

# Joint Multiple Cell-Site Processing in Wyner-like Fading Models

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# Introduction

- The main focus of information theoretic efforts is to provide ultimate performance bounds on different cellular models, and to identify efficient communications strategies coming close to these bounds
- Communications strategies: Concepts of
  - Modulation/Coding
  - Channel-Accessing/Scheduling
  - Protocols (cross-layer)
  - Cooperation/Processing
- *Joint processing* of signals related to different users is evidently the most appealing approach in either the uplink or downlink

# Introduction (*Cont'd*)

- The *uplink* channel is a *multiple access* channel (MAC):
  - *Joint processing of the received signals* at the cell-sites
  - Commonly no user cooperation is assumed
  - User cooperation is also studied lately  
[Shamai-Somekh-Simeone-Sanderovich-Zaidel-Poor JWCC '07] (and references therein).
- The *downlink* channel is a *broadcast* channel (BC):
  - *Transmitter based joint preprocessing* (complexity considerations)
  - Commonly *single-user detection* at the mobile receiver is assumed
  - User cooperation is also of interest  
[Shamai-Somekh-Simeone-Sanderovich-Zaidel-Poor JWCC '07] (and references therein).
- The MAC/BC duality principle in different frameworks provides the firm information theoretic connections between these seemingly basically different models [Vishwanath-Jindal-Goldsmith '03], [Viswanath-Tse '03], [Jindal-Vishwanath-Goldsmith '04], [Yu-Lan '04]

# Introduction (*Cont'd*)

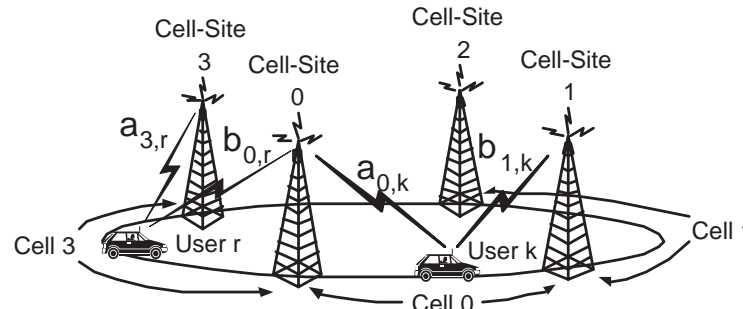
## ● Examples for multi-cell uplink analyses:

- “Older references”: [Hanly-Whiting '93], [Wyner '94], [Shamai-Wyner '97], [Somekh-Shamai '00], [Zaidel-Shamai-Verdú '01], [Zaidel-Shamai-Verdú '02], [Somekh-Zaidel-Shamai '04], [Grant-Hanly-Evans-Muller '04], [Aktas-Bacha-Evans-Hanly '04]
- Recent references: See reference list in [Shamai-Somekh-Zaidel JWCC '04], [Shamai-Somekh-Simeone-Sanderovich-Zaidel-Poor JWCC '07], [Somekh-Zaidel-Shamai IT Dec. '07], [Somekh-Simeone-Bar-Ness-Haimovich-Shamai '07].

## ● Examples for multi-cell downlink analyses:

- “Older references”: [Shamai-Zaidel '01], [Jafar-Goldsmith '02], [Jafar-Foschini-Goldsmith '04], [Dai-Molisch-Poor '04], [Zhang-Dai-Zhuo-2005], [Foschini-et-al-2005].
- Recent references: See reference list in [Shamai-Somekh-Zaidel JWCC '04], [Shamai-Somekh-Simeone-Sanderovich-Zaidel-Poor JWCC '07], [Somekh-Zaidel-Shamai IT Dec. '07], [Somekh-Simeone-Bar-Ness-Haimovich-Shamai '07].

# General System Model



- A “Wyner-type” multi-cell model with  $M$  cells ordered on a *circle*
- Motivation: symmetry properties, more amenable to analytical analysis, equivalent to linear models for  $M \gg 1$
- A fully synchronous, optimally coded system is assumed, with cell-sites located at the cells’ *boundaries*
- There are  $K$  users in each cell, and a single receive/transmit antenna at each cell-site
- Each user “sees” only the two nearest cell-sites
- Models a practical “soft-handoff” scenario at the cells’ boundaries

# Outline

- Joint multi-cell processing in the uplink is analyzed focusing on the average per-cell sum-rate capacity and spectral efficiency.
- Analytical expressions for the sum-rate capacities under various non-fading/fading conditions are derived for:
  - Intra-cell TDMA scheduling
  - A “Wide-Band” (WB) scheme (all and many ( $K \gg 1$ ) users are active simultaneously utilizing all bandwidth for coding)
  - **Asymptotic results for  $K$  users in the fading regime.**
  - **Full characterization of the randomly activated users scenario in term of per user and outage/ergodic performance.**
- For the downlink channel, assuming *individual per-cell power constraints*:
  - Sum-rate capacity is derived for non-fading channels.
  - Flat-fading channels are analyzed through upper and lower bounds.
  - The impact of uplink scheduling is demonstrated.
- Outlook and concluding remarks.

# Uplink System Model

- The discrete-time vector representation of the signals received at the  $M$  cell-sites is given, for an arbitrary time index, by

$$\mathbf{y}_{ul}[M \times 1] = \mathbf{H}_M \mathbf{x}_{ul} + \mathbf{z}_{ul}$$

where

- $\mathbf{H}_M[M \times KM]$  - Channel transfer matrix
- $\mathbf{x}_{ul}[KM \times 1] \sim \mathcal{N}_c(\mathbf{0}, P\mathbf{I}_{KM})$  - Transmitted symbols vector
- $\mathbf{z}_{ul}[M \times 1] \sim \mathcal{N}_c(\mathbf{0}, \mathbf{I}_M)$  - Circularly symmetric complex AWGN vector
- The total intra-cell transmit power is denoted by  $\bar{P} \triangleq KP$

# Uplink System Model (*Cont'd*)

- The  $M \times KM$  channel transfer matrix  $\mathbf{H}_M$  is given by:

$$\mathbf{H}_M = \begin{pmatrix} \mathbf{a}_0 & \mathbf{0} & \cdots & \mathbf{0} & \mathbf{b}_0 \\ \mathbf{b}_1 & \mathbf{a}_1 & \mathbf{0} & \cdots & \mathbf{0} \\ \mathbf{0} & \mathbf{b}_2 & \mathbf{a}_2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \mathbf{0} \\ \mathbf{0} & \cdots & \mathbf{0} & \mathbf{b}_{M-1} & \mathbf{a}_{M-1} \end{pmatrix}$$

- $\{\mathbf{a}_m[1 \times K], \mathbf{b}_m[1 \times K]\}$  - complex channel fades of the  $K$  users of the  $m$ th and  $[(m - 1) \bmod M]$ th cells, respectively, received at the  $m$ th cell-site
- i.i.d. fading among different users (ergodic with respect to the time index)
- The moments and kurtosis of the fading distribution are denoted by

$$m_1 \triangleq E\{a_{m,k}\} = E\{b_{m,k}\} \quad ; \quad m_2 \triangleq E\{|a_{m,k}|^2\} = E\{|b_{m,k}|^2\} \quad ;$$

$$m_4 \triangleq E\{|a_{m,k}|^4\} = E\{|b_{m,k}|^4\} \quad ; \quad \mathcal{K} \triangleq m_4/m_2^2 \quad \forall m, k$$



# Uplink Average Per-Cell Sum-Rate Capacity

- Underlying assumptions:
  - Full channel state information is available to the joint multi-cell receiver only
  - Users cannot cooperate their transmissions in any way
- The ergodic average per-cell sum-rate capacity of the uplink channel is given by

$$C_{\text{ul}}(\bar{P}) = \frac{1}{M} E_{H_M} \left\{ \log \det \left( \mathbf{I}_M + \frac{\bar{P}}{K} \mathbf{H}_M \mathbf{H}_M^\dagger \right) \right\}$$

- The corresponding spectral efficiency [bits/sec/Hz] is obtained through the relation

$$\bar{P} = C_{\text{ul}} \left( \frac{E_b^t}{N_0} \right) \frac{E_b^t}{N_0}$$

where  $E_b^t/N_0$  is the transmit  $E_b/N_0$  and  $C_{\text{ul}}(\bar{P}) = C_{\text{ul}}(E_b^t/N_0)$

# Uplink Average Per-Cell Sum-Rate Capacity (Cont'd)

- The uplink average per-cell sum-rate capacity is governed by the  $M \times M$  matrix  $\frac{1}{K} \mathbf{H}_M \mathbf{H}_M^\dagger$ , whose non-zero entries are explicitly expressed by

$$\left[ \frac{1}{K} \mathbf{H}_M \mathbf{H}_M^\dagger \right]_{m,m} = \frac{1}{K} \sum_{k=1}^K |a_{m,k}|^2 + \frac{1}{K} \sum_{k=1}^K |b_{m,k}|^2$$

$$\left[ \frac{1}{K} \mathbf{H}_M \mathbf{H}_M^\dagger \right]_{m, \widehat{m-1}} = \frac{1}{K} \sum_{k=1}^K a_{\widehat{m-1},k}^* b_{m,k} \quad ,$$

$$\left[ \frac{1}{K} \mathbf{H}_M \mathbf{H}_M^\dagger \right]_{m, \widehat{m+1}} = \frac{1}{K} \sum_{k=1}^K a_{m,k} b_{\widehat{m+1},k}^*$$

where

$$\widehat{m+1} \triangleq (m+1) \bmod M \quad ; \quad \widehat{m-1} \triangleq (m-1) \bmod M$$

# Extreme SNR Characterization

- The low-SNR regime is characterized through:

$$C_{ul}(E_b^t/N_0) \approx \mathcal{S}_0/3|_{\text{dB}} (E_b^t/N_0|_{\text{dB}} - E_b^t/N_{0\text{min}}|_{\text{dB}})$$

- $E_b^t/N_{0\text{min}} \triangleq \log_e 2 / \dot{C}_{ul}(0)$  (Scalar AWGN Ch.:  $E_b^t/N_{0\text{min}} = \log_e 2$ )
- $\mathcal{S}_0$  (slope)  $\triangleq -2 \left[ \dot{C}_{ul}(0) \right]^2 / \ddot{C}_{ul}(0)$  (Scalar AWGN Ch.:  $\mathcal{S}_0 = 2$ )

( $3|_{\text{dB}} = 10 \log_{10} 2$ ,  $\dot{C}_{ul}(0)$  are  $\ddot{C}_{ul}(0)$  the derivative of the capacity in nats/dimension at  $\bar{P} = 0$ )

