Mathematics at Google

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How Google started

Backrub

http://www.google.es/intl/es/about/corporate/company/history.html

- **1995**: Larry Page and Sergey Brin meet at Stanford. (Larry, 22, a U Michigan grad, is considering the school; Sergey, 21, is assigned to show him around.)
- **1996**: Larry and Sergey, now Stanford computer science grad students, begin collaborating on a search engine called BackRub. BackRub operates on Stanford servers for more than a year—eventually taking up too much bandwidth to suit the university
- **1997**: Larry and Sergey decide that the BackRub search engine needs a new name. After some brainstorming, they go with Google. The use of the term reflects their mission to organize a seemingly infinite amount of information on the web
- **1998:** In September, Google sets up workspace in Susan Wojcicki's garage. Google files for incorporation in California on September 4

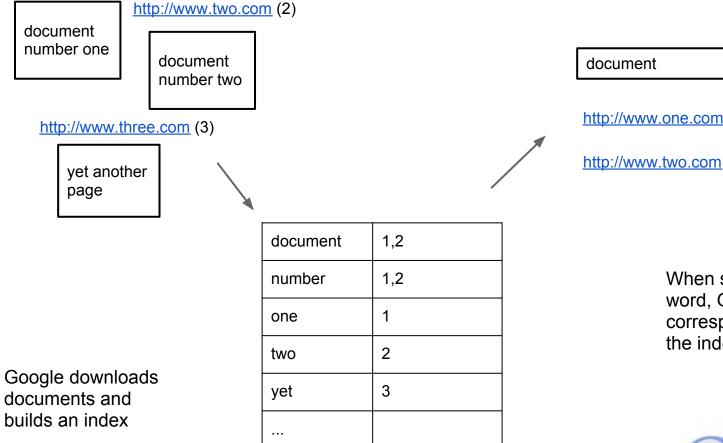


Web search (1)

- Consider the web as a collection of documents. The standard search paradigm uses an index of terms
- An index is an inverted table. In this table we have, for each term, the list of documents that contain such term
- We can use this index to search for combinations of terms by intersecting the lists of documents for each term
- The problem consists in sorting this list of documents

Web search (2)

http://www.one.com (1)





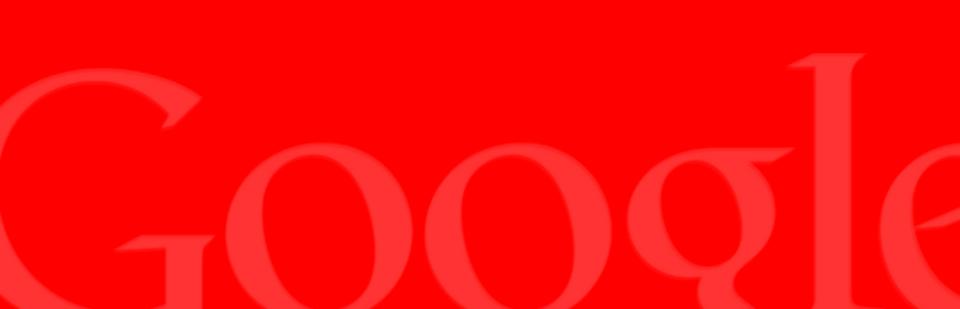
When searching for a word, Google shows the corresponding entries in the index table



Web search (3)

- Showing all documents retrieved is simply not possible
- Showing documents based on simple criteria, like date or number of occurrences of the term gives results of bad quality
- The idea of Larry Page and Sergey Brin was to use the links between documents as a signal to indicate the quality of a document. In a similar way to how citations between scientific papers work





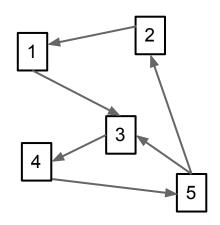
The Web as a graph

http://one.com

<h1> <h1> Web page 1 </h1> </h1> This is my web. My other page is <a href="<u>http://two.com/</u>"> here here

http://two.com/

<h1> Web page 2 </h1> This is my other web. My other page is <a href="<u>http://one.com/</u>"> here

