

# Communication Amid Uncertainty

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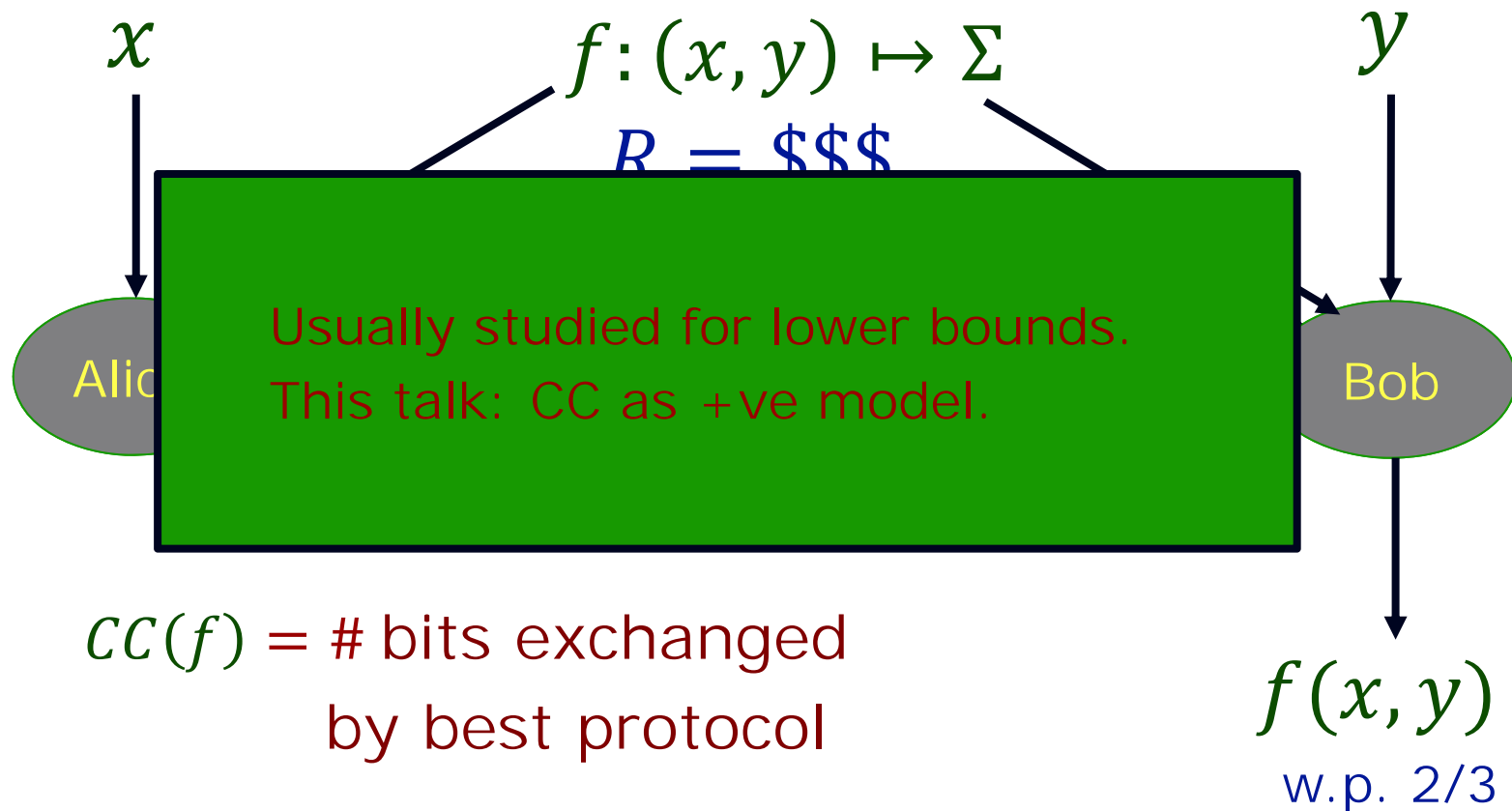
# Context in Communication

- Sender + Receiver share (huuuge) context
  - In human comm: Language, news, Social
  - In computer comm: Protocols, Codes, Distributions
  - Helps compress communication
- Perfectly shared  $\Rightarrow$  Can be abstracted away.
- Imperfectly shared  $\Rightarrow$  What is the cost?
  - How to study?



