wiki/Obama (Fukui)
                                                          Obama (Fukui)
        http://es.wikipedia.org/wiki/Republicanos por Obama
3
                                                                  Republicanos por Obama
        http://es.wikipedia.org/wiki/Engonga Obame
4
                                                          Engonga Obame
        http://es.wikipedia.org/wiki/Barack Obama
5
                                                          Barack Obama
6
        http://es.wikipedia.org/wiki/Michelle Obama
                                                          Michelle Obama
7
        http://es.wikipedia.org/wiki/Cartel %22Hope%22 de Obama Cartel "Hope" de Obama
        http://es.wikipedia.org/wiki/Transición presidencial de Barack Obama
8
                                                                                   Transición presidencial de Barack Obama
        http://es.wikipedia.org/wiki/Por_qué_Obama_ganará_en_2008_y_en_2012
9
                                                                                   Por qué Obama ganará en 2008 y en 2012
10
        http://es.wikipedia.org/wiki/Ricardo Mangue Obama Nfubea
                                                                           Ricardo Mangue Obama Nfubea
```

RANKING:

IMPORTANCE

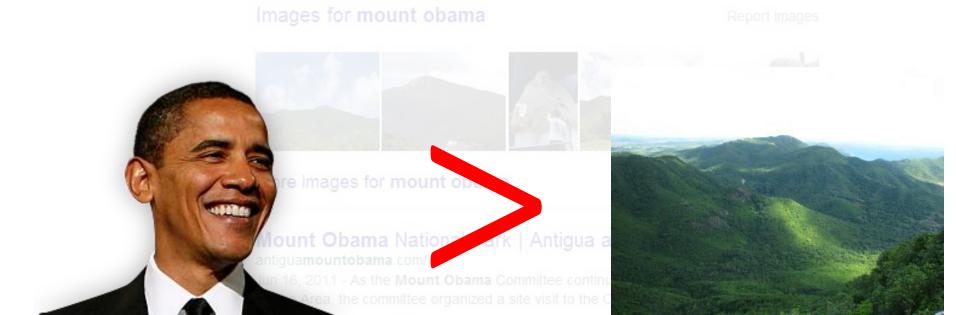
Two Sides to Ranking: Importance



About 48,100,000 results (0.26 seconds)

How could we determine that Barack Obama is more important than Mount Obama as a search result for "obama" on the Web?





Link Analysis

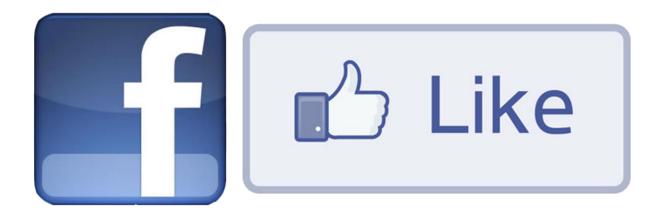
Which will have more links from other pages? The Wikipedia article for Mount Obama? The Wikipedia article for Barack Obama?





Link Analysis

- Consider links as votes of confidence in a page
- A hyperlink is the open Web's version of ...



(... even if the page is linked in a negative way.)

Link Analysis

So if we just count links to a page we can determine its importance and we are done?





