

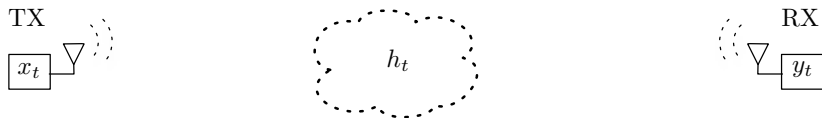
# Capacity Pre-log of Noncoherent SIMO Channels via Hironaka's Theorem

Veniamin I. Morgenshtern

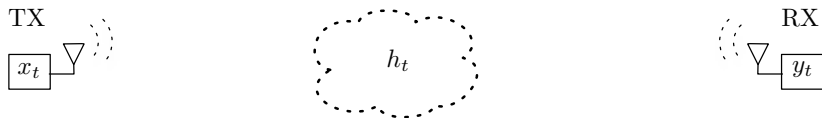
Joint work with

E. Riegler, W. Yang, G. Durisi, S. Lin, B. Sturmfels, and H. Bölcskei

# SISO Fading Channel

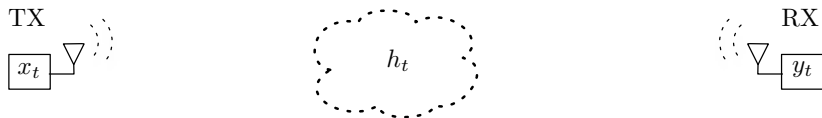


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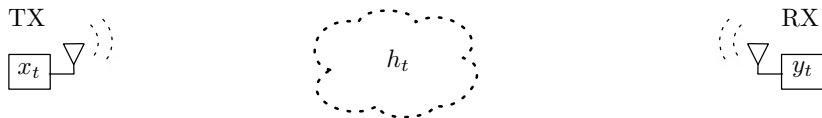
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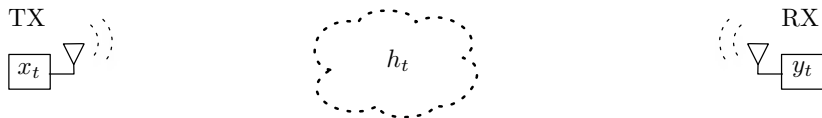


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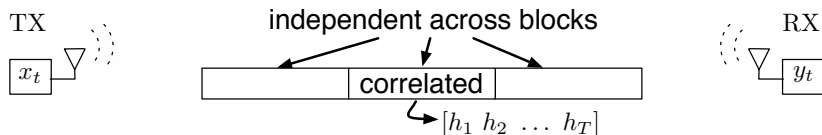
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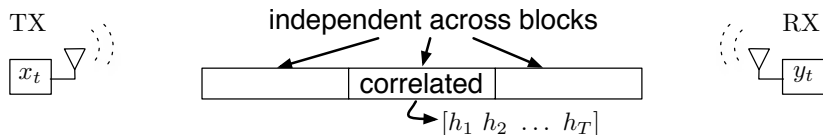
**Noncoherent setting:**  $h_t$  is not known at RX

# SISO Correlated Block-Fading Channel



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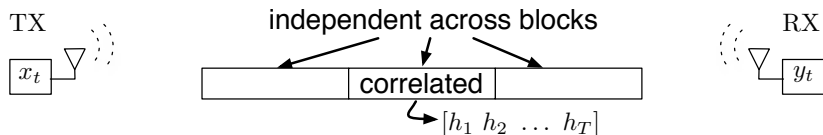


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- $[h_1 \ \dots \ h_T]^T \sim \mathcal{CN}(\mathbf{0}, \mathbf{P}\mathbf{P}^H)$
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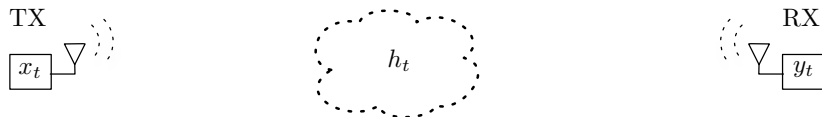


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- Channel gains in whitened form:

$$\begin{bmatrix} h_1 \\ h_2 \\ \vdots \\ h_T \end{bmatrix} = \mathbf{P} \begin{bmatrix} s_1 \\ \vdots \\ s_Q \end{bmatrix}, \quad s_q \sim \mathcal{CN}(0, 1), \text{ iid across } q$$

