

• CLASSICAL THEORY OF COMMUNICATION

→ Shannon 1948

• Classic Architecture for Reliable Communication



→ Prob. Method

→ Entropy

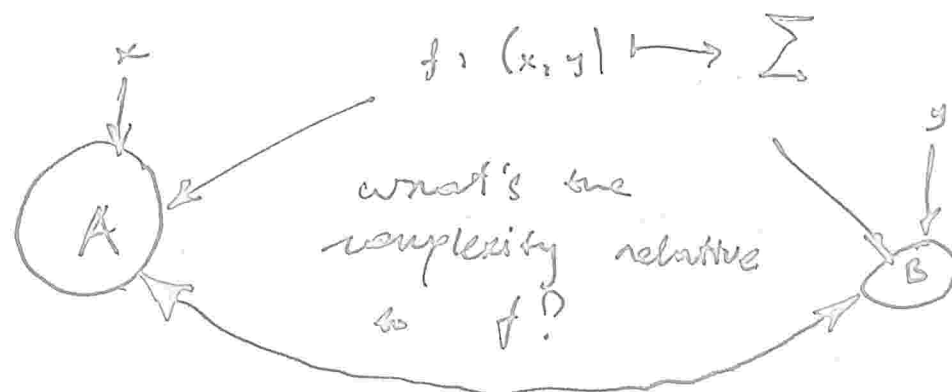
→ Mutual Information

→ Compression Schemes

→ Error-Correcting Codes

BUT DOES BOB REALLY NEED x ?

• [Yao 80] Communication Complexity with Shared Randomness



$CC(f) = \# \text{ bits exchanged by the best protocol}$

• How many bits are necessary to compute $f(x, y)$?

→ LOWER BOUNDS FOR BOOLEAN FUNCTIONS, CIRCUIT COMPLEXITY, STREAMING, ETC.

→ COMMUNICATION COMPLEXITY AS A MODEL FOR COMMUNICATION. SHANNON VS YAO?

STUDENTS DO NOT COME BLANK.
EX How much should be said in the lecture
for max. benefit?

COMM. COMPLEXITY IS ALSO MOTIVATED BY

HUMAN/HUMAN/COMPUTER/COMPUTER COMMUNICATION
WITH LARGE CONTEXT AND BRIEF COMMUNICATION.
ALSO BY THE SITUATIONS WHERE
CONTEXTS ARE IMPROPERLY SHARED.

COM 4:

EXAMPLE PROBLEMS WITH LOW COMMUNICATION
COMPLEXITY.

EASY CC PROBLEMS:

• EQUALITY TESTING

4

$$EQ(x, y) = 1 \Leftrightarrow x = y; \quad CC(EQ) = O(1) !$$

Protocol:

Fix error correcting code.

$$E: \{0, 1\}^n \rightarrow \{0, 1\}^N;$$

SHARED RANDOMNESS: $i \leftarrow [N];$

EXCHANGE $E(x)_i, E(y)_i;$

ACCEPT $1 \Leftrightarrow E(x)_i = E(y)_i.$

WHAT ARE OTHER CONSTANT-TIME EXAMPLES?

• HAMMING DISTANCE

$$H_k(x, y) = 1 \iff \Delta(x, y) \leq k;$$

$$cc(H_k) = O(k \log k) \text{ [Kwong et al]}$$

• SMALL SET INTERSECTION:

$$\cdot \Pi_k(x, y) = 1 \iff w_L(x), w_L(y) \leq k$$

$$\& \exists i: x_i = y_i = 1.$$

$$\cdot cc(\Pi_k) = O(k) \text{ [Håstad and Wigderson]}$$

poly(k) PROTOCOL:

USE COMMON RANDOMNESS
TO HASH $[n] \rightarrow [k^O]$.

• GMP (WEEK) INNER PRODUCT

$$x, y \in \mathbb{R}^n; \|x\|_2, \|y\|_2 = 1.$$

$$GIP(x, y) = 1 \text{ if } (x, y) \geq \epsilon; = 0 \text{ if } (x, y) \leq 0.$$

$$\cdot cc(GIP) = O\left(\frac{1}{\epsilon^2}\right); \text{ [Alon, Matias, Szegedy]}$$

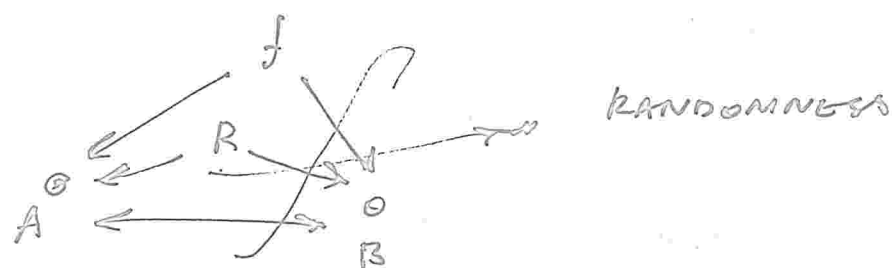
$$\cdot \text{IF } G \leftarrow N(0, 1)^n \text{ then } \mathbb{E} \left[\underbrace{\langle G, x \rangle \cdot \langle G, y \rangle}_{\text{UNBIASED ESTIMATOR}} \right] = \langle x, y \rangle$$

