

Communication complexity

Alice gets $x \in X$, Bob gets $y \in Y$
 want to compute $f(x, y)$ for
 some fixed $f: X \times Y \rightarrow Z$

Today: $X = Y = \{0, 1\}^n$ $Z = \{0, 1\}$

Only care about communication between Alice & Bob
 Computation is for free

Combinatorial rectangle $R = A \times B$; $A \subseteq X$, $B \subseteq Y$
 R monochromatic if $\exists z \forall (x, y) \in R \ f(x, y) = z$

Any deterministic protocol P for f induces
 a partition of $X \times Y$ into monochromatic rectangles

Namely, every input (x, y) ends up in exactly
 one leaf of the protocol tree

Randomized communication complexity

Alice and Bob can flip coins

- Joint random string r
- Must have $\Pr[P(x, y) = f(x, y)] \geq 1 - \epsilon$
 for all inputs; probability over random coins
- Cost of protocol: worst-case over inputs
 and random strings.

$R_\epsilon^{\text{pub}}(f) = R_\epsilon(f) =$ worst-case cost of best
 randomized protocol for computing f

How to prove lower bounds?

- No partition of inputs
- No rectangles in leaves
- No monochromaticity^(ti)

Randomized protocol becomes deterministic once random string is fixed. Hence

A public-coin randomized protocol P is a probability distribution ~~over~~ $\{P_r\}$ over deterministic protocols P_r

$$\Pr[P \text{ computes } f(x,y) \text{ correctly}] =$$

$$\Pr[\text{choosing deterministic protocol } P_r \text{ such that } P_r \text{ is correct on } (x,y)]$$

So maybe it can help to move a step back in the direction of deterministic communication. One more step.

Distributional communication complexity:

Protocols deterministic, inputs random

μ probability distribution on $X \times Y$

(μ, ϵ) - distributional communication complexity

$D_\epsilon^\mu(f)$ is cost of optimal protocol s.t.

$$\Pr_{(x,y) \sim \mu} [P(x,y) \neq f(x,y)] \leq \epsilon$$

Distributional communication complexity
exactly characterizes public-coin randomized
communication complexity

Numbering from
last lecture

THM 6 [△] [Yao '83]

$$R_{\epsilon}^{\text{pub}}(f) = \max_{\mu} D_{\epsilon}^{\mu}(f)$$

We will only need $\geq$ -direction (which we proved).

So now we're back (almost) to
deterministic setting!

- Protocol partitions input space $X \times Y$
- Partition into rectangles = for each leaf ℓ in protocol tree P , the inputs (x, y) that end up there
- No monochromaticity^{til}, but almost...
Most rectangles had better be
"almost monochrome" (where inputs
 (x, y) weighted according to μ), or
else the protocol is wrong on more
than ϵ -fraction of inputs.

Leads to notion of discrepancy

How large and almost monochrome
rectangles can f have (when we
view f as an $X \times Y$ -matrix M_f
with (x, y) -entry containing $f(x, y)$)?

DEF 7 $f: X \times Y \rightarrow \{0, 1\}$ function L17 IV
 R rectangle

μ probability distribution on $X \times Y$
The DISCREPANCY of f on R ~~under~~ ^{according to} μ is

$$\text{Disc}_\mu(R, f) = | \mu(f^{-1}(1) \cap R) - \mu(f^{-1}(0) \cap R) |$$

The DISCREPANCY OF f according to μ is

$$\text{Disc}_\mu(f) = \max_R (\text{Disc}_\mu(R, f))$$

where $\max$ taken over all rectangles $R \subseteq X \times Y$.

LEMMA 8

For every function $f: X \times Y \rightarrow \{0, 1\}$,
every probability distribution μ on $X \times Y$, and
every $\epsilon \in (0, 1/2)$ it holds that

$$D_{\frac{1}{2} - \epsilon}^\mu(f) \geq \log_2 \left(\frac{2\epsilon}{\text{Disc}_\mu(f)} \right)$$

So if we can prove strong upper bounds on
discrepancy, then we get strong lower bounds
on randomized public-coin communication cplx.

Proof Let D be a protocol of cost c that
computes f correctly with prob $\geq \frac{1}{2} + \epsilon$ (w.r.t. μ).
Then we can upper-bound the extra advantage
 ϵ over "random guessing" as follows.

