

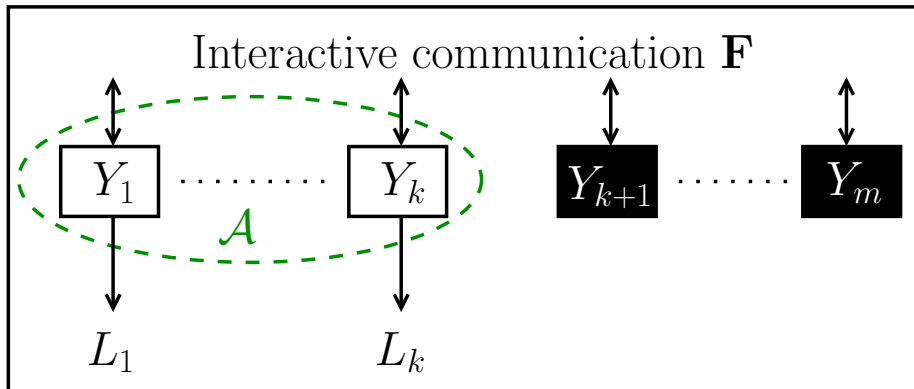
How Many Queries Will Resolve Common Randomness?

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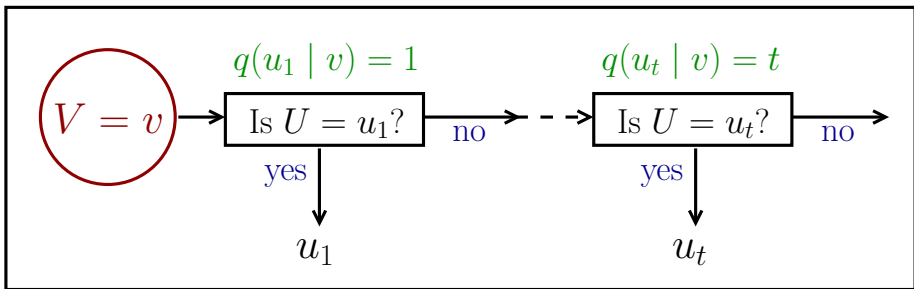
Common Randomness



Definition. L is an ϵ -common randomness for \mathcal{A} from \mathbf{F} if

$$\mathbb{P}(L = L_i(Y_i, \mathbf{F}), i \in \mathcal{A}) \geq 1 - \epsilon$$

Query Strategy



Query strategy for U given V

Massey '94, Arikan '96, Arikan-Merhav '99, Hanawal-Sundaresan '11

Query Strategy

Given rvs U, V with values in the sets \mathcal{U}, \mathcal{V} .

Definition. A **query strategy** q for U given $V = v$ is a bijection

$$q(\cdot|v) : \mathcal{U} \rightarrow \{1, \dots, |\mathcal{U}|\},$$

where the querier, upon observing $V = v$, asks the question

“Is $U = u$?”

in the $q(u|v)^{\text{th}}$ query.

$q(U|V)$: random query number for U upon observing V

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$$|\{u : q(u | v) < \gamma\}| < \gamma$$

Optimum Query Exponent

$Y_i = (X_{i1}, \dots, X_{in}) = X_i^n, \quad 1 \leq i \leq m$: i.i.d. observations

Definition. $E \geq 0$ is an ϵ -achievable *query exponent* if there exists ϵ -CR L_n for \mathcal{A} from \mathbf{F}_n such that

$$\sup_q \mathbb{P} (q(L_n | \mathbf{F}_n) < 2^{nE}) \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

where the \sup is over every query strategy for L_n given \mathbf{F}_n .