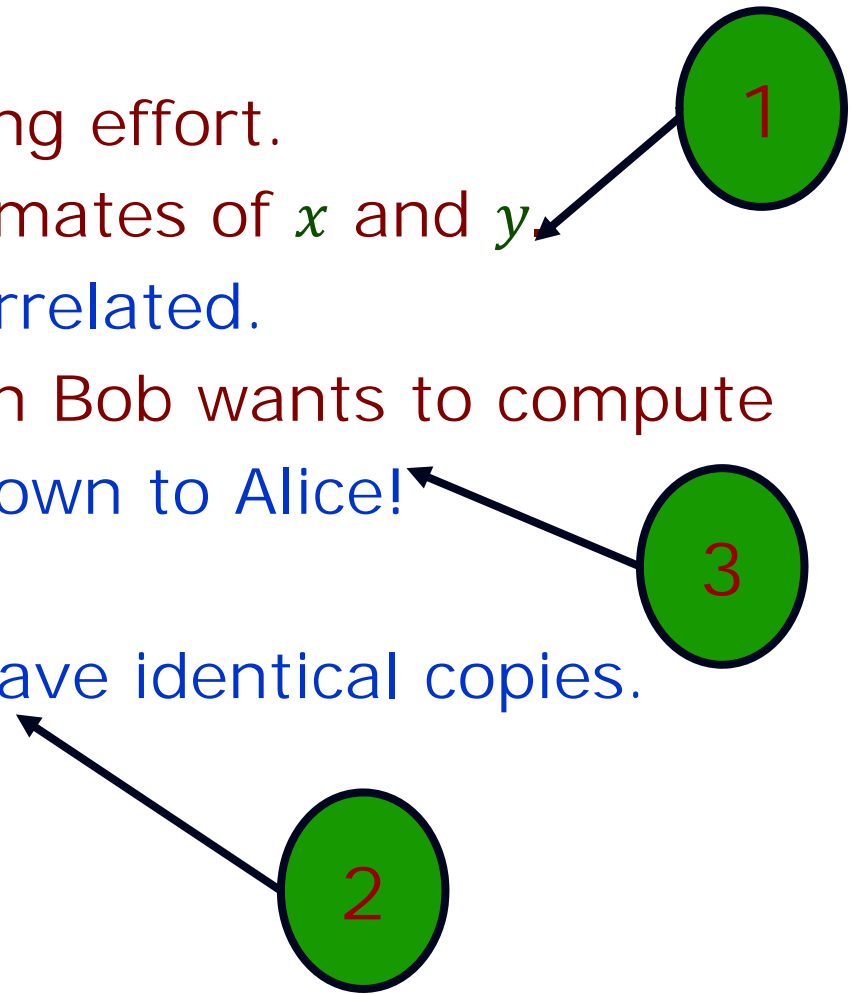
# Communication Complexity

The model (with shared randomness)



# Modelling Shared Context + Imperfection

- Many possibilities. Ongoing effort.
- Alice+Bob may have estimates of  $x$  and  $y$ 
  - More generally:  $x, y$  correlated.
- Knowledge of  $f$  – function Bob wants to compute
  - may not be exactly known to Alice!
- Shared randomness
  - Alice + Bob may not have identical copies.



# Part 1: Uncertain Compression

# Classical (One-Shot) Compression

- Sender and Receiver have distribution  $P \sim [N]$
- Sender/Receiver agree on Encoder/Decoder  $E/D$
- Sender gets  $X \in [N]$  ; Sends  $E(X)$
- Receiver gets  $Y = E(X)$  ; Decodes  $\hat{X} = D(Y)$
- Requirement:  $\hat{X} = X$  (always)
- Performance:  $\mathbb{E}_{X \leftarrow P}[|E(X)|]$
  
- Trivial Solution:  $\mathbb{E}_{X \leftarrow P}[|E(X)|] = \log N$
- Huffman Coding: Achieves  $\mathbb{E}_{X \leftarrow P}[|E(X)|] \leq H(P) + 1$

# The (Uncertain Compression) problem

[Juba, Kalai, Khanna, S.'11]

- Design encoding/decoding schemes  $(E/D)$  s.t.:
  - Sender has distribution  $P \sim [N]$
  - Receiver has distribution  $Q \sim [N]$
  - Sender gets  $X \in [N]$  ; Sends  $E(P, X)$  to receiver.
  - Receiver gets  $Y = E(P, X)$ ; Decodes  $\hat{X} = D(Q, Y)$
  - Want:  $X = \hat{X}$  (provided  $P, Q$  close),

$$\Delta(P, Q) \leq \Delta \text{ if } 2^{-\Delta} \leq \frac{\log P(x)}{\log Q(x)} \leq 2^{\Delta} \text{ for all } x$$

- Motivation: Models natural communication?

