CSE6242 / CX4242: Data & Visual Analytics

Graphs / Networks

Centrality measures, algorithms, interactive applications

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Partly based on materials by Professors Guy Lebanon, Jeffrey Heer, John Stasko, Christos Faloutsos, Parishit Ram (GT PhD alum; SkyTree), Alex Gray

Centrality = "Importance"

Why Node Centrality?

What can we do if we can rank all the nodes in a graph (e.g., Facebook, LinkedIn, Twitter)?

Why Node Centrality?

What can we do if we can rank all the nodes in a graph (e.g., Facebook, LinkedIn, Twitter)?

- Find celebrities or influential people in a social network (Twitter)
- Find "gatekeepers" who connect communities (headhunters love to find them on LinkedIn)
- What else?



Why Node Centrality?

Helps graph analysis, visualization, understanding, e.g.,

- Let us rank nodes, group or study them by centrality
- Only show subgraph formed by the top 100 nodes, out of the millions in the full graph
 - Similar to google search results (ranked, and they only show you 10 per page)
- Most graph analysis packages already have centrality algorithms implemented. Use them!

Can also compute edge centrality. Here we focus on node centrality.

Degree Centrality (easiest)

Degree = number of neighbors

- For directed graphs
 - In degree = No. of incoming edges
 - Out degree = No. of outgoing edges
- For undirected graphs, only degree is defined.
- Algorithms?
 - Sequential scan through edge list
 - What about for a graph stored in SQLite? 3, 2

1, 2

1, 3

2, 4

Computing Degrees using SQL

Recall simplest way to store a graph in SQLite:

edges(source_id, target_id)

- 1. If slow, first create index for each column
- 2. Use group by statement to find out degrees

select count(*) from edges group by source_id;

1, 2

1, 3

2, 4

3, 2

Betweenness Centrality

High betweenness = "gatekeeper"

Betweenness of a node v

 $\sum_{\substack{s \neq v \neq t \in V}} \frac{\sigma_{st}(v)}{\sigma_{st}}$

Number of shortest paths between s and t that goes through v

Number of shortest paths between s and t

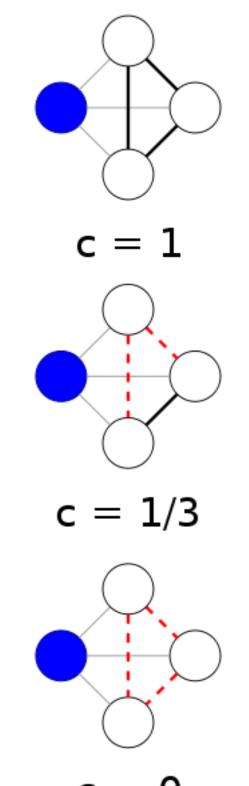
= how often a node serves as the "bridge" that connects two other nodes.

(Local) Clustering Coefficient

A node's clustering coefficient is a measure of how close the node's neighbors are from forming a clique.

- 1 = neighbors form a clique
- 0 = No edges among neighbors
- (Assuming undirected graph)

"Local" means it's for a node; can also compute a graph's "global" coefficient



Computing Clustering Coefficients...

Requires triangle counting

Real social networks have a lot of triangles

• Friends of friends are friends

Triangles are **expensive** to compute

(neighborhood intersections; several approx. algos)

Can we do that quickly?

Algorithm details: Faster Clustering Coefficient Using Vertex Covers <u>http://www.cc.gatech.edu/~ogreen3/_docs/2013VertexCoverClusteringCoefficients.pdf</u>

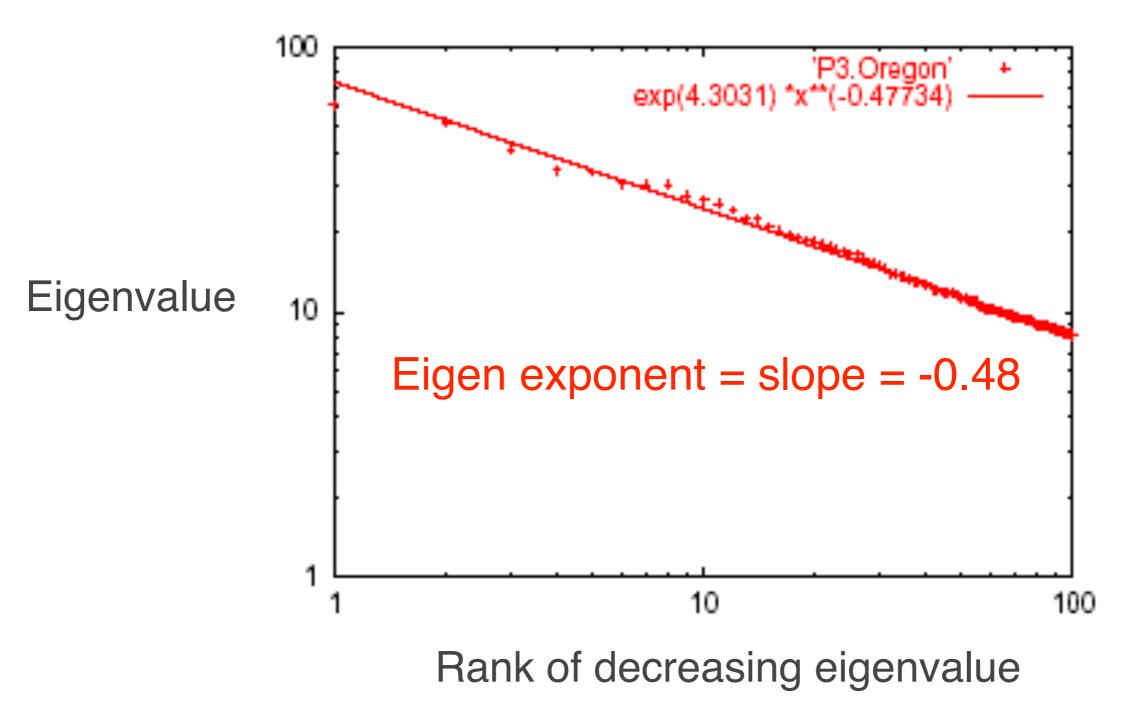
Super Fast Triangle Counting [Tsourakakis ICDM 2008]

