

## Counting Problems

Two big pieces:

1. Equivalence of counting and generating via **self reducibility**
2. Generating via Markov chains

## Volume

Outline:

- Describe problem. Membership oracle
- $\#P$  hard to volume intersection of half spaces in  $n$  dimensions
- In low dimensions, integral.
- even for convex bodies, can't do better than  $(n/\log n)^n$  ratio
- what about FPRAS?

Estimating  $\pi$ :

- pick random in unit square
- check if in circle
- gives ratio of square to circle
- Extends to arbitrary shape with “membership oracle”
- Problem: rare events.
- Circle has good easy outer box

Problem: rare events:

- In 2d, long skinny shapes
- In high  $d$ , even round shape has exponentially larger bounding box

Solution: “creep up” on volume

- modify  $P$  to contain unit sphere  $B_1$ , contained in larger  $B_2$  of radius  $r$  with  $r/r_1$  polynomial
- choose  $\rho = 1 - 1/n$ .
- Consider sequence of bodies  $\rho^i r P \cap B_2$
- note for large  $i$ , get  $P$

- but for  $i = 0$ , body contains  $B_2$
- so volume known
- so just need ratios
- At each step, need to random sample from  $\rho^i rP \cap B_2$

Sample method: random walk forbidden to leave

- MC irreducible since body connected
- ensure aperiodic by staying put with prob.  $1/2$
- markov chain is “regular graph” so uniform stationary distribution
- eigenvalues show rapid mixing: after  $t$  steps, r.p.d at most

$$\left(1 - \frac{1}{10^{17}n^{19}}\right)^t$$

- eigenvalues small because body convex: no bottlenecks.

Observations:

- Key idea of self reducibility: compare size of sequence of “related” shapes, then telescope ratios.
- Sizes compared by sampling
- Sample by markov chain
- wait: markov chain not exact?
- doesn’t matter: just get accurate to within  $(1 - 1/poly)$  in each step, product of errors still tiny.

## Application: Permanent

Counting perfect matchings

- Choose random  $n$ -edge set
- check if matching
- problem: rare event
- to solve, need sample space where matchings are dense

Idea: self reducibility by adding an edge (till reach complete graph)

- problem: don’t know how to generate random matching

Different idea: ratio of  $k$ -edge to  $k - 1$ -edge matchings

- telescope down to 1-edge matchings (self reduction)
- in **dense graphs** (degree  $n/2$ ), ratio is at most  $m^3$ .
- map each  $k$  edge matching by removing an edge:  $n^2$  to 1
- map each  $k - 1$  edge matching to  $k$ -edge matching by **augmenting path** of length at most 3.
  - take unmatched  $u$  and  $v$
  - if unmatched neighbor of  $u$  or  $v$ , done
  - by  $u$  and  $v$  have  $n/2$  neighbors, so if all matched, some neighbor  $b$  of  $u$  matched to some neighbor  $a$  of  $v$ .
  - so each size  $k$  matching “receives” at most  $m^3$  size  $k - 1$  matchings.

Generate via random walk

- based on using uniform generation to do sampling.
- applies to minimum degree  $n/2$
- Let  $M_k$  be  $k$ -edge matchings,  $\|M_k\| = m_k$
- algorithm estimates all ratios  $m_k/m_{k-1}$ , multiplies
- claim: ratio  $m_{k+1}/m_k$  polynomially bounded (dense).
- deduce sufficient to generate randomly from  $M_k \cup M_{k-1}$ , test frequency of  $m_k$
- do so by random walk of local moves:
  - with probability  $1/2$ . stay still
  - else Pick random edge  $e$
  - if in  $M_k$  and  $e$  matched, remove
  - if in  $M_{k-1}$  and  $e$  can be added, add.
  - if in  $M_k$ ,  $e = (u, v)$ ,  $u$  matched to  $w$  and  $v$  unmatched, then match  $u$  to  $w$ .
  - else do nothing
  - Note that exactly one applies
- Matrix is symmetric (undirected), so double stochastic, so stationary distribution is uniform as desired.
- In text, prove  $\lambda_2 = 1 - 1/n^{O(1)}$  on an  $n$  vertex graph (by proving expansion property)
- so within  $n^{O(1)}$  steps, rpd is polynomially small
- so can pretend stationary

Recently, extended to non-dense case.

