

# Imperfectly Shared Randomness in Communication

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**Dedicate to our SVC  
colleagues!**

**You are the best!**

# Context in Communication

- Context = Central element of Communication.
  - Shared between sender and receiver
  - Implicit. (Doesn't participate in  $n$ )
- Examples:
  - Meaning of bits (what action to take given rec'd message).
  - Shannon theory: Distribution of source, Channel behavior, Codes used.
  - Communication Complexity: Function being computed, Randomness being shared etc.
  - Human communication: Language, Grammar ...

# Uncertainty in sharing of context

- Whenever “large” amounts of information is “shared”, there must be some imperfection.
  - Online Forms: Example – my bank:
    - “Please enter your PIN now”
      - But I have an ATM PIN, a phone PIN, a transaction PIN.
  - Compression: Do sender and receiver agree perfectly on the prior?  
[Juba, Kalai, Khanna, S.’11], [Haramaty, S.’14]
  - This talk: Shared Randomness in Communication Complexity.

# Shared Randomness in CC

- Canonical example: Equality testing.
  - Alice has  $x \in \{0,1\}^n$ ; Bob has  $y \in \{0,1\}^n$  ;
  - Want to know if  $x = y$ ?
  - Deterministically: Communicate  $\Omega(n)$  bits
  - With private randomness:  $\Theta(\log n)$  bits
    - Idea: Alice encodes  $x \mapsto E(x)$ ; Picks  $i \in [N]$ ; sends  $(i, E(x)_i)$
  - With shared randomness:  $O(1)$  bits
    - Just send  $E(x)_i$
- Upshot: Randomness very helpful!

# Compression with Uncertain Priors

- [JKKS'11]:
- Alice has  $P = (P_1, \dots, P_N)$ ;  $m \leftarrow_P [N]$ ;
- Bob has  $Q = (Q_1, \dots, Q_N)$ ;  $P \approx_\Delta Q$ ;
- Both want to know  $m$ .
- State of affairs:
  - $P = Q$ : Expected comm. =  $H(P)$ . [Huffman]
  - $P \approx_\Delta Q$  + shared randomness:  $H(P) + 2\Delta$  [JKKS]
  - $P \approx_\Delta Q$  deterministically:  $O(H(P) + \Delta + \log \log N)$  [Haramaty+S.]

# Uncertain Compression (thoughts)

- Is entropy the right measure of compressibility?
  - With uncertainty?
    - Deterministically ... may be not (the  $\log \log n$ )
    - Randomized: Perfect sharing inconsistent with uncertainty!
    - Unless ... randomness is shared imperfectly!
- Motivates: Imperfectly shared randomness in CC.
- “Independently” raised and studied by [Bavarian, Gaminsky, Ito'14].

# Our Model

- General communication complexity with imperfectly shared randomness.
- Alice  $\leftarrow r$ ; and Bob  $\leftarrow s$  where  $(r, s) =$  i.i.d. sequence of correlated pairs  $(r_i, s_i)_i$ ;  $r_i, s_i \in \{-1, +1\}$ ;  $E[r_i] = E[s_i] = 0$ ;  $E[r_i s_i] = \rho$ .
- Notation:
  - $isr_\rho(f)$  = cc of  $f$  with  $\rho$ -correlated bits.
  - $psr(f)$ : perfectly shared randomness cc.
  - $priv(f)$ : cc with private randomness
- Starting point: for Boolean functions  $f$ 
  - $psr(f) \leq isr_\rho(f) \leq priv(f) \leq psr(f) + \log n$



# Results

- [Bavarian et al.]: Focus on simultaneous message model; more general correlations.
- Our focus:
  - One-way communication: Alice  $\rightarrow$  Bob; Bob outputs  $f$ .
  - Problems where difference of  $\log n$  significant.
- Results:
  - Uncertain Compression:  $O_\rho(H(P) + \Delta)$
  - Equality testing:  $O_\rho(1)$  (also [Bavarian et al.])
  - More generally:  $psr(f) \leq k \Rightarrow ow-isr(f) \leq 2^k$
  - Converse:  $\exists f$  with  $ow-psr(f) \leq k$  &  $ow-isr(f) \geq 2^k$

## Rest of the talk

- Uncertain Compression:  $O_\rho(H(P) + \Delta)$
- Equality testing:  $O_\rho(1)$  (also [Bavarian et al.])
- General upper bound:  $psr(f) \leq k \Rightarrow ow-isr(f) \leq 2^k$
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# Compression:

- [JKKS] *psr* solution: Let common randomness define “dictionary”: arbitrarily long sequences  $r_m$  for every message  $m$ .
  - Alice sends “long enough” prefix of  $r_m$
  - Bob does maximum likelihood decoding based on  $Q$ .
  - Analysis: Exercise
- Our *isr* solution:
  - Alice send longer prefix.
  - Bob does max. likelihood decoding among messages that are close enough to rec'd word.
- Moral: Protocols “natural”  $\Rightarrow$  Explains behavior?