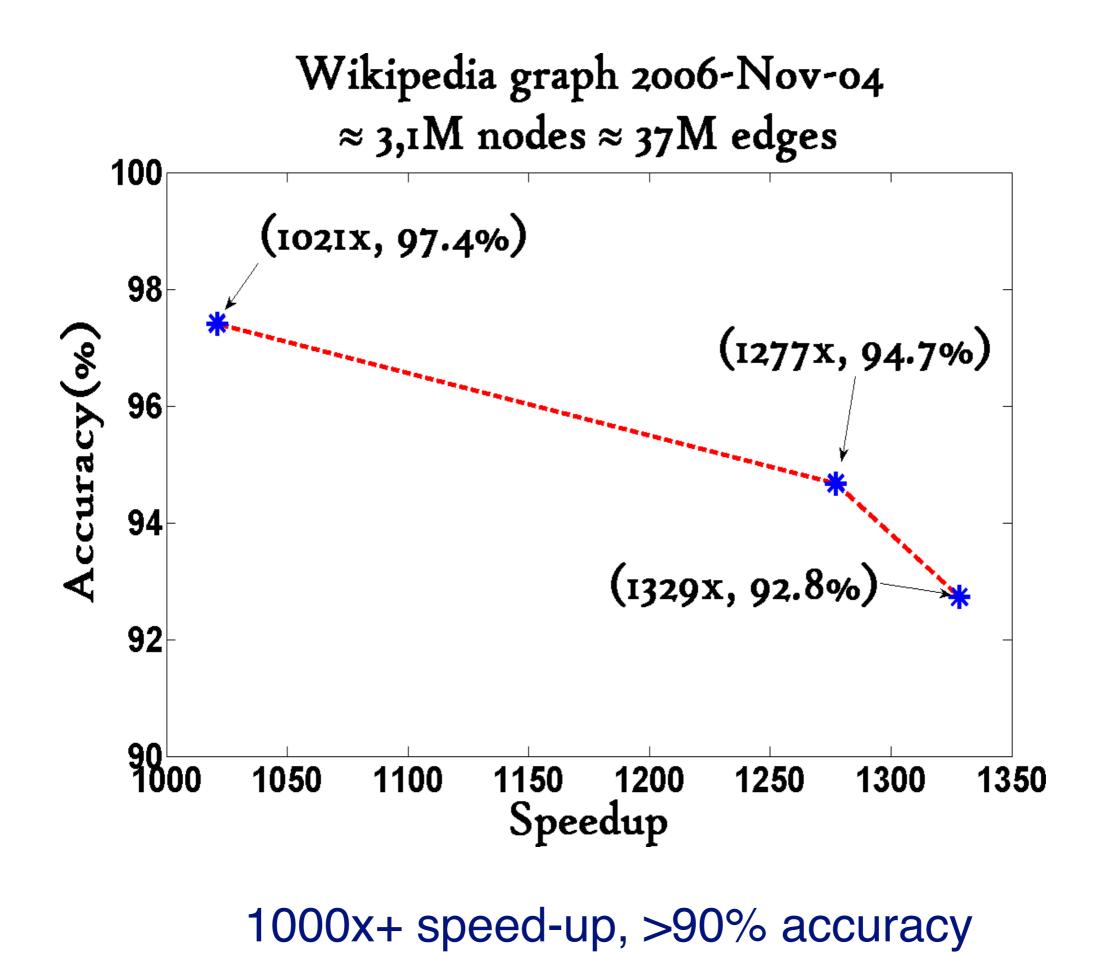


But: triangles are expensive to compute (3-way join; several approx. algos) Q: Can we do that quickly? A: Yes! #triangles = 1/6 Sum (λ_i³)

(and, because of skewness, we only need the top few eigenvalues!

Power Law in Eigenvalues of Adjacency Matrix



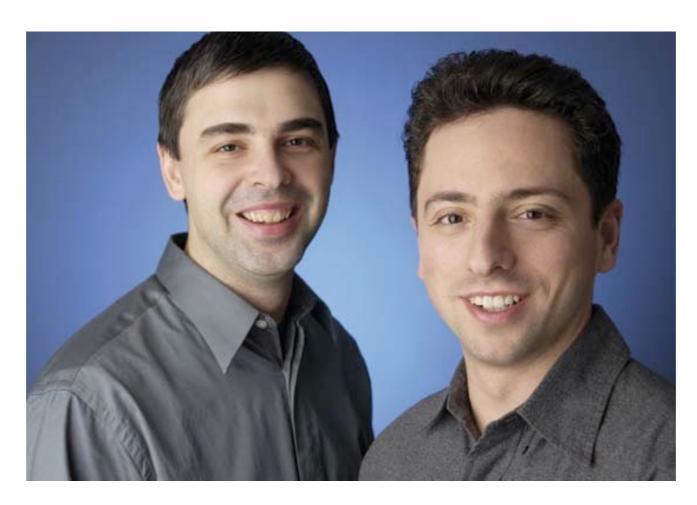


More Centrality Measures...

- Degree
- Betweenness
- Closeness, by computing
 - Shortest paths
 - "Proximity" (usually via random walks) used successfully in a lot of applications
- Eigenvector



PageRank (Google)



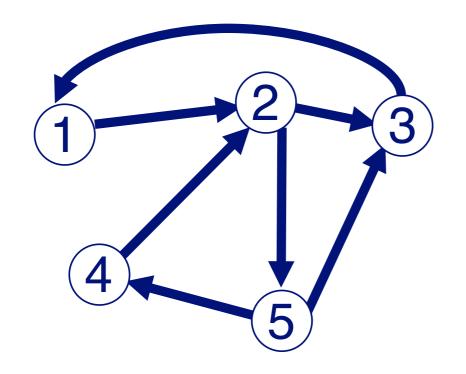
Larry Page

Sergey Brin

Brin, Sergey and Lawrence Page (1998). Anatomy of a Large-Scale Hypertextual Web Search Engine. 7th Intl World Wide Web Conf.

PageRank: Problem

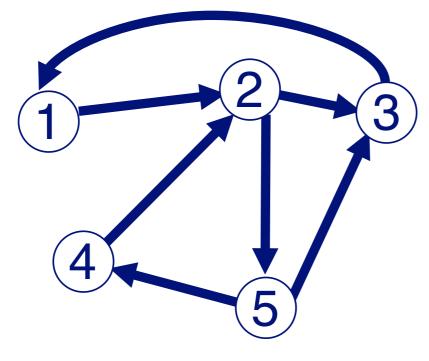
Given a directed graph, find its most interesting/central node



A node is important, if it is connected with important nodes (recursive, but OK!)