- The high-SNR regime is characterized through the affine capacity approximation:

$$C_{ul}(\bar{P}) \underset{\bar{P} \gg 1}{\approx} \mathcal{S}_\infty \left( \frac{\bar{P}|_{\text{dB}}}{3|_{\text{dB}}} - \mathcal{L}_\infty \right) = \frac{\mathcal{S}_\infty}{3|_{\text{dB}}} (\bar{P}|_{\text{dB}} - 3|_{\text{dB}} \mathcal{L}_\infty)$$

- $\mathcal{S}_\infty$  (multiplexing gain)  $\triangleq \lim_{\bar{P} \rightarrow \infty} \bar{P} \dot{C}_{ul}(\bar{P})$  (Scalar AWGN Ch.:  $\mathcal{S}_\infty = 1$ )  
( $\dot{C}_{ul}(\bar{P})$  is the derivative of the capacity in nats/dimension)
- $\mathcal{L}_\infty$  (power offset)  $\triangleq \lim_{\bar{P} \rightarrow \infty} (\log_2 \bar{P} - C_{ul}(\bar{P})/\mathcal{S}_\infty)$  ( $C_{ul}(\bar{P})$  in bits/dimension)  
(Scalar AWGN Ch.:  $\mathcal{L}_\infty = 0$ )

# Uplink - No Fading

- In the absence of fading  $a_{m,k} = b_{m,k} = 1, \forall m, k$
- The matrix  $\frac{\bar{P}}{K} \mathbf{H}_M \mathbf{H}_M^\dagger$  becomes a circulant matrix with non-zero row entries  $(\bar{P}, 2\bar{P}, \bar{P})$
- The average per-cell sum-rate capacity depends only on the total intra-cell transmit power  $\bar{P}$ :
  - Intra-cell TDMA - Single active user per cell, transmitting  $1/K$  fraction of the time, signaling with power  $\bar{P} = KP$
  - Wide-Band (WB) - All  $K$  users are simultaneously active, occupying the whole bandwidth, signaling with power  $P$
- **Proposition 1** *The average per-cell sum-rate capacity, in the absence of fading, is given for  $M \rightarrow \infty$  by*

$$C_{ul-nf}(\bar{P}) = \log \left( \frac{1 + 2\bar{P} + \sqrt{1 + 4\bar{P}}}{2} \right)$$

# Uplink - No Fading (*Cont'd*)

- **Proposition 2** *The uplink channel extreme-SNR regimes for non-fading channels are characterized for  $M \rightarrow \infty$  by*

$$\frac{E_b^t}{N_{0 \min}} = \frac{\log_e 2}{2};$$

$$S_0 = \frac{4}{3};$$

$$S_\infty = 1;$$

$$\mathcal{L}_\infty = 0.$$

- Extension to  $a_{m,k} = 1, b_{m,k} = \alpha, \forall m, k$  is straight forward [Lifand-Goldsmith '05]

$$C_{ul-nf}(\bar{P}) = \log \left( \frac{1 + (1 + \alpha^2)\bar{P} + \sqrt{1 + 2(1 + \alpha^2)\bar{P} + (1 - \alpha^2)^2\bar{P}^2}}{2} \right).$$

# Uplink - Rayleigh Fading - Intra-Cell TDMA

- The focus is on the large number of cells regime ( $M \rightarrow \infty$ )
- The setup is equivalent to a two tap time varying ISI channel [Narula '97]
- **Proposition 3** *The limiting average per-cell sum-rate capacity with intra-cell TDMA scheduling, as  $M \rightarrow \infty$ , is given by*

$$C_{tdma}(\bar{P}) = \int_1^{\infty} \log x \frac{\log_e(x) e^{-\frac{x}{\bar{P}}}}{\text{Ei}\left(\frac{1}{\bar{P}}\right) \bar{P}} dx ,$$

where  $\text{Ei}(x) = \int_1^{\infty} \frac{\exp(-xt)}{t} dt$  is the exponential integral function.

- Following [Narula '97] it can be shown that  $C_{tdma}(\bar{P}) \leq C_{ul-nf}(\bar{P})$
- **Proposition 4** *The uplink channel extreme-SNR regimes for Rayleigh-fading channels and intra-cell TDMA scheduling are characterized by*

$$\begin{aligned} \mathcal{S}_0 &= 1 \quad ; \quad E_b^t / N_{0\min} = \frac{\log_e 2}{2} \\ \mathcal{S}_{\infty} &= 1 \quad ; \quad \mathcal{L}_{\infty} \approx 0.84 . \end{aligned}$$

# Uplink - Continuous Fading - Intra-Cell TDMA

- We assume an absolute continuity condition on the fading
- **Proposition 5** *The limiting average per-cell sum-rate capacity with intra-cell TDMA scheduling, as  $M \rightarrow \infty$  exists, we denote by  $C_{tdma}(\bar{P})$  its value. Moreover,*

$$\begin{aligned} \max \left( E\{\log(1 + \bar{P} |a_m|^2)\}, E\{\log(1 + \bar{P} |b_m|^2)\} \right) &\leq C_{tdma}(\bar{P}) \\ &\leq E\{\log(1 + \bar{P}(|a_m|^2 + |b_m|^2))\}. \end{aligned}$$

- **Proposition 6** *The uplink channel high-SNR regime for continuous-fading channels and intra-cell TDMA scheduling is characterized by*

$$\mathcal{S}_\infty = 1 \quad ; \quad \mathcal{L}_\infty = -2 \max(E\{\log |a_m|\}, E\{\log |b_m|\}) .$$

- The lower bound of Proposition 5 is tight in the high-SNR regime.
- The hypothesis of i.i.d. fading among different users and absolute continuity of the fading can be relaxed.

# Proof of Proposition 6

- The techniques: certain Large Random Hermitian Jacobi Matrices.
  - Usual techniques in Large Random Matrices use the large number of random coefficients (quadratic in size of the matrix)
  - Here, the number of random coefficients is linear in the size of the matrix.
  - $H_M H_M^\dagger$  is similar to a random Schrödinger operator. We use the Thouless formula that was “rediscovered” in [Narula '97].
- The Thouless formula:
  - The eigenvalues of  $H_M H_M^\dagger$  are related to its eigenvectors.
  - Let  $(x_1, \dots, x_M)$  be an eigenvector,  $1/M \det \left( I_M + \bar{P} H_M H_M^\dagger \right)$  is related to its exponential growth, that is  $1/M \log |x_M|$ .
  - We give a proof similar to [Narula '97].



# Proof of Proposition 6 (Cont'd)

- Define  $G_M = I_M + \bar{P}H_M H_M^\dagger$ .
- [Narula '97]: to compute  $\det G_M$ , write  $G_M = L_M U_M$ , where  $L_M$  is lower triangular and  $U_M$  is upper triangular.
- We use the following decomposition:  $G_M U_M = L_M$ .
- Denote

$$G_M = \begin{pmatrix} d_1 & c_2^\dagger & 0 & 0 \\ c_2 & d_2 & c_3^\dagger & 0 \\ 0 & \ddots & \ddots & c_M^\dagger \\ 0 & 0 & c_M & d_M \end{pmatrix}$$

# Proof of Proposition 6 (Cont'd)

The decomposition  $G_M U_M = L_M$  is as follows:

$$\begin{pmatrix} d_1 & c_2^\dagger & 0 & 0 \\ c_2 & d_2 & c_3^\dagger & 0 \\ 0 & \ddots & \ddots & c_M^\dagger \\ 0 & 0 & c_M & d_M \end{pmatrix} \begin{pmatrix} x_1 & x_1 & \cdots & x_1 \\ 0 & x_2 & \cdots & x_2 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & x_M \end{pmatrix} = \begin{pmatrix} -c_2^\dagger x_2 & 0 & \cdots & 0 \\ \ddots & -c_3^\dagger x_3 & \ddots & \vdots \\ 0 & \ddots & \ddots & 0 \\ 0 & 0 & \ddots & -c_{M+1}^\dagger x_{M+1} \end{pmatrix}.$$

- $c_i = \bar{P} a_i b_{i+1}^\dagger, \quad d_i = 1 + \bar{P} (|a_i|^2 + |b_i|^2).$
- $x_0 = 0, \quad x_1 = 1$
- $c_i x_{i-1} + d_i x_i + c_{i+1}^\dagger x_{i+1} = 0.$
- $x_i$  is governed by a random second-order linear recurrence.
- $\frac{1}{M} \log |\det G_M| = \frac{1}{M} \sum_{i=2}^{M+1} \log |c_i| + \frac{1}{M} \log |x_{M+1}|.$
- It is enough to study the sequence  $x_i$ .

# Proof of Proposition 6 (*Cont'd*)

Study of a Markov chain:

- The study of the sequence with random second-order linear recurrence is equivalent to the study of a Markov chain.
- That Markov chain is closely related to the one studied in [Narula '97].
- Since the fading is absolutely continuous, we can use the theory of Harris chains, which are Markov chains in a continuous state-space with adequate regularity properties.

# Proof of Proposition 6 (Cont'd)

An alternative method, the Lyapunov exponents:

- The classical way to deal with random second-order linear recurrence is product of two-by-two random matrices:

$$\begin{pmatrix} x_i \\ x_{i+1} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -c_i/c_{i+1} & -d_i/c_{i+1} \end{pmatrix} \begin{pmatrix} x_{i-1} \\ x_i \end{pmatrix}.$$

- The relevant quantity is the Lyapunov exponent of the two-by-two random matrices:

$$\lim_{M \rightarrow \infty} \frac{1}{M} \log \det \left[ \begin{pmatrix} 0 & 1 \\ -c_M/c_{M+1} & -d_M/c_{M+1} \end{pmatrix} \cdots \begin{pmatrix} 0 & 1 \\ -c_1/c_2 & -d_1/c_2 \end{pmatrix} \right],$$

which exists by ergodic sub-additivity.

- This method can be generalized to more general Wyner-type models with larger interference range.

