# Random Walks, PageRank, and Computer Science 

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## Yahoo in 1996: Human-edited directory



Yellow Pages - People Search - City Maps -- News Headlines - Stock Quotes - Sports Scores

- Arts - - Humanities, Photography, Architecture, ...
- Business and Economy [Xtra!] - - Directory, Investments, Classifieds, ...
- Computers and Internet [Xtra!] - Internet, WWW, Software, Multimedia, ...
- Education - - Universities, K-12, Courses, ...
- Entertainment [Xtra!] - - TV, Movies, Music, Magazines, ...
- Government - - Politics [Xtra!], Agencies, Law, Military, ...
- Health [Xtra!] - - Medicine, Drugs, Diseases, Fitness, ...
- News [Xtra!] - - World [Xtra!], Daily, Current Events, ...


## Only 5 directory entries for "Linear Algebra"

## TAHOO: ©-3-3

## Top:Science:Mathematics:Linear Algebra

Search Options

- Search all of Yahoo Search only in Linear Algebra
- I Hate Linear Algebra
- Java Script Linear Algebra - calculate simple matrix operations.
- Linear Algebra WebNotes - undergraduate course at the University of Nebraska-Lincoln. Contains lectures, homework assignments, Maple examples, and a discussion page.
- Slitex Foilset CPS615 Gauss Seidel Finite Element Methods and Conjugate Gradient - discusses sequential and parallel Gauss Seidel and Jacobi iteration, finite element methods applied to Laplace's equation in 2D, and the Conjugate Gradient method.
- Solving of linear equationsystems


## Google today

- ~10,000,000 Linear Algebra pages, ranked mostly by importance

Linear algebra | Khan Academy<br>https://www.khanacademy.org/math/linear-algebra * Khan Academy *<br>Linear algebra describes things in two dimensions, but many of the concepts can be extended into three, four or more. Linear algebra implies two dimensional ...<br>Vectors and spaces - Matrix transformations - Alternate coordinate systems - Vectors<br>Linear algebra - Wikipedia, the free encyclopedia https://en.wikipedia.org/wiki/Linear_algebra • Wikipedia •<br>Linear algebra is the branch of mathematics concerning vector spaces and linear<br>mappings between such spaces. It includes the study of lines, planes, and subspaces, but is also concerned with properties common to all vector spaces.<br>List of linear algebra topics - Basis - Rank - Category:Linear algebra

Free Linear Algebra textbook
joshua.smcvt.edu/linearalgebra/ - Saint Michael's College *
Linear Algebra is a text for a first US undergraduate Linear Algebra course. It is Free.
You can use it as a main text, as a supplement, or for independent study.
Linear Algebra | Mathematics | MIT OpenCourseWare
ocw.mit.edu > Courses > Mathematics ~ MIT OpenCourseWare *
This is a basic subject on matrix theory and linear algebra. Emphasis is given to topics that will be useful in other disciplines, including systems of equations, ...

## Part I:

PageRank and Random walk

## PageRank Algorithm

- Named after Larry Page, Google cofounder and current CEO
- Determines webpages' importance only by link structure of the directed graph of webpages


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## PageRank Algorithm

- Named after Larry Page, Google cofounder and current CEO
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- Spectral graph theory!
- For simplicity, focus on undirected, regular graphs in this talk


## Random Walk

- Stochastic progress on a graph (undirected for now)


## Random Walk

- Stochastic progress on a graph (undirected for now)
- Starts from a vertex, at each time step $t$ moves to a uniformly random neighbour of the current vertex


## Distribution $\mathbf{p}_{t}$

Probability transition matrix $\mathbf{K}$
Row vector $\mathbf{p}_{t}$ : distribution at time $t$

- $\mathbf{p}_{t+1}=\mathbf{p}_{t} \mathbf{K}$
- $\mathbf{p}_{t}=\mathbf{p}_{0} \mathbf{K}^{t}$
- Does $\mathbf{p}_{t}$ converge as $t$ increases?


## Stationary distribution

If $\mathbf{p}_{t}$ converges, the limiting distribution $\mathbf{p}_{\infty}$ must be stationary

$$
\mathbf{p}_{\infty} \mathbf{K}=\mathbf{p}_{\infty}
$$

- uniform distribution always a stationary distribution (for an undirected, regular graph)


## Limiting distribution: unique?

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(Some vertex not reachable from some other vertex via intermediate vertices)


## Limiting distribution: unique?

- Not unique on disconnected graphs (Some vertex not reachable from some other vertex via intermediate vertices)
- Not unique on bipartite graphs
(Can partition all vertices into two subsets $V_{1}, V_{2}$ so that all edges only go between $V_{1}$ and $V_{2}$ )


## Unique limiting distribution

Theorem (Uniqueness)
An undirected, regular, connected, non-bipartite graph has a unique stationary distribution $\mathbf{p}_{*}$.
Further, given any initial distribution $\mathbf{p}_{0}$,

$$
\lim _{t \rightarrow \infty} \mathbf{p}_{t}=\mathbf{p}_{*} .
$$

## Eigenvalues and eigenvectors

$\lambda \in \mathbb{R}$ is an eigenvalue and $\mathbf{q} \in \mathbb{R}^{n}$ is an eigenvector of $\mathbf{K}$ if

$$
\mathbf{q} \mathbf{K}=\lambda \mathbf{q}
$$

Fact: Transition matrix $\mathbf{K}$ of an undirected graph is real symmetric
Lemma (Spectral Theorem)
Any $n \times n$ real symmetric matrix $\mathbf{K}$ has $n$ eigenvalue-eigenvector pairs

$$
\begin{gathered}
\mathbf{q}^{(1)} \mathbf{K}=\lambda_{1} \mathbf{q}^{(1)} \\
\vdots \\
\mathbf{q}^{(\mathbf{n})} \mathbf{K}=\lambda_{n} \mathbf{q}^{(\mathbf{n})}
\end{gathered}
$$

such that $\left\{\mathbf{q}^{(1)}, \ldots, \mathbf{q}^{(\mathbf{n})}\right\}$ is an orthogonal basis