PageRank: Solution

Given a directed graph, find its most interesting/central node Proposed solution: use **random walk**; most "popular" nodes are the ones with highest steady state probability (ssp)

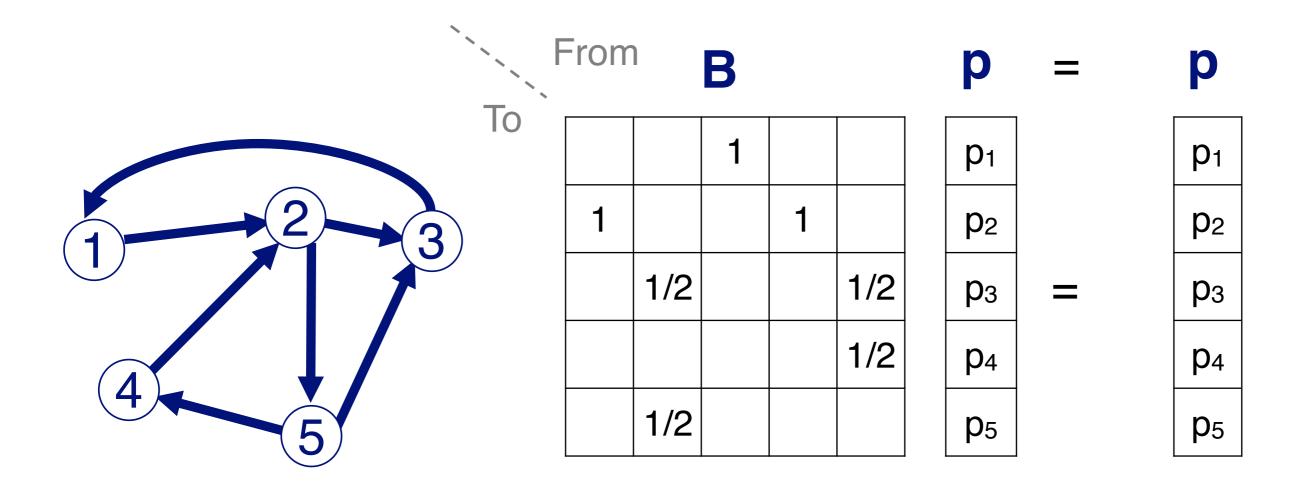


A node is important, if it is connected with important nodes (recursive, but OK!)

"state" = webpage

(Simplified) PageRank

Let **B** be the transition matrix: transposed, column-normalized



How to compute SSP: https://fenix.tecnico.ulisboa.pt/downloadFile/3779579688473/6.3.pdf http://www.sosmath.com/matrix/markov/markov.html

(Simplified) PageRank

Bp=1 * **p**

Thus, **p** is the **eigenvector** that corresponds to the highest eigenvalue (=1, since the matrix is column-normalized)

Why does such a **p** exist?

p exists if **B** is nxn, nonnegative, irreducible [Perron–Frobenius theorem]

(Simplified) PageRank

- In short: imagine a person randomly moving along the edges/links
- A node's PageRank score is the steady-state probability (ssp) of finding the person at that node

Full version of algorithm:

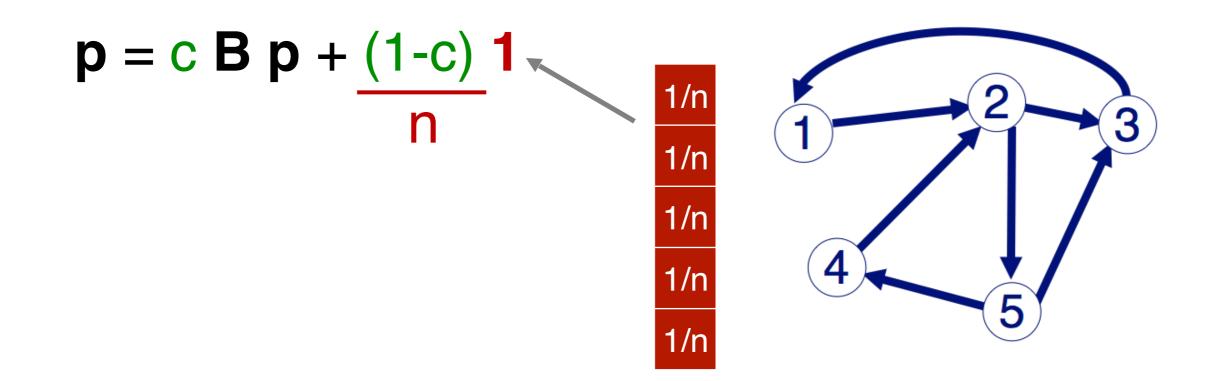
With occasional random jumps to any nodes

Why? To make the matrix irreducible.

Irreducible = from any state (node), there's non-zero probability to reach any other state (node)

Full Algorithm

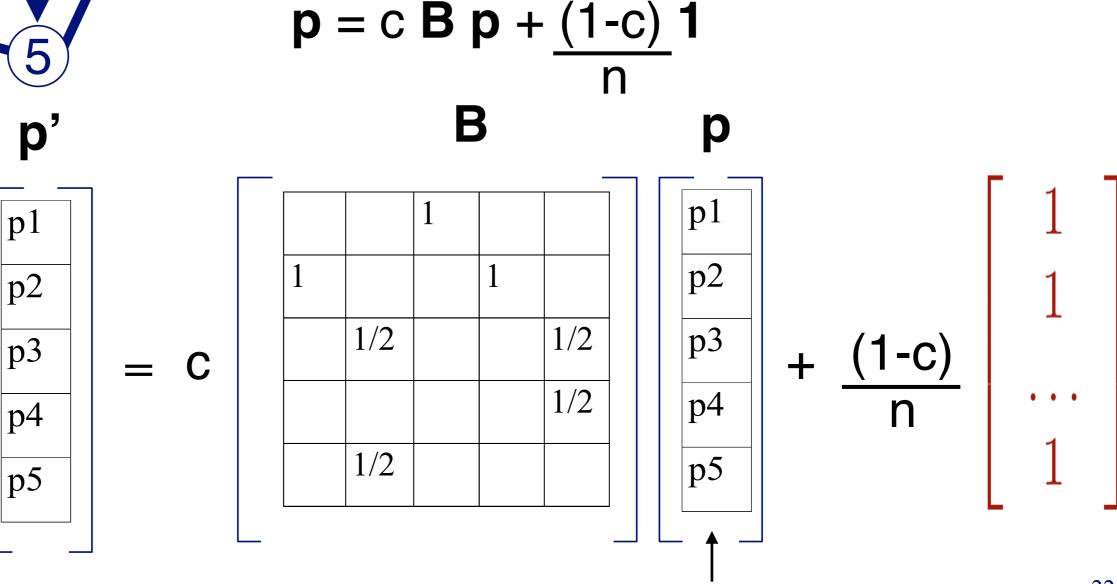
With probability 1-c, fly-out to a random node Then, we have



How to compute PageRank for huge matrix?