UNBIASED
ESTIMATOR

Summary from [Ghosh, Khramata 16]

UNCERTAINTY IN COMMUNICATION

Are there communication mechanisms that can overcome uncertainty?



MODEL: IMPERFECTLY SHAKEN RANDOMNESS

• ALICE $\leftarrow R$, BOB $\leftarrow S$ where

$(r, s) =$ INDEPENDENT SEQUENCE OF
CORRELATED PAIRS $(r_i, s_i)_i$;

$r_i, s_i \in \{-1, +1\}$; $E(r_i) = E(s_i)$;

$E(r_i s_i) = \rho \geq 0$

• NOTATION

• $\text{scr}_\rho(f) =$ cc of f with ρ -correlated pairs

• $\text{cc}(f)$ perfectly shared randomness

• $\text{priv}(f)$: cc with PRIVATE randomness

starting point: for Boolean f :

$$cc(f) \leq isrp(f) \leq priv(f) \leq cc(f) + \log k$$

Can we get away with constant communication?

What if $cc(f) \ll \log(n)$?

DISTRIBUTED PERFECT AGREEMENTS FROM ISR

• AGREEMENT DISTRIBUTION:

- Alice $\leftarrow r$, Bob $\leftarrow s$; (r, s) p-corr. indep. bits
- outputs: $A \rightarrow u$, $B \rightarrow v$
 $H_\infty(u), H_\infty(v) \geq k$
- communication = c bits
- what is max prob. r of agreement ($u=v$)
- well studied!

We cannot individually reduce randomness and compute

→ requires the use of a new model.

Results.

model first studied by [Berman, Gurevich, 1986]

- Their focus: Simultaneous communication, general models of correlation

- isr (Equality) $= O(1)$

- M's results

$$u(t) \leq k \Rightarrow \text{isr}(t) \leq 2^k$$

Converse

$$\exists t \text{ with } u(t) \leq k \text{ \& \& } \text{isr}(t) \geq 2^k$$

Equality testing proof

- encode $x \mapsto X = E(x); y \mapsto Y = E(y); X, Y \in \{-1, 1\}^N$

$$x = y \Rightarrow \langle X, Y \rangle = N$$

$$x \neq y \Rightarrow \langle X, Y \rangle \leq \frac{N}{2}$$

do not look at a particular coordinate?

Butter: how do they correlate with random Gaussian vectors

Binding or sketching protocols

A: pick Gaussians $G_1, \dots, G_t \in \mathbb{R}^N$

$\xrightarrow{\text{sends}}$ $i \in [t]$ maximizing

$\langle G_i, x \rangle$ to Bob.

Bob: accepts iff $\langle G_i, x \rangle \geq 0$.

Turns out that the same
random vectors are not necessary

\rightarrow can use slightly positively
correlated Gaussians.

Averaging Gaussians \rightarrow
same correlation as
the initial sequence.

Analysts: $O_p(1)$ bits suffice
if $G \in G_p$.

Matching lower bound:

There exists a (promise) problem \mathcal{P} :

$$cc(\mathcal{P}) \geq k$$

$$|sup(\mathcal{P})| \geq \exp(k)$$

• Gap Sparse Inner Product

• Alice gets sparse

$$x \in \{0, 1\}^n, \text{wt}(x) \approx 2^{-k} \cdot n$$

• Bob gets $y \in \{0, 1\}^n$

$$\text{Promise: } \langle x, y \rangle \geq (.9) 2^{-k} \cdot n$$

$$\text{or } \langle x, y \rangle \leq (.0) 2^{-k} \cdot n$$

Decide which.

Principles of MOO

Indomally:

- Analysis of prob. process with bits often hard
- Related processes with Gaussian variables is easy
- $[1V:]$ under suff. general conditions
prob. of event "dominant"
when switching from
bits to Gaussians

k bits of comm. with prob. sharing

→ 2^k bits with sup. sh.

And is
right

What's necessary:
introducing scale
(context very large)

What happens when randomness is private.

→ sample
the entire distribution corresponding
to messages

→ makes randomness
oblivious

It problems are solved on the sphere
⇒ easy to solve on the cube

equality testing: can we get away
with constant cost?

no shared randomness: everything test
can be done
works by adding a constant to the