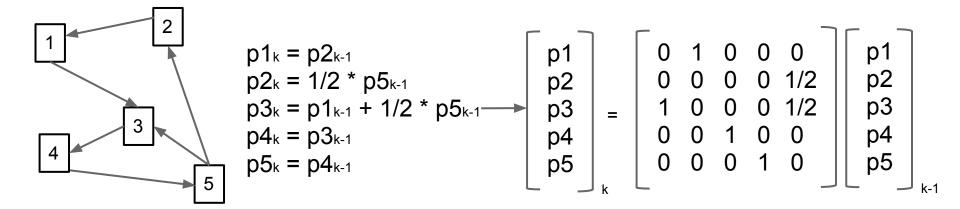




Iterative version of PageRank (1)

- PageRank is an approximation to the probability of reaching a page following links randomly
- For example: if a person is in page i with probability pi, which has links to pages {j, k} then the probability to reach j is 1/2 * pi and the probability to reach k is also 1/2 * pi
- If a page doesn't have outgoing links we assume that it links to every other page
- Initially we assume that the probability of reaching all pages is the same

Iterative version of PageRank (2)



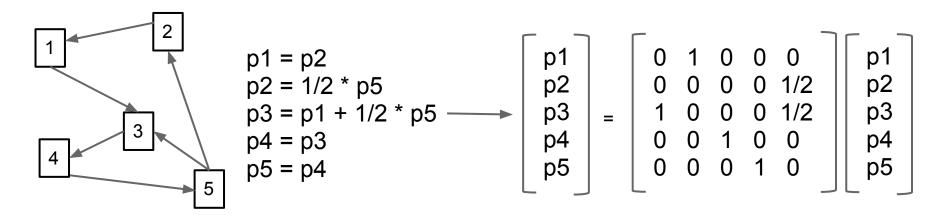
- In each phase, the probability (PageRank) is computed from the probability in the previous phase
- We can define a matrix A, which has in each position (i,j) a 0 if the page j does not link to page i, or 1/k if page j has k outgoing links and one of them is to page i
- In the first step we initialize the probabilities of all pages to the same value.
 Each subsequent step is computed according to pk = A * pk-1
- In general, after a reasonable number of iterations, we can obtain a reasonable approximation to PageRank



Algebraic version of PageRank (1)

- Consider web pages as nodes, links as edges, and the web as a directed graph
- PageRank is an estimation of the importance of each node in the graph
- If a page has k outgoing links to pages P1,... Pk, we can consider each link as a vote for page Pk
- The PageRank of page Pk, pk, is the sum of all the votes for this page. Each vote from a page Pi is weighted by the PageRank of Pi

Algebraic version of PageRank (2)



- Taking p = G . p, the PageRank vector is an eigenvector with eigenvalue 1
- G is an stochastic matrix. All elements are positive and the sum of the elements in each column is 1
- Column i contains 1/k for each one of the k outgoing links from node i
- If a node has no outgoing links, we assume that this node links to all other nodes. This is necessary for the matrix to be stochastic
- In this conditions the matrix will always have the eigenvalue 1



PageRank (1)

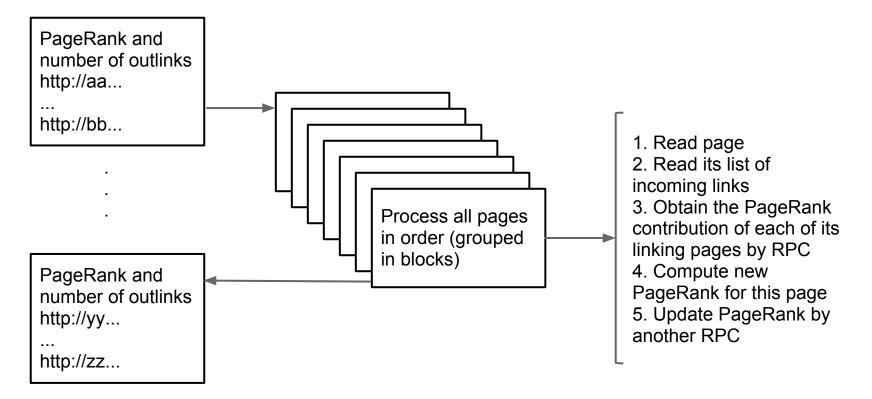
- The algorithms described before have problems when the graph is not connected. Either because it's not possible to reach a particular page by following links (in the iterative version) or because there are multiple eigenvectors for the eigenvalue 1 (in the algebraic version)
- The solution is to add a factor λ * 1/n * 1, where 1 is a matrix with ones in all positions and n is the number of nodes (And normally λ = 0.15)

