

Randomization + Derandomization?

Today

- randomized complexity classes
- derandomization via enumeration

$$\text{BPP} \subseteq \text{EXP}$$

- pairwise independence & derandomization

Max Cut Algorithm

defn. of p.i.

derandomizing max cut

Some Complexity Classes:

def. a language L is a subset of $\{0,1\}^*$

e.g. $\{x \mid x \text{ is a graph with a hamilton path}\}$

$\{x \mid x \text{ is a collection of sets that have a proper 2-coloring}\}$

def P is class of languages L
 with ptime deterministic algorithms of
 s.t. $x \in L \Rightarrow A(x) \text{ accepts}$
 $x \notin L \Rightarrow A(x) \text{ rejects}$

def RP is class of languages L
 with ptime probabilistic algorithm of
 s.t. $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] \geq \frac{1}{2}$
 $x \notin L \Rightarrow \Pr[A(x) \text{ accepts}] = 0$ } 1-sided error

def. BPP is class of languages L
 with ptime probabilistic algorithm of
 s.t. $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] \geq \frac{2}{3}$
 $x \notin L \Rightarrow \Pr[A(x) \text{ accepts}] \leq \frac{1}{3}$ } 2-sided error

Comments

- constants arbitrary -
with mult cost of $O(\log \frac{1}{\beta})$ can get error $\leq \beta$
- Clearly $P \subseteq RP \subseteq BPP$

Big Open Question :

is $P = BPP$?

do we need random coins for efficient algorithms?

Derandomization via enumeration

- Given probabilistic algorithm A & input x
 - Run A on every possible random string
of length $r(n)$
 - output majority answer
- at most time bound of A .
Is there a better bound?

Behavior

if $x \in L$, $\geq \frac{2}{3}$ of random strings cause d to accept \Rightarrow majority answer is ACCEPT
 if $x \notin L$ " " " " " " " " reject \Rightarrow " "
 " " " " " " " " REJECT

runtime

$$O(2^{r(n)} \cdot t(n)) \leq O(2^{t(n)} t(n))$$

under
time bound of ct

Corollary

$$\text{BPP} \subseteq \text{EXP}$$

$$\uparrow \text{EXP} \equiv \text{DTIME}(\bigcup_c 2^{n^c})$$

Comments: $r(n) \leq t(n)$ since can use at most 1 bit per step.

if can get better bound on $r(n)$, can improve runtime

e.g. if $r(n) = O(\log n)$,

runtime is $\text{poly}(n)$ for ptime A

- Given a problem with a randomized ptime algorithm, 1-sided error
Homework problem 3

$\Rightarrow \exists$ one random string that works for all

inputs of size n

i.e. \exists ckt (with no random bits) that work for all

inputs of size n .

- What about 2-sided error?

also true!

Pairwise independence & derandomization

- a simple randomized algorithm for MaxCut
- pairwise independent sample spaces
- derandomization

Max Cut:

given: $G = (V, E)$

output: partition V into S, T to } NP-hard

maximize $\sum_{(u,v) \in E} \{ u \in S, v \in T \}$

size of S, T cut

A randomized algorithm :

Flip n coins r_1, \dots, r_n

put vertex i on side r_i to get S, T ← i.e. add i to S

if $r_i = 0$
+ to T o.w.

Analysis:

let $1_{u,v} = 1$ if $r_u \neq r_v$ (i.e. placed on different sides so (u,v) crosses cut)
0 o.w.

so cut size
 $= \sum_{(u,v) \in E} 1_{u,v}$

$$E[\text{cut}] = E \left[\sum_{(u,v) \in E} 1_{u,v} \right]$$

$$= \sum_{(u,v) \in E} E[1_{u,v}] = \sum_{(u,v) \in E} \Pr[1_{u,v} = 1]$$

$$= \sum_{(u,v) \in E} \Pr[r_u = 1 \text{ or } r_v = 1]$$

$$= \sum_{(u,v) \in E} \left(\Pr[r_u = 1] + \Pr[r_v = 1] \right) = \frac{|E|}{2}$$

