

Outline

- set disjointness problem
- information complexity of a protocol
- $IC \leq CC$
- choice of distribution + conditional information complexity
- $CIC \leq IC$
- $CIC^n \geq n \cdot CIC'$
- $CIC' = \Omega(1)$

Set Disjointness Problem

Alice, Bob respectively have $X, Y \in \{0, 1\}^n$, want to compute $DISJ(X, Y)$

$$DISJ(X, Y) = \bigwedge_{i=1}^n (X_i \wedge Y_i)$$

$$= \begin{cases} 1 & \text{if } \exists i \text{ st. } X_i = Y_i = 1 \\ 0 & \text{o.w.} \end{cases}$$

If X is characteristic vector of set S , Y is characteristic vector of T , $DISJ(X, Y)$ computes if the intersection of $S \cap T$ is nonempty, i.e. whether the sets are disjoint or not.

Goal: lower bound the communication complexity (CC) for a randomized protocol Π for computing $DISJ(X, Y)$ using information theory.

Information ComplexityDefinitions/Preliminaries

- $f: \{0, 1\}^n \times \{0, 1\}^n \rightarrow \{0, 1\}$
- μ is a distribution on $\{0, 1\}^n \times \{0, 1\}^n$
- ϵ is an error parameter
- Π is a correct protocol for f ($\Pi(X, Y)$ outputs $f(X, Y)$ $\forall X, Y$)
- $\Pi(X, Y, R, R_A, R_B)$ is the transcript of protocol Π with inputs X, Y , private randomness R_A, R_B , and public randomness R . Notated as Π .

Def: the information cost ($IC_{\mu, \epsilon}^{\Pi}(f)$) of Π wrt μ is $I(X, Y; \Pi(R))$; intuitively this is the amount of information learned about the inputs X, Y from the transcript

Def: the ϵ -error information cost ($IC_{\mu, \epsilon}(f)$) of f wrt μ is the minimum information cost of an ϵ -error protocol Π for f wrt μ :

$$IC_{\mu, \epsilon}(f) \triangleq \min_{\Pi} \{ IC_{\mu, \epsilon}^{\Pi}(f) \}$$

Ex. protocol with nontrivial information complexity: 1-bit AND

setup: $X, Y \in \{0, 1\}$
 $f(x, y) = x \wedge y$

protocol: parameters $k \gg \frac{1}{\delta}$, $\delta \rightarrow 0$

repeat k times:

- $A \rightarrow B$ send 1 w.p. $1 - \delta$
 x w.p. δ
- $B \rightarrow A$ send 1 w.p. $1 - \delta$
 y w.p. δ

 if either bit 0, stop + output 0
else output 1

Correctness: protocol incorrect w.p. $\leq e^{-\delta k}$

information complexity < 2
since protocol ends w/ either X or Y unknown

Lemma: for any μ and $\epsilon > 0$, $CC_\epsilon(f) \geq IC_{\mu, \epsilon}(f)$

Pf: Let π^* be a protocol that solves f with error probability $\leq \epsilon$
Recall $CC_\epsilon(\pi) = |\pi|$ is the max length of $\pi(X, Y, R, R_A, R_B)$ over all inputs X, Y .
 $CC_\epsilon(f) = |\pi^*| \geq H(\pi^*) \geq I(X, Y; \pi^* | R) = IC_{\mu, \epsilon}(f)$ \square

Goal: lower bound $CC(DISJ)$ via the information complexity of $DISJ$.

Thm: $IC_{\mu, \epsilon}(DISJ) = \Omega(n)$

Proof strategy: ① $IC(DISJ^n) \geq n \cdot IC(DISJ)$ ("Direct Sum")
② $IC(DISJ) = \Omega(1)$

The Choice of Distribution and Conditional Information Complexity

Want to show $IC^n \geq n \cdot IC$ but hindered by fact that distribution μ of inputs (X, Y) is not a product distribution. To get around this, use auxiliary random variable $Z \sim \text{Unif}(\{0, 1\}^n)$. Then construct a new distribution η^* in following way:

- If $z_i = 0$, $X_i = 0$ and $Y_i \sim \text{Unif}(\{0, 1\})$
 - If $z_i = 1$, $Y_i = 0$ and $X_i \sim \text{Unif}(\{0, 1\})$
- η^n is the joint distribution (X, Y, Z) .

This new distribution has the following properties:

- $\forall X, Y \in \text{support of } \eta^n, DISJ(X, Y) = 0$
- $\forall Z \in \{0, 1\}^n, X \perp Y | Z$

Def: the conditional information cost ($CIC_{\eta, \epsilon}^\pi(f)$) of π wrt η^n is $I(X, Y; \pi | Z, R)$

Def: the conditional information complexity ($CIC_{\eta, \epsilon}(f)$) of f wrt η^n is the minimal conditional information cost of a protocol that solves f :

$$CIC_{\eta, \epsilon}(f) = \min_{\pi} \{CIC_{\eta, \epsilon}^\pi(f)\}$$

Lemma: $IC_{\mu, \epsilon}(f) \geq CIC_{\eta, \epsilon}(f)$ where $\eta^n | Z = \mu$

Pf: Note that a protocol π for f is independent of Z given X, Y . Then,
 $IC_{\mu, \epsilon}(f) = I(X, Y; \pi) \geq I(X, Y; \pi | Z) = CIC_{\eta, \epsilon}(f)$ (dropping conditioning on R)

$$\begin{aligned}
 & H(\pi) - H(\pi | X, Y) & H(\pi | Z) - H(\pi | X, Y, Z) \\
 & H(\pi | X, Y) = H(\pi | X, Y, Z) \\
 & H(\pi) \geq H(\pi | Z)
 \end{aligned}$$

\square

Direct Sum for CIC: Proof of (1)

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Want to show $CIC^n \geq n \cdot CIC'$. Divide into two parts:

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

$$(B) \quad I(X_i, Y_i; \Pi | Z) \geq CIC'_{\xi, \epsilon}(\text{DISJ}') \quad \text{where } \eta = \xi^n$$

Pf of A : $I(X, Y; \Pi | Z) = H(X, Y | Z) - H(X, Y | \Pi, Z)$

$$H(X, Y | Z) = \sum_{i=1}^n H(X_i, Y_i | Z, X_1, Y_1, \dots, X_{i-1}, Y_{i-1}) \quad (\text{chain rule})$$

$$= \sum_{i=1}^n H(X_i, Y_i | Z) \quad (X_i, Y_i \perp \{X_j, Y_j\}_{j \neq i})$$

$$H(X, Y | \Pi, Z) = \sum_{i=1}^n H(X_i, Y_i | Z, \Pi, X_1, Y_1, \dots, X_{i-1}, Y_{i-1}) \leq \sum_{i=1}^n H(X_i, Y_i | Z, \Pi) \quad (\text{conditionally reduces entropy}) \quad \square$$

Next time: Proof of (B) and (2) to complete proof of the theorem.