

Capacity Analysis of Correlated MIMO Channels

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Overview

- Background
- Analytic capacity for correlated channel
- “Large n ” capacity
- Concept: “always linear” capacity growth
- Conclusions

Background

- Capacity of i.i.d MIMO channel well studied. (Telatar, 1999)
 - t transmit, r receive
 - “linear growth” with respect to minimum number of elements $m = \min(r, t)$.
 - reasonable match physical non-line-of-sight channel.
- Correlated MIMO channel more heuristic
 - LOS components
 - Chuah *et al.* (2002) noted that *capacity growth still linear* for correlated channels, through use of Stieljes transform, although LOS introduces *equivalent power loss* due to correlation.

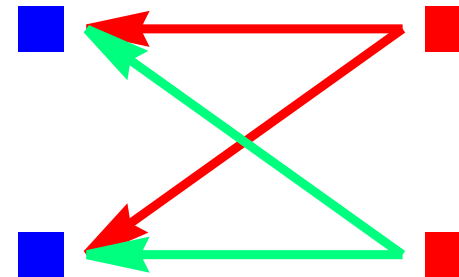
Channel model

- General linear MIMO channel model:

$$y = Hx + w$$

restriction: H random matrix, known at receiver, unknown at transmitter.

$$C = \mathbb{E} \left\{ \log_2 \det \left(I_m + HH^* \cdot \frac{P}{t\eta} \right) \right\}$$



- For our model:
 - H is Gaussian random, with covariance Σ .
 - H and Σ unknown at transmitter.

What is the analytic capacity of this channel?

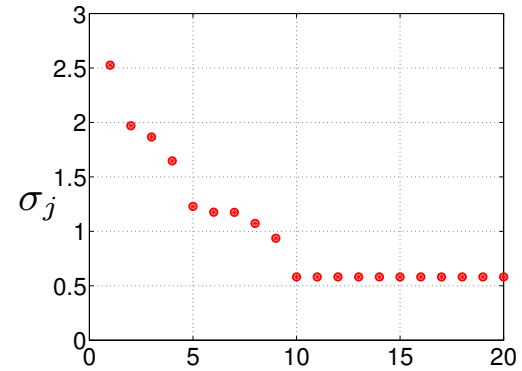
Capacity of Correlated Gaussian MIMO Channel

$$C = K_{\Sigma, m, n} \cdot \int_{\Lambda} {}_0F_0 \left(-\frac{1}{2}\Sigma^{-1}, n\Lambda \right) \cdot J(\Lambda) \cdot \sum_{i=1}^m \log \left(1 + \frac{P}{t} \lambda_i \right) d\Lambda$$

- Integral arises from definition of expectation.
- Problem: contains hypergeometric function
- Can be solved numerically, and *under certain conditions* also analytically.
 1. $\Sigma = I$: “famous” iid case, ${}_0F_0 \left(-\frac{1}{2}\Sigma^{-1}, n\Lambda \right)$ reduces to simpler form
 2. asymptotically for $n \gg m$

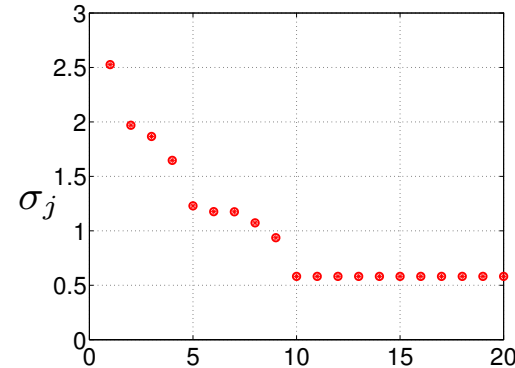
The next step... $n \gg m$

- Can assume Σ is diagonal,
 - $\Sigma = \text{diag} \{ \sigma_1, \dots, \sigma_m \}$
 - Also, $\sigma_1 > \dots > \sigma_k > \sigma_{k+1} = \dots = \sigma_m$



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- ${}_0F_0(\cdot)$ separates into two (joint) distributions
 - Normal independent distribution for distinct eigenvalues and
 - Normal distribution, with conditional density for remaining equal eigenvalues
 - **The distributions are jointly independent**
- Separate capacity integral into two parts:

$$C = C_d + C_{\text{eq}}$$

- C_d for distinct eigenvalues $\sigma_1 > \dots > \sigma_{k-1}$
- C_{eq} for equal eigenvalues $\sigma_k = \sigma_{k+1} = \dots = \sigma_m$

