

## Monte Carlo Methods

collection of tools for estimating values thru sampling & estimation

### $(\epsilon, \delta)$ approximation

A randomized alg gives an  $(\epsilon, \delta)$  approx for value  $V$  of the output  $X$  of alg satisfies

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

### Example:

Sample indep random vars whose mean is quantity you want

Let  $X_1, \dots, X_m$  iid. indicator w/  $\mu = E(X_i)$

If  $m \geq \frac{3\ln\frac{2}{\delta}}{\epsilon^2 \mu}$  then from Chernoff bounds

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

## DNF counting

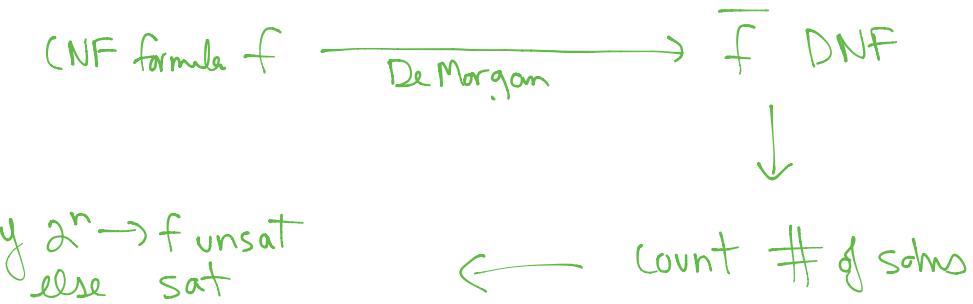
want to know # satisfying assignments

$$(\bar{x}_1 \wedge x_2) \vee (x_2 \wedge x_3) \vee (x_1 \wedge \bar{x}_2 \wedge \bar{x}_3 \wedge x_4) \vee (x_3 \wedge \bar{x}_4)$$

Observation: satisfying such a formula is easy

counting # satisfying assignments is hard

if could do this, could solve 3SAT



Problem actually  $\#P$  complete  $\rightarrow$  strong intractability

$\#P$  counting analog of NP: problems of form

compute  $f(x)$  where  $f(x)$

# solns to problem in NP

Approx counting of # solns of DNF formula  $\Phi$

Obvious approach: sample random assignments,  
estimate (fraction satisfying)  $2^n$

From  $(\varepsilon, \delta)$  calculation need  $m = \mathcal{O}\left(\frac{1}{\mu^2}\right)$

$m$  could be exponentially small e.g.  $\frac{n^2}{2^n}$

$$\Phi = C_1 \vee C_2 \vee \dots \vee C_t$$

$SC_i$ : set of assignments that satisfy clause  $i$ :

want to approx  $|\cup_i SC_i|$

Let  $V = \{(i, \vec{x}) \mid 1 \leq i \leq t, \vec{x} \in SC_i\}$

$$|V| = \sum_{i=1}^t |SC_i|$$

$$X = \{(i, \vec{x}) \mid \vec{x} \in SC_i, \vec{x} \notin SC_j \quad \forall j < i\}$$

estimate  $|X|$  by approximating  $\frac{|X|}{|U|}$ . To do this

sample elts  $(i, \vec{x}) \in U$  and determine if  $(i, \vec{x}) \in S$

Key observation:  $\frac{|S|}{|U|} \geq \frac{1}{f}$ , so  $\frac{3 + \ln(\frac{\delta}{\epsilon})}{\epsilon^2}$  samples suffice  
for  $(\epsilon, \delta)$  approx

How to pick random sample from  $U$ ?

pick  $i$  w prob  $\frac{|SC_i|}{\sum |SC_j|}$  set vars outside  $(i$  v.a.r.)

$$\Pr((i, \vec{x}) \text{ selected}) = \frac{|SC_i|}{|U|} \cdot \frac{1}{|SC_i|} = \frac{1}{|U|}$$

