

{ Csc 80030 | Lecture 8 }

PROBABILISTIC ANALYSIS &
RANDOMIZED ALGORITHMS

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PART 0

- ▶ HOMEWORK: HW4 due next week, HW5 out by sun
- ▶ PROGRAMMING ASSIGNMENT: due tonight
- ▶ TODAY: random-walk algorithms

PART 1 | undirected s-t connectivity

Undirected S-T Connectivity

- ▶ S-T CONNECTIVITY: Given a directed graph G and two vertices s and t , is there a path from s to t in G ?
- ▶ can be solved in linear time and space using BFS or DFS.
- ▶ UNDIRECTED S-T CONNECTIVITY (USTCON): Given an undirected graph G and two vertices s and t , is there a path from s to t in G ?
- ▶ TODAY: randomized algorithm for USTCON running in space $O(\log n)$.

USTCON algorithm

USTCON via random walks

Input: (G, s, t) , where $G = (V, E)$ has n vertices

1. Let $v = s$. Repeat up to n^4 times:
 - 1.1 If $v = t$, halt and accept.
 - 1.2 Else update v to be a random neighbor of v .
 2. Reject (if we haven't visited t yet).
- ▶ space $O(\log n)$ to store current vertex v and a counter for # steps
 - ▶ never accepts when there isn't a path from s to t .
 - ▶ if G is a connected d -regular graph, then a random walk of length $\tilde{O}(d^2 n^3)$ from s will hit t w.h.p.
 - ▶ replace each vertex v with a cycle of length $\deg(v)$.

random-walk matrix

- ▶ Define *random-walk matrix* of an n -vertex digraph G to be the $n \times n$ matrix M where $M_{i,j}$ is the prob. of going from vertex i to j in a single step.
- ▶ If G is d -regular, then M is $\frac{1}{d}$ times adjacency matrix of G .
- ▶ For every probability distribution $\pi \in \mathbb{R}^n$ on vertices of G , the vector πM is the probability distribution obtained after taking one step of the random walk.
- ▶ Start with π concentrated at vertex s , interested in πM^k after taking k steps on the graph.
- ▶ We will show that random walk converges to the uniform distribution $u = (\frac{1}{n} \quad \frac{1}{n} \quad \dots \quad \frac{1}{n})$ after $\text{poly}(n)$ steps.

Basic linear algebra

- ▶ A non-zero vector $v \in \mathbb{R}^n$ is an *eigenvector* of a $n \times n$ matrix M if $vM = \lambda v$ for some $\lambda \in \mathbb{R}$, the corresponding *eigenvalue*.
- ▶ EXAMPLE. the graph K_3

$$M = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix}$$

$$v_1 = (\frac{1}{3} \quad \frac{1}{3} \quad \frac{1}{3}), \quad \lambda_1 = 1;$$

$$v_2 = (1 \quad -1 \quad 0), \quad \lambda_2 = -1/2;$$

$$v_3 = (1 \quad 0 \quad -1), \quad \lambda_3 = -1/2$$

Basic linear algebra

Theorem

If M is a symmetric $n \times n$ real matrix, then M has n eigenvectors which form an orthogonal basis for $\mathbb{R}^n$.

Henceforth, M is random-walk matrix for a d -regular undirected graph.

- ▶ M is symmetric, with eigenvectors $v_1, \dots, v_n$.
- ▶ Since $uM = u$, $v_1 = u$ is an eigenvector of eigenvalue $\lambda_1 = 1$.
- ▶ Assume $|\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n|$.
- ▶ Note $\langle \pi - u, u \rangle = 0$, so we can write $\pi = u + c_2 v_2 + \dots + c_n v_n$.
- ▶ After k steps, the distribution is $\pi M^k = u + \lambda_2^k c_2 v_2 + \dots + \lambda_n^k c_n v_n$, so
$$\|\pi M^k - u\| \leq |\lambda_2|^k \|\pi - u\| \leq |\lambda_2|^k.$$

Random walk analysis

Will show on homework:

- ▶ All eigenvalues of M have absolute value at most 1.
- ▶ If G is connected and non-bipartite, then all eigenvalues (other than 1) have absolute value at most $1 - \Omega(1/(dn)^2)$.

From before,

- ▶ $\|\pi M^k - u\| \leq 2(1 - \Omega(1/(dn)^2))^k$.
- ▶ Take $k = O(d^2 n^2 \log n)$, every entry of πM^k is at least $1/2n$.
- ▶ Take $O(n)$ such walks, we will hit vertex t with probability at least $1/2$.

