

Analysis of Network Diffusion and Distributed Network Algorithms

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Overview of the 4 Sessions

- Random walks
- Percolation processes
 - Branching processes, random graphs, and percolation phenomena
- Rumors & routes
 - Rumor spreading, small-world model, network navigability
- Distributed algorithms
 - Maximal independent set, dominating set, local balancing algorithms

Random Walks

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Outline

- Basic definitions and notation
- Applications
- Two results:
 - Mixing time and convergence of random walks
 - Cover time of random walks
- Applications to clustering
- Techniques:
 - Probability theory
 - Spectral graph theory

What is a Random Walk?

- Let G be an arbitrary undirected graph
- A walk starts at an arbitrary vertex v_0
- At the start of step t , the walker moves from v_{t-1} to vertex v_t chosen uniformly at random from neighbors of v_{t-1} in G
- For all $t > 0$, v_t is a random variable

Notation

- Let G be an arbitrary undirected graph and A be its adjacency matrix
 - A_{ij} is 1 whenever there is an edge (i,j)
- Define the random walk matrix M
 - M_{ij} is $A_{ij}/\text{degree}(i)$
- Let x denote the initial probability distribution (row) vector
- After t steps, the probability distribution vector equals xM^t

Definitions

- **Stationary distribution**
 - Probability vector π such that $\pi M = \pi$
- **Hitting time h_{ij}**
 - Expected time for random walk starting from i to visit j
- **Cover time C**
 - Expected time for random walk starting from an arbitrary vertex to visit all nodes of G
- **Mixing time**
 - Time it takes for the random walk to converge to a stationary distribution

Questions of Interest

- Stationary distribution:
 - Do they always exist?
 - Is the stationary distribution unique?
 - Does a random walk always converge to a stationary distribution? If it does, what is the mixing time?
- Hitting time:
 - For a given graph G and vertices i, j , what is the hitting time h_{ij}
- Cover time:
 - For a given graph G , what is its cover time C ?

Applications

- Probabilistic process whose variants capture social and physical phenomena
 - Brownian motion in physics
 - Spread of epidemics in contact networks
 - Spread of innovation and influence in social networks
 - Connections to electrical networks
 - Markov chains arise in numerous scenarios
- Pseudo-random number generators
 - Random walk in an expander graph is an efficient way to generate pseudo-random bits from a small random seed
- Use in randomized algorithms
- Google's PageRank
 - PageRank is the probability vector of the stationary distribution of an appropriately defined random walk

Stationary Distribution and Mixing Time

- Lemma 1: A stationary distribution always exists and is unique
- For d -regular undirected graphs G , let $\lambda(G)$ denote the second largest eigenvalue of M
- **Theorem 1:** The random walk is within ε of the stationary distribution in

$$O\left(\frac{\ln(n / \varepsilon)}{1 - \lambda(G)}\right) \text{ steps}$$

- For non-regular graphs, replace M by a normalized version

Cover Time

- **Matthews Bound:**
 - Let h_{\max} be the maximum hitting time
 - The cover time is at most $h_{\max} \ln(n)$
- **Exercise:** Prove that time for a random walk to cover every vertex is $O(h_{\max} \log(n))$ whp
- **Theorem 2:** For any m -edge n -vertex undirected graph G , the cover time is $O(mn)$
- [Mitzenmacher-Upfal 04]

Lovasz-Simonovits Theorem

- Lazy random walk:
 - With probability $\frac{1}{2}$, walk stays at current node;
 - With probability $\frac{1}{2}$, does regular random walk
- Theorem 3: [LS 93] For any initial probability distribution and every t , we have

$$I_t(x) \leq \min(\sqrt{x}, \sqrt{2m-x}) \left(1 - \frac{\phi^2}{2}\right)^t + \frac{x}{2m}$$

$$p_t(e_1) \geq p_t(e_2) \geq \dots \geq p_t(e_{2m})$$

Define $I_t(k) = \sum_{i=1}^k p_t(e_i)$ and extend to interval $[0, 2m]$ by linear interpolation

Sparsest Cut

- **Conductance**: Φ measures the “expansion” of a graph

$$\phi = \min_{S \subset V, S \neq \emptyset} \frac{e(S, V - S)}{\min(e(S), e(V - S))}$$

- Finding the cut $(S, V - S)$ that yields the above minimum ratio is the sparsest cut problem
- LP rounding: Yields $O(\log(n))$ -approximation [Leighton-Rao 88, Linial-London-Rabinovich 94]
- SDP rounding: Yields $O(\sqrt{\log(n)})$ approximation [Arora-Rao-Vazirani 05]

Local Clustering

- Suppose you are given a massive graph and want to find a “good” cluster containing a given vertex v
 - Good means low conductance
- Approach: Solve the sparsest cut problem and return the cluster containing v
 - Too expensive
- Local clustering [Spielman-Teng 08, Andersen-Chung-Lang 08]:
 - Start a random walk from v , maintaining the probability vector for each vertex
 - Keep zeroing out vertices that have very low probability
 - LS Theorem helps in showing that in time nearly proportional to the size of the cluster, can achieve close to desired conductance

Take Away Messages

- Random walk and related processes
 - Arise in several scenarios
 - Are useful primitives for designing fast algorithms
 - Yield effective and practical pseudo-random sources
- Analysis tools for random walks
 - Basic probability (Markov's inequality, Chernoff-type bounds, Martingales)
 - Spectral graph theory