Split Capacity

- C_d comprises joint distribution of independent random variables:

$$C_d = \frac{1}{\sqrt{2\pi}} \sum_{i=1}^k \int \exp\left(-\frac{z^2}{2}\right) \log\left(1 + \frac{P}{t} n \lambda_i \left[z \left(\frac{n}{2}\right)^{-1/2} + 1\right]\right) dz$$

- C_{eq} has *similar* distribution to standard i.i.d. MIMO case

$$C_{eq} = \int \log\left(1 + \frac{P}{t} n \lambda \left[z \left(\frac{n}{2}\right)^{-1/2} + 1\right]\right) \cdot \sum_{j=1}^{m-k} \frac{1}{2^j j! \sqrt{2\pi}} \left[H_j\left(\frac{z}{\sqrt{2}}\right)\right]^2 e^{-z^2/2} dz$$

where $H_j(\cdot)$ is the j -th Hermite polynomial (Szegő, 1939).

Divide & Conquer

- C_d
 - requires special attention, use approximation of $n \rightarrow \infty$.
- C_{eq}
 - has same form as capacity for iid channel.

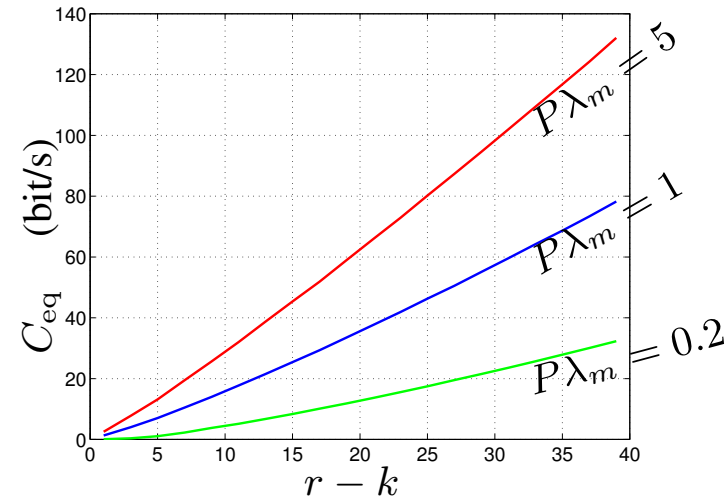


Figure 1: C_{eq} vs number of equal eigenvalues

$$C_{n \rightarrow \infty} = C_d + C_{eq}$$

$$\sim \sum_{i=1}^k \log_2 \left(1 + \frac{P}{t} n \sigma_i \right) + \int \log_2 \left(1 + \frac{P}{t} n \sigma_m \right) \cdot f(\sigma_m)$$

This limit only converges for large t .