# Uplink - Continuous Fading - WB: Finite $K > 1$

- Proposition 5 is verified in the WB scheme.
- **Proposition 7** *The uplink channel high-SNR regime for continuous-fading channels in the WB scheme is characterized by*

$$\mathcal{S}_\infty = 1 \quad ; \quad \mathcal{L}_\infty = -E \left\{ \log \left( \frac{e + |\mathbf{b}_m|^2}{K} \right) \right\} ,$$

where the random variables  $e$  and  $\mathbf{b}_m$  are independent, and the distribution of  $e$  is the unique invariant probability of the Markov chain defined by:

$$e_{n+1} = |\mathbf{a}_n|^2 \left( \frac{e_n + |\mathbf{b}_{n-1}|^2 \sin^2(\mathbf{a}_n, \mathbf{b}_{n-1})}{e_n + |\mathbf{b}_{n-1}|^2} \right) .$$

# Uplink - Continuous Fading - WB: Finite $K > 1$ (Cont'd)

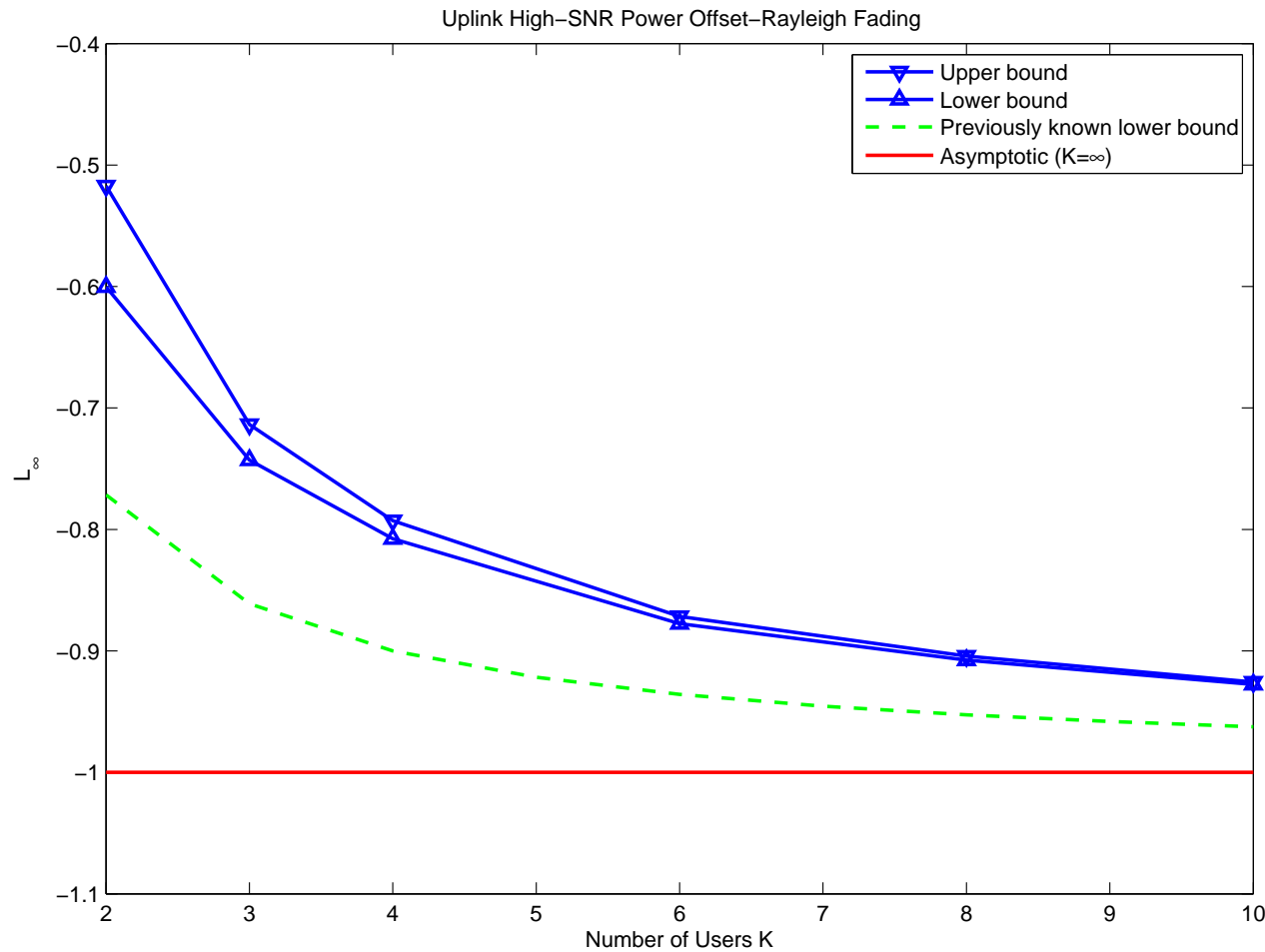
- The proof gives bounds on  $\mathcal{L}_\infty$  tighter than previously known bounds of [Liang-Goldsmith '06].
- For  $n \in \mathcal{N}$ ,

$$-E \left\{ \log \left( \frac{e_n(\infty) + |\mathbf{b}_m|^2}{K} \right) \right\} \leq \mathcal{L}_\infty \leq -E \left\{ \log \left( \frac{e_n(0) + |\mathbf{b}_m|^2}{K} \right) \right\},$$

where  $e_n(a)$  denotes the Markov chain after  $n$  steps starting from  $a$ .

- With  $n = 2$  the bounds are already very tight (see following graph).

# Uplink High-SNR Power Offset - Rayleigh Fading



- The impact of  $K$  on the High-SNR Power Offset in Rayleigh fading channels. Bounds computed with  $n = 2$ .

# Uplink - WB - Fading ( $K \gg 1$ )

- The focus is on the large number of users per cell limit,  $K \gg 1$ , while keeping  $\bar{P} = KP$  constant
- Applying the SLLN, the  $M \times M$  matrix  $\frac{\bar{P}}{K} \mathbf{H}_M \mathbf{H}_M^\dagger$  becomes a circulant matrix with non-zero row entries  $(|m_1|^2 \bar{P}, 2m_2 \bar{P}, |m_1|^2 \bar{P})$
- **Proposition 8** *The average per-cell sum-rate capacity of the WB scheme ( $K \gg 1$ ), in the presence of fading, is given for  $M \rightarrow \infty$  by*

$$C_{ul-wb}(\bar{P}) = \log \left( \frac{1 + 2\bar{P}m_2 + \sqrt{1 + 4\bar{P}m_2 + 4\bar{P}^2(m_2^2 - |m_1|^4)}}{2} \right)$$

- The result upper bounds the corresponding sum-rate capacity for any finite  $K$  [Somekh-Shamai '00]
- $m_1 = 0$  is optimum, and assuming  $m_2 = 1$  (e.g., Rayleigh fading):

$$C_{ul-wb}(\bar{P}) = \log(1 + 2\bar{P})$$



# Uplink - WB - Fading (Cont'd)

- Proposition 9** *The uplink average per-cell sum-rate capacity in the low-SNR regime with the WB scheme is characterized  $\forall K > 0$ , and  $\forall M \geq 3$ , by*

$$\frac{E_b^t}{N_{0 \min}} = \frac{\log_e 2}{2m_2} \quad ; \quad \mathcal{S}_0 = \frac{2}{\frac{\kappa}{2K} + \frac{|m_1|^4}{2m_2^2} + 1} .$$

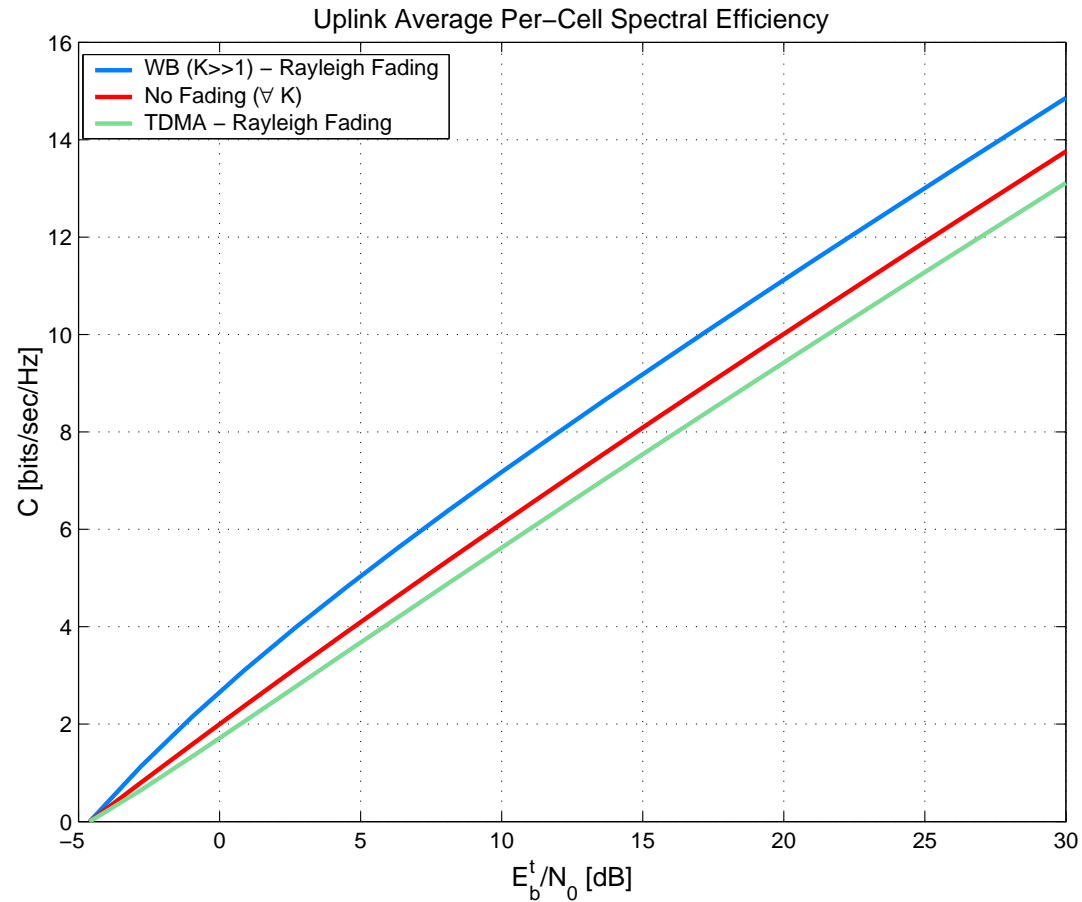
- Proposition 10** *The uplink average per-cell sum-rate capacity in the high-SNR regime with the WB scheme is characterized for  $K \gg 1$ , and  $M \rightarrow \infty$ , by*

$$\mathcal{S}_\infty = 1; \quad \mathcal{L}_\infty = -\log_2 \left( m_2 + \sqrt{m_2^2 - |m_1|^4} \right) .$$

- Proposition 11** *For Rayleigh fading,  $K \gg 1$ , and  $\forall M \geq 3$ , the uplink average per-cell sum-rate capacity in extreme-SNR regimes with the WB scheme is characterized by*