## Coupling:

### Method

- Run two copies of Markov chain  $X_t, Y_t$
- Each considered in isolation is a copy of MC (that is, both have MC distribution)
- **but** they are not independent: they make dependent choices at each step
- in fact, after a while they are almost certainly the **same**
- Start  $Y_t$  in stationary distribution,  $X_t$  anywhere
- Coupling argument:

$$\begin{aligned}\Pr[X_t = j] &= \Pr[X_t = j \mid X_t = Y_t] \Pr[X_t = Y_t] + \Pr[X_t = j \mid X_t \neq Y_t] \Pr[X_t \neq Y_t] \\ &= \Pr[Y_t = j] \Pr[X_t = Y_t] + \epsilon \Pr[X_t = j \mid X_t \neq Y_t]\end{aligned}$$

So just need to make  $\epsilon$  (which is r.p.d.) small enough.

$n$ -bit Hypercube walk: at each step, flip random bit to random value

- At step  $t$ , pick a random bit  $b$ , random value  $v$
- both chains set bit  $b$  to value  $v$
- after  $O(n \log n)$  steps, probably all bits matched.

Counting  $k$  colorings when  $k > 2\Delta + 1$

- The reduction from (approximate) uniform generation
  - compute ratio of coloring of  $G$  to coloring of  $G - e$
  - Recurse counting  $G - e$  colorings
  - Base case  $k^n$  colorings of empty graph
- Bounding the ratio:
  - note  $G - e$  colorings outnumber  $G$  colorings
  - By how much? Let  $L$  colorings in difference ( $u$  and  $v$  same color)
  - to make an  $L$  coloring a  $G$  coloring, change  $u$  to one of  $k - \Delta = \Delta + 1$  legal colors
  - Each  $G$ -coloring arises at most one way from this
  - So each  $L$  coloring has at least  $\Delta + 1$  neighbors unique to them
  - So  $L$  is  $1/(\Delta + 1)$  fraction of  $G$ .
  - So can estimate ratio with few samples
- The chain:

- Pick random vertex, random color, try to recolor
  - loops, so aperiodic
  - Chain is time-reversible, so uniform distribution.
- Coupling:
    - choose random vertex  $v$  (same for both)
    - based on  $X_t$  and  $Y_t$ , choose bijection of colors
    - choose random color  $c$
    - apply  $c$  to  $v$  in  $X_t$  (if can),  $g(c)$  to  $v$  in  $Y_t$  (if can).
    - What bijection?
      - \* Let  $A$  be vertices that agree in color,  $D$  that disagree.
      - \* if  $v \in D$ , let  $g$  be identity
      - \* if  $v \in A$ , let  $N$  be neighbors of  $v$
      - \* let  $C_X$  be colors that  $N$  has in  $X$  but not  $Y$  ( $X$  can't use them at  $v$ )
      - \* let  $C_Y$  similar, wlog larger than  $C_X$
      - \*  $g$  should swap each  $C_X$  with some  $C_Y$ , leave other colors fixed. **Result:** if  $X$  doesn't change,  $Y$  doesn't
- Convergence:
    - Let  $d'(v)$  be number of neighbors of  $v$  in opposite set, so
 
$$\sum_{v \in A} d'(v) = \sum_{v \in D} d'(v) = m'$$
    - Let  $\delta = |D|$
    - Note at each step,  $\delta$  changes by  $0, \pm 1$
    - When does it increase?
      - \*  $v$  must be in  $A$ , but move to  $D$
      - \* happens if only one MC accepts new color
      - \* If  $c$  not in  $C_X$  or  $C_Y$ , then  $g(c) = c$  and both change
      - \* If  $c \in C_X$ , then  $g(c) \in C_Y$  so neither moves
      - \* So must have  $c \in C_Y$
      - \* But  $|C_Y| \leq d'(v)$ , so probability this happens is
 
$$\sum_{v \in A} \frac{1}{n} \cdot \frac{d'(v)}{k} = \frac{m'}{kn}$$
    - When does it decrease?
      - \* must have  $v \in D$ , only one moves

- \* sufficient that pick color not in either neighborhood of  $v$ ,
- \* total neighborhood size  $2\Delta$ , but that counts the  $d'(v)$  elements of  $A$  twice.
- \* so Prob.

$$\sum_{v \in D} \frac{1}{n} \cdot \frac{k - (2\Delta - d'(v))}{k} = \frac{k - 2\Delta}{kn} \delta + \frac{m'}{kn}$$

- Deduce that expected *change* in  $\delta$  is difference of above, namely

$$-\frac{k - 2\Delta}{kn} \delta = -a\delta.$$

- So after  $t$  steps,  $E[\delta_t] \leq (1 - a)^t \delta_0 \leq (1 - a)^t n$ .
- Thus, probability  $\delta > 0$  at most  $(1 - a)^t n$ .
- But now note  $a > 1/n^2$ , so  $n^2 \log n$  steps reduce to one over polynomial chance.

Note: couple depends on state, but who cares

- From worm's eye view, each chain is random walk
- so, all arguments hold

Counting vs. generating:

- we showed that by generating, can count
- by counting, can generate: