

Announcements

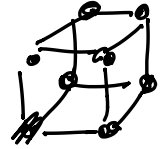
- start recording
- scribe: work w/other scribe to produce one set of notes
- My OH: Mon 12:30-1:30, Thu 9-10
- TF hybrid section/OH: Mon 2-3, Wed 2-3, Wed 5-6, Thu 5:30-6:30
- PS 1 posted, due Fri 9/25
- If Zoom gets down, check Piazza
- sync whiteboard

Agenda

- 1) Corrections + Recap
- 2) Mixing of Random walks on Undirected Graphs
- 3) Random Walks on Directed Graphs
- 4) Markov Chain Monte Carlo

Correction: eigenvalues of undirected n -cycle adjacency matrix $= \omega^{-r} + \omega^r = 2\cos\left(\frac{2\pi r n}{n}\right)$
 $r=0, \dots, n-1$

Example 1: Boolean hypercube $G = \mathbb{Z}_2^d$ $S = \{e_1, e_2, \dots, e_d\}$
 $\underbrace{\mathbb{Z}_2^d}_{\{0, 1\}^d}$ (i.e. $W = 1_S$)
 w/ addition mod 2



Fourier basis: for each $r \in \mathbb{Z}_2^d$
 $\chi_r(x) = (-1)^{\langle r, x \rangle} = (-1)^{\langle r, x \rangle \text{ mod } 2}$

$$\hat{W}_r = \sum_{s \in \mathbb{Z}_2^d} 1_S(s) \cdot (-1)^{\langle r, s \rangle}$$

$$= \sum_{i=1}^d (-1)^{\langle r, e_i \rangle} = \prod_{i=1}^d (-1)^{r_i}$$

$$= \#(0\text{'s in } r) - \#(1\text{'s in } r) = d - 2|r|$$

$$L = D - M = dI - M$$

$ r $	# e-values	e-value of M	e-value of L	e-value of W	e-value of N	Hamming weight
0	1	d	0	1	0	
1	d	d-2	2	$1 - 2/d$	$2/d$	
2	$\binom{d}{2}$	d-4	4	$1 - 4/d$	$4/d$	
3	$\binom{d}{3}$	d-6	6	$1 - 6/d$...	
...	
$d/2$	$\binom{d}{d/2}$	0	d	0	...	
...	...	-2	$d+2$	$-2/d$...	
...	
d-1	d	$-(d-2)$...	$-1 + 2/d$...	
d	1	-d	2d	-1	2	

Perron-Frobenius Thm for Symmetric Matrices

Let $M =$ symmetric, nonnegative real matrix (e.g. adjacency or normalized adj. mx of undirected G)
w/ corresponding graph G (i.e. $E = \{(a,b) : M(a,b) > 0\}$).

Eigenvalues $\mu_1 \geq \mu_2 \geq \dots \geq \mu_n$

Thm 1) \exists ~~strictly positive~~ ^{nonnegative} e-vector v_1 w/ e-value μ_1

~~(and the only nonneg. e-vectors are multiples of v_1)~~
+ strict positivity of v_1

2) $\mu_1 > \mu_2$ with strict inequality if G connected

3) $\mu_1 > -\mu_n$ with strict inequality if G connected + unimodular

Moreover, if G bipartite then $\mu_{n-i} = -\mu_i$ for $i=0, \dots, n-1$

Random Walks on Undirected Graphs $M(a,b) = w(b,a)$
 $n = \# \text{ vertices}$
 G undirected, weighted. $W = MD^{-1}$

$P_0 =$ initial probability distribution on vertices $\in \mathbb{R}^n$

$P_t =$ prob. dist. after t steps of r.w.

$$= W^t P_0$$

How to analyze?

Claim: W has a basis of e-vectors ϕ_1, \dots, ϕ_n
w/ e-values $\omega_1 \geq \omega_2 \geq \omega_3 \geq \dots \geq \omega_n$

PF: $W = D^{1/2} A D^{-1/2}$ for normalized adj. mx $A = D^{-1/2} M D^{-1/2}$
which is symmetric

$\Rightarrow A$ has orthonormal basis of e-vectors ψ_1, \dots, ψ_n
with e-values $\omega_1 \geq \omega_2 \geq \omega_3 \geq \dots \geq \omega_n$

$\Rightarrow \phi_i = D^{1/2} \psi_i$ is basis of e-vectors for W \square
(not necessarily orthogonal when G irregular)

Write $P_0 = c_1 \phi_1 + c_2 \phi_2 + \dots + c_n \phi_n$

$$W^t P_0 = c_1 \omega_1^t \phi_1 + c_2 \omega_2^t \phi_2 + \dots + c_n \omega_n^t \phi_n$$

Will be dominated by eigenvector(s) of largest magnitude

$\Rightarrow \omega_1 = 1$ because $\psi_1 = \frac{\vec{d}^{1/2}}{\|\vec{d}^{1/2}\|}$ is strictly pos e-vector
of A of e-value 1. By P-F (on each
coordinate) $|\omega_i| \leq 1$ for $i = 2, \dots, n$.

$$\phi_1 = D^{1/2} \psi_1 = \frac{\vec{d}}{\|\vec{d}\|_{1/2}}$$

$$c_1 = \underbrace{(D^{-1/2} p_0)^T}_{\vec{d}} \psi_1 = p_0^T D^{-1/2} \frac{\vec{d}}{\|\vec{d}\|_{1/2}} = \frac{p_0^T \vec{d}}{\|\vec{d}\|_{1/2}} = \frac{1}{\|\vec{d}\|_{1/2}}$$

$$c_1 \phi_1 = \frac{\vec{d}}{\|\vec{d}\|_{1/2}^2} = \frac{\vec{d}}{\sum_a d(a)} \stackrel{\text{def}}{=} \pi \quad \text{stationary dist.} \\ (W\pi = \pi)$$

