

# A CLT for Information-Theoretic Statistics of Gram Random Matrices

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# The Model: A Non-Centered Random Matrices

Consider a  $p \times n$  random matrices:

$$\Sigma_n = \frac{1}{\sqrt{n}} X_n + A_n,$$

where,

- ▶  $X_{nij}$ ,  $1 \leq i \leq p$ ,  $1 \leq j \leq n$  are i.i.d. centered with unit variance and  $\mathbb{E}|X_{11}|^{16} < \infty$ .
- ▶  $A_n$  is a  $p \times n$  deterministic matrix with uniformly bounded spectral norm.

# The Model: Information-Theoretic Statistics of Gram random matrices

Linear spectral statistics:

$$\mathcal{I}_n(\rho) = \frac{1}{p} \sum_{i=1}^p \log \left( \lambda_i^{(n)} + \rho \right),$$

where,  $\lambda_i^{(n)}$ ,  $i = 1, \dots, p$  are the eigenvalues of the Gram random matrix  $\Sigma_n \Sigma_n^*$  and  $\rho$  is a nonnegative parameter.

Objective: Understanding the asymptotic distribution of the fluctuations of  $\mathcal{I}_n(\rho)$ , when the dimensions of the matrix  $\Sigma_n$  converge to infinity at the same pace and obtain a simple form of the variance.

# Plan

Motivations: Mutual Information for Multiple Antenna Radio Channels

Asymptotic behavior of  $\mathcal{I}_n(\rho)$ : First-order results

- Fundamental system of equations

- Deterministic equivalents

Study of the fluctuations

- Definition of the variance

- The Central Limit Theorem

Outline of the proof of the CLT

- The approach: REFORM method

- Main steps of the proof

The bias

Outline of the proof of the bias term

Motivations:

## **Mutual Information for Multiple Antenna Radio Channels**

## Multi-user MIMO scheme

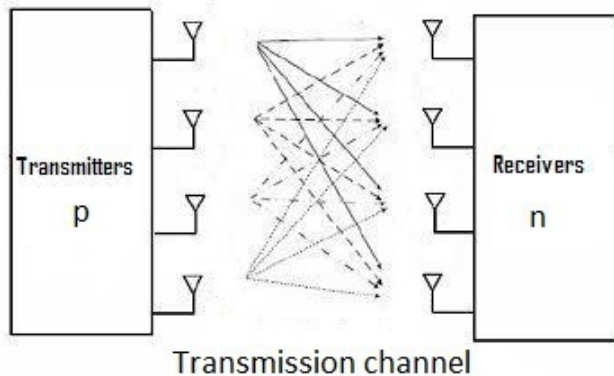


Figure: MIMO Systems

# MIMO System: Mathematical Model

The  $p$ -dimensional receiver vector  $\mathbf{r}_n$  is given by:

$$\mathbf{r}_n = \Sigma_n \mathbf{t}_n + \mathbf{b}_n,$$

where,

- ▶  $\Sigma_n$  represents the channel matrix which assumed to be **random**.
- ▶  $\mathbf{t}_n$  is the  $n$ -dimensional transmitter vector.
- ▶  $\mathbf{b}_n$  is an additive white Gaussian noise with covariance matrix  $\mathbb{E} \mathbf{b}_n \mathbf{b}_n^* = \rho I_p$ .

Performance indicator: The Mutual Information:

$$\mathcal{I}_n(\rho) = \frac{1}{p} \log \det (\Sigma_n \Sigma_n^* + \rho I_p) = \frac{1}{p} \sum_{i=1}^p \log (\lambda_i + \rho)$$

Asymptotic behavior of  $\mathcal{I}_n(\rho)$  when  $n, p \rightarrow \infty$  at the same rate ?

## Asymptotic behavior of $\mathcal{I}_n(\rho)$ : First-order results



## First-order results

Let  $f_n$  denotes the ST of  $\mu_{\Sigma_n \Sigma_n^*}$ , the spectral measure of the eigenvalues of  $\Sigma_n \Sigma_n^*$ . Then,

$$\mathcal{I}_n(\rho) = - \int_{\rho}^{\infty} f_n(-\omega) d\omega.$$

Then the asymptotic behavior of  $\mathcal{I}_n(\rho)$  is closely linked to the asymptotic behavior of  $f_n$  as  $p, n \rightarrow \infty$  with the same pace.

# State of the art

- ▶  $F^{A_n A_n^*} \rightarrow H$ ,  $H$  is a deterministic probability measure.  
Dozier and Silverstein (04):

$$F^{\Sigma_n \Sigma_n^*} \xrightarrow{\text{weakly}} F,$$

where,  $F$  is a deterministic probability measure which the Stieltjes transform is a unique solution of a given coupled equation.

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- ▶ V. L. Girko (91), Hachem-Loubaton-Najim (07) : Look for a deterministic approximation of the Stieltjes transform  $f_n$  of  $F^{\Sigma_n \Sigma_n^*}$ .  $\exists$  a  $p \times p$  deterministic valued function  $T_n(\rho)$  such that:

$$f_n(-\rho) - \frac{1}{p} \text{Tr} T_n(-\rho) \xrightarrow[n \rightarrow \infty]{a.s.} 0$$

# Fundamental equations

Theorem (Girko '91, Hachem-Loubaton-Najim '07)

*The following system of two equations*

$$\begin{cases} \delta_n(\rho) = \frac{1}{n} \operatorname{Tr} \left( \rho \left( 1 + \tilde{\delta}_n(\rho) \right) I_p + \frac{A_n A_n^*}{1 + \delta_n(\rho)} \right)^{-1} \triangleq \frac{1}{n} \operatorname{Tr} T_n(\rho) \\ \tilde{\delta}_n(\rho) = \frac{1}{n} \operatorname{Tr} \left( \rho \left( 1 + \delta_n(\rho) \right) I_n + \frac{A_n^* A_n}{1 + \tilde{\delta}_n(\rho)} \right)^{-1} \triangleq \frac{1}{n} \operatorname{Tr} \tilde{T}_n(\rho), \end{cases}$$

*admits a unique solution  $(\delta_n, \tilde{\delta}_n)$  in  $\mathcal{S}(\mathbb{R}^+)^2$ . Moreover,*

$$\int_{\mathbb{R}^+} f(\lambda) dF^{\Sigma_n \Sigma_n^*}(\lambda) - \int_{\mathbb{R}^+} f(\lambda) \pi_n(d\lambda) \xrightarrow[n \rightarrow \infty]{a.s.} 0, \quad \forall f \in \mathcal{C}_B(\mathbb{R}^+),$$

*where  $\pi_n$  is the positive measure where  $\delta_n$  is the Stieltjes transform.*

# First order result: Deterministic equivalents

## Theorem (Hachem-Loubaton-Najim '07)

Let  $V_n(\rho) = \int_{\mathbb{R}^+} \log(\lambda + \rho) \pi_n(d\lambda)$ . Then we have:

$$\mathbb{E}\mathcal{I}_n(\rho) - V_n(\rho) \xrightarrow[n, p \rightarrow \infty, \frac{p}{n} \rightarrow c > 0]{} 0.$$

Moreover,  $V_n(\rho)$  admits a closed-form expression

$$V_n(\rho) = \frac{1}{p} \sum_{i=1}^p \log \left( \rho \left( 1 + \tilde{\delta}_n \right) + \frac{\mu_{n,i}^2}{1 + \delta_n} \right) + \frac{n}{p} \log(1 + \delta_n) - \frac{\rho n}{p} \delta_n \tilde{\delta}_n,$$

where  $\mu_{n,i}$  are the singular values of the mean matrix  $A_n$ .

In the non-centered case, the first-order asymptotic study of the mutual information depends mainly on the limiting behavior of the singular values of the mean matrix  $A_n$ .

## Study of the fluctuations

## CLT for $\rho(\mathcal{I}_n(\rho) - V_n(\rho))$

In order to study the CLT for  $\rho(\mathcal{I}_n(\rho) - V_n(\rho))$  we study separately two quantities:

- ▶ The random quantity  $\rho(\mathcal{I}_n(\rho) - \mathbb{E}\mathcal{I}_n(\rho))$  from which the fluctuations arise and,
- ▶ The deterministic quantity  $\rho(\mathbb{E}\mathcal{I}_n(\rho) - V_n(\rho))$  which yields a bias.



# Asymptotic distribution of the fluctuations: Definition of the variance

Theorem (Hachem-Kharouf-Najim-Silverstein '10)

Let  $\vartheta = \mathbb{E}X_{11}^2$ ,  $\kappa = \mathbb{E}|X_{11}|^4 - 2 - \vartheta^2$  and let

$\gamma = \frac{1}{n} \text{Tr } T^2$ ,  $\tilde{\gamma} = \frac{1}{n} \text{Tr } \tilde{T}^2$ ,  $\underline{\gamma} = \frac{1}{n} \text{Tr } T \bar{T}$ ,  $\tilde{\underline{\gamma}} = \frac{1}{n} \text{Tr } \tilde{T} \tilde{\bar{T}}$ . Denote by

$$\Theta_n^2 = -\log \left( \left( 1 - \frac{1}{n(1+\tilde{\delta})} \text{Tr } TAA^*T \right)^2 - \rho^2 \gamma \tilde{\gamma} \right) \\ - \log \left( \left| 1 - \vartheta \frac{1}{n(1+\tilde{\delta})} \text{Tr } \bar{T} \bar{A} A^* T \right|^2 - |\vartheta|^2 \rho^2 \underline{\gamma} \tilde{\underline{\gamma}} \right) \\ + \kappa \frac{\rho^2}{n^2} \sum_i t_{ii}^2 \sum_j \tilde{t}_{jj}^2$$

Then  $\Theta_n^2$  is well defined.

## Some remarks

- ▶ The variance is the sum of tree terms: the first term would be the same in the Gaussian case.
- ▶ The variance depends on the singular values of the main matrix as well as on its singular vectors.
- ▶ In the circular case ( $X_{ij} \stackrel{\mathcal{D}}{=} X_{ij} e^{i\alpha}$  for all  $\alpha$ ), the second term disappears.

# Asymptotic distribution of the fluctuations: The CLT

Theorem (Hachem-Kharouf-Najim-Silverstein '10)

*The following convergence holds true:*

$$\frac{P}{\Theta_n} (\mathcal{I}_n(\rho) - \mathbb{E}\mathcal{I}_n(\rho)) \xrightarrow[\rho, n \rightarrow \infty]{\mathcal{D}} \mathcal{N}(0, 1),$$

*where  $\mathcal{D}$  stands for convergence in distribution.*

# Proof of the CLT: The approach

**REFORM** (**RE**solvent **FOR**mula and **M**artingale).

- ▶  $\mathcal{I}_n(\rho) - \mathbb{E}\mathcal{I}_n(\rho)$  as a sum of increments of martingale.
- ▶ Identification of the variance.

# CLT for martingales

## Theorem

Let  $\Gamma_1^{(n)}, \dots, \Gamma_n^{(n)}$  be a sequence of increments of martingale with respect to a given filtration  $\mathcal{F}_1^{(n)}, \dots, \mathcal{F}_n^{(n)}$ . Assume that there exists a sequence of nonnegative real numbers  $(\Theta_n^2)_n$  uniformly bounded away from zero and from infinity. Assume that:



$$\sum_{j=1}^n \mathbb{E} \left( \Gamma_j^{(n)^2} \mid \mathcal{F}_{j-1}^{(n)} \right) - \Theta_n^2 \xrightarrow[n \rightarrow \infty]{\mathcal{P}} 0.$$

▶ The Lyapunov's condition

$$\exists \alpha > 0, \quad \frac{1}{\Theta_n^{2(1+\alpha)}} \sum_{j=1}^n \mathbb{E} |\Gamma_j^{(n)}|^{2+\alpha} \xrightarrow[n \rightarrow \infty]{} 0, \quad \text{holds.}$$

Then  $\Theta_n^{-1} \sum_{j=1}^n \Gamma_j^{(n)}$  converges in distribution to  $\mathcal{N}(0, 1)$ .

## Sum of martingale differences

We have,

$$\mathcal{I}_n - \mathbb{E}\mathcal{I}_n = \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) (-\log(1 + \xi_j)) \triangleq \sum_{j=1}^n \Gamma_j,$$

where,

$$\xi_j = \frac{\eta_j^* Q_j \eta_j - \left( \frac{1}{n} \text{Tr} Q_j + a_j^* Q_j a_j \right)}{1 + \frac{1}{n} \text{Tr} Q_j + a_j^* Q_j a_j}.$$

with  $\eta_j, a_j$  are resp. the  $j$ th columns of matrices  $\Sigma_n$  and  $A_n$ ,  $Q_j$  is the resolvent of the matrix  $\Sigma_j \Sigma_j^*$  and  $\mathbb{E}_j$  stands for the conditional expectation with respect to the  $\sigma$ -algebra  $\mathcal{F}_j^{(n)} = \sigma(x_1, \dots, x_j)$ .

## Sum of the conditional variances

Some properties of the function  $\log$ ,

$$\sum_{j=1}^n \mathbb{E}_{j-1} ((\mathbb{E}_j - \mathbb{E}_{j-1}) \log(1 + \xi_j))^2 - \sum_{j=1}^n \mathbb{E}_{j-1} (\mathbb{E}_j \xi_j)^2 \xrightarrow[\mathcal{P}, n \rightarrow \infty]{} 0$$

where (recall)

$$\xi_j = \frac{\eta_j^* Q_j \eta_j - \left( \frac{1}{n} \text{Tr} Q_j + a_j^* Q_j a_j \right)}{1 + \frac{1}{n} \text{Tr} Q_j + a_j^* Q_j a_j}.$$

# Study of the sum of conditional variances

Standard calculations remain the problem to the study of the asymptotic behavior of the quantities:

$$\frac{1}{n} \text{Tr} (\mathbb{E}_j Q_n)^2 \quad \text{and} \quad a_j^* (\mathbb{E}_j Q_n)^2 a_j,$$

where  $Q_n$  is the resolvent of  $\Sigma_n \Sigma_n^*$  matrix.