## Solution (variant of Arith. Coding)

- Uses shared randomness: Sender+Receiver  $\leftarrow r \in \{0,1\}^*$
- Use  $r$  to define sequences "dictionary"
  - $r_1 [1], r_1 [2], r_1 [3], \dots$
  - $r_2 [1], r_2 [2], r_2 [3], \dots$
  - $\dots$
  - $r_N [1], r_N [2], r_N [3], \dots$
- Sender:  $r_x [1 \dots L]$
- Receiver:  $r_z [1 \dots L]$
- Want:  $L : r_z [1 \dots L] = r_x [1 \dots L] \Rightarrow Q(z) < Q(x)$ ;
  - $\Leftrightarrow (Q(z) > Q(x) \Rightarrow r_z [1 \dots L] \neq r_x [1 \dots L])$
  - $\Leftarrow (P(z) > 4^{-\Delta} P(x) \Rightarrow r_z [1 \dots L] \neq r_x [1 \dots L])$

Analysis:

$$\mathbb{E}_r [L] = 2\Delta + \log \frac{1}{P(x)}$$

$$\mathbb{E}_{x,r} [L] = 2\Delta + H(P)$$



# Implications

- Coding scheme reflects the nature of human communication (extend messages till they feel unambiguous).
- Reflects tension between ambiguity resolution and compression.
  - Larger the ((estimated) gap in context), larger the encoding length.
  - Entropy is still a valid measure!
- The “shared randomness” assumption
  - A convenient starting point for discussion
  - But is dictionary independent of context?
    - This is problematic.

# Deterministic Compression: Challenge

- Say Alice and Bob have rankings of  $N$  players.
  - Rankings = bijections  $\pi, \sigma : [N] \rightarrow [N]$
  - $\pi(i)$  = rank of  $i^{\text{th}}$  player in Alice's ranking.
- Further suppose they know rankings are close.
  - $\forall i \in [N]: |\pi(i) - \sigma(i)| \leq 2.$
- Bob wants to know: Is  $\pi^{-1}(1) = \sigma^{-1}(1)$
- How many bits does Alice need to send (non-interactively).
  - With shared randomness –  $O(1)$
  - Deterministically?
    - With Elad Haramaty:  $O(\log^* n)$

# Part 2: Imperfectly Shared Randomness

# Model: Imperfectly Shared Randomness

- Alice  $\leftarrow r$ ; and Bob  $\leftarrow s$  where  $(r, s) =$  i.i.d. sequence of correlated pairs  $(r_i, s_i)_i$ ;  $r_i, s_i \in \{-1, +1\}$ ;  $\mathbb{E}[r_i] = \mathbb{E}[s_i] = 0$ ;  $\mathbb{E}[r_i s_i] = \rho \geq 0$ .
- Notation:
  - $isr_\rho(f)$  = cc of  $f$  with  $\rho$ -correlated bits.
  - $cc(f)$ : Perfectly Shared Randomness cc.  $= isr_1(f)$
  - $priv(f)$ : cc with PRIVate randomness  $= isr_0(f)$
- Starting point: for Boolean functions  $f$ 
  - $cc(f) \leq isr_\rho(f) \leq priv(f) \leq cc(f) + \log n$   $\rho \leq \tau \Rightarrow isr_\rho(f) \geq isr_\tau(f)$
  - What if  $cc(f) \ll \log n$ ? E.g.  $cc(f) = O(1)$

# Imperfectly Shared Randomness: Results

- Model first studied by [Bavarian, Gavinsky, Ito'14] ("Independently and earlier").
  - Their focus: Simultaneous Communication; general models of correlation.
  - They show  $isr(\text{Equality}) = O(1)$  (among other things)
- Our Results: [Canonne, Guruswami, Meka, S'15]
  - Generally:  $cc(f) \leq k \Rightarrow isr(f) \leq 2^k$
  - Converse:  $\exists f$  with  $cc(f) \leq k$  &  $isr(f) \geq 2^k$

# Aside: Easy CC Problems [Ghazi,Kamath,S'15]

- Equality testing:
  - $EQ(x, y) = 1 \Leftrightarrow x = y;$
- Hamming distance:
  - $H_k(x, y) = 1 \Leftrightarrow \Delta(x, y) \leq k;$
- Small set intersection:
  - $\cap_k(x, y) = 1 \Leftrightarrow wt(x), wt(y) \leq k$
  - $CC(\cap_k) = O(k)$  [Håstad Wi]

Protocol:

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

$poly(k)$  Protocol

Use common

to hash  $[n] \rightarrow$

$$\begin{aligned}
 x &= (x_1, \dots, x_n) \\
 y &= (y_1, \dots, y_n) \\
 \langle x, y \rangle &\triangleq \sum_i x_i y_i
 \end{aligned}$$

Unstated philosophical contribution of CC a la Yao:

