

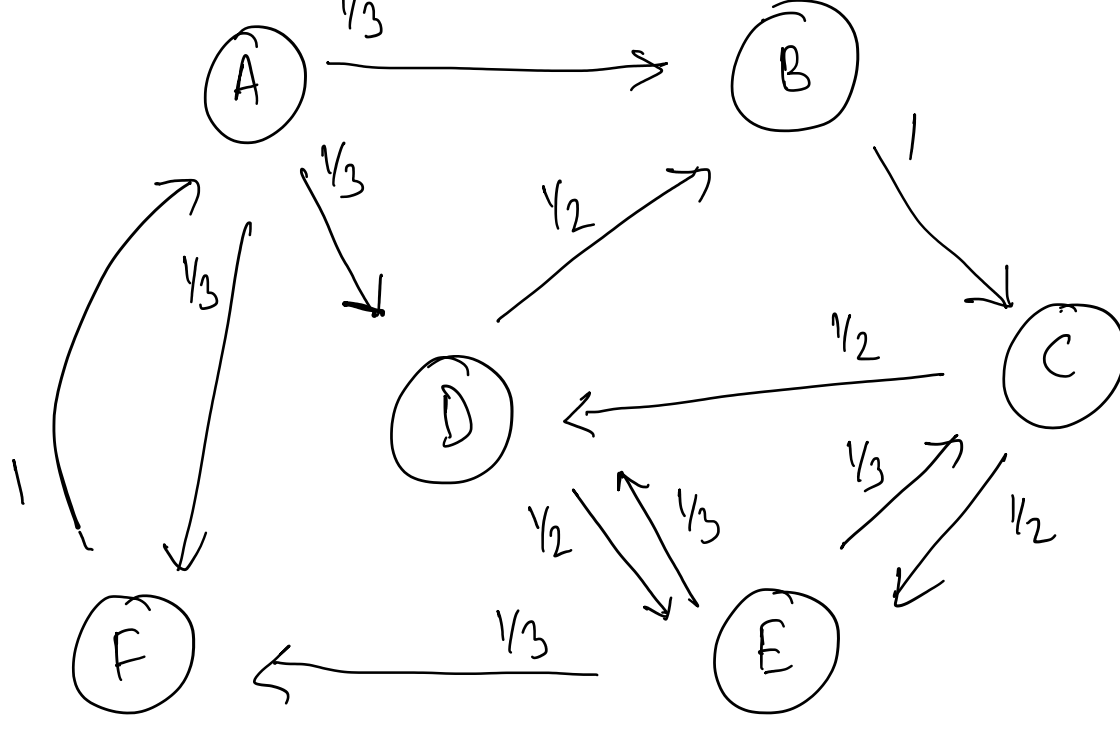
Page Rank Algorithm

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Developed by Larry Page + Sergey Brin in 1996 to determine relevance of webpages

Tiny model of the web:



Consider a user that when visiting a page, clicks a link chosen uniformly at random from the links on a page.

At any point of time, for any page P , what is the probability that the user is on page P ?

$$\text{Let } H = \begin{matrix} & \begin{matrix} A & B & C & D & E & F \end{matrix} \\ \begin{matrix} A \\ B \\ C \\ D \\ E \\ F \end{matrix} & \begin{bmatrix} & 1/3 & & 1/3 & & 1/3 \\ & & 1 & & & \\ & & & 1/2 & 1/2 & \\ & 1/2 & & & 1/2 & \\ & & 1/3 & 1/3 & & 1/3 \\ 1 & & & & & \end{bmatrix} \end{matrix}$$

For page P, P' , $H_{P,P'}$ is the probability that the user goes to P' , given that she is on page P .

Suppose $\pi(t)$ denotes the vector of probabilities of this user at time t . So $\pi_P(t)$ is prob. user is at page P at time t .

Suppose further $\pi(0) = \frac{1}{6} [1 \dots 1]^T$, i.e., the user is at a uniformly random page at $t=0$.

$$\text{Then } \pi_P(1) = \sum_{P'} \pi_{P'}(0) H_{P',P}$$

$$\begin{aligned} \Rightarrow \pi(1)^T &= \pi(0)^T H \\ \pi(2)^T &= \pi(1)^T H = \pi(0)^T H^2 \\ \dots \pi(t)^T &= \pi(0)^T H^t \end{aligned}$$

(This is what is called a Markov chain)

Is there a steady-state probability distribution?

Or will probability keep on changing at every step?

$$\text{want: } \pi \in \mathbb{R}_+^n : \pi^T = \pi^T H$$

$$\& \pi^T \mathbf{1} = 1$$

$$\text{OR: } \boxed{\pi^T (I - H) = 0}$$

If the Markov chain (in this case, H) is irreducible & aperiodic, then π exists & is unique.