PageRank (2)

- The Google matrix is:
 G = (1 λ) A + λ 1/n 1
- This matrix is also stochastic, and all elements are strictly positive
- From the Perron-Frobenius theorem, G has the eigenvalue 1 and the corresponding eigenvector has multiplicity 1
- Using the power iteration method with G it's possible to find this same eigenvalue in an iterative way



Sample implementation



Solving a linear system is difficult to parallelize. Even if the iterative method is slow, it's faster to reach an approximation



PageRank. Links

Original article about Google from Sergey Brin y Larry Page: <u>http://infolab.stanford.edu/~backrub/google.html</u>

Presentation about PageRank at Cornell University: http://www.math.cornell.edu/~mec/Winter2009/RalucaRemus/Lecture3/lecture3.html

PageRank: http://es.wikipedia.org/wiki/PageRank

Perron-Frobenius theorem: http://en.wikipedia.org/wiki/Perron%E2%80%93Frobenius_theorem

Power iteration: http://en.wikipedia.org/wiki/Power_iteration



Gallery of Mathematics

Gmail (1)

Gmail Calendar Documents	Photos Re	ader Web more -								Hiking Fan -	*				
GMail				SEARC	H MAIL	SEARCH THE	WEB Show	search options e a filter							
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COMPOSE MAIL		Jason Cornwell	» Pleas	e return my	/ stapler - Hi,	You seem to h	nave taken m	y stapler. Pl	ease,	1:10 pm					
Inbox (3)	‼⊟ ☆	Paul McDonald	» Fun l	like Yestero	day! - Thanks	for the great h	ike yesterday	/, it was awe	some	1:06 pm					
Starred 🚖 Sent Mail	‼⊟ ☆	Arielle Reinstein	» July	th weekend	d - Hi there: I	neard you'll be	around this w	weekend and	l'd lc	Jun 28					
Drafts (2)	io 🕸	JS Bach	Tonha	Ille concert l	Friday - Hey n	nan, there's a g	great concert	this Friday e	evenir	Jun 22					
+ Hiking (3)	∥⊡ ☆	Christine Chiu	» Hi Hil	ing, Looking	g for opinion o	n my diet/fitne	ss app - Hi Hi	iking, I bump	ped ir	Jun 9					
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■ Jason ■ Michael	□ ☆	Jr Wikane	» VW A	uction in Ta	coma - Hi, I w	as doing a sea	arch on Googl	le for VW's in	n Tac	Mar 6					
■ Paul	: n s²:	Google Visice	» Now	oicemail fro	m (610) 810_F	507 at 5-10 AM	A - Voicemail	l from: (610)	810-	Mar 6					
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Gmail (2)

- Spam detection is a classical example of classification using machine learning (the computer learns the algorithm from the data), in particular supervised learning (where we have previously classified data samples)
- In essence, machine learning has two phases, the training phase (when we build the classification model) and the classification phase (using the model to classify new instances)

Gmail (3)

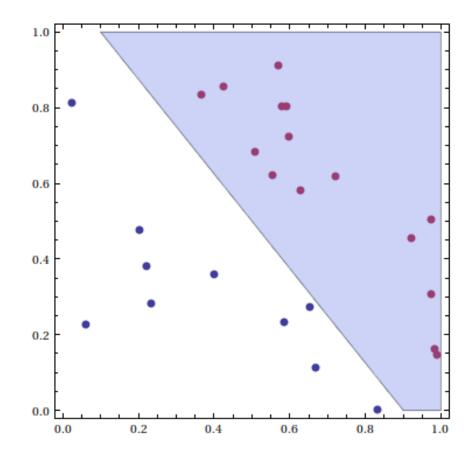
- The classification phase involves extracting the characteristics of the data instance, and then applying the model to the characteristics
- In general, the characteristics of an instance can be considered as elements in a vector of an ndimensional euclidean space for a large n (100-1000 dimensions is normal, 1M-10M is not unheard of)
- The model is a subspace of dimension n-1 which divides the original space into two disjoint subspaces



Gmail (4)

- A simple example
- From an email we can extract characteristics such as: length of the email, number of capital characters, whether the sender is in the address book, etc.
- A simple classification model is a hyperplane in the space of characteristics. Data instances on one side of the hyperplane are classified as valid emails and instances on the other side are classified as spam

Gmail (5)



Gmail (6)

Slightly more complex examples:

- Decision trees (step functions)
- Neural networks (each node of the network is a composition of a function, normally a logistic function, with a linear combination of its inputs. A network is formed by multiple levels of nodes)
- Support vector machines with a kernel function (composition of a linear function with a nonlinear transform of the original space)

Gmail (7)

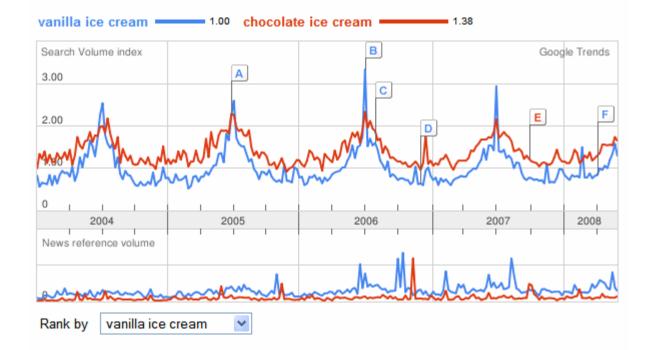
Links:

- The War Against Spam: A report from the front line
 - http://research.google.com/pubs/pub36954.html
- The Learning Behind Gmail Priority Inbox research.google.com/pubs/archive/36955.pdf
- Publications by Googlers in Artificial Intelligence and Machine Learning <u>http://research.google.</u>

com/pubs/ArtificialIntelligenceandMachineLearning.html



Google trends (1)





Google trends (2)

- Time series processing is one of the most common uses of applied mathematics. The techniques used range from regression to Fourier Analysis, hidden Markov models or self-correlation
- It is used to predict the number of search queries in a given day, number of users, income, etc. for a variety of products (thousands of daily analysis) Large-Scale Parallel Statistical Forecasting Computations in R http://research.google.com/pubs/pub37483.html

Voice search (1)



Go	ogle	
Google Search	I'm Feeling Lucky	Speak now Cancel



Voice search (2)

- Automated speech recognition (ASR) has two fundamental parts:
- First, the processing of the sound signal. Splitting it into smaller parts, applying the Fourier transform and extracting the most significant coefficients
- Second, modelling the speech using a hidden Markov model. In this model the states are the letters of the message and the sequence of events is the sound signal. The Viterbi algorithm can be used to obtain the sequence of states of maximum likelihood Google Search by Voice: A case study http://research.google.com/pubs/archive/36340.pdf

Google books (1)

el quijote de	la mancha	
Books Add	to my library 🔻 Write review	
GET PRINT BOOK	El ingenioso hidalgo Don Quijote de la Mancha	
No eBook available	Abreo Fernández de Avellaneda	
Amazon.com	****	
Barnes&Noble.com	ELINGENIOSO	
Books-A-Million	HIDALGO HIDALGO DON QUIJOTE DEL A MANCHA Biblioteca Nueva, 2000 - Literary Criticism - 789 pages	
IndieBound	Balicite de Lais Génere Carnere -	
Find in a library		
All sellers »		
All Sellers #	NERAOTECA NUEVA	
New! Shop for Books on Google Play Browse the world's largest eBookstore and start reading today on the web, tablet phone or concerdent	From inside the book el quijote de la mancha Search 100 pages matching el quijote de la mancha in this book	WARNING
tablet, phone, or ereader.	100 pages matching el quijote de la mancha in this book	
Go to Google Play Now »	Page 10	
My library	<mark>de</mark> Vega. Pero lo cierto es que Avellaneda nunca tuvo en su	0
My History	de Vega. Pero lo cierto es que Avellaneda nunca tuvo en su mira hacer lo mismo que ellos. <mark>De</mark> hecho, evitó mencionar intencionadamente <mark>la</mark> continuación más cercana en <mark>el</mark>	
Books on Google Play	intencionadamente la continuación más cercana en el	

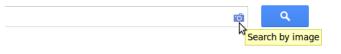
Google books (2)

- OCR techniques (optical character recognition) can be considered as a combination of image processing (obtaining individual characters images, with appropriate resolution, orientation and contrast levels) and machine learning (character classification)
- For example: An Overview of the Tesseract OCR Engine <u>http://research.google.com/pubs/archive/33418.pdf</u> Low Cost Correction of OCR Errors Using Learning in a Multi-Engine Environment
 - http://research.google.com/pubs/archive/35525.pdf Translation-Inspired OCR

http://research.google.com/pubs/pub37260.html

Image search (1)

Google



at image results with new related search previews. Learn more.





Search Google with an image instead of text.

Paste image URL | Upload an image ②

Choose File No file chosen

Den Open 🔀 <u>C</u>ancel



Search

Everything

Images

Maps

Videos

News

More

Shopping

About 3,450 results (0.85 seconds)





Find other sizes of this image: All sizes - Medium - Large

Best guess for this image: scarlett johansson

Scarlett Johansson - Wikipedia, the free encyclopedia en.wikipedia.org/wiki/Scarlett Johansson

Scarlett Johansson (born November 22, 1984) is an American actress, singer. Johansson made her film debut in North (1994) and was later ...





Image search (2)

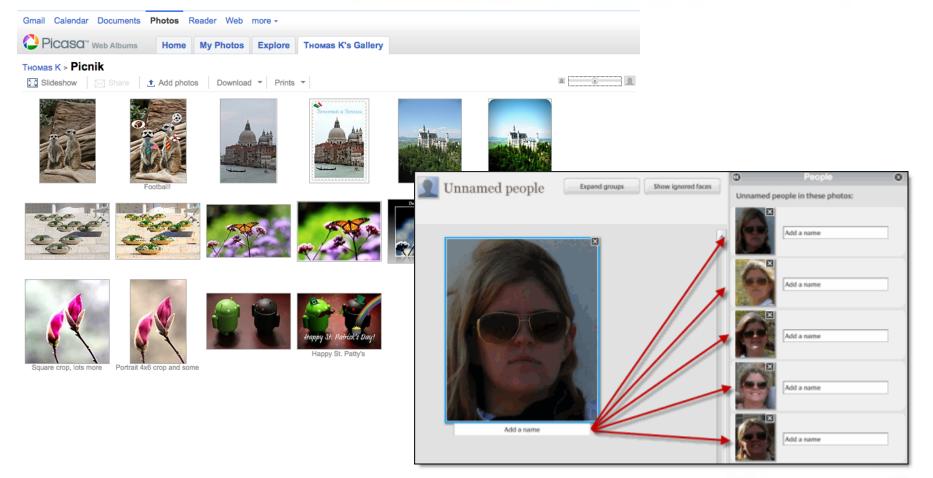
 Image search is an example of content based information retrieval (using colors, shapes, textures, etc.)