Link Spamming



The Voice of Semantic Technology Busine Big Data, Linked Data, Smart Data

Home	Events	Media	Industry Verticals		Answers	J
		Questions	Tags	Users	Badges	
			,			



[deleted] Kala Jadu Specialist +91961



black magic specialist baba ji call now +919610897260



http://www.blackmagicspecialist.net.in



java

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Claritin Clomid Combivent Confido Copegus Cordarone Coreg Coumadin Cozaar Crestor Cyklokapron Cymbalta Cystone Cytotec Danazol Deltasone Depakote Desyrel Detrol Diabecon Diakof Diarex Didronel Differin Dilantin Diovan Dostinex Elavil Elimite Emsam Endep Eurax <u> Evecare Evista Exelon Famvir Feldene Femara Femcare Flomax Flonase Flovent Fosamax Gasex</u> <u>Geodon Geriforte Herbolax High Love Himcocid Himcolin Himcospaz Himplasia Hoodia Hytrin</u> <u> Hyzaar Imdur Imitrex Inderal Ismo Isoptin Isordil Kamagra Karela Keftab Koflet Kytril Lamictal</u> amisil Lanoxin Lariam Lasix Lasuna Leukeran Levaquin Levlen Levothroid Lincocin Lioresal <u>.isinopril Liv.52 Lopid Lopressor Loprox Lotensin Lotrisone Loxitane Lozol Lukol Lynoral</u> Maxaquin Menosan Mentat Mentax Mevacor Mexitil Miacalcin Micardis Mobic Monoket Motrin Myambutol Mycelex-G Mysoline Naprosyn Neurontin Nicotinell Nimotop Nirdosh Nizoral Nolvadex Nonoxinol Noroxin Omnicef Ophthacare Oxytrol Pamelor Parlodel Paxil Penisole (a) OH Phentrimine Pilex Plan B Plavix Plendil Pletal Prandin Pravachol Prednisone Premarin Prevacid Prilosec Prinivil Procardia Prograf Prometrium Propecia Proscar Protonix Proventil Prozac Purim Purinethol Quibron-T Relafen Renalka Reosto Requip Retin-A Revia Rhinocort Rimonabant Risperdal Rocaltrol Rogaine Rumalaya Sarafem Septilin Serevent Serophene Seroquel Shallaki Shoot Sinequan Singulair Snoroff Sorbitrate Speman Starlix StretchNil Stromectol Styplon Sumycin Superman Sustiva Synthroid Tenormin Topamax Trandate Tricor Trimox Triphala Tulasi Urispas V-Gel Vantin Vasodilan Vasotec Ventolin Viramune Vytorin Xeloda Xenacore Zanaflex Zantac Zebeta Zelnorm Zerit Yerba Diet Wellbutrin SR Women Attracting Pheromones Women's Intimacy Enhancer Women's Intimacy Enhancer Cream Virility Gum Vitamin A & D Viagra + Cialis Viagra + Cialis + Levitra Viagra Jelly Viagra Soft + Cialis Soft Viagra Soft Tabs Ultimate Male Enhancer Toprol XL Touch-Up Kit Tentex Royal Tentex Forte Tiberius Erectus Zero Nicotine 2 Complete Professional Whitening Kits 2 Sets Of Moldable Mouth Trays 36 Beauty Acne-n-Pimple Cream ActoPlus Met Superloss Multi SleepWell (Herbal XANAX) Shuddha Guggulu Rythmol SR Rumalaya Forte Pulmicort Inhaler Professional Plasma Tooth Whitening Kit Premium Diet Patch Penis Growth Oil Penis Growth Pack Penis Growth Patch Penis Growth Pills Orgasm Enhancer Norpace CR Mental Booster Men Attracting Pheromones Menopause Gum Male Enhancement Oil Male Enhancement Patch Male Enhancement Pills Male Sexual Tonic InnoPran XL Hoodia Weght Loss Gum Hoodia Weight Loss Patch Human Growth Hormone Agent Glucotrol XL Green Tea Grifulvin V Gyne-Lotrimin Hair Loss Cream Herbal Maxx Herbal Phentermine Flagyl ER Female Sexual Tonic Female Viagra Epivir-HBV Diet Maxx Deluxe Handheld Plasma Whitening Tool Deluxe Whitening System With Plasma Lamp Coral Calcium Cialis Jelly Cialis Soft Tabs Calcium Carbonate Bust Enhancer Beconase AQ Anatrim Diet Pills Advair Diskus Advanced Gain Pro Breast Augmentation Breast Enhancement Breast Enhancement Gel Breast Enhancement Gum Breast Intense Buv Trazodone Buv Celebrex Buv Alprazolam Buv Tramadol Buy Fioricet Buy Soma Buy Cialis Buy Carisoprodol Buy Levitra Buy Ultram Buy Ambien Buy Viagra Buy Xanax Buy Phentermine Buy Valium Buy Diazepam Generic Celebrex Generic Alprazolam Generic Tramadol Generic Fioricet Generic Soma Generic Cialis Generic Carisoprodol Generic Levitra Generic Ultram Generic Ambien

Link Importance

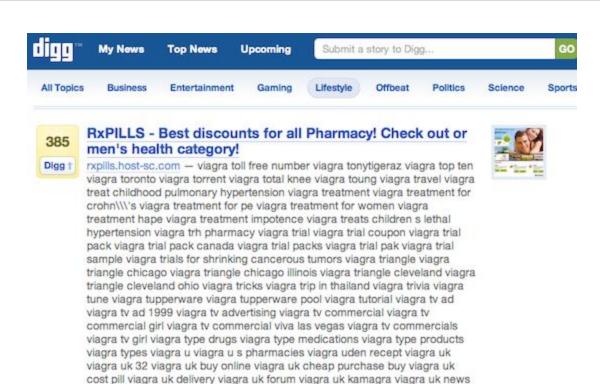
So which should count for more?







