

On the Estimation of Joint Mutual Information for Physical Layer Security

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Course Project: ECEN 670

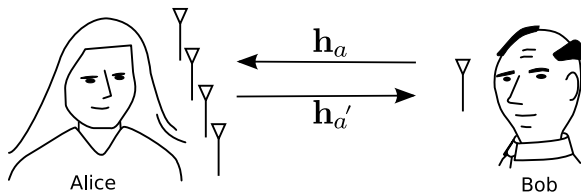
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Physical Layer Security



- ▶ The reciprocal channel ($h_a = h_{a'}$) can be used to generate the secret key.
- ▶ Requires channel to be changing with time (fading).

Available Key bits (I_K)

- ▶ How many secret key bits can be generated per observation of the channel?
- ▶ Depends on the mutual information between the two channels.
- ▶ **Mutual Information:** Amount of information shared between \hat{h}_a and $\hat{h}_{a'}$. How much information \hat{h}_a tells us about $\hat{h}_{a'}$ and vice versa.

$$I_K = I(\hat{h}_a; \hat{h}_{a'}) = \mathbf{E} \left\{ \log_2 \frac{f(\hat{h}_a, \hat{h}_{a'})}{f(\hat{h}_a)f(\hat{h}_{a'})} \right\}. \quad (1)$$

Available Key bits (I_K)

- ▶ Mutual Information can also be computed from Entropy.
- ▶ **Entropy**: Average information gained by observing a single variable

$$H(\hat{h}_a) = \int f(\hat{h}_a) \log f(\hat{h}_a) d\hat{h}_a. \quad (2)$$

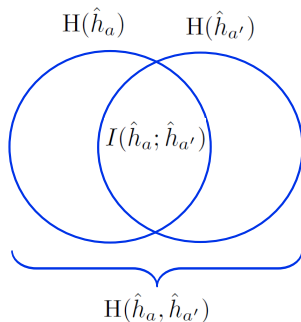
- ▶ **Joint Entropy**: Average total information gained by observing two or more variables

$$H(\hat{h}_a, \hat{h}_{a'}) = \int f(\hat{h}_a, \hat{h}_{a'}) \log f(\hat{h}_a, \hat{h}_{a'}) d\hat{h}_a d\hat{h}_{a'}. \quad (3)$$

Available Key bits (I_K)

- ▶ Mutual Information in terms of entropy is

$$I_K = H(\hat{h}_a) + H(\hat{h}_{a'}) - H(\hat{h}_a, \hat{h}_{a'}).$$

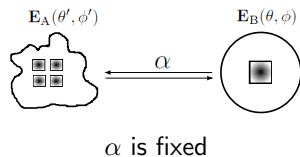
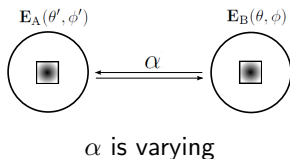


Wireless Channel

- ▶ Estimated channels at Alice and Bob are

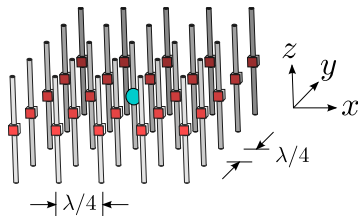
$$\hat{h}_a = \mathbf{E}_A(\theta', \phi') \alpha \mathbf{E}_B(\theta, \phi) + \epsilon_a. \quad (4)$$

$$\hat{h}_{a'} = \mathbf{E}_B(\theta, \phi) \alpha \mathbf{E}_A(\theta', \phi') + \epsilon_{a'}. \quad (5)$$



Wireless Channel using Reconfigurable Antenna

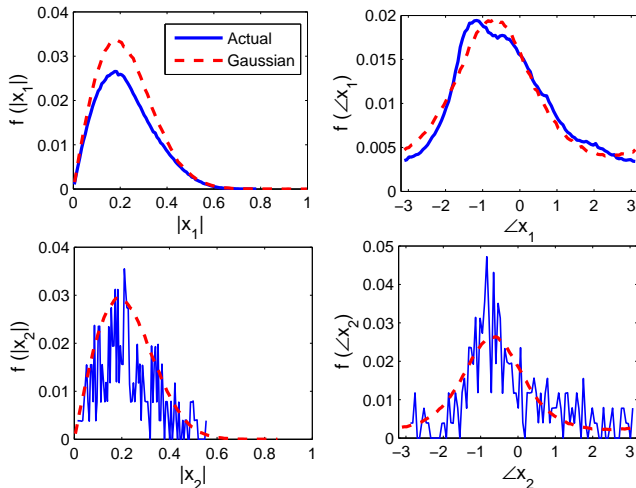
- ▶ Alice has a reconfigurable antenna.
- ▶ Each reconfigurable element (RE) is a switch.



- ▶ Channel is generated by changing the states of REs using an i.i.d uniform distribution.
- ▶ Channel distribution is unknown.