Large n vs Large t

- $n = t \rightarrow \infty$, limit converges

- growth is linear with respect to $m = r$

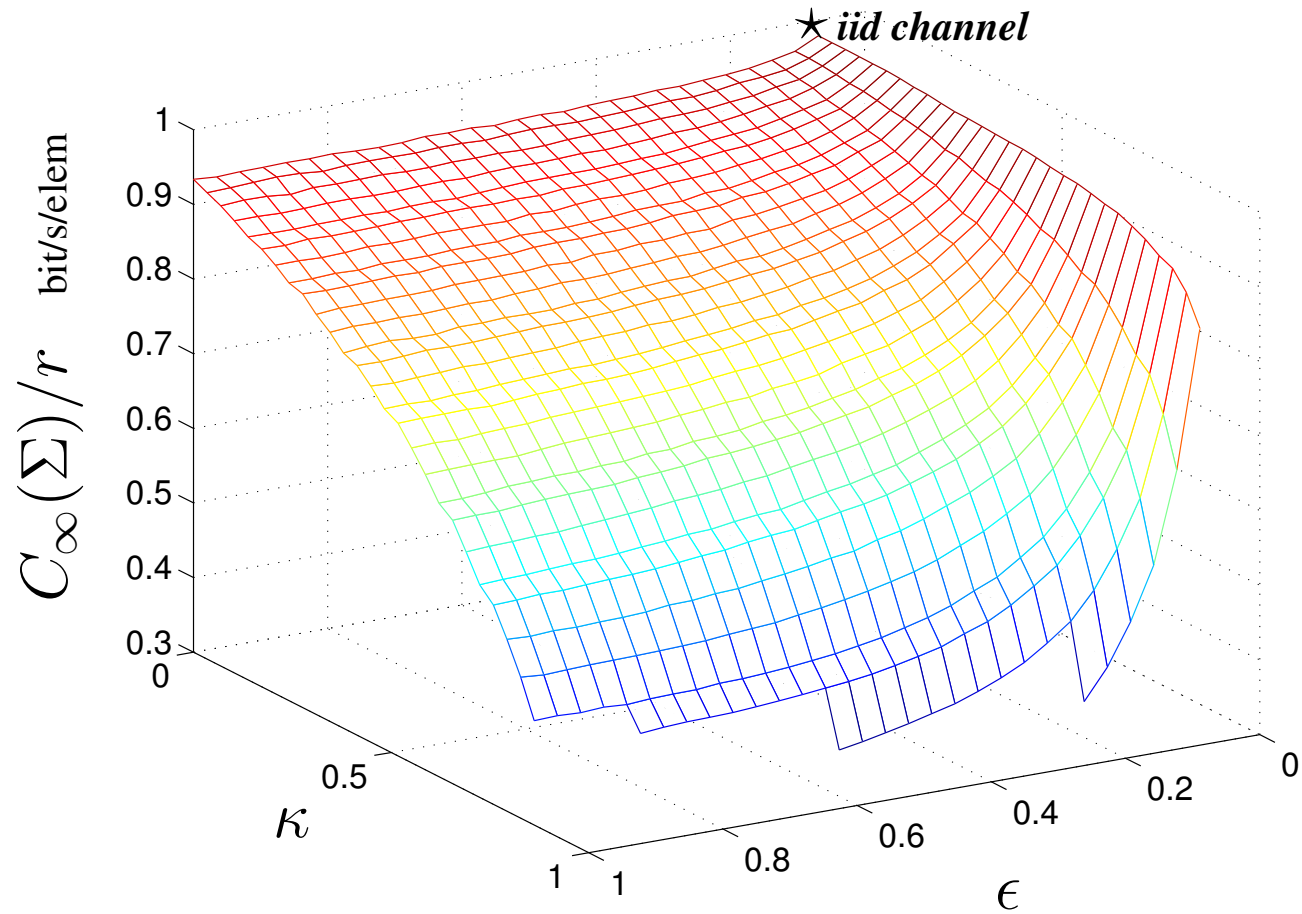
$$C_{t \rightarrow \infty} \sim \sum_{i=1}^k \log_2 \left(1 + \frac{P}{\eta} \sigma_i \right) + (r - k) \log_2 \left(1 + \frac{P}{\eta} \sigma_r \right)$$

- σ_i is the i^{th} eigenvalue of Σ
- Define equivalence class of correlation matrices $S_m \subset \mathbb{C}^{m \times m}$, with equal trace