Maybe we could consider links from some domains as having more "vote"?



viagra uk online a href viagra uk online viagra uk viagra uk purchase viagra ghuk retail price viagra ukasdasd rettfsRxPILLS - Best discountds to all

Pharmacysadasdasdailers viasd 2 days ago



PageRank

- Not just a count of inlinks
 - A link from a more important page is more important
 - A link from a page with fewer links is more important
 - ∴ A page with lots of inlinks from important pages (which have few outlinks) is more important

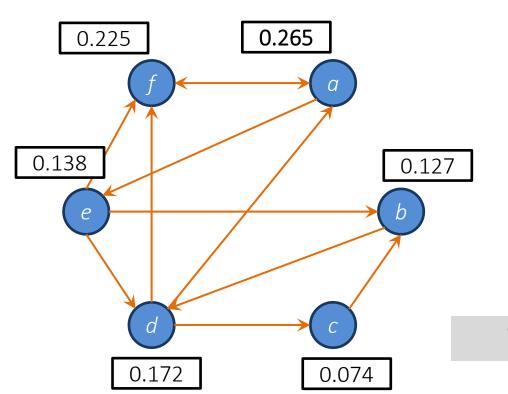
PageRank is Recursive

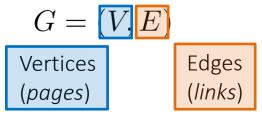
- Not just a count of inlinks
 - A link from a <u>more important</u> page is <u>more important</u>
 - A link from a page

∴ A page with lots of have few outling.



The Web: a directed graph





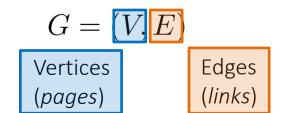
Which vertex is most important?

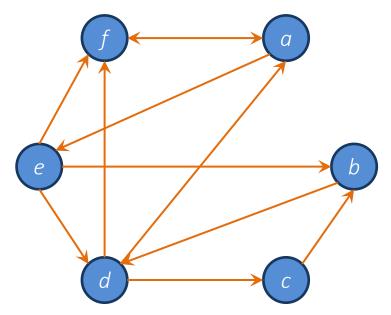