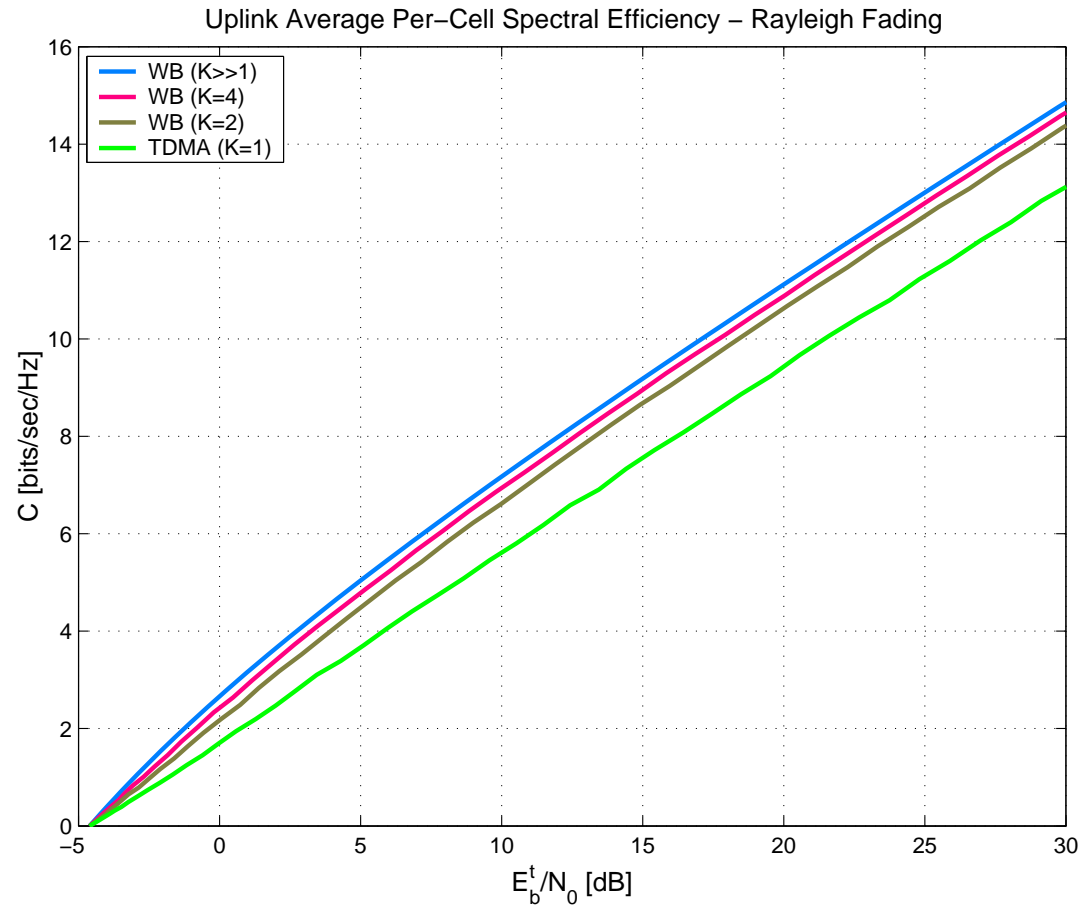
$$E_b^t/N_{0 \min} = \log_e 2/2 \quad ; \quad \mathcal{S}_0 = 2 \quad ; \quad \mathcal{S}_\infty = 1 \quad ; \quad \mathcal{L}_\infty = -1 .$$

# Uplink Average Per-Cell Spectral Efficiency



- Uplink average per-cell spectral efficiency plotted as a function of  $E_b^t/N_0$

# Uplink Average Per-Cell Spectral Efficiency (*Cont'd*)



- The impact of  $K$  on the average per-cell spectral efficiency in Rayleigh fading channels

## Uplink - Activity Controlled Users - Intra-Cell TDMA - The Non-ergodic Regime

- If the channel is random but fixed, the per-cell sum-rate capacity of the uplink channel is given by

$$C_{\text{ul}}(\bar{P}) = \frac{1}{M} \left\{ \log \det \left( \mathbf{I}_M + \frac{\bar{P}}{K} \mathbf{H}_M \mathbf{H}_M^\dagger \right) \right\},$$

which is a random variable

- Simple model of activity controlled independent users:  
 $P((a_i, b_{i+1}) = (1, \alpha)) = p$  (active user) and  $P((a_i, b_{i+1}) = (0, 0)) = 1 - p$  (silent user).
- All users broadcast at a given rate  $R_{\text{target}}$
- The multi-cell processor is fully aware of the CSI.  
In order to decode the message of user  $n$ : identify the first silent user on the left (resp. right) of user  $n$ :  $n_0$  (resp.  $n_1$ ).
- The section between  $n_0 + 1$  and  $n_1 - 1$  is isolated from the rest and can be decoded independently.

## Uplink - Activity Controlled Users - Intra-Cell TDMA - The Non-ergodic Regime

- Denote by  $L$  the number of consecutive active users around user  $n$
- $L$  is distributed, for  $l \geq 1$  as,

$$P(L = l) = l(1 - p)^2 p^{(l-1)}.$$

- **Proposition 12** *The message of the consecutive active users around user  $n$  can be reliably decoded if and only if*

$$R_{target} \leq r_L \triangleq \frac{1}{L} \log \frac{\eta_1^{L+1} - \eta_2^{L+1}}{\eta_1 - \eta_2},$$

where

$$\eta_1 = \frac{1 + \bar{P}(1 + \alpha^2) + \sqrt{(1 + \bar{P}(1 + \alpha)^2)(1 + \bar{P}(1 - \alpha)^2)}}{2},$$

$$\eta_2 = \frac{1 + \bar{P}(1 + \alpha^2) - \sqrt{(1 + \bar{P}(1 + \alpha)^2)(1 + \bar{P}(1 - \alpha)^2)}}{2}.$$

# Uplink - Activity Controlled Users - Intra-Cell TDMA - The Non-ergodic Regime

- The outage capacity is given by

$$P_{outage} = Prob(C_{ul}(\bar{P}) < R_{target}) = p^i (i(1-p) + 1),$$

$$\text{for } r_{i+1} < R_{target} \leq r_i.$$

$$P_{outage} = 0 \text{ if } R_{target} < r_\infty \text{ and } P_{outage} = 1 \text{ if } R_{target} > r_1.$$

$$r_1 = \log(1 + \bar{P}(1 + \alpha^2)),$$

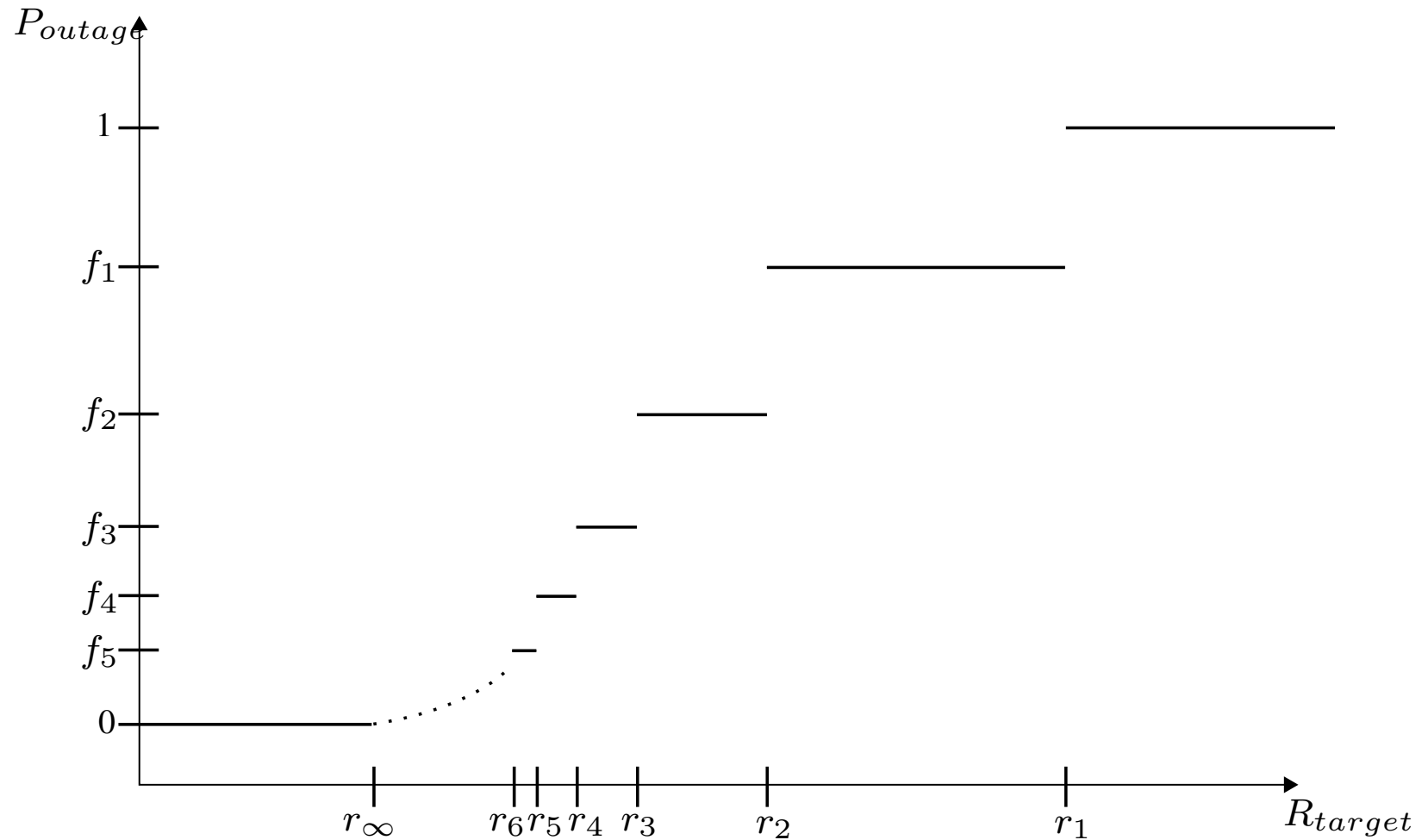
$$r_\infty = \log\left(\frac{1 + (1 + \alpha^2)\bar{P} + \sqrt{1 + 2(1 + \alpha^2)\bar{P} + (1 - \alpha^2)^2\bar{P}^2}}{2}\right).$$

- Ergodic capacity (average over long term activity of users):

$$C_{ul-ergo}(\bar{P}) = \sum_{l=1}^{\infty} (1-p)^2 p^{(l-1)} \log \frac{\eta_1^{l+1} - \eta_2^{l+1}}{\eta_1 - \eta_2}.$$

# Uplink - Activity Controlled Users - Outage probability

We define  $f_i \triangleq p^i(i(1-p) + 1)$ .



Outage probability versus rate.

# Local decoding

- Local decoding: To decode the message of user  $n$ , the multi-cell processor is only allowed to use the messages received at the antennas in a *decoding window* around user  $n$ .
- Remember that the string between  $n_0 + 1$  and  $n_1 - 1$  is isolated from the rest of the users. Therefore, if the decoding window includes  $n_0 + 1$  and  $n_1 - 1$ , local decoding is as good as global decoding.
- As the size of the decoding window grows, the probability that  $n_0 + 1$  and/or  $n_1 - 1$  are not in the decoding window converges exponentially fast to 0.
- This property can be extended to longer range inter-cell interference.
- For  $p$  small enough, this property can be extended to multi-dimensional models.