Pairwise independent random variables: definition

Pick n values x_1, \dots, x_n

each $x_i \in T$ (domain) st. $|T| = t$ (size of domain)
in some way

def. x_1, \dots, x_n independent if $\forall b_1, \dots, b_n \in T^n$

$$\Pr[x_1, \dots, x_n = b_1, \dots, b_n] = \frac{1}{t^n}$$

pairwise independent if $\forall i \neq j \quad b_i, b_j \in T^2$

$$\Pr[x_i, x_j = b_i, b_j] = \frac{1}{t^2}$$

k -wise independent if $\forall \overset{\text{distinct}}{i_1, \dots, i_k} \quad b_{i_1}, \dots, b_{i_k} \in T^k$

$$\Pr[x_{i_1}, \dots, x_{i_k} = b_{i_1}, \dots, b_{i_k}] = \frac{1}{t^k}$$

Main point:

Only use pairwise independence in max-cut algorithm
(i.e., algorithm analysis still works if random bits are
only pairwise indep.).

Derandomization of max-cut

Full enumeration:

try all 2^n possible coin tosses
pick best cut

n fully random bits \rightarrow Algorithm \rightarrow cut

gets very best cut, not just $\frac{|E|}{2}$

both work pretty well!

"Partial enumeration":

m pairwise indep random bits \rightarrow Algorithm \rightarrow cut

don't try all possible coin tosses

just a subset that satisfies pairwise independence

e.g. $r_1 \ r_2 \ r_3$

pick a row uniformly

	r_1	r_2	r_3
0	0	0	0
0	1	1	1
1	0	1	1
1	1	0	0

for $i \neq j$, $\forall b_1, b_2 \in \{0, 1\}^2$

$$\Pr[r_i = b_1 \wedge r_j = b_2] = \frac{1}{4}$$

good enough to give

$$E[\text{cut}] = \frac{|E|}{2}$$

for 3 node graphs,
only need to enumerate over 4 rows
instead of 8 rows.

Another picture

$b_1 \dots b_m$
totally independent
enumerate all 2^m choices

"randomness generator"
pick a random row

$r_1 \ r_2 \ \dots \ r_n$

pairwise independent + good enough
for our algorithm!

above example: $m=2, n=3$

CAN WE MAKE $n > m$?

enumerate all choices of $r_1 \dots r_n$

derandomize Max-Cut, given "randomness generator" taking $(\log n + 1) \Rightarrow n$ bits

• First; construct new randomized MC alg MC' :

- given $\log n$ truly random bits $b_1 \dots b_{\log n + 1}$
- use generator to construct n p.i. random bits
 $r_1 \dots r_n$
- Use r_i 's in MC alg & evaluate cutsize

• Then; derandomize via enumeration

Deterministic M-C alg:

For all choices of $b_1 \dots b_{\log n + 1}$

run MC' on $b_1 \dots b_{\log n + 1}$ & evaluate cutsize

pick best cutsize

Runtime: $\underbrace{(2^{\log n})}_{\# \text{choices of } b_i}$ \times (time for generator + time to run MC) = $\text{poly}(n)$

