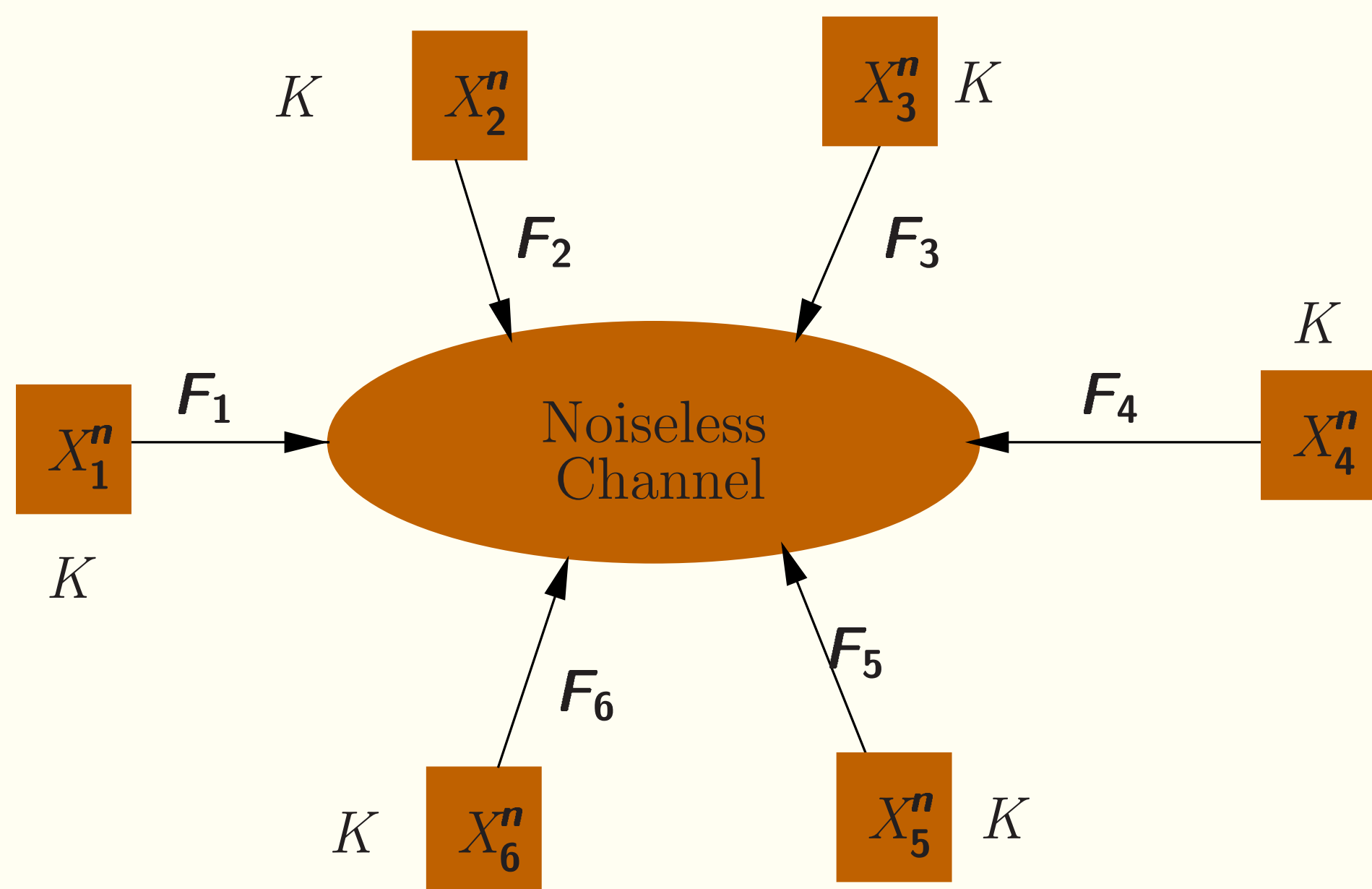


On the Communication Complexity for SK Generation in the Multiterminal Source Model

Manuj Mukherjee, Navin Kashyap
Indian Institute of Science, Bangalore, India



The Multiterminal Source Model



- ▶ A set of terminals $\mathcal{M} = \{1, 2, \dots, m\}$, with each terminal possessing a component of a **Discrete Memoryless Multiple Source**, $X_i^n, \forall 1 \leq i \leq m$.
- ▶ The terminals are allowed to communicate 'interactively'.
- ▶ $\mathbf{F} = \{F_1, F_2, \dots, F_r\}$ is the **interactive communication** taking values in \mathcal{F} . Here F_j sent by some terminal i is a function of X_i^n and all the previous communication.
- ▶ Rate of the interactive communication is $\frac{1}{n} \log |\mathcal{F}|$.
- ▶ The terminals compute a **secret key** (SK) $\mathbf{K} = \mathbf{K}(X_{\mathcal{M}}^n)$ with the aid of \mathbf{F} .
- ▶ The secret key \mathbf{K} satisfies the following property: for any $\epsilon > 0$ and for all sufficiently large n ,
 - ▷ \exists some function $g_i^{(n)}(X_i^n, \mathbf{F})$ such that $\mathcal{P}(\mathbf{K} \neq g_i^{(n)}(X_i^n, \mathbf{F})) \leq \epsilon, \forall 1 \leq i \leq m$. (**Recoverability**)
 - ▷ $I(\mathbf{K}; \mathbf{F}) \leq \epsilon$. (**Strong secrecy**)
 - ▷ $\log |\mathcal{K}| - H(\mathbf{K}) \leq \epsilon$, where \mathcal{K} is the range of \mathbf{K} . (**Uniformity**)
- ▶ If $\frac{1}{n} H(\mathbf{K}) \rightarrow \mathbf{R}$ as $n \rightarrow \infty$, then \mathbf{R} is called an **achievable secret key rate**.
- ▶ **SK capacity** $I(\mathcal{X}_{\mathcal{M}}) = \sup \mathbf{R}$.

Achieving SK Capacity

- ▶ Use a **Slepian-Wolf code** to recover $X_{\mathcal{M}}^n$ at all terminals.
- ▶ Such a code is called a **communication for omniscience**.
- ▶ The **achievable rate region**:

$$\mathcal{R}_{\text{CO}} = \{(R_1, R_2, \dots, R_m) : R_i \geq 0, \forall 1 \leq i \leq m, \sum_{j \in B} R_j \geq H(X_B | X_{B^c}), \forall B \subset \mathcal{M}, B \neq \emptyset\}$$
- ▶ Use a **balanced coloring function** on $X_{\mathcal{M}}^n$ to get \mathbf{K} .

Evaluating SK Capacity

- ▶ $I(\mathcal{X}_{\mathcal{M}}) = H(\mathcal{X}_{\mathcal{M}}) - R_{\text{CO}}$, where $R_{\text{CO}} = \min_{(R_1, R_2, \dots, R_m) \in \mathcal{R}_{\text{CO}}} \sum_{i=1}^m R_i$ is the **minimum rate of communication for omniscience**. [Csiszár & Narayan, '04]
- ▶ $I(\mathcal{X}_{\mathcal{M}}) = \min_{\mathcal{P}} \Delta(\mathcal{P})$, where $\Delta(\mathcal{P}) = \frac{1}{\ell-1} [H(X_{A_1}) + H(X_{A_2}) + \dots + H(X_{A_\ell}) - H(\mathcal{X}_{\mathcal{M}})]$ for any partition $\mathcal{P} = \{A_1, A_2, \dots, A_\ell\}$ of \mathcal{M} with $\ell \geq 2$. We denote by \mathcal{P}^* the finest optimal partition and call it the **fundamental partition**. [Chan & Zheng, '10]

Communication Complexity

- ▶ $R_{\text{SK}} = \text{Communication complexity}$, is the minimum rate of communication required to achieve SK capacity.
- ▶ $R_{\text{SK}} \leq R_{\text{CO}}$. [Csiszár & Narayan, 2004]
- ▶ If $R_{\text{SK}} = R_{\text{CO}}$, we call the source **R_{SK} -maximal**. These are thus the worst-case sources in terms of communication rates.

Lower Bound on Communication Complexity

- ▶ **Result**: $R_{\text{SK}} \geq \text{CI}(\mathcal{X}_{\mathcal{M}}) - I(\mathcal{X}_{\mathcal{M}})$. [Mukherjee & Kashyap, '16]
 - ▷ $\text{CI}(\mathcal{X}_{\mathcal{M}})$ is the **minimum rate of interactive common information**.
 - ▷ Fact: $H(\mathcal{X}_{\mathcal{M}}) \geq \text{CI}(\mathcal{X}_{\mathcal{M}}) \geq I(\mathcal{X}_{\mathcal{M}})$ and hence the lower bound is non-negative.

What is Interactive Common Information?

- ▶ $\mathbf{J} = \mathbf{J}(X_{\mathcal{M}}^n)$ is a **common randomness** if \mathbf{J} is recoverable at all terminals using some interactive communication \mathbf{F} .
- ▶ (\mathbf{J}, \mathbf{F}) is an **interactive common information** (CI) if (\mathbf{J}, \mathbf{F}) is a **Wyner common information** for $X_{\mathcal{M}}^n$.
- ▶ $\text{CI}(\mathcal{X}_{\mathcal{M}}) = \min_{(\mathbf{J}, \mathbf{F}) \text{ is CI}} \lim_{n \rightarrow \infty} \frac{1}{n} H(\mathbf{J}, \mathbf{F})$.

What is Wyner Common Information?

- ▶ $\mathbf{L} = \mathbf{L}(X_{\mathcal{M}}^n)$ is a **Wyner common information** for $X_{\mathcal{M}}^n$ if $\lim_{n \rightarrow \infty} \frac{1}{n} I(X_{\mathcal{M}}^n | \mathbf{L}) = 0$ where

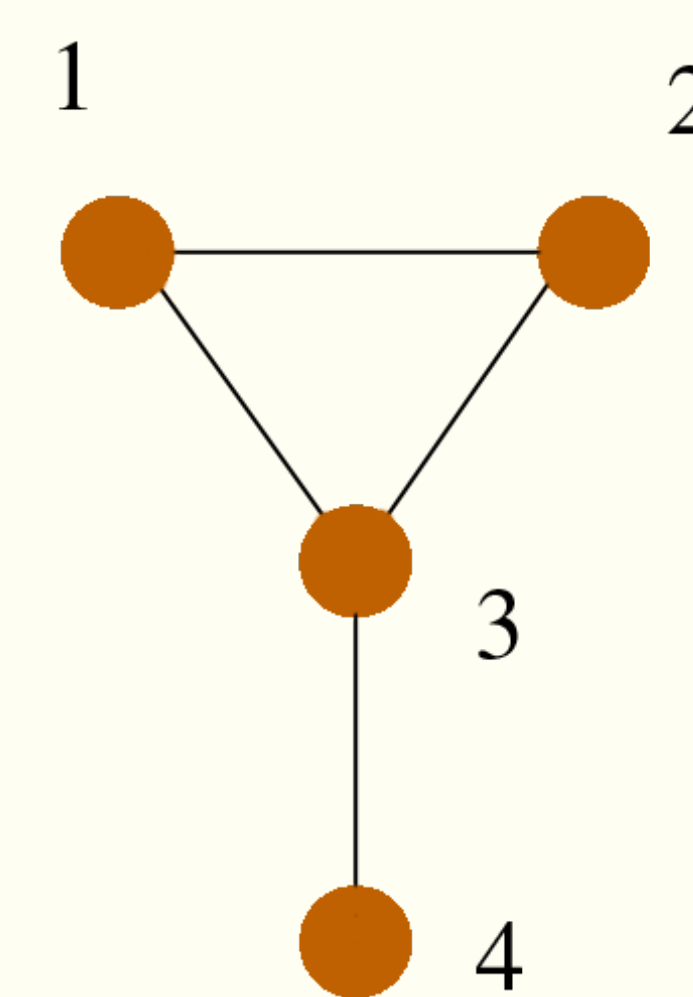
$$I(X_{\mathcal{M}}^n | \mathbf{L}) \triangleq \frac{1}{|\mathcal{P}^*| - 1} \left[\sum_{A \in \mathcal{P}^*} H(X_A^n | \mathbf{L}) - H(X_{\mathcal{M}}^n | \mathbf{L}) \right].$$

- ▶ Remark: $I(X_{\mathcal{M}}^n | \mathbf{L}) = 0$ implies conditional independence of $(X_A^n)_{A \in \mathcal{P}^*}$ given \mathbf{L} .

Evaluating $\text{CI}(\mathcal{X}_{\mathcal{M}})$: The Hypergraphical Source

- ▶ Consider a **hypergraph** $\mathcal{H} = (\mathcal{V}, \mathcal{E})$.
- ▶ $\mathcal{V} = \mathcal{M}$.
- ▶ Associate with each hyperedge $e \in \mathcal{E}$ an i.i.d. sequence of n Bernoulli $(1/2)$ random variables ξ_e^n .
- ▶ Random variables associated with distinct hyperedges in \mathcal{E} are independent.
- ▶ Define a multiterminal source as follows: $X_i^n = (\xi_e^n : e \in \mathcal{E} \text{ such that } i \in e)$.
- ▶ The multiterminal source $X_{\mathcal{M}}^n$ is known as the **hypergraphical source**.
- ▶ **Result**: For a hypergraphical source $\text{CI}(\mathcal{X}_{\mathcal{M}}) = |\mathcal{E}_{\mathcal{P}^*}|$, where $\mathcal{E}_{\mathcal{P}^*}$ is the set of hyperedges intersecting with at least two parts of \mathcal{P}^* .

The Lower Bound is Loose



- ▶ Consider the following binary hypergraphical model.
 - ▷ $m = 4$ and $\mathcal{E} = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{3, 4\}\}$.
 - ▷ $\mathcal{P}^* = \{\{1, 2, 3\}, \{4\}\}$ and $I(\mathcal{X}_{\mathcal{M}}) = 1$.
- ▶ Therefore, $\text{CI}(\mathcal{X}_{\mathcal{M}}) = 1$ and hence, $\text{CI}(\mathcal{X}_{\mathcal{M}}) - I(\mathcal{X}_{\mathcal{M}}) = 0$.
- ▶ However, $R_{\text{SK}} > 0$ as (X_1, X_2) is independent of X_4 .

Results on R_{SK} -maximality

- ▶ **Result**: A multiterminal source $X_{\mathcal{M}}$ with fundamental partition \mathcal{P}^* is R_{SK} -maximal if for all $A \in \mathcal{P}^*$ we have $H(X_A | X_{A^c}) = 0$.
- ▶ **Result**: A hypergraphical source $\mathcal{H} = (\mathcal{M}, \mathcal{E})$ is R_{SK} -maximal iff $\mathcal{E} = \mathcal{E}_{\mathcal{P}^*}$.

Examples of R_{SK} -maximal Sources

- ▶ Hypergraphical source defined on the **complete t -uniform hypergraph $K_{m,t}$** : $\mathcal{V} = \mathcal{M}$. \mathcal{E} is the set of all t -subsets of \mathcal{M} .
- ▶ Hypergraphical source defined on **Harary graphs**: These are k -regular, k -edge-connected graphs (i.e., all hyperedges are of size 2).
- ▶ Hypergraphical source defined on **Steiner Triple Systems**: $\mathcal{V} = \mathcal{M}$. \mathcal{E} consists of 3-subsets of \mathcal{M} such that every $\{i, j\}, i, j \in \mathcal{M}, i \neq j$, is a subset of exactly one $e \in \mathcal{E}$.