Content-based Multimedia Information Retrieval: State of the Art and Challenges

http://www.liacs.nl/home/mlew/mir.survey16b.pdf

 The key concept is the measure of similarity between images. For example the difference between the color histograms, or in general the difference between the characteristic vectors of the images Tour the World: building a web-scale landmark recognition engine <u>http://research.google.com/pubs/archive/35291.pdf</u> (Image search) Web-scale Image Annotation <u>http://research.google.com/pubs/archive/34669.pdf</u>

Picasa (1)





Picasa (2)

- An image is basically a set of three integer valued matrices, one for each primary color
- Digital image processing, and in particular applying a filter consists in executing a convolution operation in these matrices

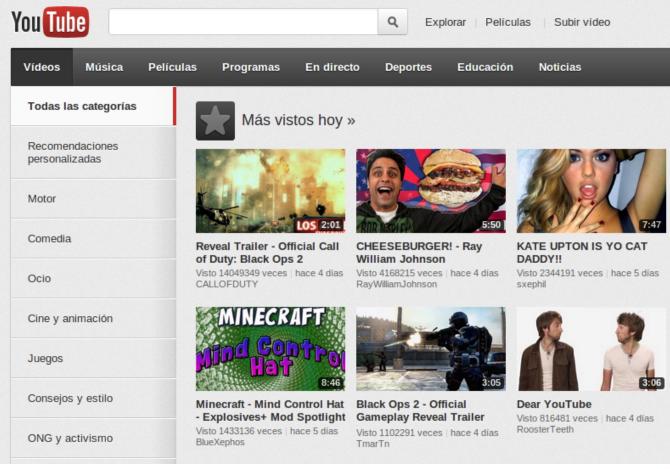
http://lodev.org/cgtutor/filtering.html

http://www.emt.jku.at/education/Inhalte/se_moderne_methoden/WS0304/Haim-Mathematics_in_Imaging.pdf

 One of the recent features in Picasa is automatic face recognition. In general face recognition is a complex problem in image processing and machine learning Handbook of Face Recognition <u>http://research.google.com/pubs/archive/36368.pdf</u> Large-Scale Manifold Learning <u>http://research.google.com/pubs/pub34395.html</u>



YouTube (1)



ace 4 dias

YouTube (2)

- There are many mathematical applications in a complex product like YouTube. For example:
- YouTube video is compressed. (<u>http://en.wikipedia.org/wiki/Data_compression</u>). Compression algorithms fundamentals (<u>http://en.wikipedia.org/wiki/Rate%E2%80%</u>
 <u>93distortion_theory</u>) come from information theory, coding theory, etc.
- Another problem is automatic event detection. For example to classify video, or to create snippets YouTubeEvent: On Large-Scale Video Event Classification <u>http://research.google.com/pubs/archive/37392.pdf</u>
 YouTubeCat: Learning to Categorize Wild Web Videos <u>http://research.google.com/pubs/archive/36387.pdf</u>

Google translate (1)

+Tú	Búsqueda	Imágenes	Maps	Play	YouTube	Noticias	Gmail	Docs	Calendar	Más -
G	oogle									Iniciar sesión
Traductor Del: inglés - detectado 👻 📥 Al: español 👻 Traducir										
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hello everybody			×	Hola a	a todo)S				
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¡Nuevo! Haz clic en las palabras anteriores para editar y ver traducciones alternativas. Descartar

Traductor de Google para empresas: Google Translator Toolkit

Traductor de sitios web

Global Market Finder



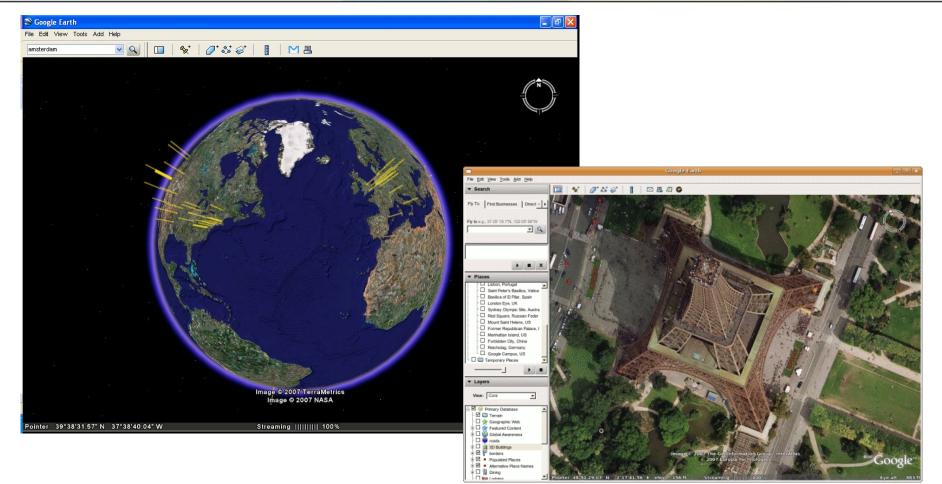
Google translate (2)