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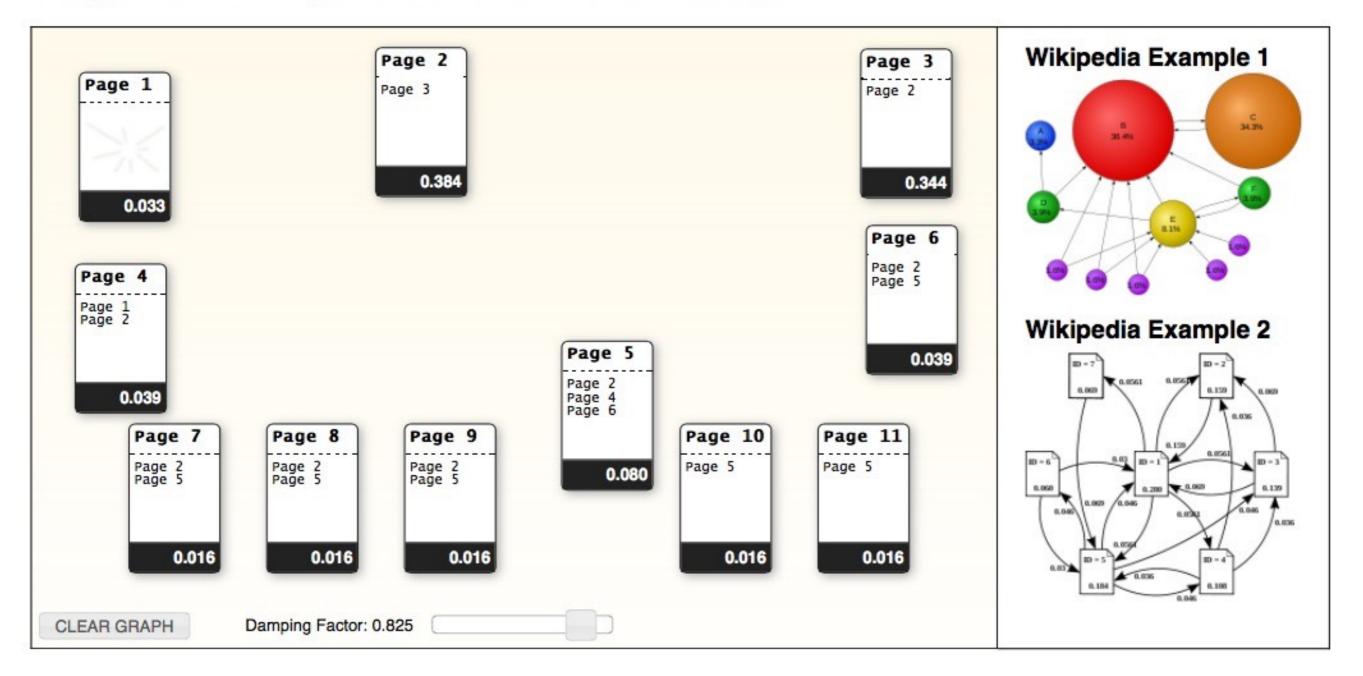
Use the power iteration method http://en.wikipedia.org/wiki/Power_iteration



Can initialize this vector to any non-zero vector, e.g., all "1"s

22

PageRank Explained with Javascript



Also great for checking the correctness of your PageRank Implementation.

http://www.cs.duke.edu/csed/principles/pagerank/

PageRank for graphs (generally)

You can run PageRank on any graphs

• All you need are the graph edges!

Should be in your algorithm "toolbox"

- Better than degree centrality
- Fast to compute for large graphs, runtime linear in the number of edges, O(E)

But can be "misled" (Google Bomb)

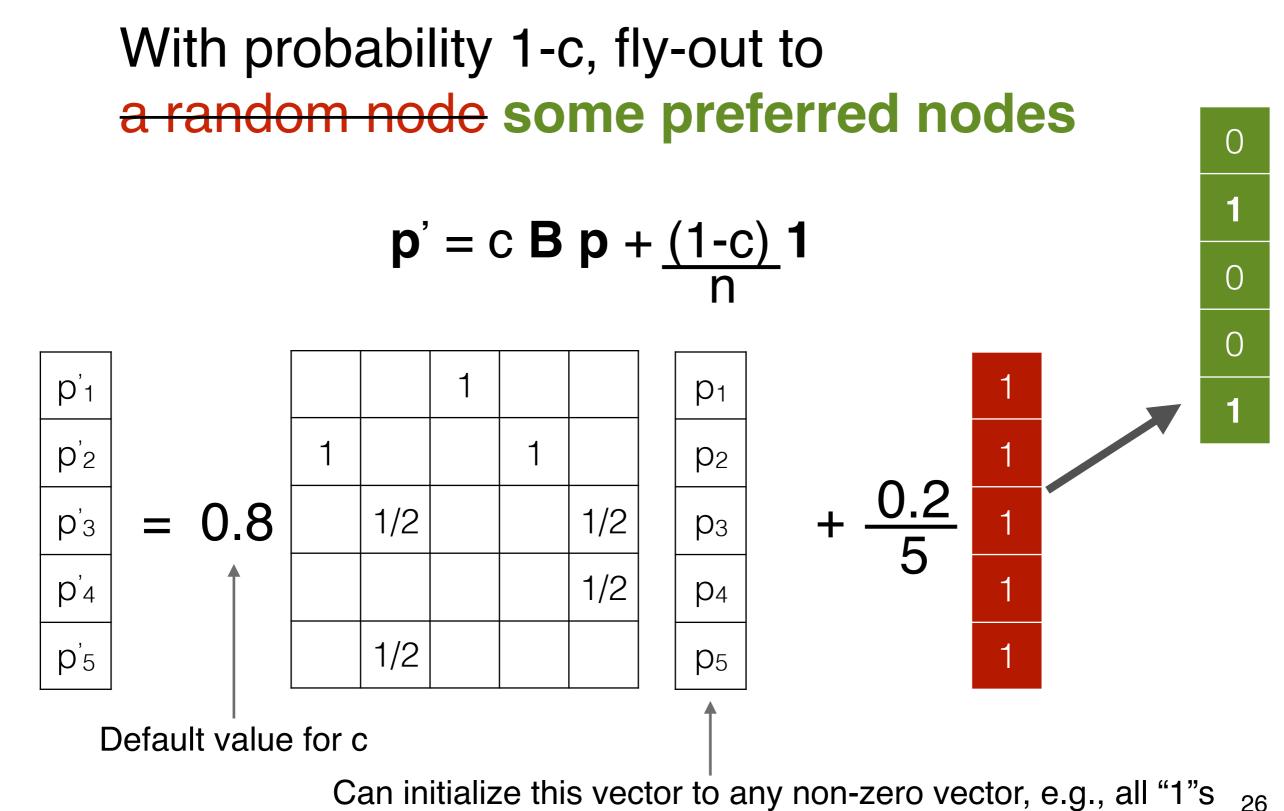
• How?