# Local decoding - a sub-optimal method

- In the case that  $n_0 + 1$  and/or  $n_1 - 1$  are not in the decoding window, we present a sub-optimal decoding method
- If, say,  $n_0 + 1$  is not in the decoding window, there is a user on the left of the window whose signal is received at only one antenna. We consider this user as interfering noise.
- **Proposition 13** *The message of user  $n$  can be decoded using this method if and only if*

$$R \leq \frac{1}{L'} \log \left( v_{L'+1} - (\alpha\rho)^2 \left( \frac{\omega_1}{1 + \rho\alpha^2} + \frac{\omega_2}{1 + \rho} \right) v_{L'} + \frac{(\alpha\rho)^4}{(1 + \rho)(1 + \rho\alpha^2)} \omega_1\omega_2 v_{L'-1} \right),$$

where

$$v_l = \frac{\eta_1^l - \eta_2^l}{\eta_1 - \eta_2},$$

$L'$  is the number of consecutive active users around user  $n$  in the decoding window and  $\omega_1 = 1$  (resp.  $\omega_2$ ) if  $n_0 + 1$  (resp.  $n_0 - 1$ ) is in the decoding window, and 0 otherwise.

# Downlink System Model

- The received  $MK \times 1$  signal vector, is given by

$$\mathbf{y}_{dl} = \mathbf{H}_M^\dagger \mathbf{x}_{dl} + \mathbf{z}_{dl}$$

- $\mathbf{H}_M [M \times KM]$  - Channel transfer matrix
- $\mathbf{x}_{dl} [M \times 1]$  - The vector of signals transmitted by the  $M$  cell-sites. An *equal individual per-cell-site power constraint* is assumed:  
$$\left[ E \left\{ \mathbf{x}_{dl} \mathbf{x}_{dl}^\dagger \right\} \right]_{(m,m)} \leq \bar{P} \quad \forall m$$
- $\mathbf{z}_{dl} [MK \times 1] \sim \mathcal{N}_c(\mathbf{0}, \mathbf{I}_{MK})$  - Circularly symmetric AWGN vector
- Full CSI is available to the joint multi-cell transmitter only
- The mobile receivers are assumed to be cognizant of their own CSI, and of the employed transmission strategy

# Yu-Lan Lagrangian Duality

- **Theorem 1** [Yu-Lan '04] *The sum capacity of the Gaussian multi-antenna broadcast channel is the same as the sum capacity of a dual MAC with a sum power constraint and with a diagonal and “uncertain” noise:*

$$C_{sum} = \min_{\Lambda} \max_{\mathcal{D}} \log \frac{\det \left( \mathbf{H} \mathcal{D} \mathbf{H}^\dagger + \Lambda \right)}{\det (\Lambda)},$$

*such that  $\mathcal{D}$  and  $\Lambda$  are  $K \times K$  and  $N \times N$  nonnegative diagonal matrices, satisfying  $\text{trace}(\mathcal{D}) \leq 1$  and  $\sum_n P_n [\Lambda]_{(n,n)} \leq 1$ , respectively.*

- $P_n$  - transmitted power by antenna  $n$
- Theorem 1 can be directly applied to the downlink channel of the circular cell-array.
- With total average power constraint  $\frac{1}{n} \sum_n P_n \leq \bar{P}$ ,  $\Lambda_{nn} = 1/\bar{P}$  and no minimum is effective.

# Dual Uplink System Model

- The dual uplink channel is described by

$$\tilde{\mathbf{y}}_{ul[M \times 1]} = \mathbf{H}_M \tilde{\mathbf{x}}_{ul} + \tilde{\mathbf{z}}_{ul},$$

- $\mathbf{H}_M[M \times KM]$  - Channel transfer matrix
- $\tilde{\mathbf{x}}_{ul[KM \times 1]}$  - The vector of transmitted symbols with constraints:
  - $E \left\{ \tilde{\mathbf{x}}_{ul} \tilde{\mathbf{x}}_{ul}^\dagger \right\} = \mathbf{D}_M = \text{diag}(\mathbf{D}_0, \mathbf{D}_1, \dots, \mathbf{D}_{M-1})$
  - $\text{trace}(\mathbf{D}_M) \leq 1$
  - $\mathbf{D}_m[K \times K]$  - Nonnegative diagonal matrix, representing the power constraints of the  $m$ th cell's users
- $\tilde{\mathbf{z}}_{ul[M \times 1]} \sim \mathcal{N}_c(0, \mathbf{\Lambda}_M)$  - The circularly symmetric zero mean AWGN
  - Equal individual power constraints  $\Rightarrow \bar{P} \text{trace}(\mathbf{\Lambda}_M) \leq 1$

# Downlink - No-Fading [Somekh-Zaidel-Shamai '07]

- For non-fading channels  $a_{m,k}, b_{m,k} = 1, \forall m, k$
- The channel transfer matrix becomes “block-circulant”
- **Proposition 14** For non-fading channels the average per-cell downlink sum-rate capacity is given for  $M \rightarrow \infty$  by

$$C_{dl-nf}(\bar{P}) = \log \left( \frac{1 + 2\bar{P} + \sqrt{1 + 4\bar{P}}}{2} \right).$$

- **Proposition 15** Extreme-SNR regimes for non-fading channels:

$$\frac{E_b^t}{N_0 \min} = \frac{\log_e 2}{2} \quad ; \quad \mathcal{S}_0 = \frac{4}{3} \quad ; \quad \mathcal{S}_\infty = 1 \quad ; \quad \mathcal{L}_\infty = 0 \quad .$$

- The above results also hold for an *overall* power constraint
- Generalization to  $b_{m,k} = \alpha$  straight forward [Jing-Tse-Hou-Soriaga-Smee-Padovani '07]

$$C_{dl-nf} = \log \left( \frac{1 + (1 + \alpha^2)\bar{P} + \sqrt{1 + 2(1 + \alpha^2)\bar{P} + (1 - \alpha^2)^2\bar{P}^2}}{2} \right).$$

# Downlink - Rayleigh Fading [Somekh-Zaidel-Shamai '07]

- **Proposition 16** For  $K \gg 1$  the downlink average per-cell sum-rate capacity for Rayleigh fading satisfies (ignoring little orders of  $\log_e K$ )

$$\log (1 + \bar{P} ((1 - \epsilon) \log_e K + 2)) \leq C_{dl}(\bar{P}) \leq \log (1 + 2\bar{P} \log_e K) ,$$

for an appropriately chosen  $\epsilon \in (0, 1)$ .

- $\log_e K$ -multi-user diversity.
- Lower bound: scheduling by a Threshold Crossing (TC) policy
  - The rate of the TC policy with  $K \gg 1$  upper bounds the achievable rate of the TC policy with any  $K$
- Upper bound: iid noise, max fading values incorporated with Hadamard and Jensen inequalities

## Downlink - Rayleigh Fading [Somekh-Zaidel-Shamai '07] (*Cont'd*)

- **Proposition 17** *The downlink channel extreme-SNR regimes for Rayleigh fading are characterized by ( $K \gg 1$ )*

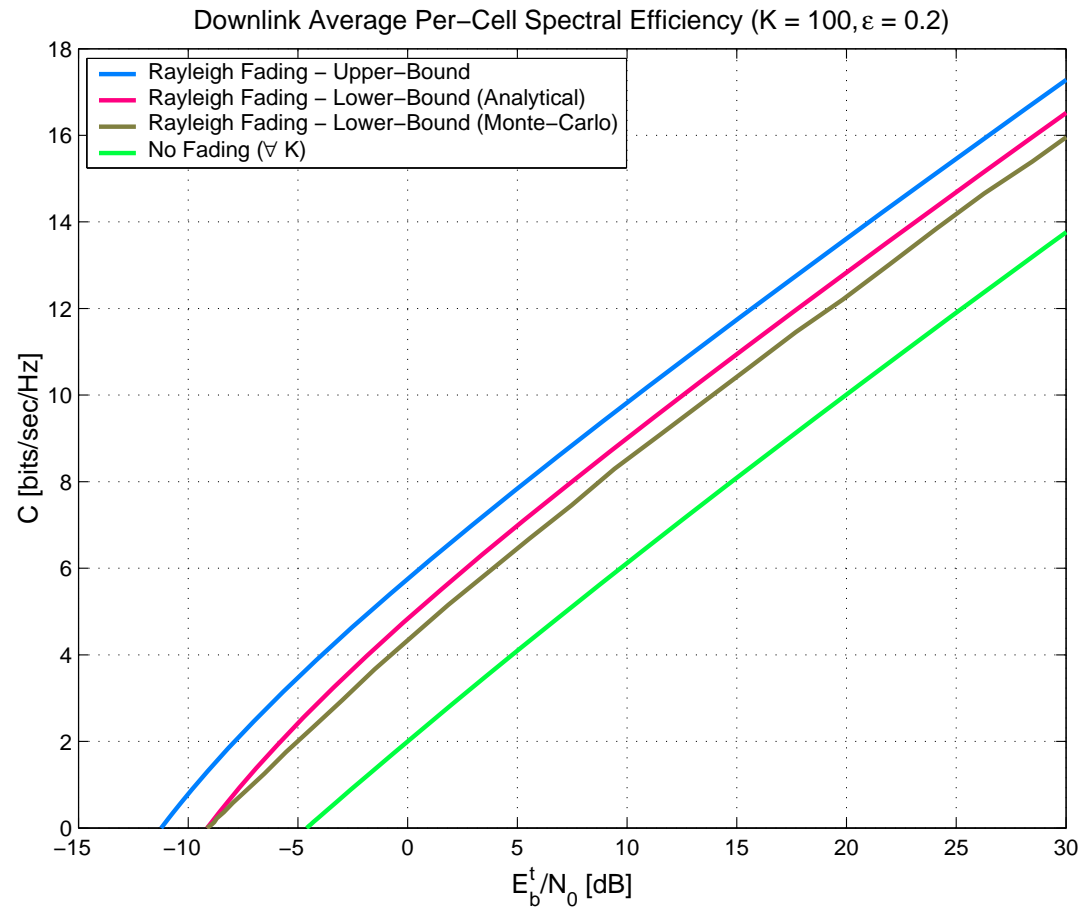
$$\frac{\log_e 2}{2 \log_e K} \leq \frac{E_b^t}{N_{0 \min}} \leq \frac{\log_e 2}{(1 - \epsilon) \log_e K + 2} ;$$

$$\mathcal{S}_0 = 2 ;$$

$$\mathcal{S}_\infty = 1 ;$$

$$-1 - \log_2 \log_e K \leq \mathcal{L}_\infty \leq -\log_2 ((1 - \epsilon) \log_e K + 2) .$$