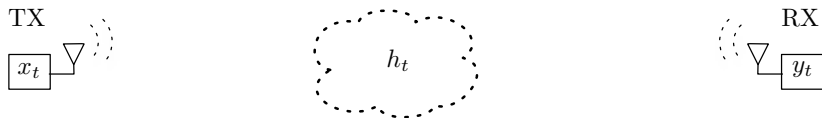
# Capacity



Fundamental problem:

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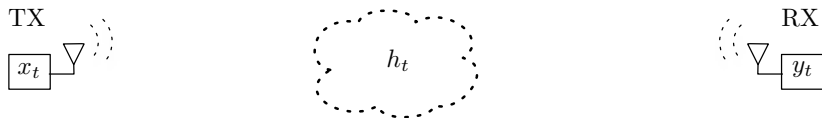
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$$C(\text{SNR}) \triangleq (1/T) \sup_{f_{\mathbf{x}}(\cdot)} I(\{x_t\}; \{y_t\}); \quad \mathbb{E} \left[ \sum_{t=1}^T |x_t|^2 \right] \leq T$$

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- Mutual information:  $I(\{x_t\}; \{y_t\}) = h(\{x_t\}) - h(\{x_t\} | \{y_t\})$
- Differential Entropy:  $h(\{y_t\})$

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## Pre-Log

$$\chi = \lim_{\text{SNR} \rightarrow \infty} \frac{C(\text{SNR})}{\log(\text{SNR})}$$

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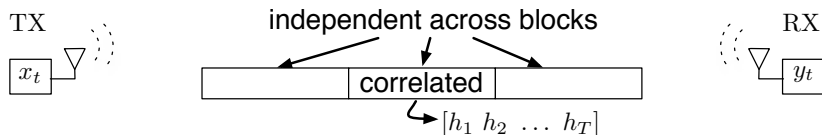
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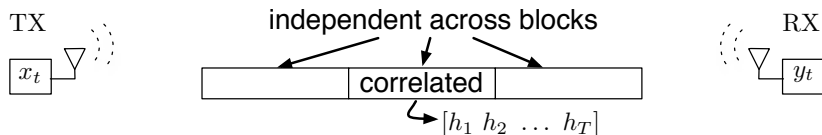
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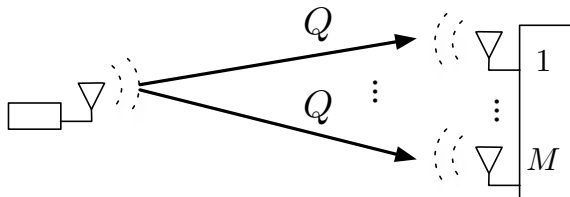


Number of unknown channel parameters per block is  $Q = \text{rank } \mathbf{P}$ :

$$[h_1 \ h_2 \ \dots \ h_T]^T = \mathbf{P}[s_1 \ s_2 \ \dots \ s_Q]^T$$

Eliminating the Noncoherent Penalty  
by Adding Receive Antennas (Only!)

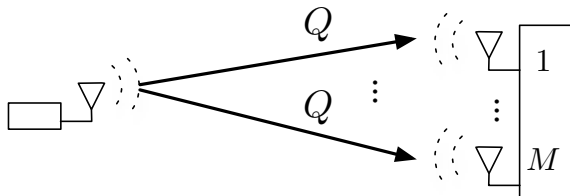
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$$y_{mt} = \sqrt{\text{SNR}} h_{mt} x_t + w_{mt} \quad (m = 1, \dots, M, t = 1, \dots, T)$$

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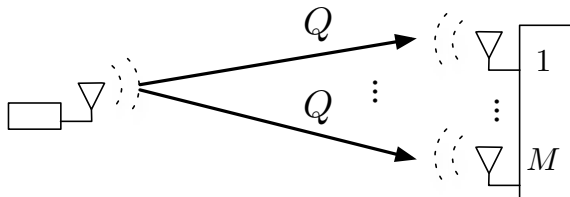
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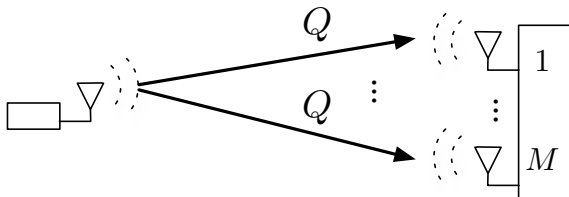


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We can actually do better!