- Compare $\Sigma \in S_m$ matrices

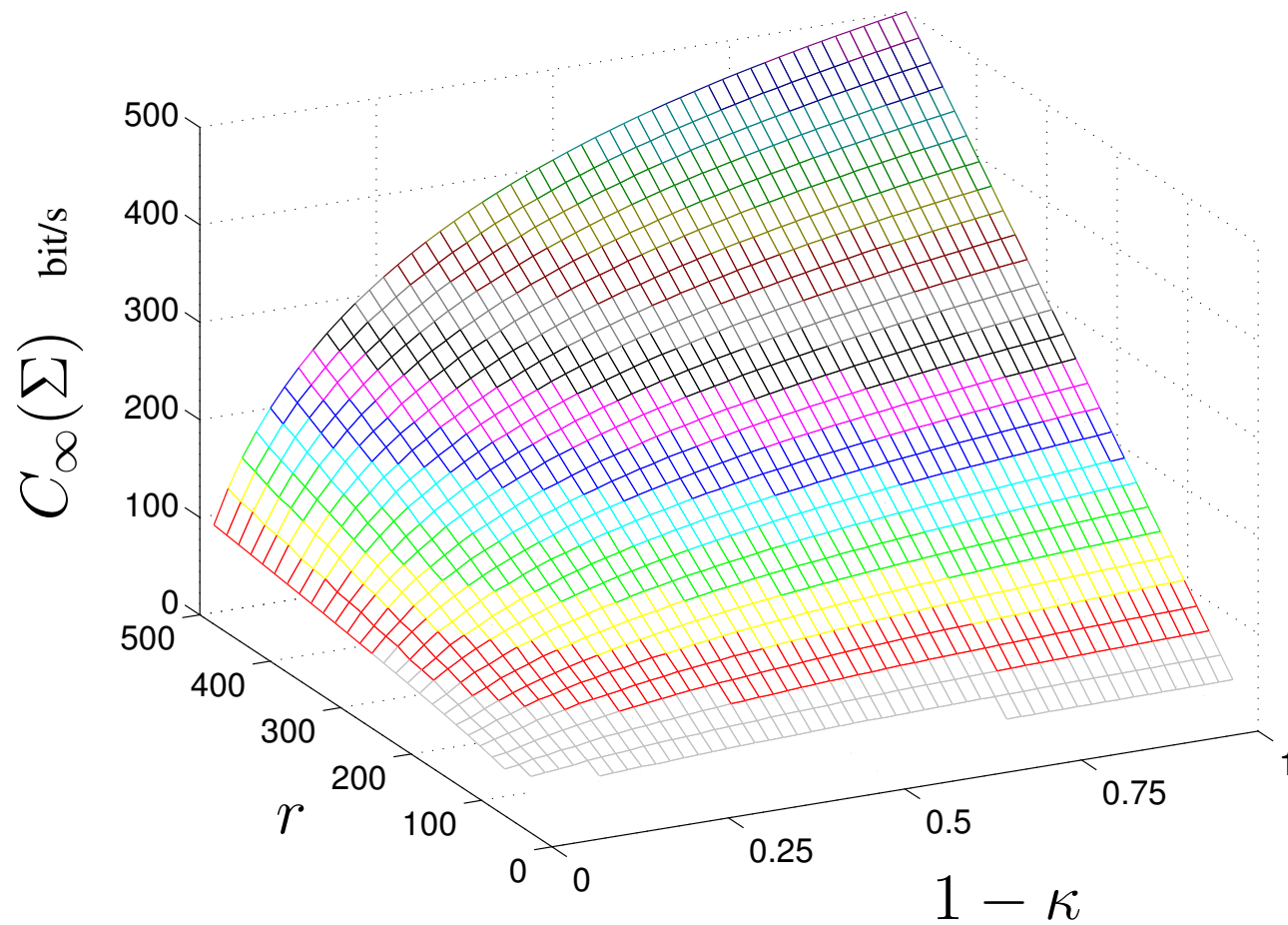
- number of distinct eigenvalues k (or $\kappa = k/r$) and
- ratio $\epsilon = \lambda_1/\lambda_m$

Capacity Surface $C_\infty(\Sigma)/r$



Capacity $C_\infty(\Sigma)/r$ for κ and ϵ

Generalized Linear Growth



Capacity $C_\infty(\Sigma)$ for κ and r , given $\epsilon = 15$

Conclusions

- Provided analytic capacity for correlated random MIMO channel.
- Provided asymptotic simplification for large t and fixed r .
 - For full rank correlation matrices Σ_m , growth is linear
 - Σ_m gives **rate of growth**
- i.i.d. channel has largest rate of growth $\alpha = \log_2(1 + P)$ for increasing r .
 - Equivalent β dB power loss for LOS channel due to ratio ϵ .

References

Chuah, C.-N., D. N. C. Tse, J. M. Kahn and R. A. Valenzuela (2002). Capacity scaling in MIMO systems under correlated fading. *IEEE Transactions on Information Theory* **48**(3), 637–650.

Szegő, G. (1939). *Orthogonal Polynomials*. American Mathematical Society. Providence, RI.

Telatar, I. E. (1999). Capacity of multi-antenna gaussian channels. *European Transactions on Telecommunications* **10**(6), 585–595.