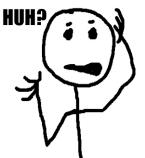
$$V = \{a, b, c, d, e, f\}$$

$$E = \{(a, e), (a, f), (b, d), (c, b), (d, a), (d, c), (d, f), (e, b), (e, d), (e, f), (f, a)\}$$

The Web: a directed graph





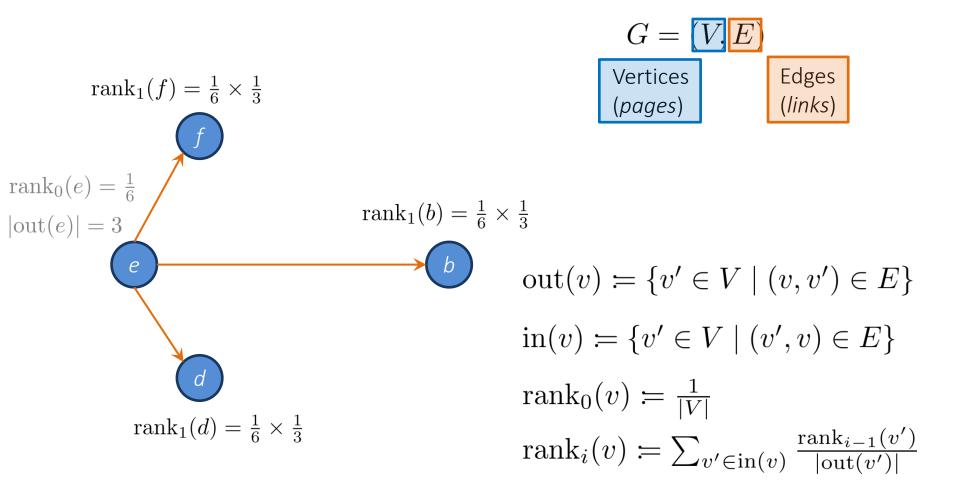


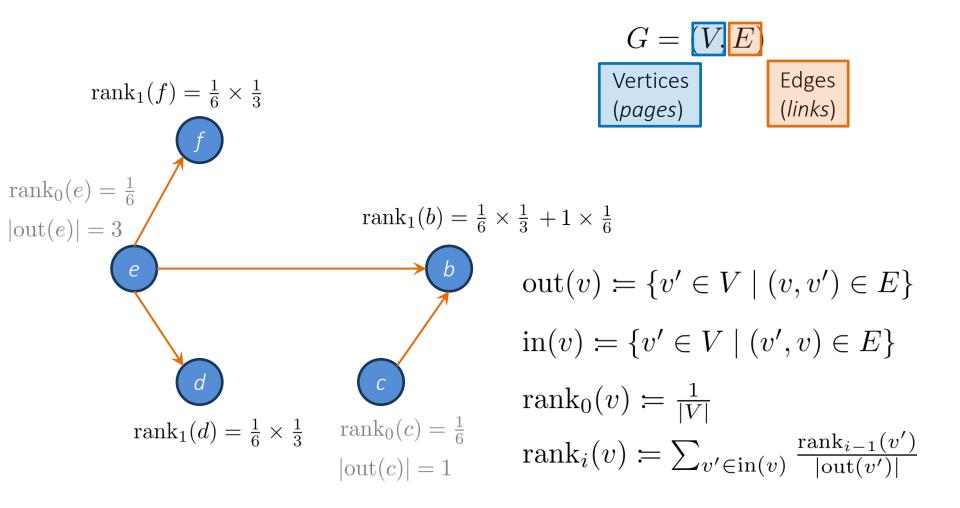
$$\operatorname{out}(v) \coloneqq \{v' \in V \mid (v, v') \in E\}$$

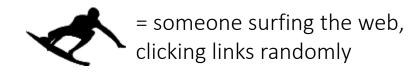
$$\operatorname{in}(v) \coloneqq \{v' \in V \mid (v', v) \in E\}$$

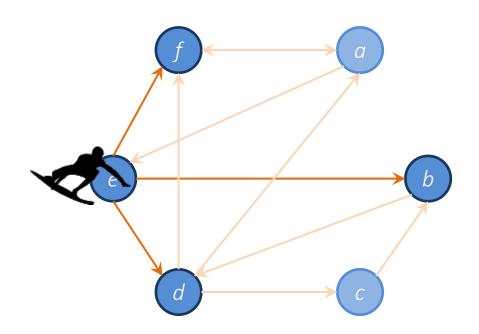
$$\operatorname{rank}_0(v) \coloneqq \frac{1}{|V|}$$

$$\operatorname{rank}_i(v) \coloneqq \sum_{v' \in \operatorname{in}(v)} \frac{\operatorname{rank}_{i-1}(v')}{|\operatorname{out}(v')|}$$

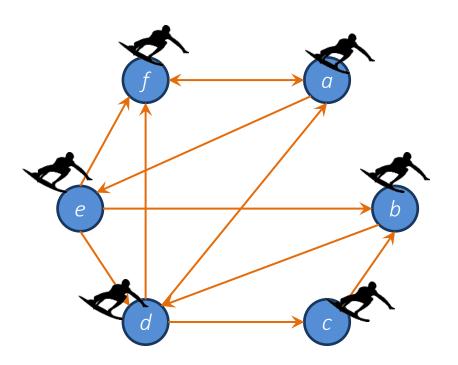


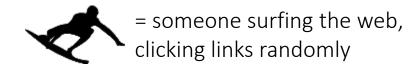




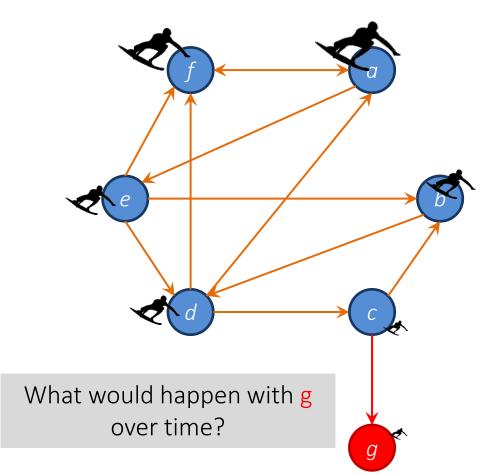


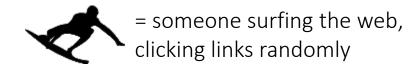
• What is the probability of being at page x after n hops?