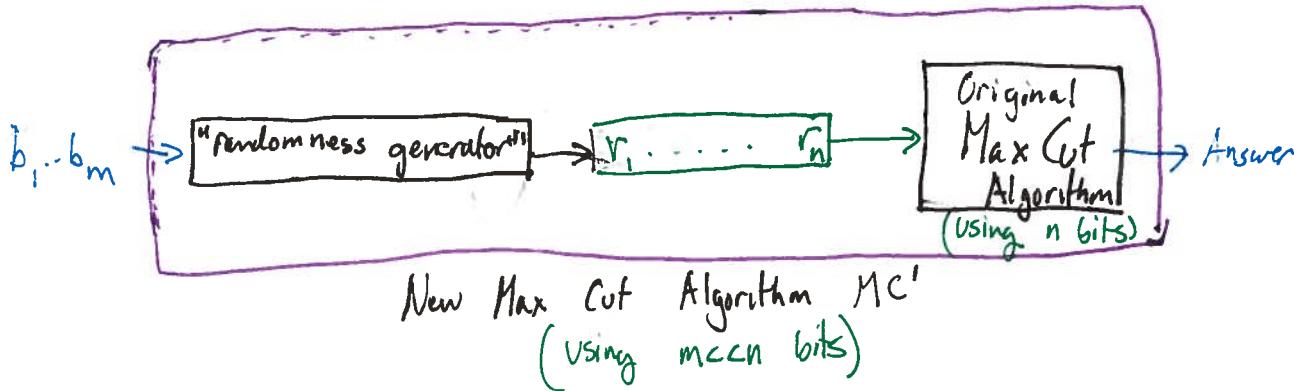
choices
of b_i 's

Comments

• no guarantee of getting OPT cut as in basic enumeration method

- generator determines a very small set of random strings, at least one of which gives a good cut

dr. 7a



do "full enumeration" derandomization
on this in $O(2^m) \times [$ time to generate
+ time to run MaxCut]

How to generate pairwise independent random variables?

dr.8

1) Bits

- choose k truly random bits $b_1 \dots b_k$

$$\forall S \subseteq [k] \text{ s.t. } S \neq \emptyset \text{ set } c_S = \bigoplus_{i \in S} b_i$$

- output all c_S

Generates $2^k - 1$ bits from k truly random bits
i.e. $m = \log n$

Generated bits are pairwise independent
proof exercise

2) Integers in $[0, \dots, q-1]$ (q prime)

trivial method that works for $q=2^l$ (note that q is not prime)

- repeat "bits" construction independently for each position in 1..l

uses $O(\log n \cdot \log q) = O(\log n)$ bits of true randomness

Somewhat better construction:

(when $n \approx q$ needs $O(\log q)$ bits of randomness)

- pick $a, b \in \mathbb{Z}_q$
- $r_i \leftarrow a \cdot i + b \pmod{q} \quad \forall i \in \{0..q\}$
- output $r_1 \dots r_q$

Useful to think of as $\xrightarrow{\text{input/output description of } a}$
fn from

$$h_{a,b} : [0..q] \rightarrow \mathbb{Z}_q$$

note: $|H| = q^2$

Family of fns $H = \{h_1, h_2, \dots\}$ for $h_i : [N] \rightarrow [M]$ is

"pairwise independent" if:

when $H \in_u H$

- (1) $\forall x \in [N], H(x) \in_u [M]$ ← any one location distributed uniformly
- (2) $\forall x_1 \neq x_2 \in [N], H(x_1) \perp H(x_2)$ independent ← any 2 are indep

equivalently: $\forall x_1 \neq x_2 \in [N]$

$$\forall y_1, y_2 \in [M]$$

$$\Pr_{H \in H} [H(x_1) = y_1 \wedge H(x_2) = y_2] = \frac{1}{M^2}$$

Notation:
 $x \in_u D$ means x
 chosen uniformly
 at random
 from D

Comments

- no single fn is p.i. - have to pick a random fn from a family
- given H & $x \in [N]$ $H(x)$ should be computable in time $\text{poly}(\log N, \log M)$ don't have to compute "all at once"
- also called "strongly 2-universal hash funcs"

Why is our example p.i.?

$$H = \{h_{a,b} \mid \mathbb{Z}_q \rightarrow \mathbb{Z}_q\} \quad (\text{recall } q \text{ is prime})$$

$$h_{a,b} = ax + b \bmod q$$

fix any $x \neq w, c, d$

$$\Pr_{a,b} [\begin{matrix} h_{a,b}(x) \\ ax + b \end{matrix} = \begin{matrix} c \\ d \end{matrix}] = \frac{1}{q^2}$$

$$\begin{pmatrix} x & 1 \\ w & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} c \\ d \end{pmatrix}$$

$w \neq x$ so $\begin{cases} \text{nonsingular} \end{cases} \Rightarrow \text{unique soln}$

how many truly random bits?

$2 \log q$ yields q p.i. random field elts.

More Comments

- can construct for all finite fields, even when domain + range have different sizes
- original motivation: hashing
hash funcs chosen from p.i. family instead of random funcs.

Why is this good?

how would you store a
random func on a domain
 $100000000\ 00000\ 0000\ 00\dots 00$?
of size 2?