Personalized PageRank

Intuition: not all pages are equal, some more relevant to some people

Goal: rank pages in a way that those more relevant to you will be ranked higher

How? Make just one small change to PageRank



Personalized PageRank

Why Learn Personalized PageRank?

For recommendation

- If I like webpage A, what else do I like?
- If I bought product A, what other products would I also buy?

Visualizing and interacting with large graphs

 Instead of visualizing every single nodes, visualize the most important ones

Very flexible — works on any graph

Related "guilt-by-association" / diffusion techniques

Personalized PageRank

(= Random Walk with Restart)

- "Spreading activation" or "degree of interest" in Human-Computer Interaction (HCI)
- Belief Propagation (powerful inference algorithm, for fraud detection, image segmentation, errorcorrecting codes, etc.)

Why are these algorithms popular?

- Intuitive to interpret uses "network effect", homophily
- Easy to implement math is relatively simple (mainly matrixvector multiplication)
- Fast

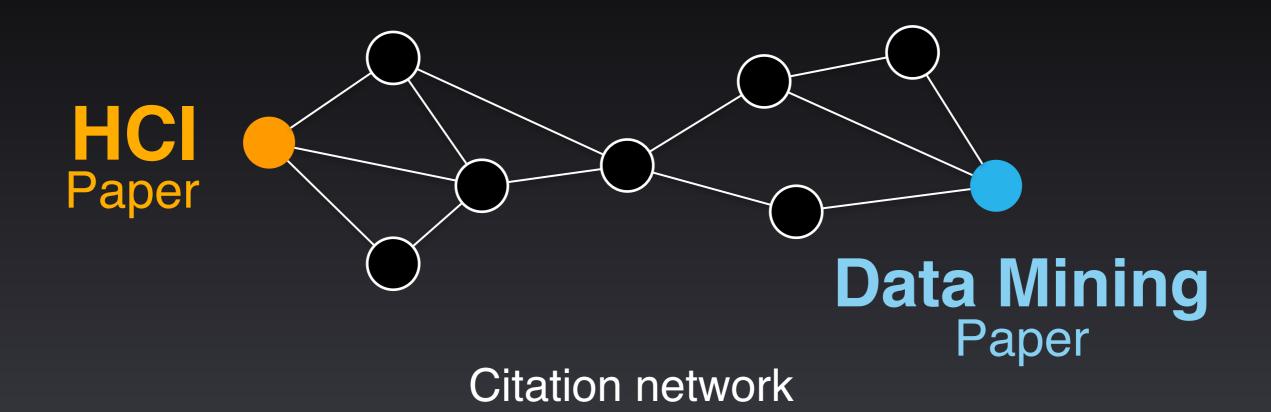
run time linear to #edges, or better

Probabilistic meaning

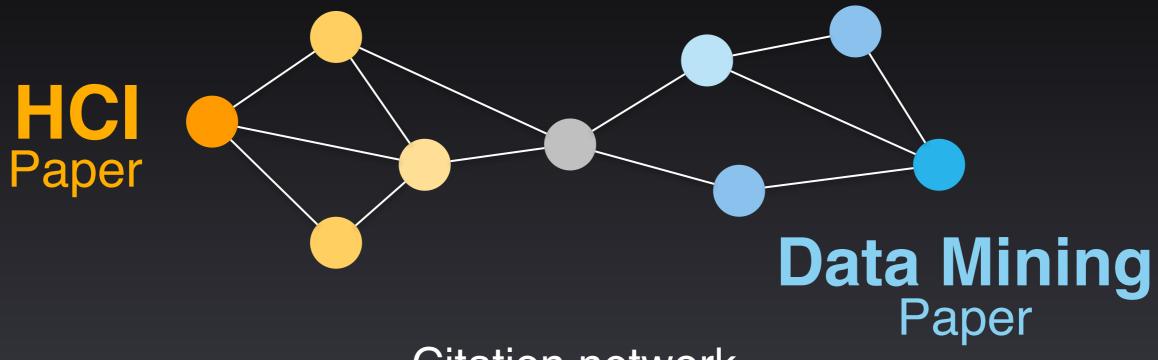
Human-In-The-Loop Graph Mining **Apolo:** Machine Learning + Visualization *CHI 2011*

Apolo: Making Sense of Large Network Data by Combining Rich User Interaction and Machine Learning

Finding More Relevant Nodes

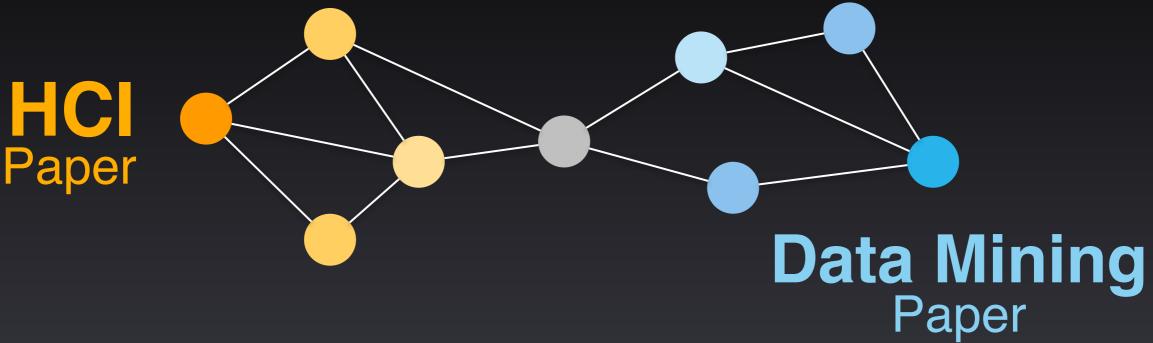


Finding More Relevant Nodes



Citation network

Finding More Relevant Nodes

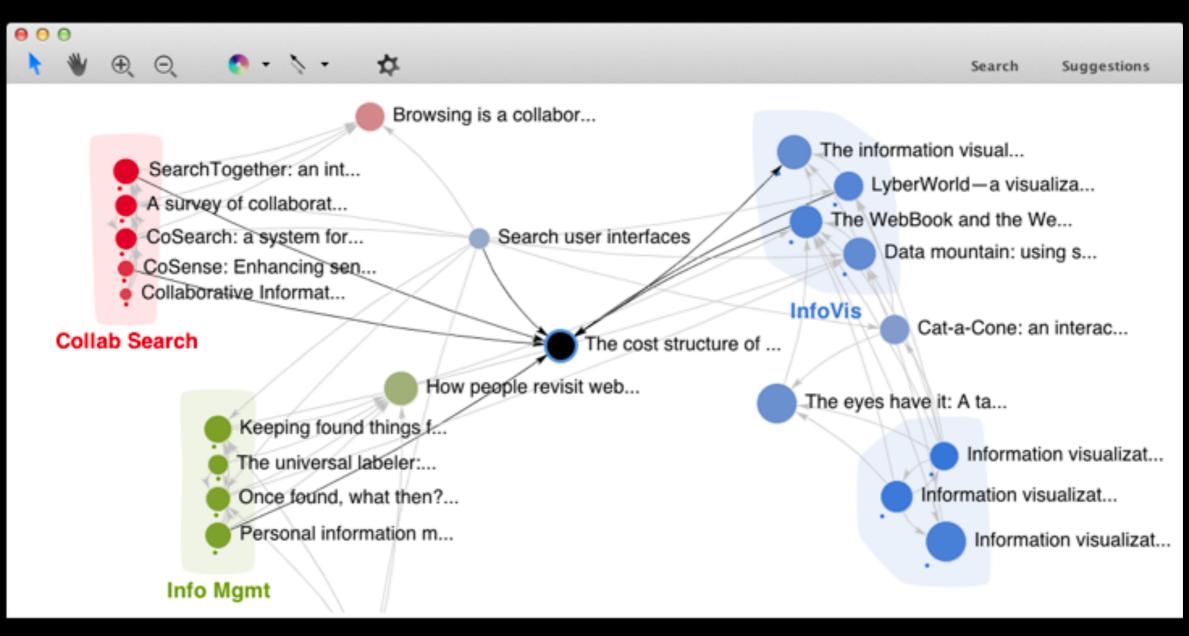


Citation network

Apolo uses guilt-by-association (Belief Propagation, similar to personalized PageRank)

Demo: Mapping the Sensemaking Literature

Nodes: 80k papers from Google Scholar (node size: #citation) Edges: 150k citations



The cost structure of sensemaking

Russell, D.M. and Stefik, M.J. and Pirolli, P. and Card, S.K.

245 citations 8 versions

PDF 1993

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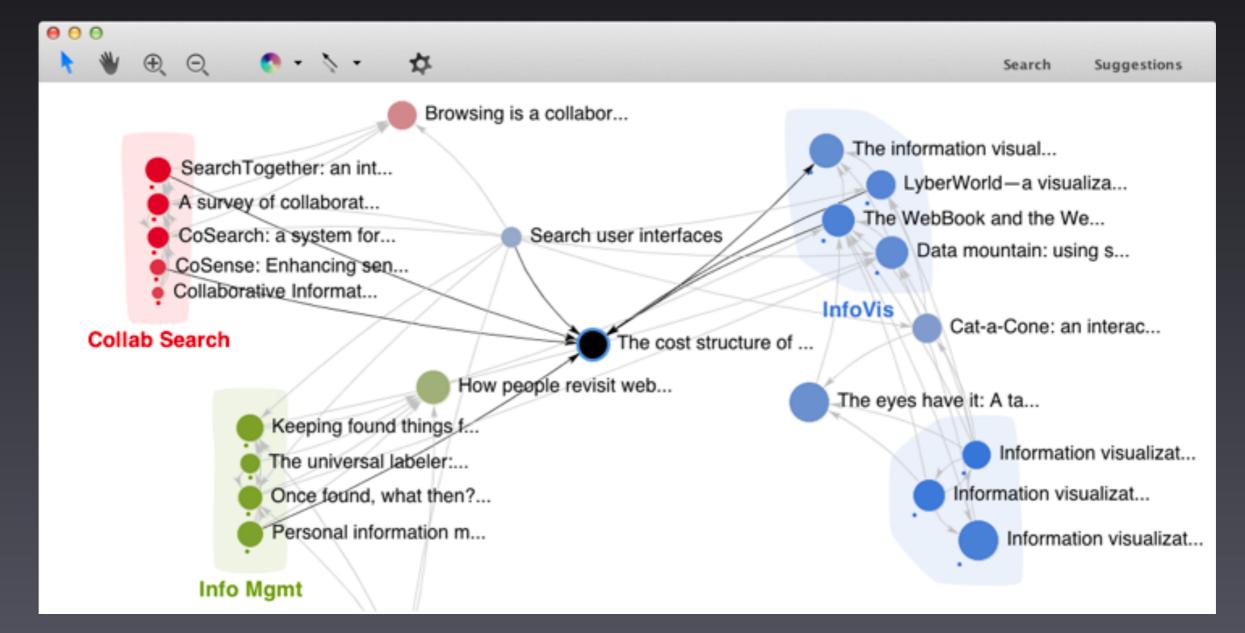
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Key Ideas (Recap)

Specify exemplars Find other relevant nodes (BP)



Apolo's Contributions 1 Human + Machine

It was like having a **partnership** with the machine.