DNF counting illustrates fundamental connection

Sampling  $\leftrightarrow$  counting

very often use "self-reducibility" from problem of size  $i \rightarrow \text{size } i+1$

approx sampling  $\rightarrow$  approx counting

illustrate reduction for counting ISs

## Counting Indep Sets

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

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

$\mathcal{L}(G_i)$ : set of ISs in  $G_i$

$$|\mathcal{L}(G)| = \frac{|\mathcal{L}(G_m)|}{|\mathcal{L}(G_{m-1})|} \times \frac{|\mathcal{L}(G_{m-1})|}{|\mathcal{L}(G_{m-2})|} \times \dots \times \frac{|\mathcal{L}(G_1)|}{|\mathcal{L}(G_0)|} \cdot |\mathcal{L}(G_0)|$$

to estimate  $|\mathcal{L}(G)|$  need good estimate for

$$r_i := \frac{|\mathcal{L}(G_i)|}{|\mathcal{L}(G_{i-1})|}$$

Alg for estimating  $r_i$ , given alg for generating random samples

$$\textcircled{1} \quad X := 0$$

$$\textcircled{2} \quad \text{repeat for } M = \left\lceil \frac{3}{(\frac{\epsilon}{2m})^2} \ln \left( \frac{2}{\delta/m} \right) \right\rceil \text{ trials}$$

- generate unit sample from  $\mathcal{L}(G_{i-1})$

- if sample is indep set in  $G_i$ ,  $X++$

$$\textcircled{3} \quad \text{return } \tilde{r}_i = \frac{X}{M}$$

$\tilde{r}_i$  is  $(\frac{\epsilon}{2m}, \frac{\delta}{m})$  approx for  $r_i$

Fact:  $r_i \geq \frac{1}{2}$

$\exists$  one edge  $(u, v)$  in  $G_i \setminus G_{i-1}$

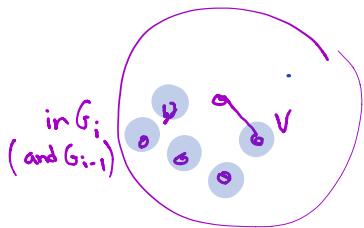
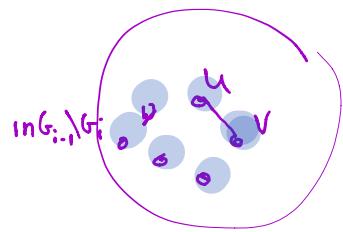
so any  $I$  in  $G_{i-1}$  that is not in  $G_i$ ;

contains both  $u$  &  $v$

$I \rightarrow I \setminus v$  in  $G_i$  &  $G_{i-1}$

$$|U(G_{i-1} \setminus G_i)| \leq |U(G_i)|$$

$$r_i = \frac{|U(G_i)|}{|U(G_{i-1})|} = \frac{|U(G_i)|}{|U(G_i)| + |U(G_{i-1} \setminus G_i)|} \geq \frac{1}{2}$$



Our estimate:  $2^n \prod_{i=1}^m \tilde{r}_i$

Truth:  $2^n \prod_{i=1}^m r_i$

Claim:  $\Pr\left(\left|\prod_{i=1}^m \tilde{r}_i - 1\right| \leq \varepsilon\right) \geq 1 - \delta$

$$\Pr\left(|\tilde{r}_i - r_i| \leq \frac{\varepsilon}{2m} r_i\right) \geq 1 - \frac{\delta}{m}$$

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

$$1 - \frac{\varepsilon}{2m} \leq \frac{\tilde{r}_i}{r_i} \leq \left(1 + \frac{\varepsilon}{2m}\right) \quad \text{II}$$

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

we only need to figure out how to generate uniform sample from

$$\mathcal{N}(G_{i-1}) / V_i$$

actually almost uniform sample suffices

How to sample elts from a universe  $\mathcal{N}$  according to some  
distr  $\Pi$ ?

Cool idea: design MC whose state space is  $\mathcal{N}$   
that has stationary distr  $\Pi$

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

2 key questions:

- ① how to design chain w/ right  $\pi$ ?
- ② how to bound mixing time?

