

Monte Carlo Methods

collection of tools for estimating values thru sampling & estimation

(ϵ, δ) Approximation

A randomized alg gives an (ϵ, δ) approx for value V if the output X of the alg satisfies

$$\Pr(|X - V| > \epsilon | V|) \leq \delta$$

Monte Carlo Thm

Let X_1, X_2, \dots, X_m iid. Bernoulli with $E(X_i) = \mu$

If $m \geq \frac{3 \ln(\frac{2}{\delta})}{\epsilon^2 \mu}$ then

$$\Pr\left(\left|\frac{1}{m} \sum_{i=1}^m X_i - \mu\right| \geq \epsilon \mu\right) \leq \delta$$

Pf: Chernoff bounds

#P complexity class associated with counting solns to problems in NP

#P complete problems:

indep sets in a graph

satisfying assignments to DNF formula

perfect matchings in a bipartite graph

Fully polynomial randomized approx scheme FPRAS

a randomized alg for which, given an input x and any parameters ϵ & δ with $0 < \epsilon, \delta < 1$, the alg outputs an (ϵ, δ) approx to $V(x)$ in time poly in $\frac{1}{\epsilon}$, $\ln \frac{1}{\delta}$ & size of input

Note: suffices to take $\delta = \frac{1}{4}$

because easy to boost error prob:

Run $k = 16 \log\left(\frac{2}{\delta}\right)$ trials w. error prob $\frac{1}{4} \Rightarrow y_1, y_2, \dots, y_k$

Let $m = \text{median}(y_1, \dots, y_k)$

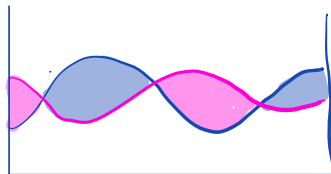
Then $\Pr(m \notin (1 \pm \epsilon)V(x)) \leq \delta$ by Chernoff

Want to sample from some set (e.g. ISs of graph G)
where Π is desired distribution $\Pr_{\Pi} = \Pr(\text{output IS } I)$
(often Π uniform distn)

Definition

total variation distance between 2 distns on same sample space \mathcal{X}

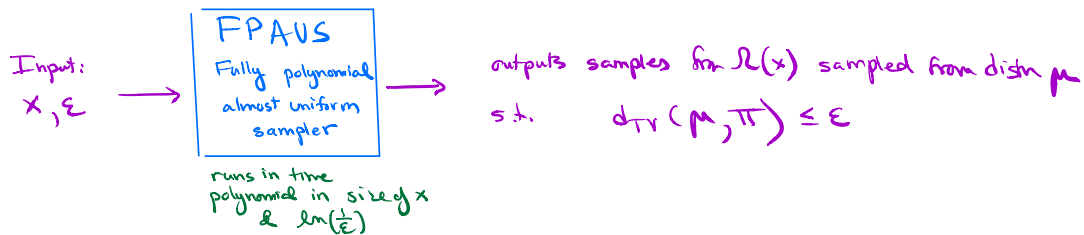
$$\|D_1 - D_2\|_{TV} = \frac{1}{2} \sum_{x \in \mathcal{X}} |D_1(x) - D_2(x)| = \max_{A \subseteq \mathcal{X}} |D_1(A) - D_2(A)|$$



$\|D_1 - D_2\|_{TV} = \text{pink area} = \text{blue area}$

For sampling problems, we seek

FPAUS - fully polynomial almost uniform sampler



Approximate counting
FPRAS



Approximate Sampling
FPAUS

self reducible problems
randomized reduction from
problems of size k
to problems of size $k-1$

Lemma Given FPAUS for sampling indep sets of graph G ,
we can construct FPRAS for estimating $|\Omega(G)|$
ISs.

Pf Want estimate $\bar{\Omega}$ of $|\Omega(G)|$
 $\Pr(|\bar{\Omega} - |\Omega(G)|| \geq \epsilon |\Omega(G)|) \leq \delta$

$G=(V,E)$ e_1, e_2, \dots, e_m arbitrary ordering of edges

$E_i = \{e_1, e_2, \dots, e_i\}$ $G_i = (V, E_i)$ $G_m = G$ $G_0 = (V, \emptyset)$

$\Omega(G_i)$: set of ISs in G_i

$$|\Omega(G)| = \frac{|\Omega(G_m)|}{|\Omega(G_{m-1})|} \times \frac{|\Omega(G_{m-1})|}{|\Omega(G_{m-2})|} \times \dots \times \frac{|\Omega(G_1)|}{|\Omega(G_0)|} \times |\Omega(G_0)| \rightarrow 2^n$$

$$\text{Let } r_i = \frac{|\Omega(G_i)|}{|\Omega(G_{i-1})|}$$

$$|\Omega(G)| = \prod_{i=1}^m r_i \cdot 2^n$$

Observation: $\frac{1}{2} \leq r_i \leq 1$

G_i has one extra
edge, say (u,v)

$$\Omega(G_i) \subseteq \Omega(G_{i-1})$$

Can't sample from uniform distn on $\mathcal{L}(G_{T-1})$
 but can use FPAUS

Two errors we need to bound:

① FPAUS \neq exact sampler so avg value $\neq r_i$:

But using, say, $\frac{\epsilon}{6m}$ sampler $|E(\tilde{r}_i) - r_i| \leq \frac{\epsilon}{6m}$