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$$E^*(\epsilon) \triangleq \sup \{ E : E \text{ is an } \epsilon\text{-achievable query exponent} \}$$

$$E^* \triangleq \inf_{0 < \epsilon < 1} E^*(\epsilon) : \text{optimum query exponent}$$

Main Result

Theorem

For $0 < \epsilon < 1$, the optimum query exponent E^* equals

$$E^* = E^*(\epsilon) = H(X_{\mathcal{M}}) - \max_{\lambda \in \Lambda(\mathcal{A})} \sum_{B \in \mathcal{B}} \lambda_B H(X_B | X_{B^c}).$$

$$\mathcal{B} = \{B \subsetneq \mathcal{M} : B \neq \emptyset, \mathcal{A} \not\subseteq B\}$$

$\Lambda(\mathcal{A}) =$ set of all $\{\lambda_B \in [0, 1] : B \in \mathcal{B}\}$ such that

$$\sum_{B \in \mathcal{B} : B \ni i} \lambda_B = 1, \quad i \in \mathcal{M}$$

$\lambda \in \Lambda(\mathcal{A})$ is a *fractional partition* of \mathcal{M}

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For $m = 2$: The expression on the right = $I(X_1 \wedge X_2)$

Strong Converse for Secret Key Capacity

Secret Key Capacity

Definition. $C(\epsilon)$ is the supremum over rates of rv $K \in \mathcal{K}$ s.t.

- (i) K is an ϵ -CR for \mathcal{A} from \mathbf{F}
- (ii) K is almost independent of \mathbf{F} :

$$n_{\text{Svar}}(K; \mathbf{F}) = n \left\| P_{K, \mathbf{F}} - U_{\mathcal{K}} \times P_{\mathbf{F}} \right\|_1 \rightarrow 0$$

Secret key capacity C is defined as $\inf_{0 < \epsilon < 1} C(\epsilon)$

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Optimum Query Exponent and SK Capacity

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For $0 < \epsilon < 1$, the optimum query exponent E^ equals*

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Achievability: $E^*(\epsilon) \geq C(\epsilon)$ - Easy

Converse: $E^*(\epsilon) \leq C$ - Main contribution

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Theorem (Strong converse for SK capacity)

For $0 < \epsilon < 1$, the ϵ -SK capacity is given by

$$C(\epsilon) = E^* = C.$$

Proof of Achievability

Query Strategies and Conditional Probabilities

Lemma. The rvs U, V , satisfy

$$P \left(\left\{ (u, v) : P_{U|V}(u|v) \leq \frac{1}{\gamma} \right\} \right) \approx 1. \quad (*)$$

Then for every query strategy q for U given V ,

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► Proof of $C(\epsilon) \leq E^*(\epsilon)$

$U = \text{SK of rate } R, V = \mathbf{F} \Rightarrow (*) \text{ holds with } \gamma \approx 2^{nR}$

Proof of Converse

Proof of Converse for $\mathcal{A} = \mathcal{M}$

Alternative Expression for C when $\mathcal{A} = \mathcal{M}$

[Csiszár-Narayan '04] observed that for $\mathcal{A} = \mathcal{M}$

$$C \leq \frac{1}{k-1} D \left(P_{X_{\mathcal{M}}} \left\| \prod_{i=1}^k P_{X_{\pi_i}} \right. \right),$$

for every partition $\pi = \{\pi_1, \dots, \pi_k\}$ of \mathcal{M} .

Alternative Expression for C when $\mathcal{A} = \mathcal{M}$

[Chan-Zheng '10] showed that for $\mathcal{A} = \mathcal{M}$

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We shall show

$$E^*(\epsilon) \leq E_{\pi} = \frac{1}{|\pi| - 1} D \left(P_{X_{\mathcal{M}}} \left\| \prod_{i=1}^{|\pi|} P_{X_{\pi_i}} \right. \right), \quad \text{for every } \pi.$$

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Roughly: For an ϵ -CR L for \mathcal{M} from \mathbf{F} , there exists q_0 s.t.

$$P(q_0(L | \mathbf{F}) \leq 2^{nE_{\pi}}) > 0, \quad \text{for every } \pi$$

A General Converse

For rvs Y_1, \dots, Y_k , let L be an ϵ -CR for $\{1, \dots, k\}$ from \mathbf{F} .

Theorem

Let θ be such that

$$\mathbb{P} \left(\left\{ (y_1, \dots, y_k) : \frac{\mathbb{P}_{Y_1, \dots, Y_k}(y_1, \dots, y_k)}{\prod_{i=1}^k \mathbb{P}_{Y_i}(y_i)} \leq \theta \right\} \right) \approx 1.$$

Then, there exists a query strategy q_0 for L given \mathbf{F} such that

$$\mathbb{P} \left(q_0(L \mid \mathbf{F}) \lesssim \theta^{\frac{1}{k-1}} \right) \geq (1 - \sqrt{\epsilon})^2 > 0.$$

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► Proof of $E^*(\epsilon) \leq E_\pi$

Choose $Y_i = X_{\pi_i}^n$ for $i \in \{1, \dots, k = |\pi|\}$.

Proof Outline for the General Converse

We show: \exists a subset \mathcal{I}_0 of values of \mathbf{F} and sets $\mathcal{L}(\mathbf{i}) \subseteq \mathcal{L}$ s.t.

$$|\mathcal{L}(\mathbf{i})| \lesssim \theta^{\frac{1}{k-1}} \quad \text{and} \quad P_{L|\mathbf{F}}(\mathcal{L}(\mathbf{i}) | \mathbf{i}) > 0, \quad \mathbf{i} \in \mathcal{I}_0$$
$$P_{\mathbf{F}}(\mathcal{I}_0) > 0$$

Lossless Data Compression:

Find small cardinality sets with large $P_{L|\mathbf{F}}$ probabilities

Small Cardinality Sets with Large Probabilities

Rényi entropy of order α of a probability measure μ on \mathcal{U} :

$$H_\alpha(\mu) \triangleq \frac{1}{1-\alpha} \log \sum_{u \in \mathcal{U}} \mu(u)^\alpha, \quad 0 \leq \alpha \neq 1$$

Lemma. There exists a set $\mathcal{U}_\delta \subseteq \mathcal{U}$ with $\mu(\mathcal{U}_\delta) \geq 1 - \delta$ s.t.

$$|\mathcal{U}_\delta| \lesssim \exp(H_\alpha(\mu)), \quad 0 \leq \alpha < 1.$$

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Conversely, for any set $\mathcal{U}_\delta \subseteq \mathcal{U}$ with $\mu(\mathcal{U}_\delta) \geq 1 - \delta$,

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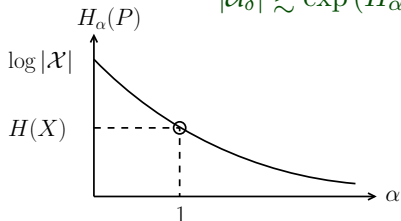
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Digression: Lossless Source Coding

Given probability measures μ_n on finite sets \mathcal{U}_n , $n \geq 1$.

$$R^*(\delta) \triangleq \inf\{R : \mu_n(\mathcal{V}_n) \geq 1 - \delta, \limsup(1/n) \log |\mathcal{V}_n| \leq R\}$$

Proposition. For each $0 < \delta < 1$,

$$\lim_{\alpha \downarrow 1} \limsup_n \frac{1}{n} H_\alpha(\mu_n) \leq R^*(\delta) \leq \lim_{\alpha \uparrow 1} \limsup_n \frac{1}{n} H_\alpha(\mu_n).$$

If μ_n is an i.i.d. probability measure on $\mathcal{U}_n = \mathcal{U}^n$, then

$$R^*(\delta) = H(\mu_1), \quad 0 < \delta < 1.$$

Abstract Alphabet and Communication

Let θ be such that

$$\mathbb{P} \left(\left\{ y^k : \frac{d P_{Y_1, \dots, Y_k}(y^k)}{d \prod_{i=1}^k P_{Y_i}}(y^k) \leq \theta \right\} \right) \approx 1.$$

Then, there exists a query strategy q_0 for L given \mathbf{F} such that

$$\mathbb{P} \left(q_0(L | \mathbf{F}) \lesssim \theta^{\frac{1}{k-1}} \right) > 0.$$

- ▶ Upper bound on $E^*(\epsilon)$ for jointly Gaussian rvs
- ▶ Strong converse for Gaussian secret key capacity

Summary

Main Result: $E^* = E^*(\epsilon) = C(\epsilon) = C$

- ▶ Largest rate SK makes the task of querying eavesdropper the most onerous.
- ▶ We proved a strong converse for the SK capacity,
- ▶ And a converse for general alphabet and communication.
- ▶ Rényi entropy can be interpreted as an answer to a lossless source coding problem.