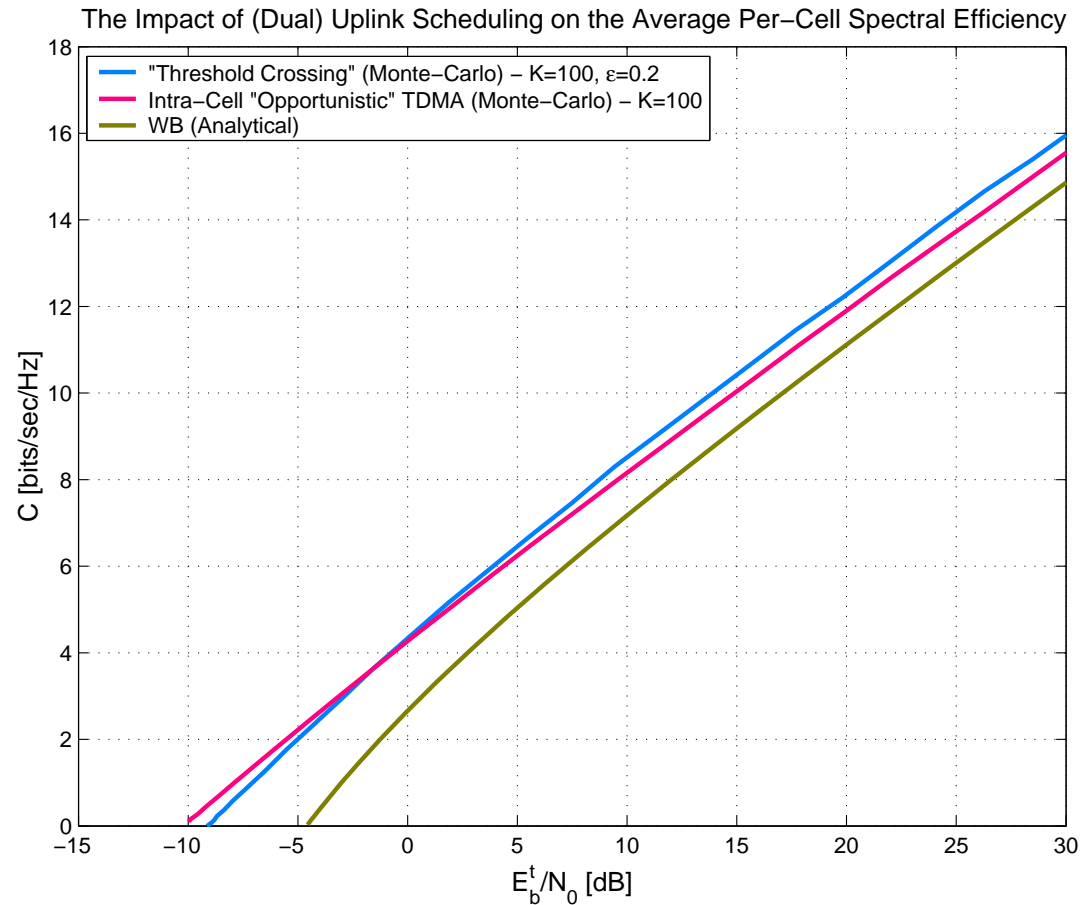
# Downlink Average Per-Cell Spectral Efficiency



- Downlink average pre-cell spectral efficiency ( $K = 100, \epsilon = 0.2$ )



# The Impact of (Dual) Uplink Scheduling



- The impact of (dual) uplink scheduling on the average per-cell spectral efficiency ( $K = 100$ )

# Outlook: Downlink and Fading: Finite (not too large) $K$

- Duality for capacity region, sum-rate:

$$C_{sum} = \min_{\Lambda} \max_{\mathcal{D}} \log \frac{\det \left( H \mathcal{D} H^\dagger + \Lambda \right)}{\det (\Lambda)},$$

such that  $\mathcal{D}$  and  $\Lambda$  are  $K \times K$  and  $N \times N$  nonnegative diagonal matrices, satisfying  $\text{trace}(\mathcal{D}) \leq 1$  and  $\sum_n P_n [\Lambda]_{(n,n)} \leq 1$ , respectively.

- Jacobi type (finite diagonal) large random matrices.  
Elements of  $\mathcal{D}$  depend on  $H$ .
- Lower bound for total power constraint.  $\Lambda = \frac{1}{P} \mathbf{I}$ . Let  $\mathcal{D}$  be  $\mathbf{I}$  (no maximization) and apply the uplink results.
- Scheduling users by activity factor (no information to transmit to users) breaks the system into separate downlinks.  
Generalization of the analysis of the uplink.

# Concluding Remarks

- The impact of joint multiple cell-site processing was demonstrated here through a simple analytically tractable circular (Wyner-like) multi-cell model
- The model represents a practical “soft handoff” scenario
- Analytical results for general fading are obtained using large random Jacobi matrices theory
- Joint (cooperative) multi-cell processing (without spreading or time sharing) eliminates out-of-cell interference, traditionally a limiting factor
- Uplink channel:
  - In the absence of fading, a WB scheme and intra-cell TDMA scheduling are equivalent, and the average per-cell sum-rate capacity approaches the sum-capacity of *a single isolated cell* in the high-SNR regime
  - In flat-fading channels the WB scheme outperforms intra-cell TDMA scheduling
  - Flat-fading enhances performance already for a small  $K$  (for  $K > 2$  and  $K > 1$  in the low-SNR and high-SNR regimes respectively)
  - Fading produces the array diversity gain factor of 2 for  $K \gg 1$
  - Analytical treatment for activity controlled users in the non-ergodic regime

# Concluding Remarks (*Cont'd*)

- Downlink channel:
  - In the absence of fading the downlink and uplink are equivalent
  - For Rayleigh fading and  $K \gg 1$ , a multi-user diversity gain factor of  $\log_e K$  is observed, in addition to the array diversity gain factor of 2
  - The gain is due to cooperation, incorporating the available CSI in the transmitting and receiving ends
  - Restricting joint processing to  $N$  cells can be shown to reduce the number of degrees of freedom by no more than a factor of  $N/(N + 1)$
- The uplink-downlink duality principle guarantees the same uplink multiuser diversity features with proper scheduling
- Can random matrices theory give general answers in the Wyner-type downlink by application of duality (with per-cell power constraint) ?
- Joint multi-cell processing is a key tool in enhancing performance of cellular systems

# Spare Slides

# Downlink Average Per-Cell Sum-Rate Capacity

- Using Theorem 1 the average per-cell sum-rate capacity is given by [Yu-Lan '04]:

$$C_{dl}(\bar{P}) = E_{H_M} \left\{ \frac{1}{M} \min_{\Lambda_M} \max_{\mathcal{D}_M} \log \frac{\det \left( \mathbf{H}_M \mathcal{D}_M \mathbf{H}_M^\dagger + \Lambda_M \right)}{\det (\Lambda_M)} \right\}$$

- The optimization is over all nonnegative diagonal matrices:
  - $\mathcal{D}_M$  [ $MK \times MK$ ], s.t.  $\text{trace}(\mathcal{D}_M) \leq 1$
  - $\Lambda_M$  [ $M \times M$ ], s.t.  $\text{trace}(\Lambda_M) \leq 1/\bar{P}$
- The corresponding spectral efficiency [bits/sec/Hz] is obtained through the relation

$$\bar{P} = C_{dl} \left( \frac{E_b^t}{N_0} \right) \frac{E_b^t}{N_0}$$

where  $E_b^t/N_0$  is the transmit  $E_b/N_0$  and  $C_{dl}(\bar{P}) = C_{dl}(E_b^t/N_0)$ .

# Downlink Average Per-Cell Sum-Rate Capacity (Cont'd)

- The non-zero entries of  $\mathbf{H}_M \mathbf{D}_M \mathbf{H}_M^\dagger$  may be explicitly expressed as

$$\left[ \mathbf{H}_M \mathbf{D}_M \mathbf{H}_M^\dagger \right]_{m,m} = \sum_{k=1}^K d_{m,k} |a_{m,k}|^2 + \sum_{k=1}^K d_{\widehat{m-1},k} |b_{m,k}|^2$$

$$\left[ \mathbf{H}_M \mathbf{D}_M \mathbf{H}_M^\dagger \right]_{m,\widehat{m-1}} = \sum_{k=1}^K d_{\widehat{m-1},k} a_{\widehat{m-1},k}^* b_{m,k}$$

$$\left[ \mathbf{H}_M \mathbf{D}_M \mathbf{H}_M^\dagger \right]_{m,\widehat{m+1}} = \sum_{k=1}^K d_{m,k} a_{m,k}^* b_{\widehat{m+1},k}$$

where

$$\widehat{m+1} \triangleq (m+1) \bmod M \quad ; \quad \widehat{m-1} \triangleq (m-1) \bmod M .$$

# Proof of Proposition 14

- Three observations:
  - The minmax optimization problem is a function of the *total* intra-cell transmit power in the dual uplink  $\Rightarrow$  We particularize to intra-cell TDMA ( $K = 1$ )
  - The minmax optimization problem is convex in  $\Lambda_M$  and concave in  $\mathcal{D}_M$  [Diggavi-Cover '01]
  - For any circulant matrix  $\mathbf{A}_{M[M \times M]}$ , and diagonal matrix  $\mathbf{B}_{M[M \times M]}$ , the quantity  $\det \left( \mathbf{I}_M + \mathbf{A}_M \mathbf{B}_M \mathbf{A}_M^\dagger \right)$  is invariant to cyclic shifts of the diagonal entries of  $\mathbf{B}_M$ .
- Upper Bound:
  - Choose  $\Lambda_M = \Lambda_M^0 = \frac{1}{M\bar{P}} \mathbf{I}_M$
  - Let  $\mathcal{D}_M^*$  be a solution to the remaining maximization problem



# Proof of Proposition 14 (Cont'd)

- Upper Bound (Cont'd):
  - The corresponding upper bound is unaffected by any cyclic shift of the diagonal entries of  $\mathcal{D}_M^*$
  - A “mixing” matrix of all  $M$  cyclic shifts  $\frac{1}{M} \sum_{m=0}^{M-1} \mathcal{D}_M^*(m)$  is also a solution to the maximization problem
  - $\mathcal{D}_M^* = \frac{1}{M} \mathbf{I}_M$  is thus a maximum achieving solution  
 $\Rightarrow$  yields and upper bound
- Lower Bound:
  - Choose  $\mathcal{D}_M = \mathcal{D}_M^0 = \frac{1}{M} \mathbf{I}_M$
  - Using similar arguments it follows that  $\Lambda_M^* = \frac{1}{M\bar{P}} \mathbf{I}_M$  is a solution to the remaining minimization problem
- The two bounds agree with  $\frac{1}{M} \log \det \left( \mathbf{I}_M + \bar{P} \mathbf{H}_M \mathbf{H}_M^\dagger \right)$
- The proof is completed by applying well known results on the eigenvalues of circulant matrices [Gray '72]

# Local Cell-Site Cooperation in the Downlink

- Underlying assumptions:
  - Joint pre-processing for a cluster of no more than  $N$  cells
  - Multiple reduced complexity transmitters are used (each with full CSI regarding its own cluster)
  - It is assumed that  $M \gg N$
- **Proposition 18** *The downlink average per-cell sum-rate capacity, with pre-processing restricted to  $N$  cells, is lower bounded for non-fading channels by*

$$C_{rp,dl}(\bar{P}) \geq \frac{1}{N+1} \sum_{n=0}^{N-1} \log \left( 1 + 2\bar{P} \left( 1 + \cos \left( 2\pi \frac{n}{N} \right) \right) \right) .$$

*The high-SNR slope (“pre-log”, “multiplexing gain”) is lower bounded by*

$$\mathcal{S}_{\infty rp} \geq \begin{cases} \frac{N}{N+1} & N \text{ odd,} \\ \frac{N-1}{N+1} & N \text{ even.} \end{cases}$$

# Restricted Cell-Site Cooperation in the Downlink (*Cont'd*)

- **Proposition 19** *The downlink average per-cell sum-rate capacity, with pre-processing restricted to  $N$  cells, is lower bounded for Rayleigh fading by*

$$C_{rp,dl}(\bar{P}) \geq \frac{N}{N+1} \log \left( 1 + \bar{P} \left( (1 - \epsilon) \log_e K + 2 \right) \right), \quad 0 < \epsilon < 1.$$

*The high-SNR slope (“pre-log”, “multiplexing gain”) is lower bounded by*

$$\mathcal{S}_{\infty rp} \geq \frac{N}{N+1}.$$

- It can be concluded that restricting joint pre-processing to no more than  $N$  cells, reduces the number of degrees of freedom by no more than a factor of  $N/(N+1)$
- The relatively low performance degradation due to restricted processing is a result of the particular system model (each user only “sees” two cell-sites)
- Even when no CSI whatsoever is available, one loses no more than  $1/2$  of the total degrees of freedom (inter-cell time sharing). In this case 2-fold diversity factor can be obtained via two-cell-site cooperation in both uplink (each user is received by two cell-sites), and downlink (Alamouti)

# Proof of Propositions 18 & 19

## ● Step 1:

- Assume no signals are transmitted to users located in cells indexed by integer multiples of  $N + 1 \Rightarrow$  Lower bound
- This generates clusters of  $N$  cells with  $N + 1$  cell-sites
- No inter-cluster interference
- Each cluster can be viewed as a *linear* cell-array
- Consider the dual uplink following [Yu-Lan '04]

## ● Step 2:

- Reduce the number of observables in the dual uplink by adding the signal received at the  $(N + 1)$ th cell-site to the one received at the first cell-site
- This yields a lower bound to the sum-rate capacity (data processing inequality)

# Proof of Propositions 18 & 19 (*Cont'd*)

- Step 3:
  - Observe that the resulting sum-rate capacity equals the capacity with full joint pre-processing for a *circular* array of  $N$  cells
- Step 4:
  - Multiply the result by  $N/(N + 1)$  to account for the fact that users of the  $(N + 1)$ th cell and its multiples are ignored

# Proof of Proposition 16

- Lower (Achievable Rate) Bound:
  - Choose a particular *dual uplink* input covariance matrix complying to a “Threshold Crossing” (TC) scheme
  - Only users received at both cell-sites with fade power levels exceeding some constant  $L$  (i.e.,  $|a_{m,k}|^2, |b_{\widehat{m+1},k}|^2 \geq L$ ), are allowed to transmit
  - As  $K \rightarrow \infty$  the number of active users per cell crystallizes to  $K_0 \triangleq K e^{-2L}$ , transmitting at power  $1/(K_0 M)$
  - $L$  should be chosen so that  $K_0 \rightarrow \infty$  as  $K \rightarrow \infty$
  - Let  $K_0 = K e^{-2L} = K^\epsilon \Rightarrow L = \frac{1-\epsilon}{2} \log_e K$ , where  $0 < \epsilon < 1$
  - $\Lambda_M = \frac{1}{M\bar{P}} \mathbf{I}_M$  is the solution to the remaining minimization problem (Jensen)
  - This achievable rate, obtained for  $K \gg 1$ , upper bounds the achievable rate of the TC policy with any  $K$

# Proof of Proposition 16 (Cont'd)

## ● Upper Bound:

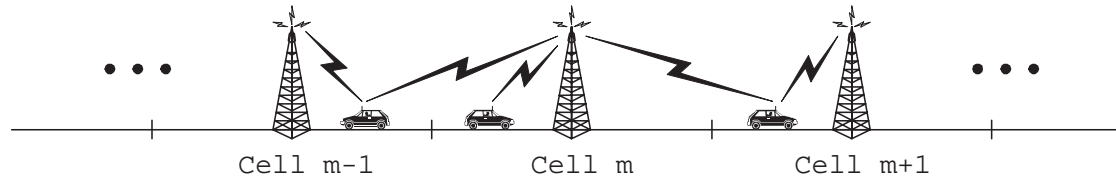
- Choose a particular noise covariance matrix  $\mathbf{\Lambda}_M^0 = \frac{1}{M\bar{P}}\mathbf{I}_M$
- Apply the Hadamard inequality to the resulting  $\max \log \det(\cdot)$  problem
- Bound the channel fades by the strongest fading gain (over all intra-cell users) received at each cell-site
- Observe that the maximum of  $K$  i.i.d.  $\chi^2(2)$  distributed random variables behaves like  $f(K) \triangleq \log_e K + O(\log_e \log_e K)$  for  $K \gg 1$  [Sharif-Hassibi '05]
- Apply Jensen's inequality to the resulting bound

# Proof of Proposition 16 (Cont'd)

- Crystallization Argument (Lower Bound):
  - The number of users scheduled for transmission in each cell (in an arbitrary fading block) is a Binomial r.v. with probability of “success”
$$p = e^{-2L} = \frac{1}{K^{(1-\epsilon)}}$$
  - For  $K \gg 1$  the Binomial distribution is well approximated by a Normal distribution  $\mathcal{N}(m, \sigma^2)$  where  $m = Kp = K^\epsilon$  and  $\sigma^2 = Kp(1-p)$
  - But  $\frac{\sigma^2}{m^2} = \frac{1-p}{Kp} = \frac{1}{K^\epsilon} - \frac{1}{K} \xrightarrow{K \rightarrow \infty} 0, \forall \epsilon > 0$   
 $\Rightarrow$  The probability mass is concentrated around the mean value  $m$
  - Since  $m = K^\epsilon \xrightarrow{K \rightarrow \infty} \infty$ , it is concluded that the number of users scheduled for transmission is approximately  $K^\epsilon \gg 1$  with high probability



# Wyner '94



- A fully synchronous optimally coded system, with cells ordered in an *infinite* linear array (the two-dimensional hexagonal array was also considered)
- Only adjacent cell interference is present and characterized by a single parameter  $0 \leq \alpha \leq 1$
- Non-fading channels
- A “wideband” transmission scheme: All bandwidth is available for coding (as opposed to *random* spreading)
- Joint processing of the signals received at all system cell-sites

# Wyner '94 (Cont'd)

- Received signal at the  $n$ th cell-site:

$$y_n = \alpha \sum_{k=1}^K x_{n-1,k} + \sum_{k=1}^K x_{n,k} + \alpha \sum_{k=1}^K x_{n+1,k} + z_n$$

- $\{x_{n,k}\} \sim \mathcal{N}(0, \bar{P})$ ,  $\{z_n\} \sim \mathcal{N}(0, 1)$
- Maximum reliable equal rate per user with optimum joint processing:

$$C_{\text{opt}} = \frac{1}{2K} \int_0^1 \log \left( 1 + (1 + 2\alpha \cos 2\pi\theta)^2 K \bar{P} \right) d\theta$$

- Maximum reliable equal rate per user with MMSE processing:

$$C_{\text{ms}} = -\frac{1}{K} \log \left( \int_0^1 \frac{1}{1 + (1 + 2\alpha \cos 2\pi\theta)^2 K \bar{P}} d\theta \right)$$

# Wyner '94 (Cont'd)

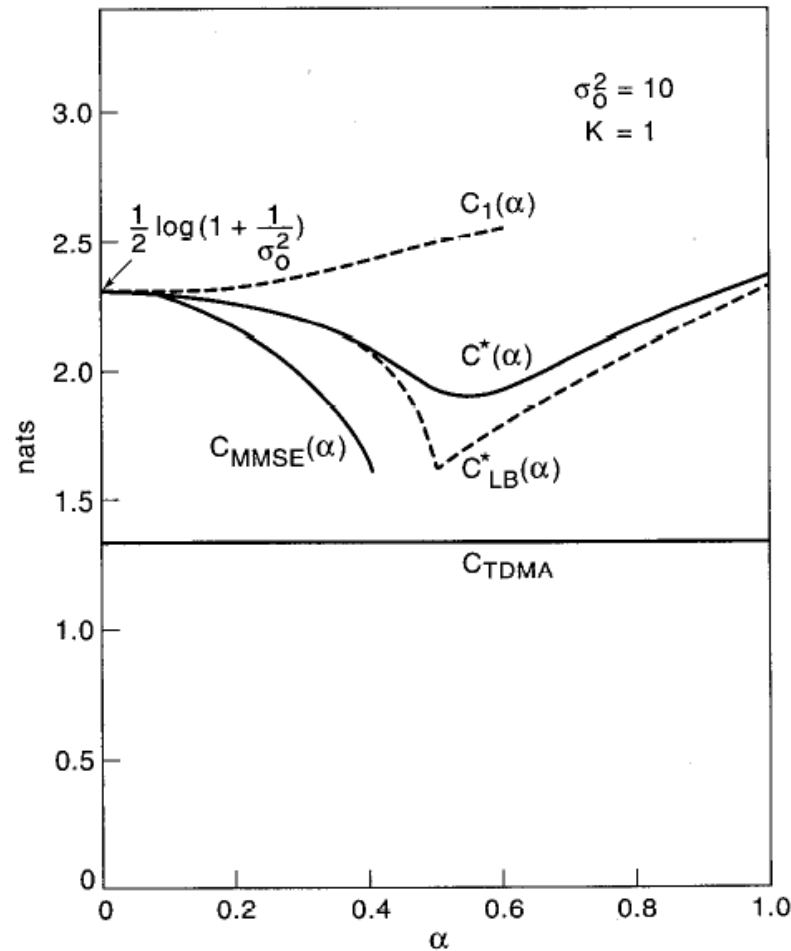


Fig. 1. Plots of  $C^*(\alpha)$ ,  $C_1(\alpha)$ ,  $C_{\text{TDMA}}$ ,  $C_{\text{MMSE}}(\alpha)$ , and  $C_{\text{LB}}^*(\alpha)$  versus  $\alpha$  for  $\sigma_0^2 = 0.01$ ,  $K = 1$  (linear array).  $C_{\text{LB}}^*(\alpha)$  is the lower bound of (2.5).