Wireless Channel using Reconfigurable Antenna

- ▶ $x_1 = h_a$ for $N_{RE} = 24$, $x_2 = h_a$ for $N_{RE} = 8$.



Available Key bits Computation

► Approaches

- Gaussian Approximation ($I_{K,GA}$)
- Numerical Computation ($I_{K,NC}$)
- Histogram based Approximation ($I_{K,HA}$)
- Gaussian Mixtures based Approximation ($I_{K,GM}$)

► Channels

- By assuming h_a is Gaussian
- By computing h_a for $N_{RE} = 24$
- By computing h_a for $N_{RE} = 8$

Gaussian Approximation

- ▶ Closed form solution of entropy exists [1]

$$\begin{aligned} I_{K,GA} &= H(\hat{h}_a) + H(\hat{h}_{a'}) - H(\hat{h}_a, \hat{h}_{a'}), \\ &= \log_2(\pi e) \sigma_{\hat{h}_a}^2 + \log_2(\pi e) \sigma_{\hat{h}_{a'}}^2 - \log_2(\pi e)^2 |\hat{\mathbf{R}}_{h_a h_{a'}}|. \end{aligned} \quad (6)$$

Channel	$I_{K,GA}$ (in bits)
Gaussian	2.5266
$N_{RE} = 24$	2.3074
$N_{RE} = 8$	2.0643

Numerical Computation

- ▶ Mutual information needs to be computed numerically [2]

$$I_{K,NC} = I(\hat{h}_a; \hat{h}_{a'}) = \mathbf{E} \left\{ \log_2 \frac{f(\hat{h}_a, \hat{h}_{a'})}{f(\hat{h}_a)f(\hat{h}_{a'})} \right\}. \quad (7)$$

- ▶ The individual pdfs $f(\hat{h}_a)$ and $f(\hat{h}_{a'})$ can be expressed in terms of the conditional pdfs

$$\begin{aligned} f(\hat{h}_a) &= \int f(\hat{h}_a|h_a)dh_a = \mathbf{E}_{h_a} f(\hat{h}_a|h_a), \\ &= \mathbf{E}_{h_a} f_n((\hat{h}_a - h_a)/\sigma_a^2). \end{aligned} \quad (8)$$

Similarly,

$$f(\hat{h}_{a'}) = \mathbf{E}_{h_a} f_n((\hat{h}_{a'} - h_a)/\sigma_{a'}^2). \quad (9)$$

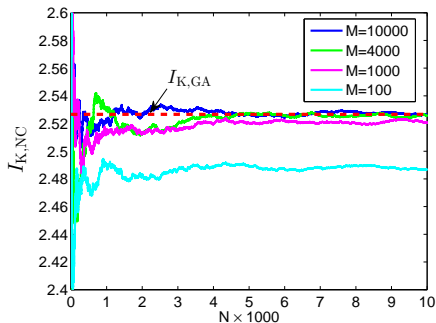
Numerical Computation

- ▶ Joint pdf $f(\hat{h}_a, \hat{h}_{a'})$ can be expressed as the product of two noise pdfs

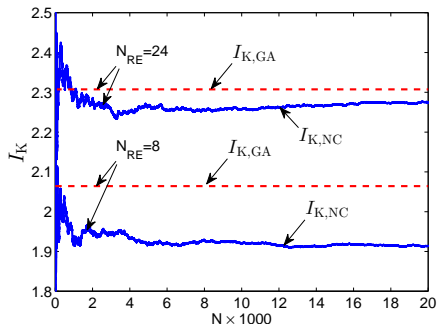
$$\begin{aligned} f(\hat{h}_a, \hat{h}_{a'}) &= \int f(\hat{h}_a, \hat{h}_{a'} | h_a) dh_a & (10) \\ &= E_{h_a} f(\hat{h}_a, \hat{h}_{a'} | h_a) = E_{h_a} \{f(\hat{h}_a | h_a) f(\hat{h}_{a'} | h_a)\} \\ &= E_{h_a} \{f_n((\hat{h}_a - h_a)/\sigma_a^2) f_n((\hat{h}_{a'} - h_a)/\sigma_{a'}^2)\}. \end{aligned}$$

- ▶ The convergence of the numerical computation will depend on N and M , which are the number of sample points in the outer and inner expectations.

