

Random Matrix Theory for MIMO Communications

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Technische Universität München (TUM)

Technical University of Munich

- Established in 1868
- Located in Munich, Bavaria, Germany
- 12 departments → Electrical Engineering and Information Technology
- Approximately 21,600 students (2006)
- 4,160 academic staff-395 Professors
- Consistent ranking amongst the best universities in Germany (DAAD)



Lehrstuhl für Netzwerktheorie und Signalverarbeitung (NWS)

Institute for Circuit Theory and Signal Processing

- Head: Professor Josef A. Nossek
- Webpage: www.nws.ei.tum.de
- **Research areas: Consistent modeling of physical layer**
 - Processing of quantized signals
 - Cellular systems
 - Filter-bank based multicarrier systems
 - Cross-layer optimization
 - Array processing for multipath and interference mitigation for Global Navigation Satellite Systems



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 - Random matrix theory for wireless communications



Research collaborators-Some wonderful people!!

- Prof. Akbar M. Sayeed (University of Wisconsin-Madison)
- Prof. Peter J. Smith (University of Canterbury, New Zealand)
- Dr. Matthew R. McKay (Hong Kong University of Science and Technology, Hong Kong)
- Prof. George K. Karagiannidis (Aristotle University of Thessaloniki, Greece)
- Dr. David I. Laurenson (University of Edinburgh, UK)
- Dr. Cheng-Xiang Wang (University of Heriot-Watt, UK)



Presentation outline

- Overview of MIMO technology
- General applications of random matrix theory (RMT) to MIMO communications
- Generic framework for the standard condition number (SCN) distribution of Wishart matrices
- Conclusions
- Future research - Open problems



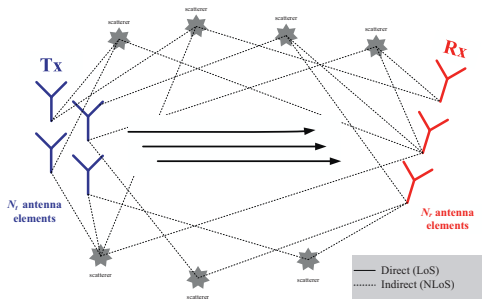
Overview of MIMO technology

- Developed by Telatar and Foschini in mid-90s
- Multiple antenna elements at both the transmitter/receiver
- Array (beamforming) gain, spatial diversity
- **Additional selectivity domain (spatial) → multiplexing gains**



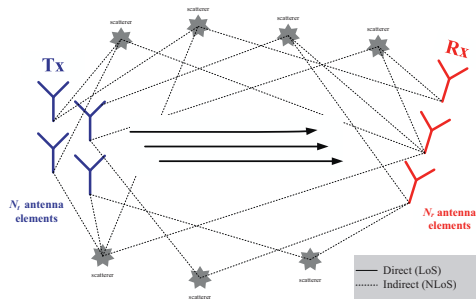
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- High degree of multipath activity (i.i.d. Rayleigh fading) → linear capacity increase with $s = \min(N_t, N_r)$

Mathematical background

- MIMO system with N_t transmit and N_r receive antennas
- $s = \min(N_t, N_r)$ and $t = \max(N_t, N_r)$
- Complex input-output relationship

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{n} \quad (1)$$

- MIMO channel matrix response $\mathbf{H} \in \mathbb{C}^{N_r \times N_t}$



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$$\mathbf{W} = \begin{cases} \mathbf{H}\mathbf{H}^\dagger, & \text{if } N_r \leq N_t \\ \mathbf{H}^\dagger\mathbf{H}, & \text{if } N_r > N_t \end{cases} \quad (2)$$



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is Hermitian, positive semi-definite and **random**.



Applications of RMT to MIMO communications - Example 1

- MIMO ergodic capacity assuming uniform power allocation

$$C = \mathcal{E} \left[\log_2 \left(\det \left(\mathbf{I}_{N_r} + \frac{\text{SNR}}{N_t} \mathbf{H} \mathbf{H}^\dagger \right) \right) \right] \quad (3)$$

$$= \mathcal{E} \left[\sum_{k=1}^s \log_2 \left(1 + \frac{\text{SNR}}{N_t} \lambda_k \right) \right] \quad (4)$$

$$= \int_0^\infty \log_2 \left(1 + \frac{\text{SNR}}{N_t} \lambda \right) p(\lambda) d\lambda \quad (5)$$

where $\lambda_1 \geq \lambda_2 \geq \dots \lambda_s \geq 0$ are the real, non-negative, ordered eigenvalues of \mathbf{W} .

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- High-SNR ergodic capacity ($\text{SNR} \rightarrow \infty$)

$$C = s \log_2 (\text{SNR}/N_t) + \frac{1}{\ln 2} \mathcal{E} \left[\ln(\det(\mathbf{H} \mathbf{H}^\dagger)) \right] \quad (6)$$

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- Transmitted symbol x , $E[|x|^2] = 1$
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- SNR at the output of combiner

$$\gamma = \bar{\gamma} \mathbf{w}^\dagger \mathbf{H}^\dagger \mathbf{H} \mathbf{w} \quad (9)$$

$$= \bar{\gamma} \mathbf{w}_{opt}^\dagger \mathbf{H}^\dagger \mathbf{H} \mathbf{w}_{opt} = \bar{\gamma} \lambda_1 \quad (10)$$

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Derive analytical closed-form expressions for the most important MIMO features

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Derive analytical closed-form expressions for the most important MIMO features

- Ergodic/outage capacity
- Higher-order capacity moments (variance, skewness,...)
- Symbol error rate (SER)-Outage probability of SM-MIMO
- Singular value decomposition (SVD) MIMO
- Performance of beamforming/MRC schemes
- Asymptotic characterization of MIMO systems (i.e. number of antennas grows infinitely large)



MIMO fading models and Wishart matrices

- What is the most appropriate model for the channel fading statistics?
- We need something simple and realistic at the same time!
- Physical measurement campaigns and theoretical studies have demonstrated that the channel statistics can be efficiently modeled via the **complex normal distribution**.



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In any case, the correlation matrix \mathbf{W} follows a complex Wishart distribution. This is some very good news for MIMO people!



MIMO fading models and Wishart matrices

Uncorrelated Rayleigh fading MIMO model

$$\mathbf{H} = \mathbf{H}_w \quad (11)$$

where \mathbf{H}_w is an $(N_r \times N_t)$ matrix whose entries are complex zero-mean unity variance RVs, i.e. $\sim \mathcal{CN}(0, 1)$.

- The matrix \mathbf{W} is **uncorrelated central Wishart** with t degrees of freedom (DoF), $\mathbf{W} \sim \mathcal{CW}_s(t, \mathbf{I}_s)$.



MIMO fading models and Wishart matrices

- Semi-correlated Rayleigh fading MIMO model

$$\mathbf{H} = \begin{cases} \boldsymbol{\Sigma}_s^{1/2} \mathbf{H}_w, & \text{if } N_r \leq N_t \\ \mathbf{H}_w \boldsymbol{\Sigma}_s^{1/2}, & \text{if } N_r > N_t. \end{cases} \quad (12)$$

with $\boldsymbol{\Sigma}_s \in \mathbb{C}^{s \times s}$ being a positive definite matrix containing the variances of the entries of \mathbf{H} on its main diagonal.

- The matrix \mathbf{W} is **semi-correlated central Wishart** with t DoF, $\mathbf{W} \sim \mathcal{CW}_s(t, \boldsymbol{\Sigma}_s)$.



MIMO fading models and Wishart matrices

- Uncorrelated Ricean fading MIMO model

$$\mathbf{H} = \underbrace{\sqrt{\frac{K_r}{K_r + 1}} \mathbf{H}_L}_{\text{LoS component}} + \underbrace{\sqrt{\frac{1}{K_r + 1}} \mathbf{H}_W}_{\text{Scattered waves}} \quad (13)$$

where K_r is the Ricean K -factor.

- The matrix \mathbf{W} is **uncorrelated non-central Wishart** with t DoF, $\mathbf{W} \sim \mathcal{CW}_s \left(t, 1/(K_r + 1) \mathbf{I}_s, K_r/(K_r + 1) \mathbf{H}_L \mathbf{H}_L^\dagger \right)$.