- There are multiple techniques for automatic translation. One of them consists in parsing the text into an abstract representation and then transforming this representation into the destination language. But this requires knowledge about the structure of language
- The method used at Google relies on an immense amount of data to build a statistical model of the translation

Large Language Models in Machine Translation http://research.google.com/pubs/archive/33278.pdf



Google Earth (1)





Google Earth (2)

- The fundamentals are the 3D Euclidean geometry, topography and photogrammetry, fusion of 2D and 3D data, etc. All these are well understood areas
- The greatest contributions from Google are in the issues that come up with huge amounts of data, applying these techniques at Web scale



AdWords (1)

Google	best laptops 2012	Ŷ	٩	Sign in
Search	About 199,000,000 results (0.19 seconds)			٥
Everything Images Maps Videos	Ad related to best laptops 2012 Why The new MacBook Air - The future of the notebook apple.com www.apple.com/macbookair Available now. Learn more. 211 people +1'd this page \Rightarrow Why you'll love a Mac - OS X Lion - Great Mac apps	this ad?	Ads - Why these ads? <u>Windows Recommended PCs</u> windows.microsoft.com/ Answer 5 Quick Questions & We'll Recommend the Best PCs For You.	_
News Shopping More	Laptop Computer 2012 Best Laptop Computers Compare Laptop computers.toptenreviews.com > Computers Compare the best laptop computers. Side-by-side comparisons of features and prices of top rated laptops. Easily see which laptop computer stands above the → Dell XPS 15z - Laptop Configurator - HP Pavilion dm4x - Sony S Series		Best-Selling Laptops www.bestbuy.com/Laptops Get Connected with a Laptop. Free Shipping + In-Store Pickup! 330 NE Northgate Way, Seattle, WA	,
Seattle, WA Change location Any time Past hour	Best & Worst Laptop Brands 2012 www.laptopmag.com/mobile-life/best-brands-2012.aspx Mar 14, 2012 – The biggest laptop-makers ranked from software to support. Find out which brand makes the best laptops. → Overall Scorecard - Acer/Gateway - ASUS - Dell/Alienware		New 2012 Dell Laptops www.dell.com/Laptops dell.com is rated ★★★★ Find the Latest 2012, Performance Laptops w/ Intel® Core™ at Dell! 1,724 people +1'd Dell	

Google

AdWords (2)

AdWords uses an auction algorithm. Each advertiser makes a bid for the ad inventory

Hal Varian. Online Ad Auctions:

http://people.ischool.berkeley.edu/~hal/Papers/2009/online-ad-auctions.pdf

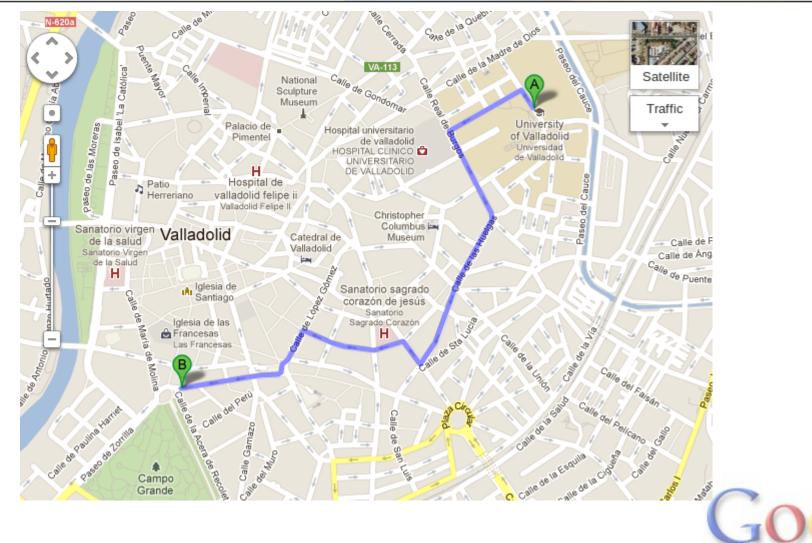
 Auction theory studies different bidding strategies and their effectiveness. It's an applied branch of Game Theory