(called the Fundamental Theorem of Markov Chains)

Ok, but will we ever reach π ?

π is called a stationary distribution

Yes, if H is irreducible & aperiodic.

Then for any starting distribution $\pi(0)$,

$$\pi^T = \lim_{n \rightarrow \infty} \pi(0)^T H^n$$

(and we can even obtain bounds on the convergence)

Lemma 1: Let H be the transition matrix for a aperiodic, irreducible Markov Chain. The the largest eigenvalue $\lambda_1 = 1$.

Proof: By the Fundamental Theorem, we know that $\exists \pi$ s.t. $\pi^T H = \pi^T$, hence \exists an eigenvector with value 1. So we only need to show that all eigenvalues have value ≤ 1 .

(try & prove this yourself)

(i) Fact: Left eigenvectors = right eigenvectors.

Hence we only need to show that if $\exists x, \lambda$ s.t.

$$Hx = \lambda x, \text{ then } \lambda \leq 1$$

(ii) Consider the matrix $H^2 = H \times H$.

Can check that H^2 is also a transition matrix:

- all entries ≥ 0 , &

- each row sums to 1

Thus so is H^k for all $k > 0$.

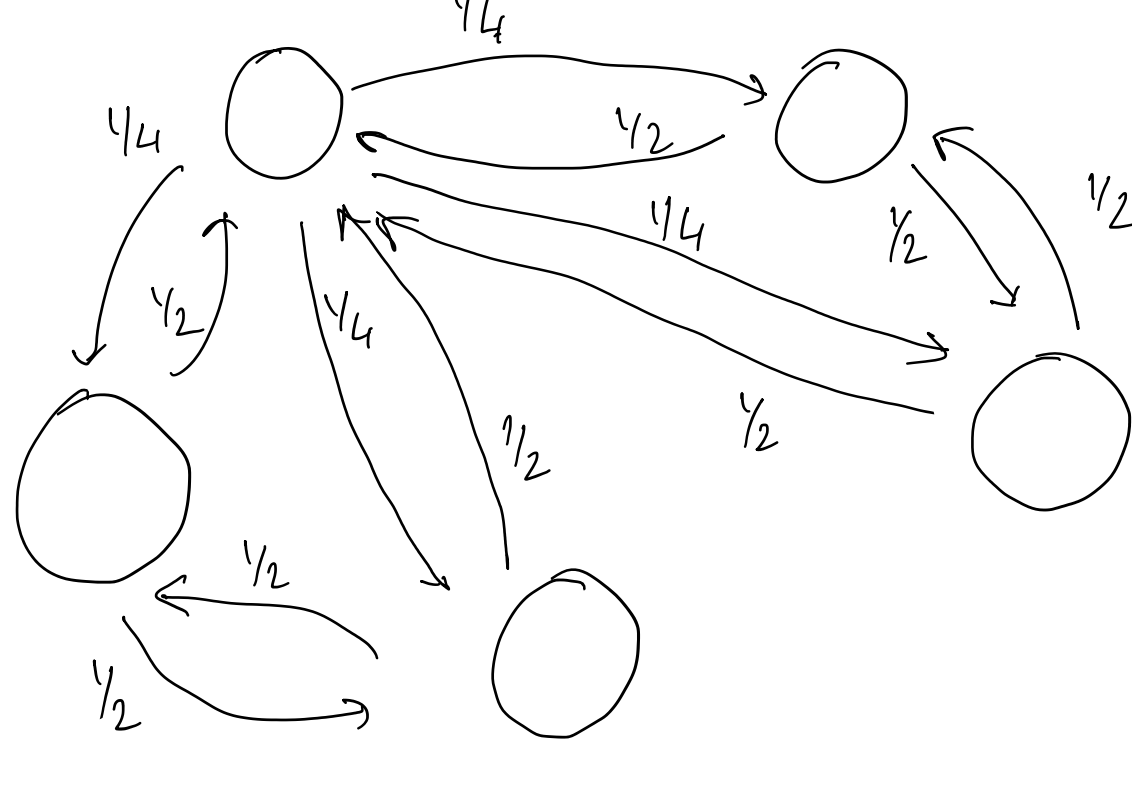
Now suppose $\exists x, \lambda$ s.t. $Hx = \lambda x$, $\lambda > 1$.

$$\text{then } H^k x = \lambda^k x$$

but each entry on the left is bounded,

while $\lambda^k x$ grows unboundedly. \square

Exercise: Consider a Markov chain with bidirectional edges, & from a vertex, an edge is chosen uniformly at random.



Then prove that the stationary distribution is:

$$\pi_v = \frac{\deg(v)}{2|E|}$$