## Proof

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Proof.
By spectral theorem, $\left\{\mathbf{q}^{(1)}, \ldots, \mathbf{q}^{(\mathbf{n})}\right\}$ forms a basis. Expand

$$
\begin{gathered}
\mathbf{p}_{0}=\sum_{i=1}^{n} \alpha_{i} \mathbf{q}^{(\mathbf{i})} . \\
\mathbf{p}_{t}=\mathbf{p}_{0} \mathbf{K}^{t}=\sum_{i=1}^{n} \alpha_{i} \mathbf{q}^{(\mathrm{i})} \mathbf{K}^{t} .
\end{gathered}
$$

## Proof (continued)

$$
\mathbf{p}_{t}=\mathbf{p}_{0} \mathbf{K}^{t}=\sum_{i=1}^{n} \alpha_{i} \mathbf{q}^{(\mathbf{i})} \mathbf{K}^{t}
$$

Since

$$
\mathbf{q}^{(\mathbf{i})} \mathbf{K}^{t}=\lambda_{i} \mathbf{q}^{(\mathrm{i})} \mathbf{K}^{t-1}=\lambda_{i}^{2} \mathbf{q}^{(\mathbf{i})} \mathbf{K}^{t-2}=\cdots=\lambda_{i}^{t} \mathbf{q}^{(\mathrm{i})}
$$

the top equation becomes

$$
\mathbf{p}_{t}=\sum_{i=1}^{n} \alpha_{i} \lambda_{i}^{t} \mathbf{q}^{(\mathbf{i})}
$$

## Graph spectrum

Assume eigenvalues are sorted

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## Proposition

$\lambda_{n}=-1$ if and only if bipartite graph

## Proof (final bits)

$$
\mathbf{p}_{t}=\sum_{i=1}^{n} \alpha_{i} \lambda_{i}^{t} \mathbf{q}^{(\mathbf{i})}
$$

For regular, connected, bipartite graph, $\left|\lambda_{2}\right|<1, \ldots,\left|\lambda_{n}\right|<1$. Hence $\mathbf{p}_{t} \rightarrow \alpha_{1} \mathbf{q}^{(1)}$.
We are done if $\alpha_{1}=1$. To determine $\alpha_{1}$, recall $\mathbf{p}_{0}=\sum_{i=1}^{n} \alpha_{i} \mathbf{q}^{(\mathbf{i})}$. Taking inner product with $q^{(1)}$ :

$$
\mathbf{p}_{0} \cdot \mathbf{q}^{(\mathbf{1})}=\sum_{i=1}^{n} \alpha_{i} \mathbf{q}^{(\mathbf{i})} \cdot \mathbf{q}^{(\mathbf{1})}=\alpha_{1} \mathbf{q}^{(\mathbf{1})} \cdot \mathbf{q}^{(\mathbf{1})}
$$

Since $q^{(1)}$ must be (a scalar multiple of) the uniform distribution,

$$
\mathbf{p}_{0} \cdot \mathbf{q}^{(1)}=1 / n \quad \text { and } \quad \mathbf{q}^{(1)} \cdot \mathbf{q}^{(1)}=1 / n,
$$

hence $\alpha_{1}$ must be 1 .

## Directed graphs and PageRank

## Theorem (Uniqueness)

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$$

- A similar theorem (suitably modified) holds for directed, non-regular graphs: Perron-Frobenius theorem
- Limiting distribution $\mathbf{p}_{*}$ not necessarily uniform
- PageRank iteratively computes the distribution $\mathbf{p}_{t}=\mathbf{p}_{0} \mathbf{K}^{t}$ from an arbitrary initial distribution $\mathbf{p}_{0}$


## Part II: <br> Connections to Theoretical Computer Science

## Spectral graph theory and expanders

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## Spectral graph theory and expanders

- Spectral graph theory: study of graph eigenvalues $\lambda_{1}, \ldots, \lambda_{n}$ and graph properties
- Graphs with $\lambda_{2}$ much smaller than $\lambda_{1}=1$ are called expanders Valuable to computer science
- For $d$-regular graphs, how small can $\lambda_{2}$ be? Recent breakthrough: Yale theoretical computer scientists (Marcus, Spielman, and Srivastava) constructed bipartite graphs for any degree $d$ with

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\max \left\{\left|\lambda_{2}\right|,\left|\lambda_{n-1}\right|\right\} \leqslant 2 \sqrt{d-1} / d
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Smallest possible (Alon-Boppana)

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Smallest possible (Alon-Boppana)
Their novel techniques also resolve 54-year-old Kadison-Singer problem in Mathematics and engineering

## Matrix multiplication and computational complexity

- Given two matrices $\mathbf{A}$ and $\mathbf{B}$ of size $n$, compute $\mathbf{A B}$
- Recall (AB) ${ }_{i j}=\sum_{k} \mathbf{A}_{i k} \mathbf{B}_{k j}$

Straightforward algorithm requires roughly $n^{3}$ elementary operations

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- Strassen algorithm: roughly $n^{\log _{2} 7} \approx n^{2.807}$ elementary operations
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- Is $n^{2}$ possible?

If so, potentially very useful If not, why not?

