

Some random matrix results with application to the multiple access channel

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The study of the capacity bounds for the K-user multiple access channel (MAC), reveals that certain random matrix theory results are of practical relevance to the problem. The result from the point-to-point block Rayleigh fading channel indicate that isotropically distributed input signals are capacity achieving in the high-SNR regime. The derivation of the mutual information obtained with these input signals in the MAC case requires analysis of the eigenvalues of Gram matrices of the type $\mathbf{V}\mathbf{V}^H$, where $\mathbf{V} \in \mathbb{C}^{K \times T}$ and the rows of \mathbf{V} are random vectors which are uniformly distributed on a unit sphere in \mathbb{C}^T .

System model

- K single-antenna users;
- one receiver with $N \geq K$ receive antennas;

The system model is the following:

$$\mathbf{Y} = \mathbf{S}\mathbf{X} + \mathbf{W},$$

where $\mathbf{X} \in \mathbb{C}^{K \times T}$, $\mathbf{Y} \in \mathbb{C}^{N \times T}$, $\mathbf{S} \in \mathbb{C}^{N \times K}$ with i.i.d $CN(0,1)$ entries and $\mathbf{W} \in \mathbb{C}^{N \times T}$ with i.i.d $CN(0,1)$ entries.

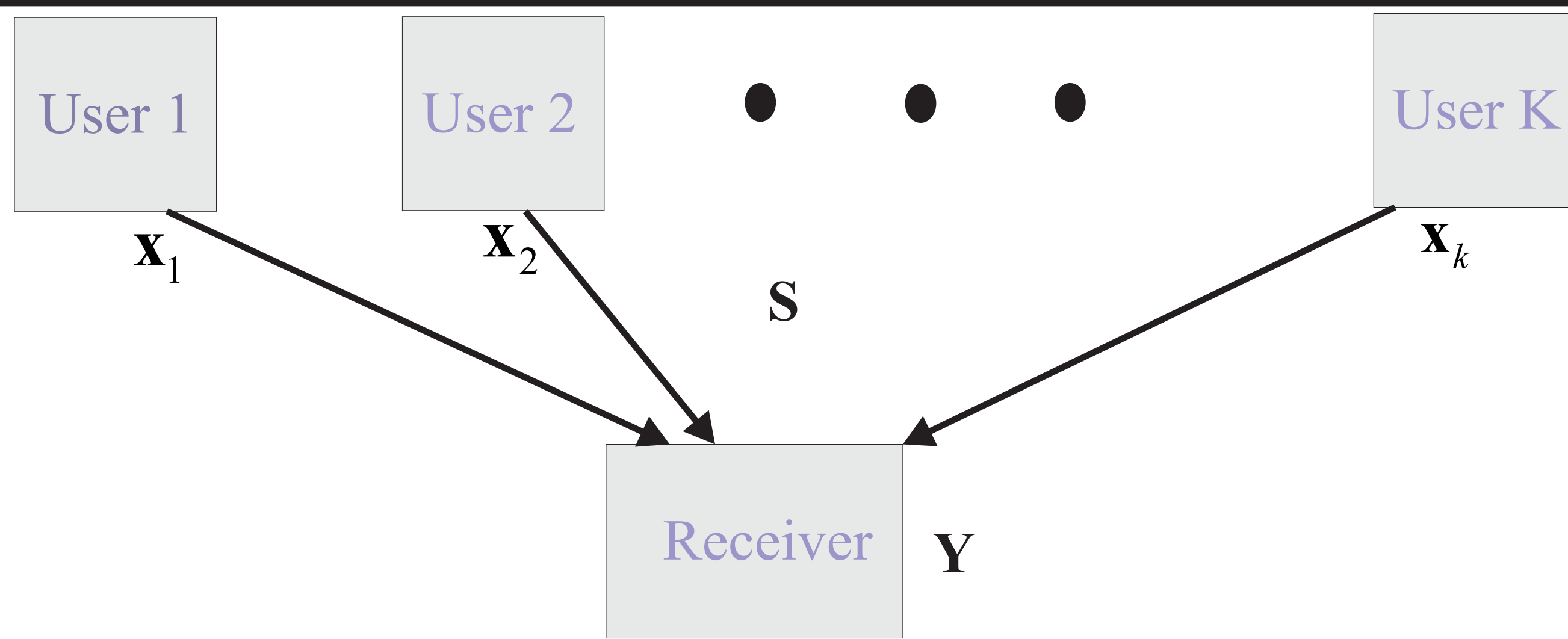


Figure 1: System model

Non-coherent communication

- The channel \mathbf{S} is unknown;
- Communication based on subspaces;
- Conjecture: The capacity achieving inputs are of the form:

$$\mathbf{x}_l = \sqrt{\frac{rT}{K}} \mathbf{v}_l,$$

where the vectors \mathbf{v}_l are independent and uniformly distributed on the unit sphere \mathbb{C}^T .

- r - SNR per receive antenna.

Computation of mutual information

We are interested in the mutual information of non-coherent MAC:

$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Y} | \mathbf{X}),$$

in high-SNR regime.