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# Equality Testing

- 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:
  - Using ideas from low-distortion embeddings ...
  - Alice: Picks Gaussian  $G \in \mathbb{R}^n$ , sends  $\langle G, X \rangle$
  - Bob: compares  $\langle G, X \rangle$  with  $\langle G, Y \rangle$
  - (mod analysis):  $O_\rho(1)$  bits suffice if  $G \approx_\rho G'$
  - [Bavarian et al.] Alternate protocol.

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# General Communication

- Idea: All communication  $\leq$  Inner Products
- Example: One-way communication  $k$  bits:
  - 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  $f_R$ ).
  - $f_R(i_R) = \langle x_R, y_R \rangle$
- Gaussian protocol estimates inner products to within relative error  $\epsilon$  with  $O\left(\frac{1}{\epsilon^2}\right)$  communication.



## Rest of the talk

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# Main Technical Result: Matching lower bound

- There exists promise problem  $f$  s.t.
  - $ow-psr(f) \leq k$
  - $ow-isr_\rho(f) \geq \exp(k)$
- The Problem:
  - Gap Sparse Inner Product (G-Sparse-IP).
  - Alice gets sparse  $x \in \{0,1\}^n$ ;  $wt(x) \approx 2^{-k} \cdot n$
  - Bob gets  $y \in \{-1, +1\}^n$
  - Promise:  $\langle x, y \rangle \geq \left(\frac{1}{3}\right) 2^{-k} \cdot n$  or  $\langle x, y \rangle \leq 0$ .
  - Decide which.

## *psr* Protocol for G-Sparse-IP

- Idea:  $x_i \neq 0 \Rightarrow y_i$  correlated with answer.
- Use (perfectly) shared randomness to find random index  $i$  s.t.  $x_i \neq 0$ .
- Shared randomness:  $i_1, i_2, i_3, \dots$  uniform over  $[n]$
- Alice  $\rightarrow$  Bob: smallest index  $j$  s.t.  $x_{i_j} \neq 0$ .
- Bob: Accept if  $y_{i_j} = 1$
- Expect  $j \approx 2^k$ ;  $psr \leq k$ .

# ISR lower bounds

- Challenge: Usual CC lower bounds give a distribution and prove lower bound against it.
  - G-Sparse-IP has a low-complexity protocol for every input, with shared randomness.
  - Thus for every distribution, there exists a deterministic low-complexity protocol!
  - So usual method can't work ...
- 
- Need to fix strategy first and then "tailor-make" a hard distribution for the strategy ...

# ISR lower bound for GSIP: Overview

- Strategies: Alice  $f_r(x) \in [\ell]$ ; Bob  $g_s(y) \in \{0,1\}^\ell$ ;
- Two possibilities:
  - Case 1: Alice's strategy and Bob's strategy have common highly "influential coordinate":
    - ( $i$  s.t. flipping  $x_i$  changes Alice's message etc.)
    - Leads to protocol for "agreement distillation" [We prove this is impossible.]
  - Case 2: Strategies have no common influential variable:
    - Invariance Principle  $\Rightarrow$  Solves some Gaussian problem
    - Lower bound for Gaussian problem. (Details shortly)

# Case 1: Agreement Distillation

- Problem: Charlie  $\leftarrow r$ ; Dana  $\leftarrow s$ ;  $(r, s)$   $\rho$ -correlated
- Goal: Charlie outputs  $u$ ; Dana outputs  $v$ ;  
$$H_\infty(u), H_\infty(v) \geq k; \quad \Pr[u = v] \geq \gamma$$
- Lemma: With zero communication  $\gamma = 2^{-\Omega(k)}$ ;
- Proof: “Small-set expansion of noisy hypercube”
  - See, e.g., [Analysis of Boolean functions, O’Donnell]
- Corollary: For  $c$  bits of communication,  
$$c \geq \epsilon \cdot k + \log \gamma$$

# Completing Case 1

- $\text{Bad} \triangleq \{i \mid \Pr_r[\text{Inf}_i(f_r) \geq \text{high}] \geq \text{large}\}$   
 $\cup \{i \mid \Pr_s[\text{Inf}_i(g_s) \geq \text{high}] \geq \text{large}\}$
- Fact: (for our defn of influence) any function has bounded number of high influence variables.
- (By Fact + Markov) Can assume  $|\text{Bad}| \leq \epsilon \cdot n$ .
- Distributions on Yes and No instances:
  - No:  $x$  random sparse  $\in \{0,1\}^n$ ;  $y \leftarrow_U \{-1,1\}^n$
  - Yes: Same as No on Bad coordinates.
    - On rest,  $y_i$  is more likely to be  $+1$  if  $x_i = 1$ .