Investigation of the standard condition number (SCN)

- Ratio of the largest to the smallest eigenvalue

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- Applications in the area of wireless communications
 - Quantifies the performance of linear detectors (ZF, MMSE)
 - Adaptive MIMO transmission/Adaptive decoding
 - Level of multipath activity



Generic framework for the CDF of the SCN-part A

A1 Definition of the joint eigenvalue PDF

$$f(\boldsymbol{\lambda}) = K |\boldsymbol{\Phi}(\boldsymbol{\lambda})| \times |\boldsymbol{\Psi}(\boldsymbol{\lambda})| \prod_{\ell=1}^S \xi(\lambda_{\ell}). \quad (15)$$



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A2 Leibniz determinant formula (use all possible permutations of the matrix elements)

$$|\boldsymbol{\Phi}(\boldsymbol{\lambda})| = |\phi_j(\lambda_i)| = \sum_{\alpha} (-1)^{\alpha} \prod_{i=1}^s \phi_{\alpha_i}(\lambda_i). \quad (16)$$

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A3 Joint eigenvalue PDF becomes

$$f(\boldsymbol{\lambda}) = K \sum_{\alpha} (-1)^{\alpha} |\psi_j(\lambda_i) \phi_{\alpha_i}(\lambda_i) \xi(\lambda_i)|, \quad 1 \leq i, j \leq s. \quad (17)$$



Generic framework for the CDF of the SCN-part B

B1 Integral definition of the SCN CDF, $F_z(x)$

$$F_z(x) = \Pr(z < x) = \int_0^\infty \left[\int_{\lambda_2}^{x\lambda_s} \int_{\lambda_3}^{x\lambda_s} \cdots \int_{\lambda_s}^{x\lambda_s} f(\lambda_1, \lambda_2, \dots, \lambda_s) \right. \\ \left. d\lambda_{s-1} \dots d\lambda_2 d\lambda_1 \right] d\lambda_s. \quad (18)$$



Generic framework for the CDF of the SCN-part B

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B2 Exploit the inherent symmetry of the joint eigenvalue PDF to permute the multiple integral $(s - 1)$ times

$$F_Z(x) = \frac{1}{(s-1)!} \int_0^\infty \int_{\lambda_s}^{x\lambda_s} \int_{\lambda_s}^{x\lambda_s} \cdots \int_{\lambda_s}^{x\lambda_s} f(\boldsymbol{\lambda}) d\lambda_1 \dots d\lambda_{s-1} d\lambda_s \quad (19)$$



Generic framework for the CDF of the SCN-part C

C1 After some crazy algebra, we end up with

$$F_z(\mathbf{x}) = K \sum_{\ell=1}^s \int_0^{\infty} \left| \left[\begin{array}{ll} \int_{\lambda_s}^{x\lambda_s} \phi_i(u)\psi_j(u)\xi(u)du, & i \neq \ell \\ \phi_i(\lambda_s)\psi_j(\lambda_s)\xi(\lambda_s), & i = \ell \end{array} \right] \right| d\lambda_s$$



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C2 Closed-form expressions for $\int_{\lambda_s}^{\mathbf{x}\lambda_s} \phi_i(u)\psi_j(u)\xi(u)du$

Eg. Uncorrelated central case:

$$\gamma(t - \mathbf{s} + i + j - 1, \mathbf{x}\lambda_s) - \gamma(t - \mathbf{s} + i + j - 1, \lambda_s)$$

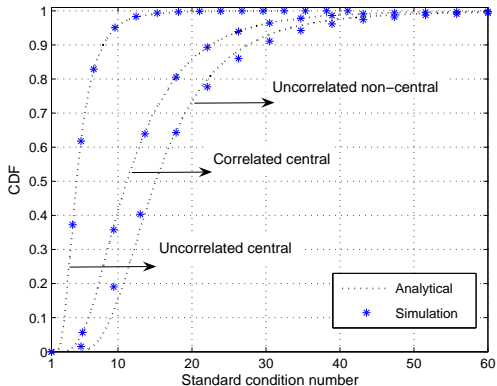


Analytical evaluation of the generic framework

- SCN CDF of an (8×3) MIMO system under uncorrelated Rayleigh, semi-correlated Rayleigh ($\rho = 0.6$) and uncorrelated Ricean ($K_r = 3$ dB) fading.

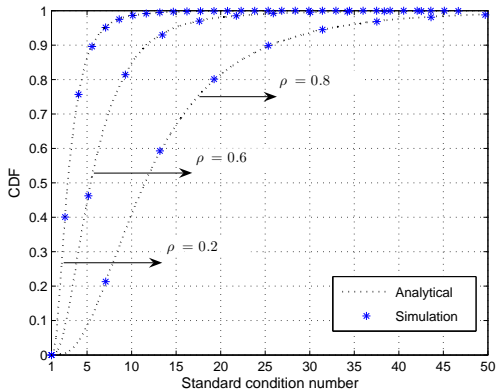
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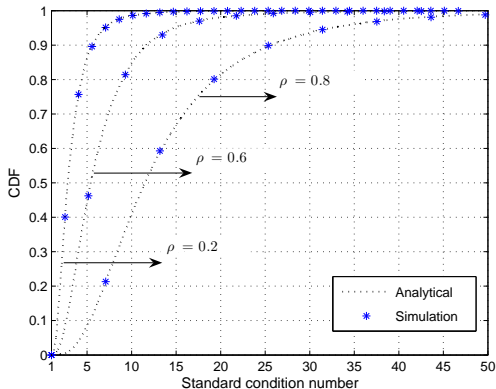
Simulation results

SCN CDF of a (6×2) semi-correlated Rayleigh system against the spatial correlation coefficient, ρ .