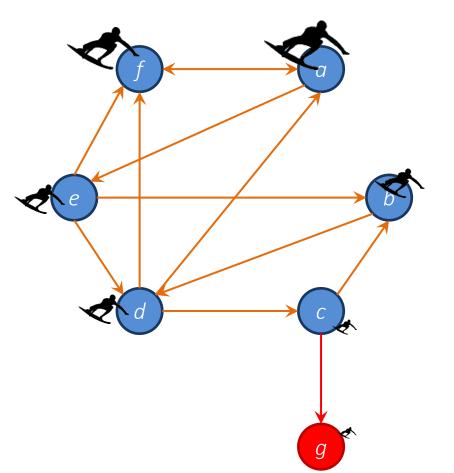


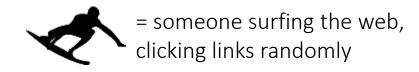
- What is the probability of being at page x after n hops?
- *Initial state:* surfer equally likely to start at any node



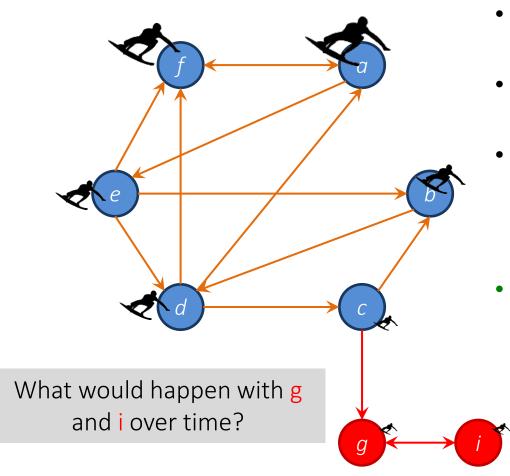


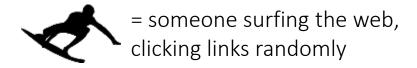
- What is the probability of being at page x after n hops?
- Initial state: surfer equally likely to start at any node
- PageRank applied iteratively for each hop: score indicates probability of being at that page after that many hops



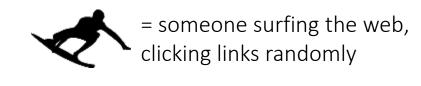


- What is the probability of being at page x after n hops?
- Initial state: surfer equally likely to start at any node
- PageRank applied iteratively for each hop: score indicates probability of being at that page after that many hops
- If the surfer reaches a page without links, the surfer randomly jumps to another page

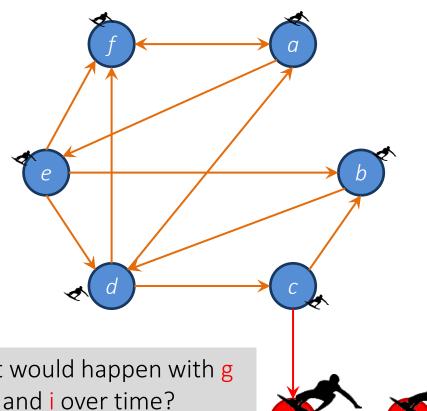




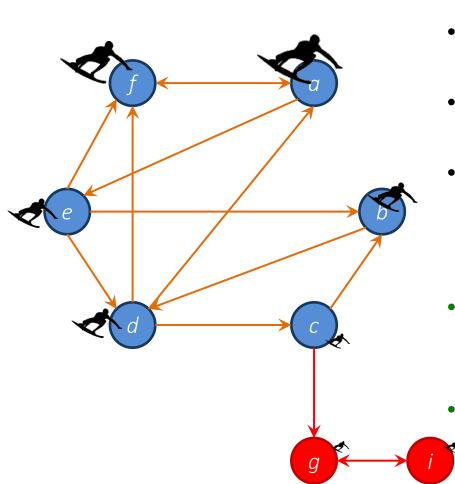
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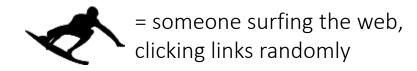


- What is the probability of being at page x after n hops?
