

LECTURE 12TODAYSET DISJOINTNESS

- ① INFORMATION COMPLEXITY.
- ② DISJOINTNESS
- ③ ~~ANSWER~~ → HELLINGER DISTANCE

See also
notes for
L10, L11

DISJOINTNESSAlice $x \in \{0,1\}^n$ Bob $y \in \{0,1\}^n$

$$\text{DISJ}^n(x, y) = 1 \text{ iff } \exists i \text{ s.t. } x_i = y_i = 1.$$

Challenge

- Hardness needs distributions where $x \perp y$.
- Exercise: Prove that if $\mu = \mu_x \times \mu_y$
then \exists protocol with expected comm. $\tilde{O}(\sqrt{n})$.
& error $\leq \epsilon$.

- Today: "Information Complexity" Approach

Warning: Distribution ✓

But not distributional lower bounds.

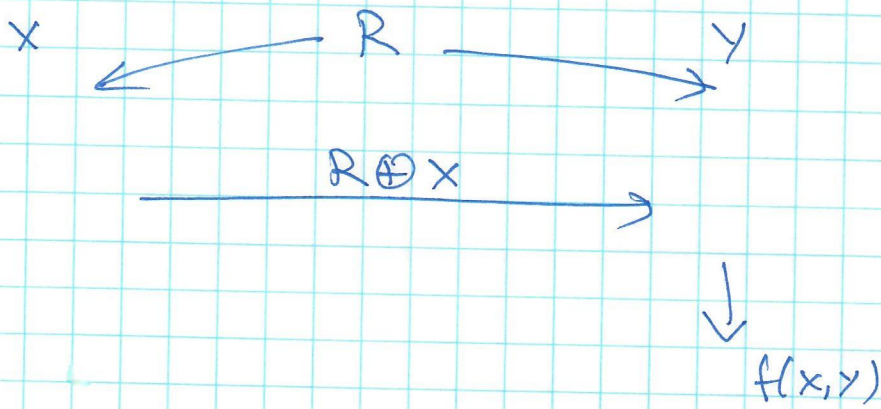
Given problem $(f = \text{Disj}^n)$, consider randomized ϵ -error protocol Π ; & distribution μ on inputs (x, y)

Information Complexity of Π w.r.t μ is

$$IC_{\mu}(\Pi) \triangleq I((X, Y); \Pi) \quad \text{when } (X, Y) \sim \mu$$

if $\# \text{bits}(\Pi) \leq k \Rightarrow IC_{\mu}(\Pi) \leq k \Rightarrow$ lower bounds on IC are what we seek

Warnings: Randomness:



IC = ? should

$$I((X, Y); R \oplus x) = 0!$$

Lesson: Should condition on R

$$IC_{\mu}(\Pi) = I((X, Y); \Pi | R) ?$$

↑
shared randomness.

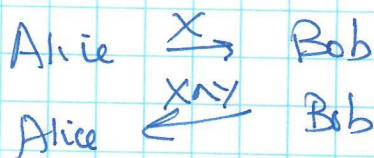
Private Randomness?

Should be usable to reduce information content!

Example: $IC_{unif}(AND)$

$$\left[IC_{unif}(f) \cong \min_{\Pi: \Pi \text{ C-computable}} \{ IC_{unif}(\Pi) \} \right]$$

Naive Protocol:



$$IC(\text{Naive}) = \frac{1}{2} \cdot 1 + \frac{1}{2} \cdot 2 = \frac{3}{2}$$

Sophisticated Protocol: "continuous time version"

- if $X=0$ Alice sends X to Bob at random time $t \in [0, 1]$
(unless Bob speaks before).
- if $X=1$ Alice sends X to Bob at time $t=1$
(unless Bob speaks before).

Similar for Bob.

$$IC(\text{Soph}) = \frac{1}{4} \cdot 2 + \frac{3}{4} \cdot 1 = \frac{5}{4}$$

④

Want to Prove $IC(DIST^n) = \Omega(n)$

On what μ ?

μ is a dist on $(X, Y, Z) : X, Y, Z \in \{0, 1\}^n$

- sampled as follows.

- $Z \sim \text{Unif}(\{0, 1\}^n)$

- for $i = 1$ to n

if $Z_i = 0$ then $X_i = 0$ $Y_i \sim \text{Unif}(\{0, 1\})$

if $Z_i = 1$ then $X_i \sim \text{Unif}(\{0, 1\})$, $Y_i = 0$

- $D_{\mu, 0}(DIST^n) = 0$! [answer always 0].

- But perfect for us ... will see.

Want to Bound $V_{\mu}(DIST^n)$.

But will lower bound $I((XY); \Pi | Z, R) \leftarrow CIC_{\mu}(DIST^n)$
(for protocol Π with public randomness R).

Claims: ① $CIC(DIST^n) \geq n \cdot CIC(DIST^1)$ [TODAY]

② $CIC(DIST^1) = \Omega(1)$ [Not Trivial !!]