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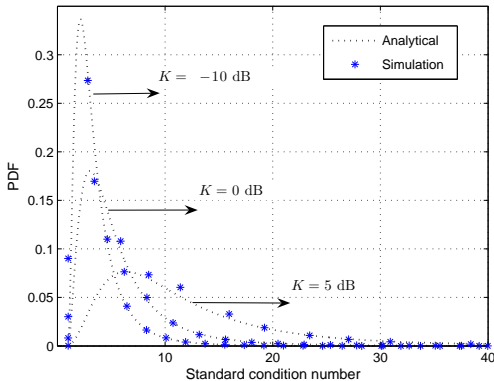
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Spatial correlation \rightarrow reduces the effective channel rank
(identical spatial characteristics of the impinging multipaths).

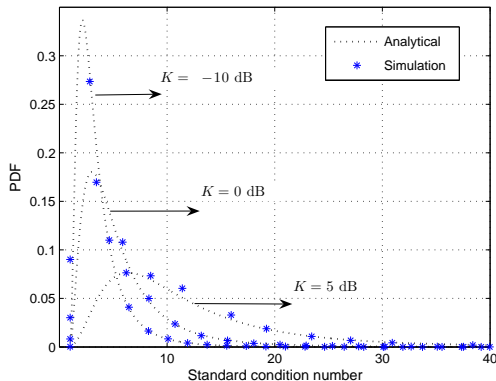
Simulation results

SCN PDF of a (2×5) uncorrelated Ricean system against the Ricean K -factor



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SCN PDF of a (2×5) uncorrelated Ricean system against the Ricean K -factor



As K gets larger the dynamic range of the SCN increases too → excessive spatial correlation between the LoS rays' phases.

Conclusions

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- Generic framework for the condition number statistics of three different classes of complex Wishart matrices (Uncorrelated/ Semi-correlated central and uncorrelated non-central).
- Validated via numerical integration for an arbitrary ($N_r \times N_t$) MIMO system.
- Implications of the model parameters (spatial correlation, Ricean K -factor) on the condition number performance.



Future work - Open problems

- Application of RMT to cognitive radio characterization (e.g. spectrum sensing schemes)
- Extension of known RMT results from point-to-point MIMO to multiuser scenarios
- Design of practical adaptive subchannel selection algorithms for performance enhancement of MIMO beamforming schemes
- Performance analysis of MIMO systems assuming different fading models (e.g. Nakagami- m , Weibull...)
- Consider the impact of practical impairments (e.g. channel estimation error and feedback delay)

Related publications

- **M. Matthaiou**, M. R. McKay, P. J. Smith, and J. A. Nосsek, “On the condition number distribution of complex Wishart matrices,” *in press IEEE Transactions on Communications*.
- **M. Matthaiou**, P. de Kerret, G. K. Karagiannidis, and J. A. Nосsek, “Mutual information statistics and beamforming performance analysis of optimized LoS MIMO systems,” submitted to *IEEE Transactions on Communications*.
- **M. Matthaiou**, Y. Kopsinis, D. I. Laurenson, and A. M. Sayeed, “Upper bound for the ergodic capacity of dual MIMO Ricean systems: Simplified derivation and asymptotic tightness,” *IEEE Transactions on Communications*, vol. 57, no. 12, pp. 3589–3596, December 2009.
- **M. Matthaiou**, D. I. Laurenson, and C. -X. Wang, “On analytical derivations of the condition number distributions of dual non-central Wishart matrices,” *IEEE Transactions on Wireless Communications*, vol. 8, no. 3, pp. 1212-1217, March 2009.
- **M. Matthaiou**, A. Pitarokoilis, and J. A. Nосsek, “Mutual information statistics of optimized LoS MIMO systems,” to appear *IEEE International Conference on Communications (ICC)*, May 2010.



Thank you for your attention!