## Outline of the proof

A good comprehension of the asymptotic behavior of these terms requires a specific study of bilinear forms of type  $u_n^* Q(\rho) v_n$  where at least  $u_n$  or  $v_n$  is a given column of the deterministic mean matrix  $A_n$ .

If  $u_n$  and  $v_n$  are deterministic, Hachem-Loubaton-Najim-Vallet (preprint'10)

$$u_n^* Q(\rho) v_n \approx u_n^* T(\rho) v_n$$

# Asymptotic behavior of the bias:

Theorem (Hachem-Kharouf-Najim-Silverstein '10)

We have,

$$\rho(\mathbb{E}\mathcal{I}_n(\rho) - V_n(\rho)) - \mathcal{B}_n(\rho) \xrightarrow{\rho, n \rightarrow \infty} 0$$

where,

$$\mathcal{B}_n(\rho) = \kappa \text{Cte}(\rho, \delta, \tilde{\delta})$$

$$\kappa = \mathbb{E}|X_{11}|^4 - 2 - \vartheta^2.$$

# Outline of the proof of the bias term

The bias term is given by

$$\begin{aligned}\chi_n(\rho) &= p(\mathbb{E}\mathcal{I}_n(\rho) - V_n(\rho)) \\ &= p \int_{\rho}^{\infty} \frac{d}{d\omega} \mathbb{E} \log \det (\Sigma_n \Sigma_n^* + \omega I_p) d\omega \\ &\quad - p \int_{\rho}^{\infty} \frac{d}{d\omega} \left( \int_{\mathbb{R}^+} \log(\lambda + \omega) \pi_n(d\lambda) \right) d\omega \\ &= \int_{\rho}^{\infty} \text{Tr} (\mathbb{E}Q_n(\omega) - T_n(\omega)) d\omega.\end{aligned}$$

Then it remains to study the asymptotic behavior of  $\text{Tr} (\mathbb{E}Q_n(\omega) - T_n(\omega))$ . We prove,

$$\text{Tr} (\mathbb{E}Q_n(\omega) - T_n(\omega)) - \kappa \text{Cte}(\rho, \delta, \tilde{\delta}) \xrightarrow[n \rightarrow \infty]{} 0$$

Case of a non-centered separable random matrix model

## The non-centered separable case

$$\Sigma_n = \frac{1}{\sqrt{n}} D_n^{1/2} X_n \tilde{D}_n^{1/2} + A_n,$$

where,  $D_n^{1/2}$  and  $\tilde{D}_n^{1/2}$  are resp.  $p \times p$  and  $n \times n$  deterministic diagonal matrices with nonnegative entries.

First-order asymptotic behavior

$$V_n(\rho) = \frac{1}{p} \log \det T_n^{-1}(\rho) + \frac{1}{p} \log \left( I_n + \delta_n \tilde{D}_n \right) - \frac{\rho n}{p} \delta_n \tilde{\delta}_n,$$

where,  $\delta_n = \frac{1}{n} \text{Tr } T(\rho)$  and  $\tilde{\delta}_n(\rho) = \frac{1}{n} \text{Tr } \tilde{T}(\rho)$ , with

$$T_n(\rho) = \left( \rho \left( I_p + \tilde{\delta}_n D_n \right) + A_n \left( I_n + \delta_n \tilde{D}_n \right)^{-1} A_n^* \right)^{-1}$$
$$\tilde{T}_n(\rho) = \left( \rho \left( I_n + \delta_n \tilde{D}_n \right) + A_n^* \left( I_p + \tilde{\delta}_n D_n \right)^{-1} A_n \right)^{-1}$$

# The non-centered separable case

The variance:

$$\Theta_n^2 = -\log(\Omega_n(\rho) - \rho^2 \gamma \tilde{\gamma}) - \log(\bar{\Omega}_n(\rho) - |\vartheta|^2 \rho^2 \underline{\gamma} \tilde{\gamma}) \\ + \kappa \frac{\rho^2}{n^2} \sum_i d_i^2 t_{ii}^2 \sum_j \tilde{d}_j^2 \tilde{t}_{jj}^2$$

where:

$$\Omega_n(\rho) = \left( 1 - \frac{1}{n} \text{Tr} D_n^{1/2} T_n A_n \left( I_n + \delta \tilde{D}_n \right)^{-1} \tilde{D}_n \left( I_n + \delta \tilde{D}_n \right)^{-1} A_n^* T_n D_n^{1/2} \right)$$

and

$$\bar{\Omega}_n(\rho) = \left| 1 - \vartheta \frac{1}{n} \text{Tr} D_n^{1/2} \bar{T}_n \bar{A}_n \left( I_n + \delta \tilde{D}_n \right)^{-1} \tilde{D}_n \left( I_n + \delta \tilde{D}_n \right)^{-1} A_n^* T_n D_n^{1/2} \right|$$

Thank you !