 In particular, AdWords uses a generalized second price auction

http://en.wikipedia.org/wiki/Generalized_second-price_auction Adwords, An Algorithmic Perspective http://paul.rutgers.edu/~mangesh/cs514/notes/pres3.pdf



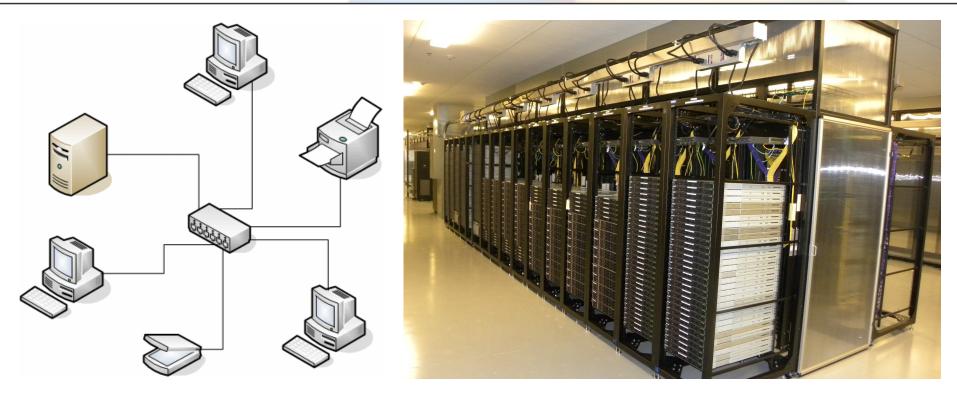
Google Maps (1)



Google Maps (2)

- Google Maps uses many basic algorithms from Graph Theory. For example, to find the shortest path between two nodes in a graph (Dijkstra) in order to get driving directions
- One unique problem is that the graphs used in Google Maps contain millions of nodes, but the algorithms have to run in milliseconds. A technique used to improve performance is graph hierarchies http://algo2.iti.kit.edu/schultes/hwy/esaHwyHierarchies.pdf

Distributed systems (1)





Distributed systems (2)

- There are many statistical techniques used to model availability of computing resources. This is similar to quality control techniques in other industries
- For example, hidden Markov models: Availability in Globally Distributed Storage Systems <u>http://research.google.com/pubs/pub36737.html</u>
 Designs, Lessons and Advice from Building Large Distributed Systems <u>http://www.cs.cornell.edu/projects/ladis2009/talks/dean-keynote-ladis2009.pdf</u>
- Queue Theory can be used to model the execution of batch jobs in a distributed system

Distributed systems (3)

- A classical example is the application of Graph Theory to network links between data centers or in general computer networks
- A network is modelled as a graph in which links can fail with some probability
- It's interesting to study which graph topologies provide the best fault tolerance, bandwidth (graph connectivity) or latency (diameter) for the lowest cost (less number of links)

Distributed systems (4)

- An especially interesting topic in Graph Theory with multiple applications to computing (not just at Google) is the concept of Ramanujan graph
- Ramanujan graphs are an example of expander graphs, which have topologies with properties that make them very useful
- They are also used to build sorting networks (AKS) which can sort n items in time log(n)
- Ramanujan graphs are those that satisfy the equivalent of Riemann's Hypothesis for the Ihara zeta function http://en.wikipedia.org/wiki/lhara_zeta_function







More links

Publications by Googlers <u>http://research.google.com/pubs/papers.html</u>

Course on the Web graph http://www.math.ryerson.ca/~abonato/webgraph.html