Would like to do this for set disjointness $\text{DISJ}(x, y) = [x \cap y = \emptyset]$.
Unfortunately, this doesn't work. We'll prove a
lower bound for a somewhat related function instead.

$$2\varepsilon = \left(\frac{1}{2} + \varepsilon\right) - \left(\frac{1}{2} - \varepsilon\right)$$

$$\leq P_{\mu}[\mathcal{P}(x,y) = f(x,y)] - P_{\mu}[\mathcal{P}(x,y) \neq f(x,y)]$$

$$= \sum_{\text{leaf } \ell \text{ of protocol } \mathcal{P}} \left(P_{\mu}[(x,y) \in R_{\ell} \text{ and } \mathcal{P}(x,y) = f(x,y)] - P_{\mu}[(x,y) \in R_{\ell} \text{ and } \mathcal{P}(x,y) \neq f(x,y)] \right) \quad (*)$$

For each leaf ℓ the deterministic protocol $\mathcal{P}$ gives a specific output 0 or 1. Hence, we can upper bound

$$\begin{aligned} (*) &\leq \sum_{\ell} \left| P_{\mu}[(x,y) \in R_{\ell} \text{ and } f(x,y) = 1] - P_{\mu}[(x,y) \in R_{\ell} \text{ and } f(x,y) = 0] \right| \\ &= \sum_{\ell} \left| \mu(f^{-1}(1) \cap R_{\ell}) - \mu(f^{-1}(0) \cap R_{\ell}) \right| \quad (\dagger) \end{aligned}$$

By definition, each term in $(\dagger)$ is bounded from above by $\text{Disc}_{\mu}(f)$. sets in partition of $X \times Y$

At most 2^c leaves = terms. So:

$$(*) \leq 2^c \cdot \text{Disc}_{\mu}(f)$$

That is,

$$2^c \geq \frac{2\varepsilon}{\text{Disc}_{\mu}(f)}$$

or taking logarithms

$$c \geq \log_2 \left(\frac{2\varepsilon}{\text{Disc}_{\mu}(f)} \right)$$

Rectangle with high discrepancy



Inner product

[217 VI]

$$IP(x, y) = \sum_{i=1}^n x_i y_i \pmod{2}$$

THM 4 [Chor & Goldreich '85]

$$R_{\epsilon}^{pub}(IP) = \Omega(n)$$

Follows immediately if we can prove strong upper bound on discrepancy. Will use uniform distribution $\mathcal{U}$ over $X \times Y = \{0,1\}^n \times \{0,1\}^n$

LEMMA 10

$$Disc_{\mathcal{U}}(IP) = 2^{-\Omega(n)}$$

so we set $\mu = \mathcal{U}$

We will conclude our discussion of communication complexity by proving this lemma.

Let R' be any fixed rectangle

$$R' = \{ (x, y) \mid x \in X', y \in Y' \}$$

for some $X' \subseteq X, Y' \subseteq Y$

$$\text{Write } R'(x, y) = \begin{cases} 1 & \text{if } (x, y) \in R' \\ 0 & \text{otherwise} \end{cases}$$

$$X'(x) = \begin{cases} 1 & \text{if } x \in X' \\ 0 & \text{otherwise} \end{cases}$$

$$Y'(y) = \begin{cases} 1 & \text{if } y \in Y' \\ 0 & \text{otherwise} \end{cases}$$

Switch notation to

$$IP: \{0,1\}^n \times \{0,1\}^n \rightarrow \{\pm 1\}$$

$$IP(x,y) = (-1)^{\sum x_i y_i}$$

Then

$$d := \mathbb{E}_{x,y} (R', IP) =$$

$$= \left| \mathbb{P}_{x,y} [IP(x,y) = 1 \text{ and } R'(x,y) = 1] - \mathbb{P}_{x,y} [IP(x,y) = -1 \text{ and } R'(x,y) = 1] \right|$$

$$= \left| \mathbb{E}_{x,y} [IP(x,y) \cdot R'(x,y)] \right|$$

$$= \left| \mathbb{E}_{x,y} [(-1)^{x_1 y_1 + \dots + x_n y_n} X'(x) Y'(y)] \right|$$

$$\leq \mathbb{E}_x \left[\underbrace{|X'(x)|}_{\leq 1} \cdot \left| \mathbb{E}_y [(-1)^{x_1 y_1 + \dots + x_n y_n} Y'(y)] \right| \right]$$

$$\leq \mathbb{E}_x [Z(x)] \quad (*)$$

$$\text{for } Z(x) = \left| \mathbb{E}_y [(-1)^{\sum x_i y_i} Y'(y)] \right|$$

We know that

$$\mathbb{E}[Z^2] \geq (\mathbb{E}[Z])^2$$

$$\left(\begin{array}{l} \text{E.g. observe that } 0 \leq \mathbb{E}[(Z - \mathbb{E}[Z])^2] \\ = \mathbb{E}[Z^2] - (\mathbb{E}[Z])^2 \end{array} \right)$$