Numerical Computation



Gaussian distribution for h_a



Arbitrary distribution for h_a

Histogram based Approximation

- ▶ Mutual information is computed by estimating the pdfs using multi-dimensional histograms

$$I_{K,HA} = I(\hat{h}_a; \hat{h}_{a'}) = \mathbf{E} \left\{ \log_2 \frac{f(\hat{h}_a, \hat{h}_{a'})}{f(\hat{h}_a)f(\hat{h}_{a'})} \right\}. \quad (11)$$

- ▶ Estimated channel pdfs can be expressed in terms of convolution as

$$f(\hat{h}_a) = f(h_a) * f(\epsilon_a). \quad (12)$$

$$f(\hat{h}_{a'}) = f(h_{a'}) * f(\epsilon_{a'}). \quad (13)$$

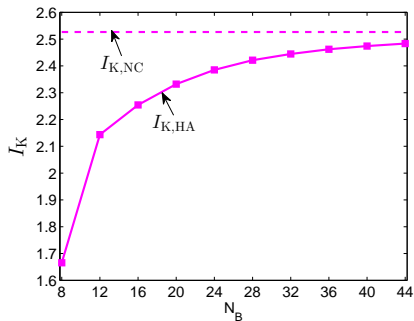
Histogram based Approximation

- ▶ The joint pdf $f(\hat{h}_a, \hat{h}_{a'})$ can be expressed as

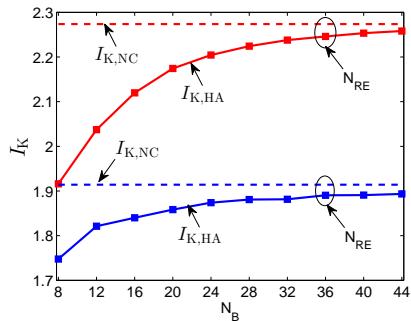
$$f(\hat{h}_a, \hat{h}_{a'}) = f(h_a, h_{a'}) * f(\epsilon_a, \epsilon_{a'}). \quad (14)$$

- ▶ 2-D and 4-D convolution and histograms are used for computing individual and joint pdfs.
- ▶ Number of bins (N_B) along each dimension is variable.

Histogram based Approximation



Gaussian distribution for h_a

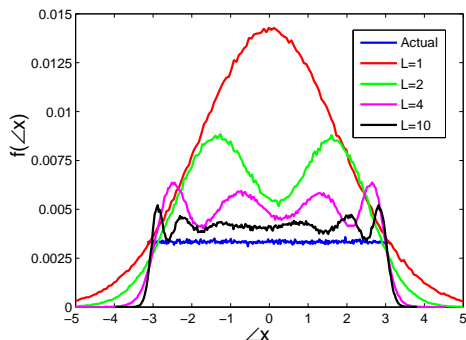


Arbitrary distribution for h_a

Gaussian Mixture based Approximation

- ▶ A given distribution can be expressed in terms of a mixture of several Gaussian distributions

$$\hat{f}(\angle x) = \sum_{i=1}^L w_i \cdot \mathcal{N}(\angle x, \mu_i, \sigma_i^2). \quad (15)$$



Gaussian Entropy Computation

- ▶ Mutual information is computed using the entropy as

$$I_{K,GM} = H(\hat{h}_a) + H(\hat{h}_{a'}) - H(\hat{h}_a, \hat{h}_{a'}). \quad (16)$$

- ▶ The entropy for a random vector \underline{x} of size P with pdf $f(\underline{x})$ is given by

$$H(\underline{x}) = \int f(\underline{x}) \cdot \log f(\underline{x}) d\underline{x}. \quad (17)$$

where

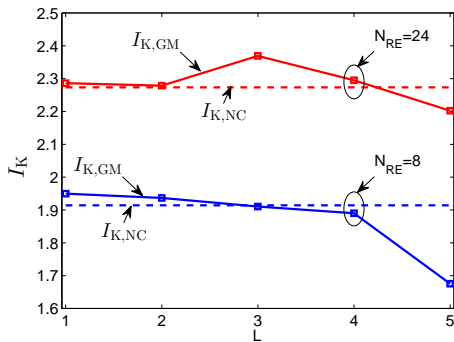
$$f(\underline{x}) = \sum_{i=1}^L w_i \cdot \mathcal{N}(\underline{x}, \underline{\mu}_i, \mathbf{C}_i). \quad (18)$$

- ▶ $\log f(\underline{x})$ is estimated using Taylor series [3]

$$\log f(\underline{x}) = \sum_{k=0}^R \frac{1}{k!} ((\underline{x} - \underline{\mu}_i) \odot \Delta)^k \log f(\underline{x})|_{\underline{x}=\underline{\mu}_i} + O_R. \quad (19)$$

Gaussian Entropy Computation

- ▶ For Gaussian distribution $I_{K,GM} = 2.5216$ when $L = 1$.






Arbitrary distribution for h_a

Conclusion

- ▶ Comparison of different techniques for arbitrary channel distribution.
- ▶ Case A: h_a is generated using $N_{RE} = 24$
- ▶ Case B: h_a is generated using $N_{RE} = 8$

Method Used	I_K for Case A	I_K for Case B	Time
Gaussian Approx	2.3074	2.0643	≤ 1
Numerical Computation	2.2737	1.9142	1920
Histogram based Approx	2.2581	1.8908	13289
Gaussian mixtures	2.2787	1.9106	209

References

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