1. Derivation of $h(\mathbf{Y})$

In the high-SNR approximation, for $h(\mathbf{Y})$ we obtain:

$$\begin{aligned} h(\mathbf{Y}) \geq & (TK - K^2 + NK) \log_2 \left(\frac{rT}{K} \right) + \log_2 |G(T, K)| \\ & + (T - K) \mathbb{E} \left[\log_2 \det(\mathbf{S}^H \mathbf{S}) \right] \\ & + (T - K + N) \mathbb{E} \left[\log_2 \det(\mathbf{\Sigma} \mathbf{\Sigma}^H) \right] \\ & + (NT - KT + K^2) \log_2(pe), \end{aligned}$$

where $\mathbf{V} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}_1^H$ is the SVD decomposition.

2. Derivation of $h(\mathbf{Y} | \mathbf{X})$

For $h(\mathbf{Y} | \mathbf{X})$ we obtain:

$$\begin{aligned} h(\mathbf{Y} | \mathbf{X}) = & NK \log_2 \left(\frac{rT}{K} \right) + N \mathbb{E} [\log_2 \det(\mathbf{\Sigma} \mathbf{\Sigma}^H)] \\ & + TN \log_2(pe) \end{aligned}$$

Mutual information

For $I(\mathbf{X}; \mathbf{Y})$ we have:

$$\begin{aligned} I(\mathbf{X}; \mathbf{Y}) \geq & K \left(1 - \frac{K}{T} \right) \log_2 \frac{Tr}{K} + \frac{1}{T} \log_2 |G(T, K)| \\ & - K \left(1 - \frac{K}{T} \right) \log_2(pe) + \left(1 - \frac{K}{T} \right) \mathbb{E} \left[\log_2 \det(\mathbf{S}^H \mathbf{S}) \right] \\ & + \left(1 - \frac{K}{T} \right) \mathbb{E} \left[\log_2 \det(\mathbf{\Sigma} \mathbf{\Sigma}^H) \right]. \end{aligned}$$

Of interest is to evaluate

$$\Delta = \mathbb{E} [\log_2 \det(\mathbf{\Sigma} \mathbf{\Sigma}^H)],$$

where $\mathbf{V} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}_1^H$ is the SVD decomposition.

First, we note that $\det(\mathbf{\Sigma} \mathbf{\Sigma}^H) = \det(\mathbf{V} \mathbf{V}^H)$. We recall that the rows of $\mathbf{V} \in \mathbb{C}^{K \times T}$ are independent and uniformly distributed on the unit sphere in \mathbb{C}^T .

Using the classical QR decomposition of \mathbf{V} , we get

$$\det(\mathbf{V} \mathbf{V}^H) = \prod_{k=1}^K R_{kk}^2.$$

In the Wishart case, when the row vectors of \mathbf{V} are chosen independently from normal distribution, the variables R_{kk}^2 are independent and chi-square distributed with respective parameters $b'(T - k + 1), k = 1, \dots, K$.

In the Gram case, when the row vectors of \mathbf{V} are chosen independently and uniformly distributed on the unit sphere in \mathbb{C}^T , the variables R_{kk}^2 are independent and beta distributed with respective parameters

$(b'(T - k + 1), b'(k - 1)), k = 2, \dots, K$.

$b' = \frac{b}{2}$, where $b = 1, 2$, or 4 , which corresponds to the classical matrix models (real, complex and quaternionic).

Let we denote

$$G_{T,K} = \ln \det(\mathbf{V} \mathbf{V}^H) = \ln \prod_{k=1}^K R_{kk}^2 = \sum_{k=2}^K \ln R_{kk}^2$$

Our case corresponds to the Gram case. Using the result from the Gram case, we obtain

$$\mathbb{E}[G_{T,K}] = - \sum_{k=1}^K \sum_{i=1}^{\infty} \frac{\Gamma(T)}{\Gamma(T - k + 1) \Gamma(k - 1)} \frac{\beta(k - 1 + i, N - k + 1)}{i},$$

where $\beta(.,.)$ is beta function.

When $N \rightarrow \infty$, and the ratio K/N is fixed, $K/N = c, c \in (0, 1)$,

$$\limsup_{n \rightarrow \infty} \left| \frac{1}{K} \mathbb{E}[G_{T,K}] + J \left(1 - \frac{K}{T} \right) \log_2 e \right| = 0,$$

where

$$J(u) = u \log u - u + 1, \text{ for } u > 0$$

$$J(u) = 1, \text{ for } u = 0$$

$$J(u) = +\infty, \text{ for } u < 0$$

Finally we have

$$\Delta = \log_2 e \mathbb{E}[G_{T,K}].$$

We derive some results for the eigenvalues of Gram matrices of the type $\mathbf{V} \mathbf{V}^H$, where $\mathbf{V} \in \mathbb{C}^{K \times T}$ and the rows of \mathbf{V} are random vectors which are uniformly distributed on a unit sphere \mathbb{C}^T . These matrices are of relevance to the derivation of the mutual information of the K-user MAC, obtained with isotropically distributed unitary input signals. As result, we pave the way for the capacity characterization of the non-coherent K-user MAC in the high-SNR regime.