Thm: $\max_{j>1} |\omega_j| = \max_{\text{prob. dist } P} \frac{\|W P - \pi\|_{\pi^{-1}}}{\|P - \pi\|_{\pi^{-1}}} \stackrel{\text{def}}{=} \omega_{\pi}$

where $\|x\|_{\pi^{-1}} \stackrel{\text{def}}{=} \sqrt{\sum_{a=1}^n \frac{x(a)^2}{\pi(a)}}$

Proof: $\max_{j>1} |\omega_j| \stackrel{ps1}{=} \max_{x \perp \psi_1} \frac{\|Ax\|}{\|x\|} \quad (A = D^{-1/2} W D^{1/2})$

$$= \max_{y: D^{1/2} y \perp \psi_1} \frac{\|D^{-1/2} W y\|}{\|D^{-1/2} y\|} \quad (x = D^{-1/2} y)$$

$$= \max_{y: \sum_a y_a = 0} \frac{\|W y\|_{\vec{d}^{-1}}}{\|y\|_{\vec{d}^{-1}}} \quad \left. \begin{array}{l} \uparrow \\ P = \pi + cy \end{array} \right\}$$

$$= \max_{\text{prob dist } P} \frac{\|W(P - \pi)\|_{\vec{d}^{-1}}}{\|P - \pi\|_{\vec{d}^{-1}}}$$

$$= \max_{\text{prob dist } P} \frac{\|W P - \pi\|_{\pi^{-1}}}{\|P - \pi\|_{\pi^{-1}}}$$

By def: $\|W^t P_0 - \pi\|_{\pi^{-1}} \leq \omega_{\pi}^t \cdot \|P_0 - \pi\|_{\pi^{-1}}$

- initial distance $\|P_0 - \pi\|_{\pi^{-1}} \leq \|P_0\|_{\pi^{-1}} \leq \sqrt{\frac{1}{\pi_{\min}}}$

- final ℓ_1 distance (= 2 · total variation distance)

$$\|W^t P_0 - \pi\|_1 = \sum_a \frac{|W^t P_0(a) - \pi(a)|}{\pi(a)^{1/2}} \cdot \pi(a)^{1/2}$$

$$\leq \left(\sum_a \frac{(W^t P_0(a) - \pi(a))^2}{\pi(a)} \right)^{1/2} \cdot 1 = \|W^t P_0 - \pi\|_{\pi^{-1}}$$

Cauchy-Schwarz

Thm: for every start distribution P_0

$$\|W^t P_0 - \pi\|_1 \leq \omega_{\pi}^t \cdot \|P_0\|_{\pi^{-1}} \leq \omega_{\pi}^t \cdot \frac{1}{\sqrt{\pi_{\min}}}$$

where $\pi_{\min} = \frac{d_{\min}}{\sum_a d(a)} \in \left[\frac{d_{\min}}{d_{\max} n}, \frac{1}{n} \right]$

Cor: "Mixing time" to get to within ℓ_1 dist. ϵ

$$t = O\left(\frac{\log\left(\frac{nd_{\max}}{\epsilon d_{\min}}\right)}{1 - \omega_{\pi}} \right)$$

Q: how to ensure $\omega_{\pi} < 1$?

A: $\omega_{\pi} = \max\{\omega_2, -\omega_n\}$ $\omega_2 < 1$ iff G connected
 $\omega_n > -1$ iff G nonbipartite

Lazy Random Walk : $\tilde{W} = \frac{1}{2}W + \frac{1}{2}I$

$$\tilde{\omega}_j = \frac{1+\omega_j}{2} \geq 0, \text{ so } \tilde{\omega}_\pi = \tilde{\omega}_2 = \frac{1+\omega_2}{2}$$

$$\text{and } 1 - \tilde{\omega}_\pi = \frac{1-\omega_2}{2} = \frac{\lambda_2}{2}$$

$\lambda_2 = 2^{\text{nd}}$ smallest e-value of $N = I - A$ normalized laplacian

Exercise: bound mixing time of lazy + non-lazy random walks on hypercube combinatorially and using e-values, and compare.

Let $k = \text{dimension of hypercube}$

$$\omega_1 = 1 \quad \omega_2 = 1 - 2/k \quad \omega_n = -1 \quad \pi = \text{uniform on } 2^k \text{ vertices}$$

Non-lazy $\omega_\pi = 1 \Rightarrow \text{no mixing}$

Lazy: $\tilde{\omega}_\pi = \frac{1+\omega_2}{2} = 1 - \frac{1}{k}$ ϵ const.

$$t_{\text{mix}} = O\left(\frac{\log(1/\epsilon)}{1 - \tilde{\omega}_\pi}\right) = O(k^2)$$

Combinatorially $t_{\text{mix}} = O(k \log k)$
(time to select all coordinates)

Converse (maybe ps 2): \forall undirected G \exists start vertex s
such that for $p_0 = \delta_s$:

$$t = \Omega\left(\frac{c_\pi}{1 - c_\pi} \cdot \log\left(\frac{1}{\epsilon}\right)\right)$$

steps are needed to get to within
 ℓ_1 distance ϵ of π .

Also: for every undirected, connected, nonbipartite G ,
 $c_\pi \leq 1 - 1/\text{poly}(n \cdot d_{\max})$ (cf. ps 1)
so mixing time $\leq \text{poly}(n \cdot d_{\max})$.

Random Walks on Directed Graphs

Q: what changes?

Q: what happens in non-connected digraphs?