CMPSCI 711: More Advanced Algorithms Lowerbounds 2: Information Statistics

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Information Statistics Approach

- Information statistics approach is based on analyzing the "information revealed" about the input from the messages.
- Useful for proving bounds on complicated functions in terms of simpler problems, e.g., proving a bound on

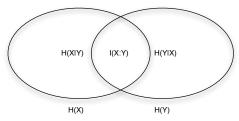
$$DISJ_t(M) = \bigvee_{j \in [n]} AND_t(M_{1,j}, \ldots, M_{t,j})$$

by first establishing a bound on AND_t .

▶ We'll first give some definitions and then run through an example.

Information Theory Definitions

- Let X and Y be random variables.
- Entropy: $H(X) := \sum_{i} -\mathbb{P}[X = i] \lg \mathbb{P}[X = i]$
- Conditional Entropy: $H(X|Y) := \mathbb{E}_{y \sim Y}[H(X|Y = y)] \leq H(X)$
- Mutual Information: I(X : Y) = H(X) H(X|Y)



Useful Facts:

- If X takes at most 2^{ℓ} values, then $H(X) \leq \ell$.
- Chain rule: H(XY) = H(X) + H(Y|X).
- Subadditivity: $H(XY) \le H(X) + H(Y)$; equality if independent.

Mutual Information

Lemma If X and Y are independent, then $I(XY : Z) \ge I(X : Z) + I(Y : Z)$. Proof.

$$I(XY : Z) = H(XY) - H(XY|Z) = H(X) + H(Y) - H(XY|Z) \geq H(X) + H(Y) - H(X|Z) - H(Y|Z) = I(X : Z) + I(Y : Z)$$

Information Cost

- Suppose you have a protocol Π for a two-party communication problem P in which Alice and Bob have random inputs X and Y.
- ▶ Let *M* be the (random) message sent by Alice and define:

 $cost(\Pi) = max |M|$

and

$$\operatorname{icost}(\Pi) = I(M : X)$$

Note $\operatorname{icost}(\Pi) = I(M : X) \le H(M) \le \operatorname{cost}(\Pi)$.

Example: Indexing

- ▶ We'll prove a lower bound on the information cost of INDEX where $X \in_R \{0,1\}^n$ in terms a simpler problem "ECHO"
- ECHO: Alice has a single bit B ∈_R {0,1} and Bob wants to output B with probability at least 1 − δ.
- ► A protocol Π_{INDEX} for INDEX yields a protocol $\Pi_{\text{ECHO},i}$ for ECHO:
 - 1. Given *B*, Alice picks $X_j \in_R \{0,1\}$ for $j \neq i$ and generates:

$$X = (X_1, X_2, \ldots, X_{i-1}, B, X_{i+1}, \ldots, X_n)$$

- 2. She sends the message M she'd have sent in Π_{INDEX} if she'd had X.
- 3. Bob receives message and outputs the value he'd have returned in Π_{INDEX} had his input been *i*.

Relating Information Cost of INDEX and ECHO

Since X_1, X_2, \ldots, X_n are independent:

$$\begin{aligned} \operatorname{cost}(\Pi_{\mathrm{INDEX}}) &\geq \operatorname{icost}(\Pi_{\mathrm{INDEX}}) \\ &= I(X_1 X_2 \dots X_n : M) \\ &\geq I(X_1 : M) + I(X_2 : M) + \dots + I(X_n : M) \\ &= \operatorname{icost}(\Pi_{\mathrm{ECHO},1}) + \operatorname{icost}(\Pi_{\mathrm{ECHO},2}) + \dots + \operatorname{icost}(\Pi_{\mathrm{ECHO},n}) \end{aligned}$$

• Lemma: Any protocol solving ECHO with probability $\geq 1 - \delta$, needs

 $\mathrm{icost}(\Pi_{\mathrm{ECHO},i}) \geq 1 - H_2(\delta)$

where $H_2(p) = -p \lg p - (1-p) \lg (1-p)$. • Hence, $cost(\Pi_{INDEX}) \ge (1 - H_2(\delta))n$.

Proof of Lemma

1. Fano's inequality: Let A and B be random variables. If you can guess B correctly with probability at least $1 - \delta$ given A, then

 $H(B|A) \leq H_2(\delta)$.

Let A = M be message and B be the bit needing echoed.
Hence,

$$\mathrm{icost}(\Pi_{\mathrm{ECHO}}) = H(B) - H(B|M) \geq 1 - H_2(\delta)$$

Outline for $DISJ_t$ Lower Bound

• Express DISJ_t in terms of AND_t where AND_t $(x_1, \ldots, x_t) = \prod_i x_i$:

$$\operatorname{DISJ}_t(M) = \bigvee_{j \in [n]} \operatorname{AND}_t(M_{1,j}, \ldots, M_{t,j})$$

- Consider a random input *M* to DISJ_t where M_{Djj} ∈_R {0,1} for D_j ∈_R [t]. All other entries are 0.
- Let T = (T₁,..., T_{t-1}) be the messages sent in a t-party protocol and define the information cost of a protocol as:

$$\operatorname{icost}(\Pi|D) = I(T:M|D) \quad ext{ where } \quad D = (D_1,\ldots,D_t) \;.$$

► A protocol for $DISJ_t$ yields *n* different protocols $\Pi_{AND_t,i}$ for AND_t :

$$\mathrm{icost}(\Pi_{\mathrm{DISJ}_t}|D) \geq \sum_{i \in [n]} \mathrm{icost}(\Pi_{\mathrm{AND}_t,i}|D) \;.$$

• Result follows by showing $icost(\Pi_{AND_t,i}|D) = \Omega(1/t)$.