Communication with a focus ("only need to determine  $f(x, y)$ ")  
 can be more effective (shorter than  $|x|, H(x), H(y), I(x; y) \dots$ )

# Equality Testing (our proof)

- Key idea: Think inner products.
  - Encode  $x \mapsto X = E(x); y \mapsto Y = E(y); X, Y \in \{-1, +1\}^N$ 
    - $x = y \Rightarrow \langle X, Y \rangle = N$
    - $x \neq y \Rightarrow \langle X, Y \rangle \leq N/2$
- Estimating inner products:
  - Building on sketching protocols ...
  - Alice: Picks Gaussians  $G_1, \dots, G_t \in \mathbb{R}^N$ ,
  - Sends  $i \in [t]$  maximizing  $\langle G_i, X \rangle$  to Bob.
  - Bob: Accepts iff  $\langle G'_i, Y \rangle \geq 0$
  - Analysis:  $O_\rho(1)$  bits suffice if  $G \approx_\rho G'$

Gaussian  
Protocol

# General One-Way Communication

- Idea: All communication  $\leq$  Inner Products
- (For now: Assume  $\text{one-way-cc}(f) \leq k$ )
  - For each random string  $R$ 
    - Alice's message =  $i_R \in [2^k]$
    - Bob's output =  $f_R(i_R)$  where  $f_R: [2^k] \rightarrow \{0,1\}$
    - W.p.  $\geq \frac{2}{3}$  over  $R$ ,  $f_R(i_R)$  is the right answer.



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- Vector representation:
  - $i_R \mapsto x_R \in \{0,1\}^{2^k}$  (unit coordinate vector)
  - $f_R \mapsto y_R \in \{0,1\}^{2^k}$  (truth table of  $f_R$ ).
  - $f_R(i_R) = \langle x_R, y_R \rangle$ ; Acc. Prob.  $\propto \langle X, Y \rangle$ ;  $X = (x_R)_R$ ;  $Y = (y_R)_R$
  - Gaussian protocol estimates inner products of unit vectors to within  $\pm\epsilon$  with  $O_\rho\left(\frac{1}{\epsilon^2}\right)$  communication.

# Two-way communication

- Still decided by inner products.
- Simple lemma:
  - $\exists K_A^k, K_B^k \subseteq \mathbb{R}^{2^k}$  convex, that describe private coin k-bit comm. strategies for Alice, Bob s.t. accept prob. of  $\pi_A \in K_A^k, \pi_B \in K_B^k$  equals  $\langle \pi_A, \pi_B \rangle$
- Putting things together:

Theorem:  $cc(f) \leq k \Rightarrow isr(f) \leq O_\rho(2^k)$

# Part 3: Uncertain Functionality

# Model

- Bob wishes to compute  $f(x, y)$ ; Alice knows  $g \approx f$ ;
- Alice, Bob given  $g, f$  explicitly. (Input size  $\sim 2^n$ )
- Modelling Questions:
  - What is  $\approx$ ?
  - Is it reasonable to expect to compute  $f(x, y)$ ?
    - E.g.,  $f(x, y) = f'(x)$ ? Can't compute  $f(x, y)$  without communicating  $x$
- Answers:
  - Assume  $x, y \sim \{0,1\}^n \times \{0,1\}^n$  uniformly.
  - $f \approx_\delta g$  if  $\delta(f, g) \leq \delta$ .
  - Suffices to compute  $h(x, y)$  for  $h \approx_\epsilon f$

# Results - 1

- Thm [Ghazi, Komargodski, Kothari, S.]:  $\exists f, g, \mu$  s.t.  $cc_{\mu,1}^{1way}(f), cc_{\mu,1}^{1way}(g) = 1$  and  $\delta_\mu(f, g) = o(1)$ ; but uncertain communication =  $\Omega(\sqrt{n})$ ;
- Thm [GKKS]: But not if  $x \perp y$  (in 1-way setting).
  - (2-way, even 2-round, open!)
- Main Idea:
  - Canonical 1-way protocol for  $f$ :
    - Alice + Bob share random  $y_1, \dots, y_m \in \{0,1\}^n$ .
    - Alice sends  $f(x, y_1), \dots, f(x, y_m)$  to Bob.
    - Protocol used previously ... but not as "canonical".
  - Canonical protocol robust when  $f \approx g$ .

# Conclusions

- Positive view of communication complexity:  
Communication with a focus can be effective!
- Context Important:
  - New layer of uncertainty.
  - New notion of scale (context LARGE)
    - Importance of  $o(\log n)$  additive factors.
- Many “uncertain” problems can be solved without resolving the uncertainty (which is a good thing)
- Many open directions+questions

**Thank You!**