 - *Initial state:* surfer equally likely to start at any node
 - PageRank applied iteratively for each hop: score indicates probability of being at that page after than many hops
 - If the surfer reaches a page without out-links, the surfer randomly jumps to another page



What would happen with g

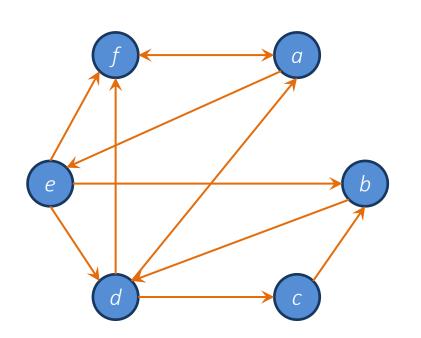


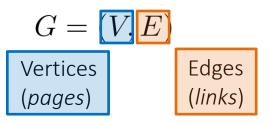


- What is the probability of being at page x after n hops?
- Initial state: surfer equally likely to start at any node
- PageRank applied iteratively for each hop: score indicates probability of being at that page after than many hops
- If the surfer reaches a page without out-links, the surfer randomly jumps to another page
 - The surfer will jump to a random page at any time with a probability 1 d ... this avoids traps and ensures convergence!

PageRank Model: Final Version

The Web: a directed graph





$$\operatorname{out}(v) \coloneqq \{v' \in V \mid (v, v') \in E\}$$

$$\operatorname{in}(v) \coloneqq \{v' \in V \mid (v', v) \in E\}$$

$$\operatorname{rank}_0(v) \coloneqq \frac{1}{|V|}$$

$$V' \coloneqq \{v \in V : |\operatorname{out}(v)| = 0\}$$

$$V'' \coloneqq \{v \in V : |\operatorname{out}(v)| \neq 0\}$$

d is the follow-a-link probability typically (d = 0.85)

$$\operatorname{rank}_{i}(v) \coloneqq d \times \sum_{u \in \operatorname{in}(v)} \frac{\operatorname{rank}_{i-1}(u)}{|\operatorname{out}(u)|} + \sum_{v' \in V'} \frac{\operatorname{rank}_{i-1}(v')}{|V|} + \underbrace{(1-d) \times \sum_{v'' \in V''} \frac{\operatorname{rank}_{i-1}(v'')}{|V|}}_{|V|}$$

PageRank: Benefits



- ✓ More robust than a simple link count
- ✓ Fewer ties than link counting
- ✓ Scalable to approximate (for sparse graphs)
- ✓ Convergence guaranteed

Two Sides to Ranking: Importance

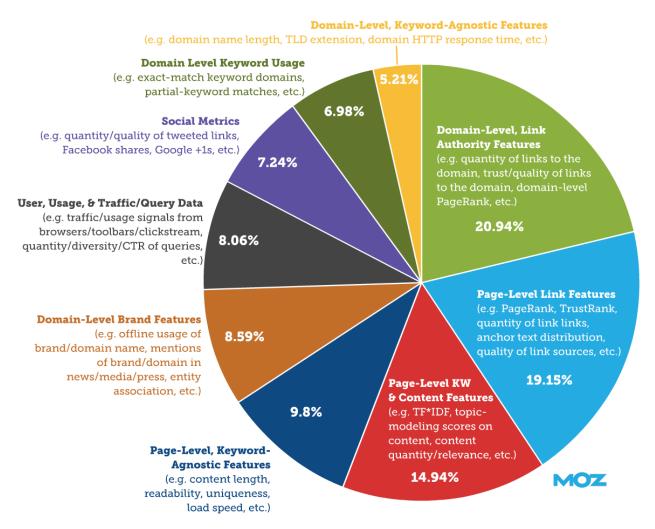


How does google really rank? An educated guess

How Modern Google ranks results (maybe)

Weighting of Thematic Clusters of Ranking Factors in Google

(based on survey responses by 128 SEO professionals in June 2013)



How Modern Google ranks results (maybe)



Why so secretive?





Ranking: Science or Art?