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$$CIC(DISS^n) \triangleq I((X, Y); \Pi | Z, R)$$

$$(1.1) \quad I((X, Y); \Pi | Z, R) \geq \sum_{i=1}^n I((X_i, Y_i); \Pi | Z, R)$$

[Simple Inf. Th.]

$$(1.2) \quad I((X_i, Y_i); \Pi | Z, R) \geq CIC(DISS^1)$$

$$= I(\underbrace{(X', Y')}_{\substack{\text{1 bit } X', Y', Z'}}; \Pi' | Z', R')$$

$$(1.1) \quad I((X, Y); \Pi | Z, R) = H((X, Y) | Z, R) - H((X, Y) | \Pi, Z, R)$$

$$- H((X, Y) | Z, R) = \sum_{i=1}^n H((X_i, Y_i) | Z, R) = \sum_{i=1}^n H((X_i, Y_i) | Z_i)$$

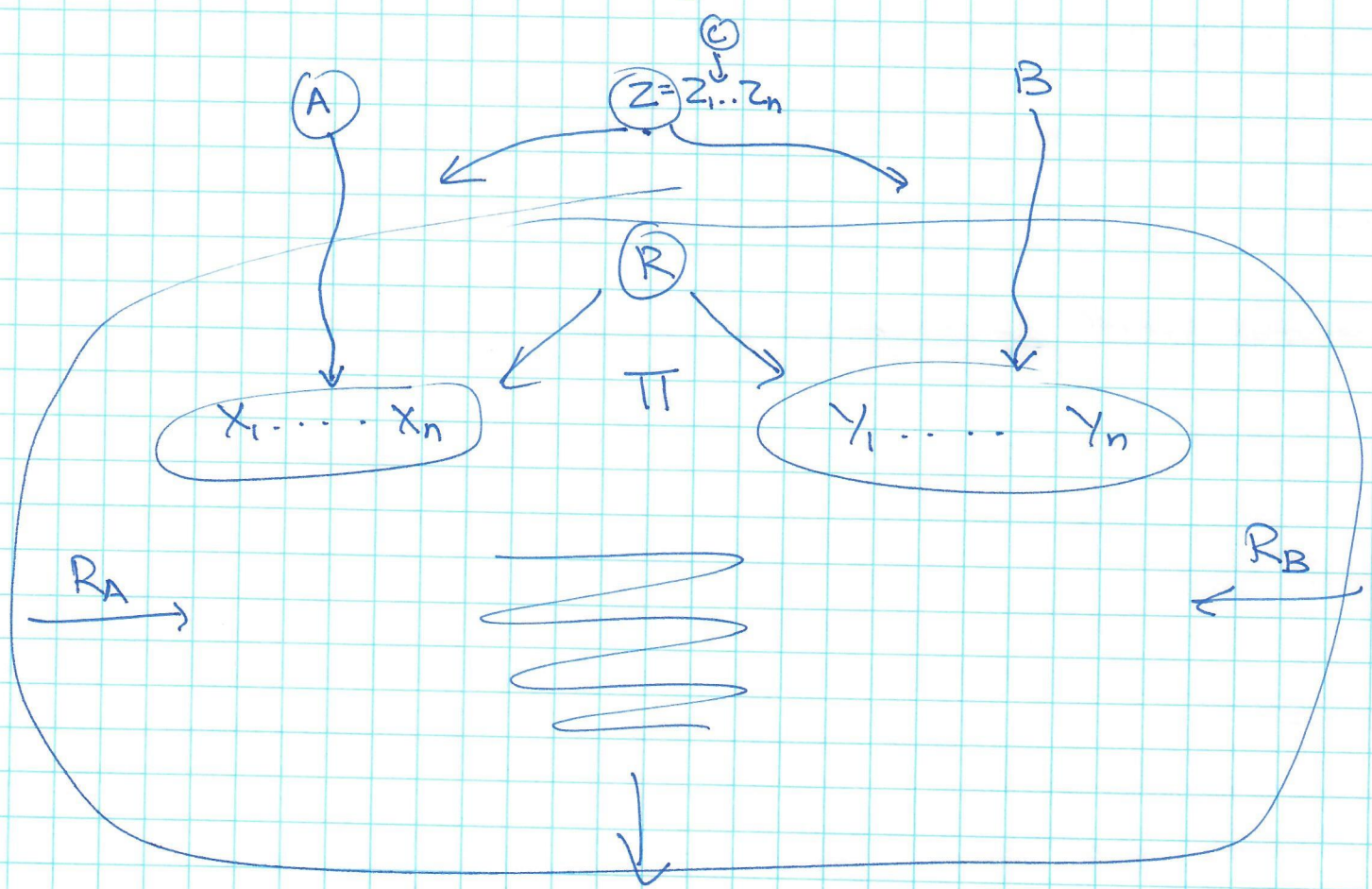
$$- H((X, Y) | \Pi, Z, R) = \sum_{i=1}^n H((X_i, Y_i) | \Pi, Z, R, (X, Y)_{\setminus i})$$

$$\leq \sum_{i=1}^n H((X_i, Y_i) | (\Pi, Z, R))$$



1.2 Non-Trivial

- Need to take protocol Π that reveals little about (x_i, y_i) while computing DISJ
- Convert to protocol Π' that reveals little about A, B while computing $A \wedge B$.



$$\text{DISJ}(x_1 \dots x_n, y_1 \dots y_n) = \text{AND}(A, B)?$$

get lower bound on $I((x_i, y_i) | \Pi, R, z) \leftarrow$
 need lower bound on $I(A, B | C, \Pi, \tilde{R}) \leftarrow$ are these same?