2 Personalized Landscape

Apolo 2009

Cluster Data Add Group

Recommendations

End User Programming

End users creating effective softw... End user software engineering: chi... Invited research overview: end-us... Brad A. Myers

Margaret M. Burnett Mary Beth Rosson Andrew Jensen Ko Alan F. Blackwell

Text Entry

In-stroke word completion. Integrating isometric joysticks into... Eyes on the road, hands on the whe... An alternative to push, press, and t... Maximizing the guessability of symb... Few-key text entry revisited: mnem... Text entry from power wheelchairs: ... Joystick text entry with date stamp, ... Hudding on the stamp, ...

Not Interested

Automatically generating user inte...

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Decision-Theoretic User Daniel S. Weld

Krzysztof Z. Gajos Automatically generating Exploring the design space Predictability and accuracy

Brad

Brad A. Myers

The garnet user interface developm... Using HCI Techniques to Design a M... Creating charts by demonstration. The Amulet User Interface Developm... Easily Adding Animations to Interfac... Simplifying video editng using metad... SILVER: simplifying video editing wit...

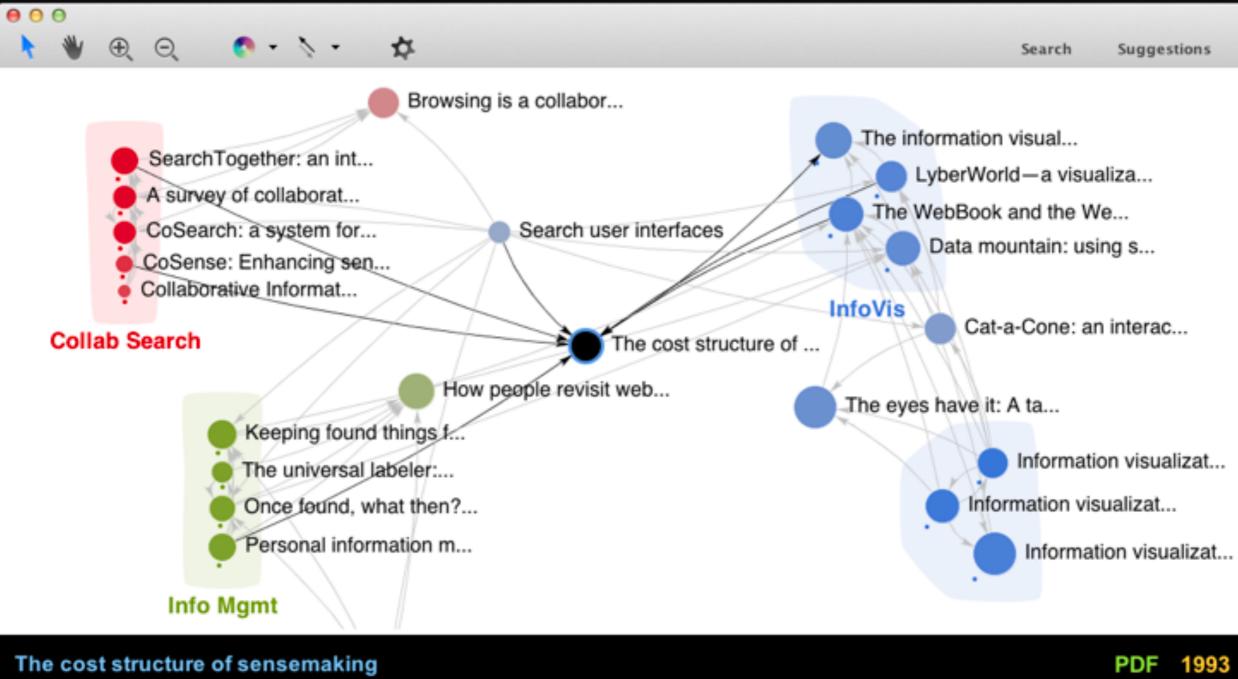
Interface Generation

Huddle: automatically generating i... UNIFORM: automatically generatin... Demonstrating the viability of auto... Jeffrey Nichols Brandon Rothrock Duen Horng Chau -

Apolo 2010

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Sensemaking: Bringing theories an 0 Sharma, f 2006 Data manipulation services in the F 7 Asdooriar 1990 Considerations for information env. 78 Furnas, G 1998 ~ Improve results C	The cost structure of sensemaking Russell, D.M., Stefik, M.J., Pirolli, P., Card, S.K. Cited by 188 booktitle Proceedings of the INTERACT'93 and CH1'93 cor	spees jo ov 0 1 2 3 4 5 6 7

Apolo 201122,000 lines of code. Java 1.6. Swing.Uses SQLite3 to store graph on disk



Russell, D.M. and Stefik, M.J. and Pirolli, P. and Card, S.K.

245 citations 8 versions

googlescholar.db

38

User Study

Used citation network

Task: Find related papers for 2 sections in a survey paper on *user interface*

- Model-based generation of UI
- Rapid prototyping tools

Past, Present and Future of User Interface Software Tools

Brad Myers, Scott E. Hudson, and Randy Pausch

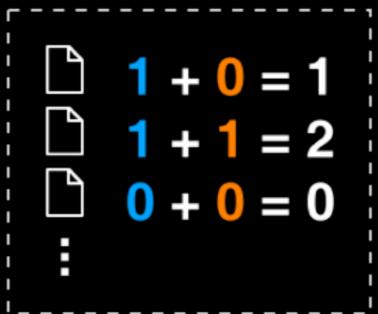
Human Computer Interaction Institute School of Computer Science Carnegie Mellon University