Combining this with (*) we get

$$d^2 \leq (\mathbb{E}_x [Z(x)])^2 \\ \leq \mathbb{E}_x [Z^2(x)]$$

$$= \mathbb{E}_x \left[\mathbb{E}_u [(-1)^{\sum x_i u_i} Y'(u)] \cdot \mathbb{E}_v [(-1)^{\sum x_i v_i} Y'(v)] \right] \\ = \mathbb{E}_x \left[\mathbb{E}_{u,v} \left[(-1)^{\sum (u_i + v_i) x_i} \underbrace{Y'(u) Y'(v)}_{\leq 1} \right] \right]$$

~~switch order of summation~~

$$\leq \mathbb{E}_{u,v} \left[\mathbb{E}_x \left[(-1)^{(u_1 + v_1)x_1 + \dots + (u_n + v_n)x_n} \right] \right] (**)$$

This is starting to look vaguely familiar...

If $u \neq v$ then

$$\mathbb{E}_x \left[(-1)^{\sum (u_i + v_i) x_i} \right] \\ = \frac{1}{2^n} \sum_{x \in \{0,1\}^n} (-1)^{\sum (u_i + v_i) x_i} = 0$$

This is just problem 3 on pset 3 the hint in with slightly different notation.

If $u = v$ then

$$\mathbb{E}_x \left[(-1)^{\sum (u_i + v_i) x_i} \right] = \mathbb{E}_x \left[(-1)^{\sum 0 \cdot x_i} \right] = 1$$

But this happens only with probability 2^{-n} .

Hence, from $(**)$ we get

$$d^2 = \left[\text{Disc}_n(R', IP) \right]^2 \leq 2^{-n}$$

and for any R' and hence

$$\text{Disc}_n(IP) \leq 2^{-n/2}$$

which proves the lemma.

Unfortunately, discrepancy is provably too weak a tool to prove lower bounds for DISJ

Lower bound $\boxed{R_\epsilon^{\text{pub}}(\text{DISJ}) = \Omega(n)}$

can be proven by other means, but this is beyond the scope of this course.

Now we have

$$R_\epsilon^{\text{pub}}(IP) \geq D_\epsilon^n(IP)$$

$$\geq \log \left(\frac{1 - 2\epsilon}{\text{Disc}_n(IP)} \right)$$

$$\geq \log \left(\frac{1 - 2\epsilon}{2^{-n/2}} \right)$$

$$= \Omega(n).$$

Computational complexity theory:

L17 X

How hard are different problems

(1) to solve

(2) in a computational model

(3) wrt some specified resources

(1) SOLVE

We looked mostly not at computing general functions but at deciding problems / languages
= computing Boolean functions.

No big deal.

(a) Exact, ^{correct} solution

(b) Correct solution with high probability

(c) Approximately correct solution

(d) Combination of (b) & (c)

(2) MODEL

(a) Turing machines

- (i) vanilla
- (ii) + randomness
- (iii) + nondeterminism
- (iv) + oracles
- (v) + advice

(b) Circuits

(c) Interactive proofs

No trust between computationally bounded verifier and all-powerful prover

(d) Proof verification

- (i) randomized $\leadsto$ PCP
- (ii) deterministic $\leadsto$ proof cplx

(3) RESOURCES

L17 X

- (a) time
- (b) space (memory)
- (c) randomness
- (d) depth (of circuits)
- (e) queries to input
- (f) communication
 - (i) total # bits
 - (ii) # rounds
- (g) secrecy / security

WHAT WE DIDN'T COVER?

Most of computational complexity theory...
(Well, not really.)

Notable omissions (incomplete list)

- Quantum computation
- Algebraic computation models (circuits & others)
- Complexity of counting (how many satisfying assignments does F have)
- Average-case complexity (not worst-case)
- Derandomization
- Pseudorandomness
- Learning theory
- Kolmogorov complexity
- Limitations to lower bound techniques (Natural proofs Razborov Rudich)
- Streaming algorithms
- "Big data" computation (map reduce and similar)
- And the list just goes on...

But I hope you liked our selection of topics ... :-)

PLEASE LET US KNOW WHAT YOU THINK by filling in the web-based COURSE EVALUATION! (Should be out next week.)

AND GOOD LUCK with problem sets 3 and 4!