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

states: indep sets

$X_t$  some indep set

choose vertex  $v$  u.a.r. from  $V$

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

if  $v \notin X_t$  & can be added w/o violating independence

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

otherwise  $X_{t+1} = X_t$

- irreducible
- if  $\exists$  edge, then aperiodic  $(u, v) \rightarrow P_{u,v} > 0$
- stationary distn uniform (chain doubly stochastic)  
 $P_{I,i}$  either  $\frac{1}{n}$  or 0

general technique, given  $\mathcal{L}$  and a connected graph on  $\mathcal{L}$   
 to define transition probs so that will have stationary  
 distn  $\pi$

### Metropolis Algorithm

Input:  $\mathcal{L}$ , connected  $G = (\mathcal{L}, E)$ ,  $\pi$  s.t.  $\sum_i \pi_i = 1$

Let  $\Delta = \max$  degree in graph

$$P_{xy} = \begin{cases} \frac{1}{2\Delta} \min\left(1, \frac{\pi_y}{\pi_x}\right) & x \neq y, y \in N(x) \\ 0 & x \neq y, y \notin N(x) \\ 1 - \sum_{y \neq x} P_{xy} & x = y \end{cases}$$

} nice fact: only depends on ratios

Claim

$$\pi_x P_{xy} = \pi_y P_{yx} \quad \forall x \neq y \quad \Rightarrow \pi \text{ is stationary distn}$$

assume wlog  $\pi_x \leq \pi_y$

$$\pi_x \left(\frac{1}{2\Delta}\right) = \pi_y \left(\frac{1}{2\Delta} \frac{\pi_x}{\pi_y}\right)$$

Example: suppose want to sample independent sets

according to  $\pi(I) = \frac{\gamma^{|I|}}{Z}$

$$Z = \sum_I \gamma^{|I|}$$

$$P_{I,I'} = \frac{1}{2n} \min(1, \gamma) \quad (\text{same graph as above})$$

important connections to hardcore lattice gas model in statistical physics

### Bounding mixing time

① spectral gap, conductance, expansion ...

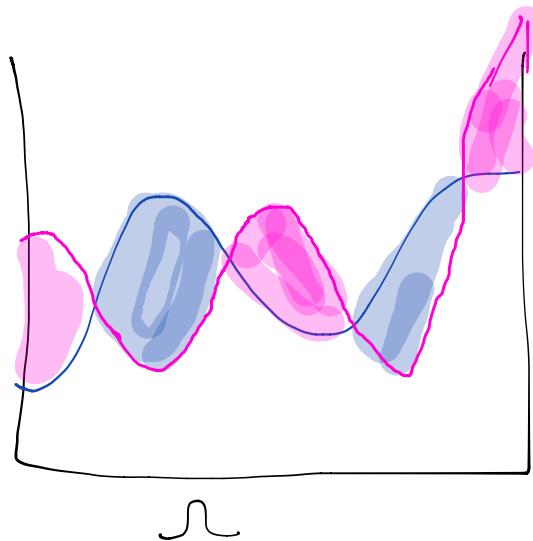
② coupling

## Coupling

Total variation distance between 2 dist's on same sample space  $\mathcal{S}$

$\text{space } \mathcal{S}$

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



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

Common defn of mixing time  $T(\varepsilon)$

$$T(\varepsilon) = \min \left\{ t \mid \|P^t, \pi\|_{TV} \leq \varepsilon \right\}$$

say MC is rapidly mixing if  $T(\varepsilon)$  poly in  $\log |\mathcal{S}|$  &  $\log(\frac{1}{\varepsilon})$

(we know this is related to spectral gap)

Coupling: simple & elegant approach to bounding mixing times

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

- ① each of  $X_t$  &  $Y_t$  in isolation is faithful copy of MC
- ② If  $X_t = Y_t$  then  $X_{t+1} = Y_{t+1}$

### Coupling Lemma

Let  $Z_t = (X_t, Y_t)$  be a coupling

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

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

Then  $T(\varepsilon) \leq T$ .

