https://vvtesh.sarahah.com/

Information Retrieval

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So much of life, it seems to me, is determined by pure randomness. – **Sidney Poitier.**



Venkatesh Vinayakarao (Vv)

A Random Surfer Model

• A random surfer may start from any node with 1/3 probability



 Can you represent this graph using Adjacency Matrix?

A Random Surfer

• May start from any node with 1/3 probability



 Can you represent this graph using Adjacency Matrix?

$$\mathbf{A} = \begin{bmatrix} 0 \ 1 \ 0 \\ 1 \ 0 \ 1 \\ 0 \ 1 \ 0 \end{bmatrix}$$

A Random Surfer

• May start from any node with 1/3 probability



• May also teleport to any node with α probability

$$\mathbf{A} = \begin{bmatrix} 0 \ 1 \ 0 \\ 1 \ 0 \ 1 \\ 0 \ 1 \ 0 \end{bmatrix}$$

How can you compute the transition probabilities?

Transition Probabilities

• We can convert the Adjacency Matrix (A) to Transition Probability Matrix (P)



- If the random surfer is at 1,
 - and he did not teleport
 - Probability = 1α
 - and he teleports
 - he may reach state3 with probability $\alpha/3$
 - and may reach state 2 with probability $\alpha/3$
- Transition Probability from 1 is $(\alpha/3, (1 \alpha) + \alpha/3, \alpha/3)$

Quiz

• If the teleportation probability, $\alpha = 0.5$, Calculate the transition probability matrix for this network.



Quiz

• If the teleportation probability, $\alpha = 0.5$, Calculate the transition probability matrix (P) for this network.



Which page is more popular?

• If a random surfer at 1 can reach (1,2,3) with probabilities (1/6, 2/3, 1/6), where will he end up in the next time slot if choses to continue his walk?



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• If a random surfer at 1 can reach (1,2,3) with probabilities (1/6, 2/3, 1/6), where will he end up in the next time slot if choses to continue his walk?



Steady State Probability

- If the random surfer keeps walking, the probabilities tend to converge!
 - Since we have an Ergodic Markov Chain!
 - A markov chain is ergodic if every state is reachable from every other state (not necessarily in a single jump).
- In our case, we should get (5/18, 8/18, 5/18)
- So, page 2 gets the highest rank.

HITS

Hyperlink-Induced Topic Search.

Another approach to finding popular pages

Hubs and Authorities

Topic specific page rank can be useful!

Query: I wish to learn about leukemia

nature	Condon About Blog Latest news and research from Nature com on the topic of Leukaemia Frequency about 4 posts per week. Blog nature.com/subjects/leukaemia Facebook fans 907,612. Twitter followers 1,574,259.
View Latest Posts	5 *
Subscribe newsle	itter
Enter email	Continue OR G Continue with Google
2. St. Baldrid	K's Foundation - Childhood Cancer Research Foundation Global About Blog St. Baldricks is a childhood cancer charity funding the most promising cures for kids with cancer. Read St. Baldrick's blog to learn more about childhood



Hub Page <u>https://blog.feedspot.com/leukemia_blogs/</u> Authority Page www.cancer.gov

Hubs and Authorities Score

• **Objective**: Assign a hub score and an authority score for each page.



Hub Score (say, h(v) = 0.002) Authority Score (say, a(v) = 0.17)

Hub Page

- A good hub page points to many good authorities
- A good authority page is pointed-to by many hubs



HITS Algorithm

- For all web pages, initialize hub score (h(v)) and authority score (a(v)) to 1. Here, v is a web page.
- $v \rightarrow y$ denotes a link from v to y.
- Iteratively update h(v) and a(v). $h(v) = \sum_{v \to y} a(y)$ $a(v) = \sum_{y \to v} h(y)$ $\vec{h} = A \vec{a}$ $\vec{a} = AT \vec{h}$ What happens on repeated updates? Rewriting in matrix form A is the adjacency matrix

Use your mathematical hat!

HITS Algorithm



Hyperlink Induced Topic Search



*Words in this graph are the anchor texts on the links

HITS Example



Thank You