## Completing Case 1 (contd.)

- Agreement strategy for Charlie + Dana:
  - Charlie:  $i \in [n]$  – Bad s.t.  $\text{Inf}_i(f_r)$  high.
  - Dana:  $j \in [n]$  – Bad s.t.  $\text{Inf}_j(g_s)$  high.
- Analysis:
  - $H_\infty(i), H_\infty(j)$  large since  $i, j \notin \text{Bad}$ .
  - $i = j?$ : Case 1 assumption.
  
- Combined with lower bound for agreement distillation, implies Case 1 can't occur



## Case 2: No common influential variable

- Key Lemma: Fix  $r, s$ ; let  $f = f_r$  and  $g = g_s$ .  
If  $\ell$  small ( $2^{2^k}$ ) and  $f, g$  distinguish Yes/No then  $f, g$  have common influential variable.
- Idea: Use “Invariance Principle”:
  - Remarkable theorem: Mossel, O’Donnell, Oleskiewicz; Mossel++;
  - Informal form:  $f, g$  low-degree polynomials with no common influential variable  $\Rightarrow$   
 $\text{Exp}_{x,y}[f(x)g(y)] \approx \text{Exp}_{X,Y}[f(X)g(Y)]$ 
    - where  $x, y$  Boolean  $n$ -wise product dist.
    - and  $X, Y$  Gaussian  $n$ -wise product dist.

# The Gaussian-IP Problem

- Suppose we can get the “perfect” invariance theorem for us ...
- Would transform:  
Sol’n for G-Sparse-IP  $\rightarrow$  Sol’n for G-Gaussian-IP
  - Alice, Bob get Gaussian vectors  $X, Y \in \mathbb{R}^n$
  - Yes:  $\langle X, Y \rangle \geq 2^{-k}$  ; No:  $\langle X, Y \rangle \leq 0$
- Hope: Non-sparse  $\Rightarrow \geq 2^k$  communication
  - Formally [Bar Yossef et al.]: Can reduce “indexing” to G-Gaussian-IP.

# Invariance Principle + Challenges

- Informal Invariance Principle:  $f, g$  low-degree polynomials with no common influential variable  
 $\Rightarrow \text{Exp}_{x,y}[f(x)g(y)] \approx \text{Exp}_{X,Y}[f(X)g(Y)]$ 
  - where  $x, y$  Boolean  $n$ -wise product dist.
  - and  $X, Y$  Gaussian  $n$ -wise product dist
- Challenges [+ Solutions]:
  - Our functions not low-degree [Smoothing]
  - Our functions not real-valued
    - $g: \{-1,1\}^n \rightarrow \{0,1\}^\ell$ : [Truncate range to  $[0,1]^\ell$ ]
    - $f: \{0,1\}^n \rightarrow [\ell]$ : [???, [work with  $\Delta(\ell)$ ]]

# Invariance Principle + Challenges

- Informal Invariance Principle:  $f, g$  low-degree polynomials with no common influential variable  
 $\Rightarrow \text{Exp}_{x,y}[f(x)g(y)] \approx \text{Exp}_{X,Y}[f(X)g(Y)]$
- Challenges
  - Our functions not low-degree [Smoothing]
  - Our functions not real-valued [Truncate]
  - Quantity of interest is not  $f(x) \cdot g(y) \dots$ 
    - [Can express quantity of interest as inner product. ]
  - ... (lots of grunge work ...)
- Get a relevant invariance principle (next)

# Invariance Principle for (one-way) CC

- Thm:  $\exists$  transformations  $T_1, T_2$  s.t.  
if  $f: \{0,1\}^n \rightarrow \Delta(\ell)$  and  $g: \{-1,1\}^n \rightarrow [0,1]^\ell$   
have no common influential variable, then  
 $F = T_1 f: \mathbb{R}^n \rightarrow \Delta(\ell)$  and  $G = T_2 g: \mathbb{R}^n \rightarrow [0,1]^\ell$  satisfy  
 $\text{Exp}_{x,y}[\langle f(x), g(y) \rangle] \approx \text{Exp}_{X,Y}[\langle F(X), G(Y) \rangle]$
- Main differences:  $f, g$  vector-valued.
- Functions are transformed:  $f \mapsto F; g \mapsto G$
- Range is preserved exactly  $(\Delta(\ell); [0,1]^\ell)$ !
  - So  $F, G$  are still communication strategies!

# Summarizing

- $k$  bits of comm. with perfect sharing  
→  $2^k$  bits with imperfect sharing.
- This is tight (for one-way communication)
  - Invariance principle for communication
  - Agreement distillation
  - Low-influence strategies

# Conclusions

- Imperfect agreement of context important.
  - Dealing with new layer of uncertainty.
  - Notion of scale (context LARGE)
- Many open directions + questions:
  - Imperfectly shared randomness:
    - One-sided error?
    - Does interaction ever help?
    - How much randomness?
    - More general forms of correlation?

**Thank You!**