### Proof:

Consider coupling w/  $Y_0$  chosen according to  $\Pi, x_0$

arbitrary

$$\begin{aligned}\forall A \subseteq \mathcal{S} \quad \Pr(X_T \in A) &> \Pr(X_T = Y_T \cap Y_T \in A) \\ &= 1 - \Pr(X_T \neq Y_T \cup Y_T \notin A) \\ &\geq 1 - \Pr(Y_T \notin A) - \Pr(X_T \neq Y_T)\end{aligned}$$

$$\begin{aligned}
 &= \Pr(Y_T \in A) - \Pr(X_T \neq Y_T) \\
 &\quad \downarrow Y_0 \sim \pi \quad \downarrow \text{assumption} \\
 &= \pi_A - \varepsilon
 \end{aligned}$$

Similarly  $\Pr(X_T \notin A) > \pi_{A^c} - \varepsilon$

$$\Pr(X_T \in A) \leq \pi_A + \varepsilon$$

Examples:

① Random walk on hypercube  $N=2^n$  nodes

at each step choose random coordinate  $i$ , random bit  $b \in \{0,1\}$

change  $i^{\text{th}}$  bit to  $b$

Coupling: at each step choose same  $i$  &  $b$

one RW starts at  $\pi$

how many steps till walks meet?

(2) indep sets of fixed size  $k$

- here distances may  $\uparrow$ ...

Chain: choose vertex  $v \in X_+$  u.a.r. & a vertex  $w \in V$  u.a.r.

if  $w \notin X_+ \& X_+-v+w$  indep,  $X_{++} := X_+-v+w$

else  $X_{++} = X_+$

$n$  # vertices

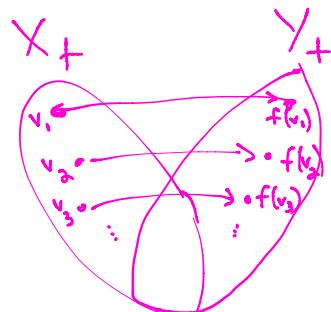
irreducible, aperiodic,  
 $\pi$  uniform

$\Delta$  max deg

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

coupling uses arbitrary bijection  $f$  between

vertices of  $X_+ - Y_+$  &  $Y_+ - X_+$



$X_+$

choose  $v \in X_+$  u.a.r.  
 $w \in V$  u.a.r.

$Y_+$  starts in  $\pi$

$v \in Y_+$  use  $(v, w)$

$v \notin Y_+$  use  $(f(v), w)$

both behave like original chain

$$d_+ = |X_+ - Y_+|$$

can change by at most 1 each step

chains meet when  $d_+ = 0$

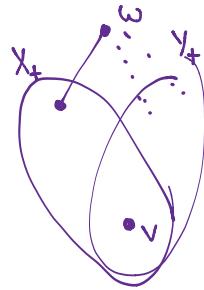
claim:  $d_+$  has negative drift so this doesn't take too long

$$d_{++1} = d_+ + 1$$

$$\Pr(d_{++1} = d_+ + 1 \mid d_+ > 0) \leq \left(\frac{k-d_+}{k}\right) \left(\frac{2d_+(\Delta+1)}{n}\right)$$

$v \in X_+ \cap Y_+$

$w \in \text{neighborhood of one or other}$   
or symm diff.

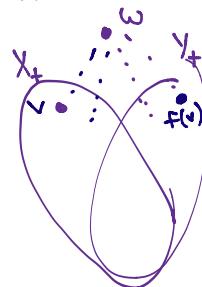


$$d_{++1} = d_+ - 1$$

$$\Pr(d_{++1} = d_+ - 1 \mid d_+ > 0) \geq \frac{d_+}{k} \cdot \frac{n - (k+d-2)(\Delta+1)}{n}$$

$v \in X_+ \setminus Y_+$

$w \notin X_+ \cup Y_+$



$$|X_+ \cup Y_+| = k+d$$

or neighborhood of either

$$E(d_{++1} \mid d_+) = \Pr(d_{++1} = d_+ + 1)(d_+ + 1) + \Pr(d_{++1} = d_+ - 1)(d_+ - 1)$$

$$\leq d_+ \underbrace{\left(1 - \frac{n - (3k-3)(\Delta+1)}{kn}\right)}_{\alpha < 1}$$

by Induction  $E(d_+) \leq d_0 \alpha^+ \rightarrow 0$

$$\Pr(d_+ \geq 1) \leq E(d_+) \rightarrow 0$$

