

## **Key Questions in Information Theory**

Some key questions at the heart of information theory.

### Key Question 1: Blocklength in Information Theory



### $0001101 \dots 10110$

### What if blocklength is 1?

**One-shot information theory** [2, 3]: network is only used **once**!

- Error probability cannot be driven to zero!
- No law of large number  $\rightarrow$  no typicality!
- No time-sharing!
- No memoryless/ergodic assumption!
- **Objective:** One-shot achievabilities that can imply existing (first/second order) asymptotic/finite-blocklength bounds?

### Key Question 2: Noisy network coding



### **Noisy network coding** [4]:

- What is the capacity of a noisy network?
- What coding scheme can achieve the capacity?

### Key Question 3: Unified Coding Scheme



### A unified coding schemes [1]:

- A unified node operation in networks?
- Unify channel coding, source coding, and coding for computing?
- Automated machine-proving tools?

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# **One-Shot Coding over General Noisy Networks**

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 $(X_N, Y_N) \leftarrow M_N$ 

## A Unified One-Shot Coding Scheme

To answer the key questions on the left, our scheme [5] combines:

- 1. One-shot/finite-blocklength network information theory
- 2. Noisy network coding
- 3. Unified scheme (source coding/channel coding/coding for computing)

### Main Theorem

For any acyclic discrete network  $(P_{Y_i|X^{i-1},Y^{i-1}})_{i\in[N]}$ , we provide a one-shot achievability result: For any collection of indices  $(a_{i,j})_{i \in [N], j \in [d_i]}$  where  $(a_{i,j})_{j \in [d_i]}$  is a sequence of distinct indices in [i - 1] for each *i*, any sequence  $(d'_i)_{i \in [N]}$  with  $0 \leq d'_i \leq d_i$  and any collection of conditional distributions  $(P_{U_i|Y_i,\overline{U}'_i}, P_{X_i|Y_i,U_i,\overline{U}'_i})_{i\in[N]}$  (where  $\overline{U}_{i,\mathcal{S}} := (U_{a_{i,j}})_{j\in\mathcal{S}}$  for  $\mathcal{S} \subseteq [d_i]$ and  $\overline{U}'_i := \overline{U}_{i,[d']}$ , which induces the joint distribution of  $X^N, Y^N, U^N$ (the "ideal distribution"), there exists a public-randomness coding scheme  $(P_W, (f_i)_{i \in [N]})$  such that the joint distribution of  $\tilde{X}^N, \tilde{Y}^N$  induced by the scheme (the "actual distribution") satisfies

$$\delta_{\mathrm{TV}} \Big( P_{X^N, Y^N}, P_{\tilde{X}^N, \tilde{Y}^N} \Big) \leq \mathbf{E} \Big[ \min \Big\{ \sum_{i=1}^N \sum_{j=1}^{d_i} B_{i,j}, 1 \Big\} \Big],$$
  
$$\gamma_{i,i} := \prod_{i=1}^{d_i} \inf (\ln |\mathcal{U}_{i,i}| + 1) \text{ and }$$

where 
$$\gamma_{i,j} := \prod_{k=j+1}^{a_i} (\ln |\mathcal{U}_{a_{i,k}}| + 1)$$
, and  
 $B_{i,j} := \gamma_{i,j} \prod_{i=1}^{d_i} \left( 2^{-\iota(\overline{U}_{i,k};\overline{U}_{i,[d_i]\setminus[j..k]},Y_i) + \iota(\overline{U}_{i,k};\overline{U}'_{a_{i,k}},Y_{a_{i,k}})} + \mathbf{1}\{k > j\} \right).$ 

### **Techniques**

**Poisson functional representation [3]**: Let  $\mathbf{U} := (Z_u)_{u \in \mathcal{U}}$  be i.i.d. Exp(1) random variables. Given a distribution P over finite  $\mathcal{U}$ ,

k=j

 $\mathbf{U}_P := \operatorname{argmin}_u$ 

- 2. Each node is associated with an exponential process.
- **Exponential Process Refinement**: For  $Q_{V,U}$  over a finite  $\mathcal{V} \times \mathcal{U}, \forall v \in \mathcal{V}$ ,

$$\begin{split} \mathbf{E} \bigg[ \frac{1}{Q_{V,U}^{\mathbf{U}}(v,\mathbf{U}_P)} \bigg| \mathbf{U}_P \bigg] &\leq \frac{\ln |\mathcal{U}| + 1}{Q_V(v)} \left( \frac{P(\mathbf{U}_P)}{Q_{U|V}(\mathbf{U}_P|v)} + 1 \right). \\ Q_{U,V} \text{ (prior)} \longrightarrow \text{refine by } \mathbf{U} \text{ (soft decoding)} \longrightarrow Q_{V,U}^{\mathbf{U}} \text{ (posterior)} \end{split}$$