② With samples we get approx to $E(\tilde{r}_i)$

since r_i big ($\geq \frac{1}{2}$), $E(\tilde{r}_i)$ big \Rightarrow

Monte Carlo Thm $\Rightarrow O\left(\frac{m^2}{\epsilon^2} \ln\left(\frac{1}{\delta}\right)\right)$ samples suffic

From these bound $R = \prod_{i=1}^m \frac{r_i}{2^n}$

Claim: If \tilde{r}_i is $\left(\frac{\epsilon}{2m}, \frac{\delta}{m}\right)$ approx to r_i $\forall i$
 then X is (ϵ, δ) approx to $|L(G)|$

$\prod_{i=1}^m \tilde{r}_i \approx 2^n$

Pr: $\Pr(|\tilde{r}_i - r_i| \leq \frac{\epsilon}{2m} r_i) \geq 1 - \frac{\delta}{m} \quad \forall i$

$\Rightarrow r_i \left(1 - \frac{\epsilon}{2m}\right) \leq \tilde{r}_i \leq r_i \left(1 + \frac{\epsilon}{2m}\right) \quad \text{w.p. } \geq 1 - \frac{\delta}{m}$

$\Rightarrow \left(1 - \frac{\epsilon}{2m}\right)^m \leq \frac{\tilde{r}_i}{r_i} \leq \left(1 + \frac{\epsilon}{2m}\right)^m$

$1 - \epsilon \leq \left(1 - \frac{\epsilon}{2m}\right)^m \leq \frac{\tilde{r}_i}{r_i} \leq \left(1 + \frac{\epsilon}{2m}\right)^m \leq 1 + \epsilon \quad \text{w.p. } \geq 1 - \delta$

$(1 - \epsilon) \prod_{i=1}^m r_i \leq \prod_{i=1}^m \tilde{r}_i \leq (1 + \epsilon) \prod_{i=1}^m r_i \quad \text{w.p. } \geq 1 - \delta$

$\equiv \Pr\left(\left|\prod_{i=1}^m \tilde{r}_i - \prod_{i=1}^m r_i\right| \geq \epsilon \prod_{i=1}^m r_i\right) \leq \delta$

How to sample elts from a unimod Ω with $|\Omega|=n$ according to distn $\Pi=(\pi_1, \dots, \pi_n)$?

Cool idea: design MC whose state space is Ω that has stationary distn Π

- simulate MC until it "mixes"
- use state at that time as sample

2 key questions:

- ① how to design chain w/ right Π ?
- ② how to bound mixing time?

Example: sampling indep sets uniformly from $G=(V, E)$

states: indep sets

X_t : some indep set

MC:

choose vertex v u.a.r. from V

if $v \in X_t$, then $X_{t+1} = X_t \setminus v$

if $v \notin X_t$ and can be added without violating independence

then $X_{t+1} = X_t \cup v$

otherwise $X_{t+1} = X_t$

Claim:

- MC irreducible
- if \exists edge then aperiodic
- stationary distn uniform (chain doubly stochastic)
 $P_{i,j} = P_{j,i} \quad [1/n \text{ or } 0]$

General technique: given Ω and a connected graph on Ω define transition probs so that will have stationary distn Π

Metropolis Algorithm

Input: Ω ; connected graph $G=(\Omega, E)$, Π s.t. $\sum_{i \in \Omega} \pi_i = 1$, $\pi_i > 0$

Bounding mixing time

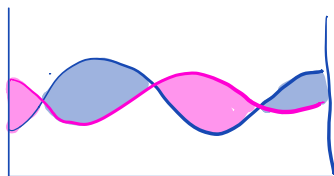
- ① Spectral gap, conductance, expansion
- ② coupling

⋮

Coupling

total variation distance between 2 distns on some sample space Ω

$$\|D_1 - D_2\|_{TV} = \frac{1}{2} \sum_{x \in \Omega} |D_1(x) - D_2(x)| = \max_{A \subseteq \Omega} |D_1(A) - D_2(A)|$$



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Ω

Common defn of mixing time $\tau(\epsilon)$

$$\tau(\epsilon) = \min \{t \mid \|P^t - \pi\|_{TV} \leq \epsilon\}$$

Say MC is rapidly mixing if $\tau(\epsilon)$ polynomial in $\log(|\Omega|)$ and $\log(\frac{1}{\epsilon})$

(we know this is related to spectral gap)

Coupling simple & elegant approach to bounding mixing time

Given a MC on \mathcal{L} , a coupling is a MC on $\mathcal{L} \times \mathcal{L}$ defining stochastic process (X_t, Y_t) s.t.

① each X_t & Y_t in isolation is faithful copy of MC

② If $X_t = Y_t$ then $X_{t+1} = Y_{t+1}$

Coupling Lemma Let (X_t, Y_t) be a coupling

Suppose $\exists T$ s.t. $\forall x, y$

$$\Pr(X_T \neq Y_T \mid X_0 = x, Y_0 = y) \leq \epsilon$$

Then $\tau(\epsilon) \leq T$

Examples:

- ① Random walk on hypercube $N=2^n$ nodes
in each step, choose random coordinate i , random bit $b \in \{0,1\}$
change i^{th} bit to b

② Indep sets of fixed size k here distances may \uparrow

Claim: choose vertex $v \in X_t$ u.a.r. & a vertex $w \in V$ u.a.r.,

If $w \notin X_t$ & $X_t - v + w$ indep then $X_{t+1} = X_t - v + w$
else $X_{t+1} = X_t$

Claim: MC rapidly mixing if $k \leq \frac{n}{3\Delta+3}$