## Main Result [VM et. al., 2012]

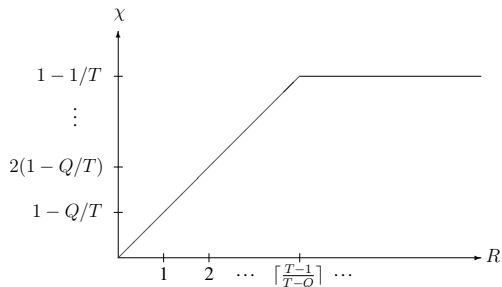
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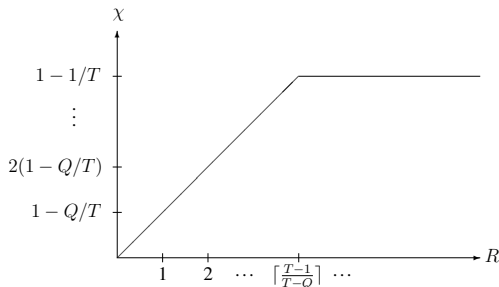
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Multiple antennas at the receiver can recover degrees of freedom

# Implications

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The “channel identification penalty” vanishes if a second receive antenna is used

# Linear Algebra: Guessing Pre-Log



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Rule of thumb:

$$\chi = \frac{\text{number of RVs } x_t \text{ that can be identified uniquely from } \{\hat{y}_{mt}\}}{T}$$

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- $\Rightarrow \chi_{\text{SISO}} = 3/3 = 1$  [Telatar, 99]

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- $\Rightarrow \chi_{\text{SISO}} = (3 - 2)/3 = 1/3 = 1 - Q/T$  [Liang & Veeravalli, 04]

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# Equations for General $\mathbf{P}$ ( $T = 3, Q = 2, M = 2$ )

$$\underbrace{\begin{bmatrix} p_{11} & p_{12} & 0 & 0 & 0 & 0 \\ p_{21} & p_{22} & 0 & 0 & \hat{y}_{12} & 0 \\ p_{31} & p_{32} & 0 & 0 & 0 & \hat{y}_{13} \\ 0 & 0 & p_{11} & p_{12} & 0 & 0 \\ 0 & 0 & p_{21} & p_{22} & \hat{y}_{22} & 0 \\ 0 & 0 & p_{31} & p_{32} & 0 & \hat{y}_{23} \end{bmatrix}}_{\mathbf{B}} \begin{bmatrix} s_{11} \\ s_{12} \\ s_{21} \\ s_{22} \\ -z_2 \\ -z_3 \end{bmatrix} = \begin{bmatrix} \hat{y}_{11} \\ 0 \\ 0 \\ \hat{y}_{21} \\ 0 \\ 0 \end{bmatrix}$$

Solution is unique iff  $\mathbf{B}$  is full-rank

# Information Theory

# Vector Notations ( $T = 3, Q = 2, M = 2$ )

$$\underbrace{\begin{bmatrix} y_{11} \\ y_{12} \\ y_{13} \\ y_{21} \\ y_{22} \\ y_{23} \end{bmatrix}}_{\mathbf{y}} = \sqrt{\text{SNR}} \underbrace{\begin{bmatrix} s_{11}x_1 \\ s_{12}x_2 \\ (s_{11} + s_{12})x_3 \\ s_{21}x_1 \\ s_{22}x_2 \\ (s_{21} + s_{22})x_3 \end{bmatrix}}_{\hat{\mathbf{y}}} + \underbrace{\begin{bmatrix} w_{11} \\ w_{12} \\ w_{13} \\ w_{21} \\ w_{22} \\ w_{23} \end{bmatrix}}_{\mathbf{w}}$$

$$\mathbf{P} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\mathbf{s} = [s_{11} \ s_{12} \ s_{21} \ s_{22}]^T$$

$$\mathbf{x} = [x_1 \ x_2 \ x_3]^T$$

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Change of Variables Lemma:

$$h(\hat{\mathbf{y}} | x_1) = h(\mathbf{s}, x_2, x_3 | x_1) + \mathbb{E}_{\mathbf{s}, \mathbf{x}} \log \left| \det \frac{\partial \hat{\mathbf{y}}}{\partial (\mathbf{s}, x_2, x_3)} \right|$$