### References

- [1] Si-Hyeon Lee and Sae-Young Chung. A unified random coding bound. IEEE Transactions on Information Theory, 64(10):6779-6802, 2018.
- [2] Cheuk Ting Li and Venkat Anantharam. A unified framework for one-shot achievability via the poisson matching lemma. IEEE Transactions on Information Theory, 67(5):2624–2651, 2021.
- [3] Cheuk Ting Li and Abbas El Gamal. Strong functional representation lemma and applications to coding theorems. IEEE Transactions on Information Theory, 64(11):6967–6978, 2018.
- [4] Sung Hoon Lim, Young-Han Kim, Abbas El Gamal, and Sae-Young Chung. Noisy network coding. IEEE Transactions on Information Theory, 57(5):3132–3152, 2011.
- [5] Yanxiao Liu and Cheuk Ting Li. One-shot coding over general noisy networks. *arXiv preprint* arXiv:2402.06021, 2024.

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$${}^{\prime} \frac{Z_u}{P(u)}.$$

The main theorem can be applied to any combination of source coding, channel coding and coding for computing. Note  $\iota(x; y|z) := \log \frac{P(x,y|z)}{(P(x|z)P(y|z))}$ .

### **Channel Coding with State Info at Encoder**



### Source Coding with Side Info at Decoder



### Multiple Access Channel

It recovers the asymptotic capacity region.

### **One-Shot Relay Channel**



- Let  $U_1 := (X, M), U_2 := U$ , Main Theorem gives a compress-forward bound:  $P_{e} \leq \mathbf{E} \left[ \min \left\{ \gamma \mathsf{L} 2^{-\iota(X;U,Y)} \left( 2^{-\iota(U;Y)+\iota(U;Y_{r})} + 1 \right), 1 \right\} \right]$
- where  $\gamma = \ln |\mathcal{U}| + 1$ ,  $(X, Y_{r}, U, X_{r}, Y) \sim P_{X} P_{Y_{r}|X} P_{U|Y_{r}} \delta_{x_{r}(Y_{r}, U)} P_{Y|X, Y_{r}, X_{r}}$ .
- It is a one-shot version of relay-with-unlimited-look-ahead. • If Y = (Y', Y'') and  $P_{Y|X,X_r,Y_r} = P_{Y'|X,Y_r}P_{Y''|X_r}$ , it is a one-shot version of primitive relay channel.
- By message splitting, we can also have a partial-decode-forward bound.

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### Examples

$$\xrightarrow{X} P_{Y|X,S} \xrightarrow{Y} \text{Dec} \longrightarrow \widehat{M}$$

- Fix  $P_{U|S}$  and  $x: \mathcal{U} \times \mathcal{S} \to \mathcal{X}$ . For  $M \sim \text{Unif}[L], S \sim P_S$ , let  $U_1 = (U, M)$ ,  $P_e := \mathbf{P}(\tilde{X}_2 \neq M) \le \mathbf{E} \left[ \min \left\{ \mathsf{L} 2^{-\iota(U;Y) + \iota(U;S)}, 1 \right\} \right].$
- It recovers asymptotic capacity, attains the best known second-order result.

- Fix  $P_{U|X}$  and  $z: \mathcal{U} \times \mathcal{Y} \to \mathcal{Z}$ . For  $X \sim P_X, T \sim P_{T|X}, M \in [L]$ ,  $P_e := \mathbf{P}\{d(X, \tilde{Z}) > \mathsf{D}\} \le \mathbf{E}\Big[\min\Big\{\mathbf{1}\{d(X, Z) > \mathsf{D}\} + \mathsf{L}^{-1}2^{-\iota(U;T) + \iota(U;X)}, 1\Big\}\Big].$ It recovers asymptotic capacity, and also covers coding for computing.
- For MAC  $P_{Y|X_1,X_2}$  and  $M_j \sim \text{Unif}[L_j]$  for j = 1, 2, with  $\gamma := \ln(L_1|\mathcal{X}_1|) + 1$ ,  $P_{e} \leq \mathbf{E} \Big[ \min \Big\{ \gamma \mathsf{L}_{1} \mathsf{L}_{2} 2^{-\iota(X_{1}, X_{2}; Y)} + \gamma \mathsf{L}_{2} 2^{-\iota(X_{2}; Y | X_{1})} + \mathsf{L}_{1} 2^{-\iota(X_{1}; Y | X_{2})}, 1 \Big\} \Big].$