PART 2 | randomized 2-SAT

Satisfiability problems

- ▶ general satisfiability problem (SAT):

- ▶ input: Boolean formula given as AND of OR's: e.g.

$$\phi(x_1, x_2, x_3, x_4) = (x_1 \vee \bar{x}_2 \vee x_4) \wedge (\bar{x}_1 \vee \bar{x}_3) \wedge (x_3 \vee x_4)$$

- ▶ n variables x_1, x_2, x_3, x_4 ; literals e.g. $x_1, \bar{x}_2$; m clauses e.g. $(x_3 \vee x_4)$
 - ▶ assignment e.g. (T, F, F, T)
 - ▶ ϕ is satisfiable if there exists an assignment for which ϕ evaluates to T
 - ▶ SAT problem: decide if the input ϕ is satisfiable

- ▶ 2-satisfiability problem (2-SAT):

- ▶ restricted version of SAT where each clause contains exactly 2 literals
 - ▶ $\phi(x_1, x_2, x_3, x_4) = (x_1 \vee \bar{x}_2) \wedge (\bar{x}_1 \vee \bar{x}_3) \wedge (x_1 \vee x_2) \wedge (x_4 \vee \bar{x}_3) \wedge (x_4 \vee \bar{x}_1)$

2-satisfiability algorithm RAND2SAT

1. Start with an arbitrary assignment
2. Repeat up to $2mn^2$ times, terminating if all clauses are satisfied:
 - 2.1 Choose an arbitrary (first) clause that is not satisfied
 - 2.2 Choose uniformly at random one of the literals in the clause and switch the value of the variable
3. If a satisfying assignment has been found, return it
4. Otherwise, return “unsatisfiable”

2-satisfiability algorithm: example

► $\phi(x_1, x_2, x_3, x_4) = (x_1 \vee \bar{x}_2) \wedge (\bar{x}_1 \vee \bar{x}_3) \wedge (x_1 \vee x_2) \wedge (x_4 \vee \bar{x}_3) \wedge (x_4 \vee \bar{x}_1)$

1. assignment $A_0 = (F, T, T, F)$

unsatisfied clauses: $x_1 \vee \bar{x}_2$, $x_4 \vee \bar{x}_3$

switch x_1 in $C_0 = x_1 \vee \bar{x}_2$

2. assignment $A_1 = (T, T, T, F)$

unsatisfied clauses: $\bar{x}_1 \vee \bar{x}_3$, $x_4 \vee \bar{x}_3$, $x_4 \vee \bar{x}_1$

switch x_3 in $C_1 = \bar{x}_1 \vee \bar{x}_3$

3. assignment $A_2 = (T, T, F, F)$

unsatisfied clauses: $x_4 \vee \bar{x}_1$

switch x_4 in $C_2 = x_4 \vee \bar{x}_1$

4. assignment $A_3 = (T, T, F, T)$

unsatisfied clauses: none

2-satisfiability algorithm: analysis

- ▶ suppose ϕ is satisfiable with an assignment S
 - ▶ e.g. ϕ as before, S is (T, T, F, T)
- ▶ let X_i be the number of variables with same values in A_i and S
 - ▶ e.g. $X_0 = 1, X_1 = 2, X_2 = 3, X_3 = 4$
 - ▶ if $X_i = n$, then algorithm terminates with satisfying assignment
 - ▶ how long does it take for X_i to reach n ?
- ▶ compute transition probabilities for $X_0, X_1, X_2, \dots$
 - ▶ if we pick a variable where A_i and S agree, $X_{i+1} = X_i - 1$
 - ▶ if we pick a variable where A_i and S disagree, $X_{i+1} = X_i + 1$

2-satisfiability algorithm: analysis

- ▶ if $X_i = 0$, then $\Pr[X_{i+1} = 1 \mid X_i = 0] = 1$
- ▶ if $X_i = j$, where $1 \leq j \leq n - 1$
 - ▶ A_i and S disagree on at least one variable in clause C_i
 - ▶ $\Pr[\text{pick a variable where } A_i \text{ and } S \text{ disagree}] \geq 1/2$

$$\Pr[X_{i+1} = j + 1 \mid X_i = j] \geq 1/2$$

$$\Pr[X_{i+1} = j - 1 \mid X_i = j] \leq 1/2$$

2-satisfiability algorithm: analysis

- ▶ consider Markov chain $Y_0, Y_1, Y_2, \dots$ with states $0, 1, 2, \dots, n$

$$Y_0 = X_0$$

$$\Pr[Y_{i+1} = 1 \mid Y_i = 0] = 1$$

$$\Pr[Y_{i+1} = j + 1 \mid Y_i = j] = 1/2$$

$$\Pr[Y_{i+1} = j - 1 \mid Y_i = j] = 1/2$$

- ▶ expected time to reach n in Y is $\geq$ expected time to reach n in X
- ▶ let h_j be expected # of steps it takes to reach n from j

2-satisfiability algorithm: analysis

- ▶ computing $h_0, h_1, \dots, h_n$
 - ▶ base case: $h_n = 0$ and $h_0 = h_1 + 1$
 - ▶ for $1 \leq j \leq n - 1$: $h_j = \frac{1}{2}(h_{j-1} + 1) + \frac{1}{2}(h_{j+1} + 1)$
 - ▶ solving $n + 1$ equations in $n + 1$ unknowns, we obtain $h_j = n^2 - j^2$ and $h_0 = n^2$
 - ▶ expected time to reach n in X is at most n^2
- ▶ RAND2SAT
 - ▶ if ϕ is unsatisfiable, always returns the right answer “unsatisfiable”
 - ▶ if ϕ is satisfiable, returns a satisfying assignment with prob $\geq 1 - 2^{-m}$

THE END | next, ...