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# Resolution of Singularities

How can we show that

$$\mathbb{E}_{\mathbf{s}, \mathbf{x}} \log \left| \det \frac{\partial \hat{\mathbf{y}}}{\partial (\mathbf{s}, x_2, x_3)} \right| > -\infty?$$

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- $J_1(\mathbf{x})$  and  $J_3(\mathbf{x})$  are diagonal matrices
- $\det J_2(\mathbf{s})$  is a homogeneous **polynomial**:

$$\det J_2(\lambda \mathbf{s}) = \lambda^D \det J_2(\mathbf{s}), \quad \forall \lambda \in \mathbb{C}$$



# Resolution of Singularities (Cont'd)

Polar coordinates:  $\mathbf{s} \rightarrow (r, \theta)$

$$\begin{aligned} & \left| \int_{\mathbb{C}^{RQ}} \exp(-\|\mathbf{s}\|^2) \log |\det J_2(\mathbf{s})| \, d\mathbf{s} \right| \\ & \leq \left| \int_{\mathbb{C}^{RQ}} \exp(-\|s\|^2) \log |\det J_2(\mathbf{s}/\|s\|^2)| \, d\mathbf{s} \right| + O(1) \\ & \leq \int_0^\infty \exp(-r^2) r^{2D-1} \, dr \times \int_{\Delta} |\log |f(\theta)|| \, d\theta + O(1) \end{aligned}$$

where

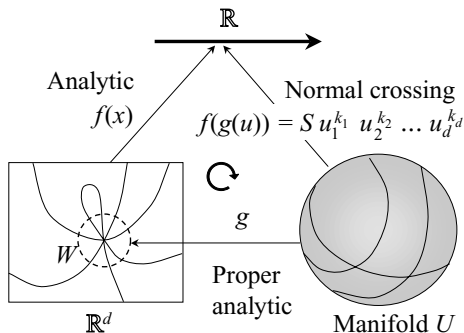
- $\Delta = [0, \pi]^{2D-2} \times [0, 2\pi]$  is a compact set
- $f$  is a **real analytic function**

# Resolution of Singularities (Cont'd)

Hironaka's Theorem implies:

If  $f \not\equiv 0$  is a real analytic function, then

$$\int_{\Delta} |\log |f(\theta)|| d\theta < \infty.$$



## The Technical Condition on $\mathbf{P}$ : not Just Rank

$$f \neq 0 \text{ iff } \begin{bmatrix} p_{11} & p_{12} & 0 & 0 & 0 & 0 \\ p_{21} & p_{22} & 0 & 0 & p_{21} & 0 \\ p_{31} & p_{32} & 0 & 0 & 0 & p_{31} \\ 0 & 0 & p_{11} & p_{12} & 0 & 0 \\ 0 & 0 & p_{21} & p_{22} & p_{22} & 0 \\ 0 & 0 & p_{31} & p_{32} & 0 & p_{32} \end{bmatrix} \text{ is full-rank}$$

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For example,  $\mathbf{P} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}$  satisfies this condition

# Connection to Linear Algebra

$$\begin{bmatrix} p_{11} & p_{12} & 0 & 0 & 0 & 0 \\ p_{21} & p_{22} & 0 & 0 & p_{21} & 0 \\ p_{31} & p_{32} & 0 & 0 & 0 & p_{31} \\ 0 & 0 & p_{11} & p_{12} & 0 & 0 \\ 0 & 0 & p_{21} & p_{22} & p_{22} & 0 \\ 0 & 0 & p_{31} & p_{32} & 0 & p_{32} \end{bmatrix} \text{ is full-rank}$$

$$\text{iff } \begin{bmatrix} p_{11} & p_{12} & 0 & 0 & 0 & 0 \\ p_{21} & p_{22} & 0 & 0 & \hat{y}_{12} & 0 \\ p_{31} & p_{32} & 0 & 0 & 0 & \hat{y}_{13} \\ 0 & 0 & p_{11} & p_{12} & 0 & 0 \\ 0 & 0 & p_{21} & p_{22} & \hat{y}_{22} & 0 \\ 0 & 0 & p_{31} & p_{32} & 0 & \hat{y}_{23} \end{bmatrix} = \mathbf{B} \text{ is full-rank a.e.}$$

# Open problems

- MIMO
- Stationary Channel Model

Thank you