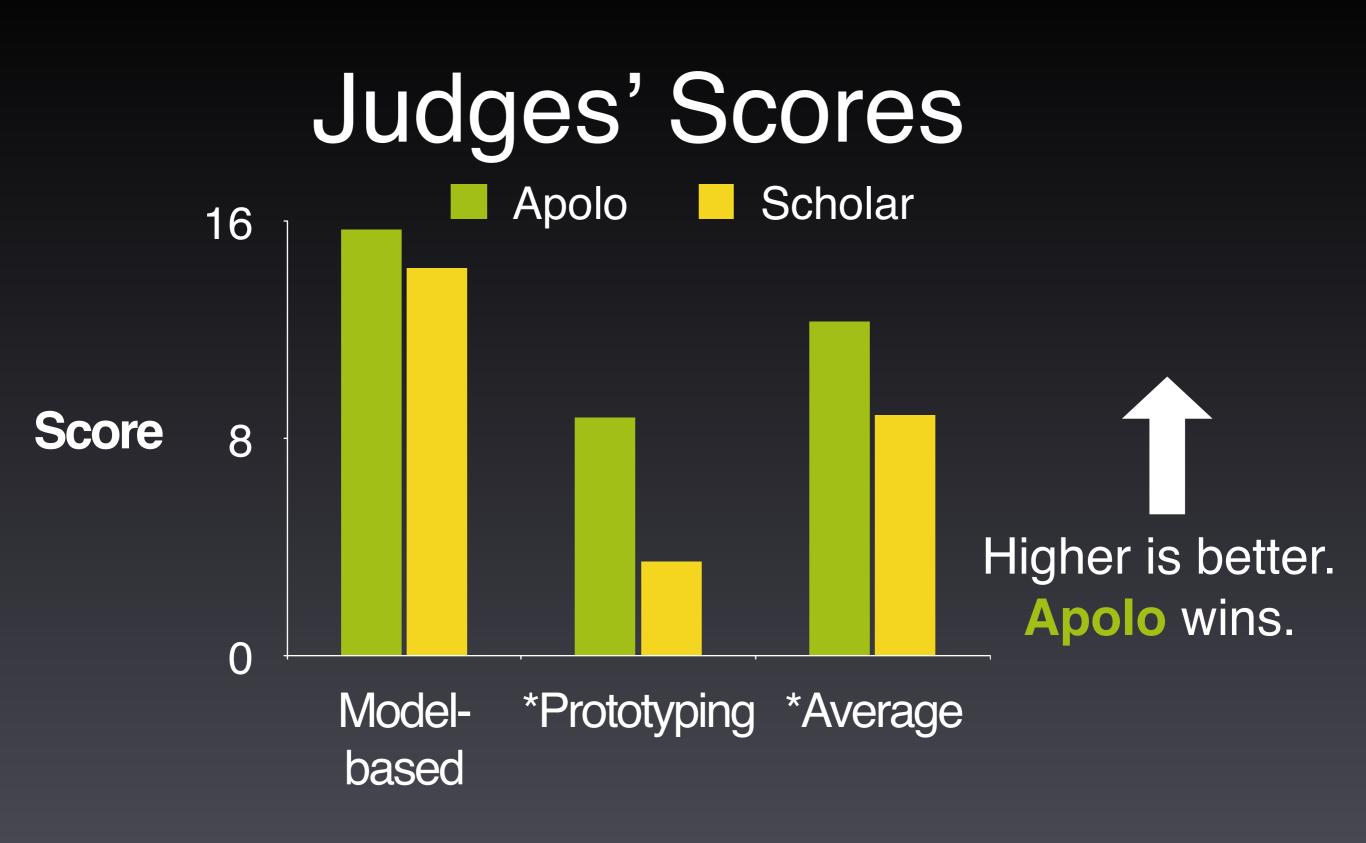
Apolo Google Scholar

Between subjects design Participants: grad student or research staff

Google Scholar Apolo 6 R **•** • ſ٩Ì ſ٩) (^m) "Model-based" "Prototyping" [予] [予] ſ٩) ſ٩Ì Ŕ ſ٩ ſ٣Ĵ ſ٩) ſ٩) ٢Ŋ 10 papers for each section

> Expert judges rated papers





* Statistically significant, by two-tailed t test, p <0.05

Practitioners' guide to building (interactive) applications

What kinds of **prototypes**?

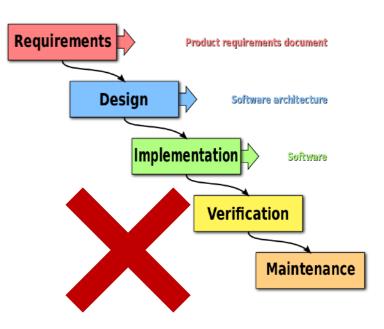
- Paper prototype, Io-fi prototype, high-fi prototype
 Important to involve **REAL users** as early as possible
 - Recruit your friends to try your tools
 - Lab study (controlled, as in Apolo)
 - Longitudinal study (usage over months)
 - Deploy it and see the world's reaction!
- To learn more:
 - CS 6750 Human-Computer Interaction
 - CS 6455 User Interface Design and Evaluation

Practitioners' guide to building (interactive) applications

Think about scalability early

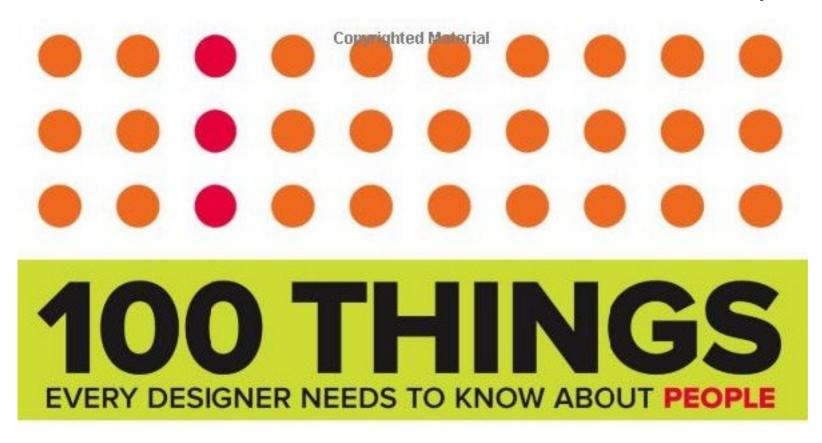
- Identify candidate scalable algorithms early on
- Use **iterative** design approach, as in Apolo and industry
 - · Why? It's hard to get it right the first time
 - Create prototype, evaluate, modify prototype, evaluate, ...
 - Quick evaluation helps you identify important fixes early — save you a lot of time overall

Waterfall model (software engineering)



If you want to know more about people...

http://amzn.com/0321767535



SUSAN M. WEINSCHENK, Ph.D.