- A. D. Wyner, "Shannon-theoretic approach to a Gaussian cellular multiple-access channel," *IEEE Trans. Inform. Theory*, vol. 40, pp. 1713–1727, Nov. 1994.

# Somekh-Shamai '00

- Wyner's *infinite* linear array model
- Flat-fading channels (i.i.d., strictly stationary and ergodic, unit power fading processes)
- Received signal at the  $n$ th cell-site:

$$y_n = \alpha \sum_{k=1}^K b_{n,k} x_{n-1,k} + \sum_{k=1}^K a_{n,k} x_{n,k} + \alpha \sum_{k=1}^K c_{n,k} x_{n+1,k} + z_n$$

- $\{a_{n,k}\}, \{b_{n,k}\}, \{c_{n,k}\}$ : independent fading coefficients
- $\{x_{n,k}\} \sim \mathcal{N}_c(0, \bar{P}), \{z_n\} \sim \mathcal{N}_c(0, 1)$
- Information theoretic upper and lower bounds for the maximum reliable equal rate per user are obtained

# Somekh-Shamai '00 (Cont'd)

- Maximum reliable asymptotic ( $K \gg 1$ ) equal rate per user with optimum joint processing (upper bound for any finite  $K$ ):

$$C_{\text{opt}} = \frac{1}{K} \int_0^1 \log \left( 1 + K \bar{P} \left( \sigma_a^2 (1 + 2\alpha^2) + (1 - \sigma_a^2) (1 + 2\alpha \cos(2\pi\theta))^2 \right) \right) d\theta$$

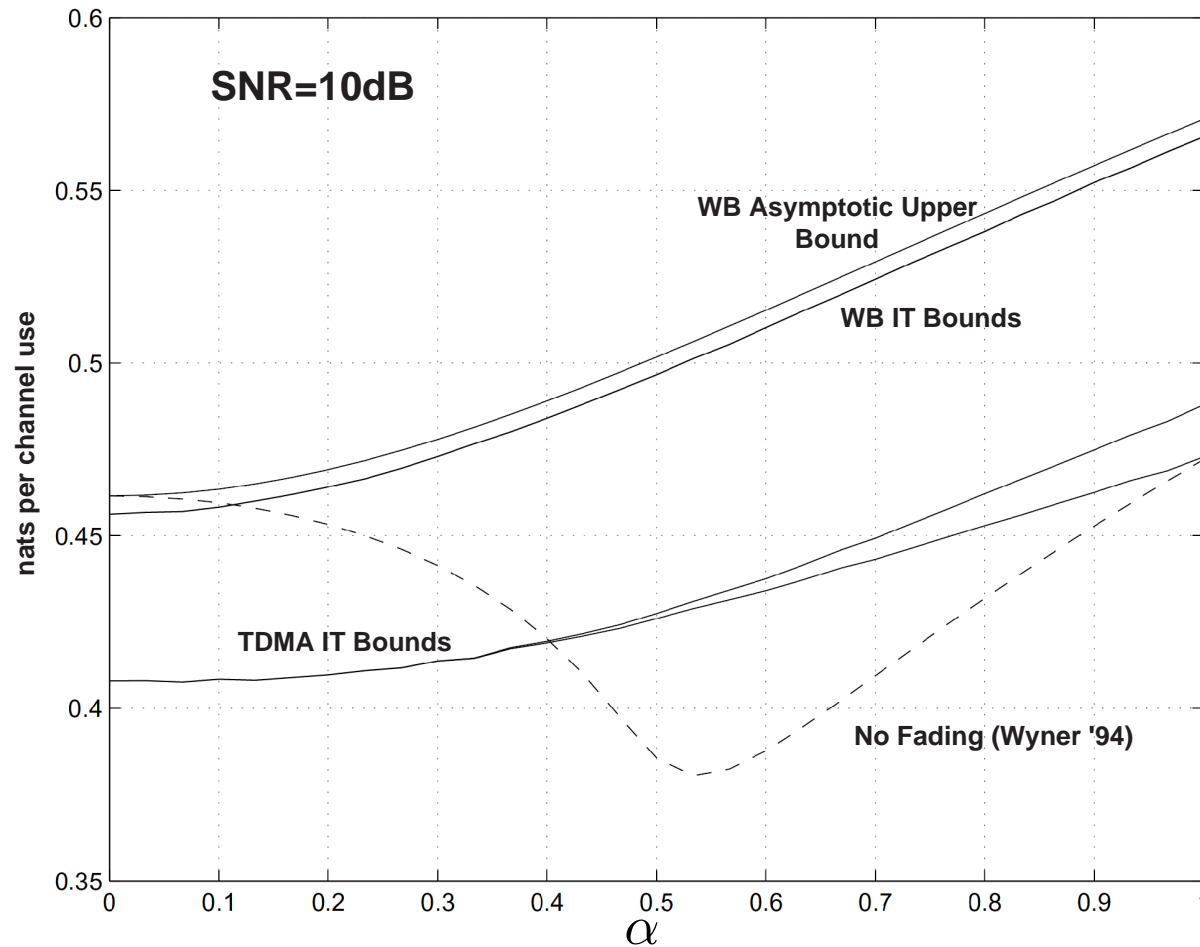
where  $\sigma_a^2 = \sigma_b^2 = \sigma_c^2 = E \{ |a|^2 \} - |E \{ a \}|^2$

- For any zero mean fading process (e.g., Rayleigh fading):

$$C_{\text{opt}} = \frac{1}{K} \log (1 + K \bar{P} (1 + 2\alpha^2))$$

- The result is identical to the single isolated cell setup with a scaled transmit power of  $\bar{P}(1 + 2\alpha^2)$

# Somekh-Shamai '00 (Cont'd)

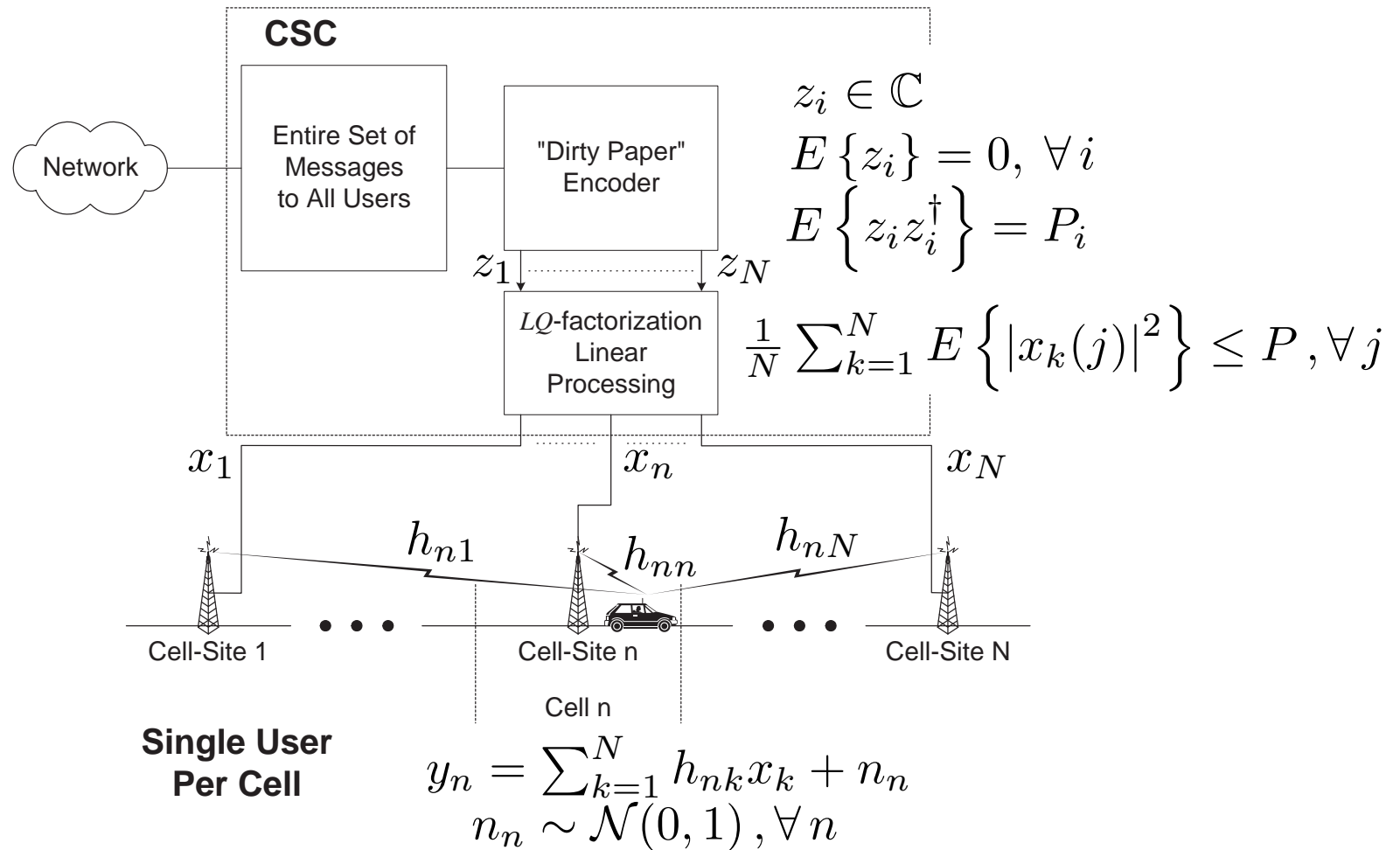


- O. Somekh and S. Shamai (Shitz), “Shannon-theoretic approach to Gaussian cellular multi-access channel with fading,” *IEEE Trans. Inform. Theory*, vol. 46, pp. 1401–1425, July 2000.

# Shamai-Zaidel '01

- Wyner's infinite linear cell-array model (applied to the downlink channel)
- Single active user per cell (TDMA, orthogonal CDMA)
- Single antennas at both cell-sites and user terminals
- No cooperation of user receivers
- All cell-sites are connected to a central cell-site controller with full CSI, that knows the messages to be transmitted to all users
- A joint  $LQ$ -factorization based linear pre-processing is employed, combined with a dirty-paper encoding scheme (LP-DP), putting the whole complexity burden on the transmitting end
- An *overall* system average power constraint  $\bar{P}$  is assumed

# Shamai-Zaidel '01 (Cont'd)



- Schematic description of the cellular system model
- S. Shamai (Shitz) and B. M. Zaidel, "Enhancing the cellular downlink capacity via co-processing at the transmitting end," in *Proc. of IEEE VTC 2001 Spring*, vol. 3, (Rhodes, Greece), pp. 1745–1749, May 6–9, 2001.



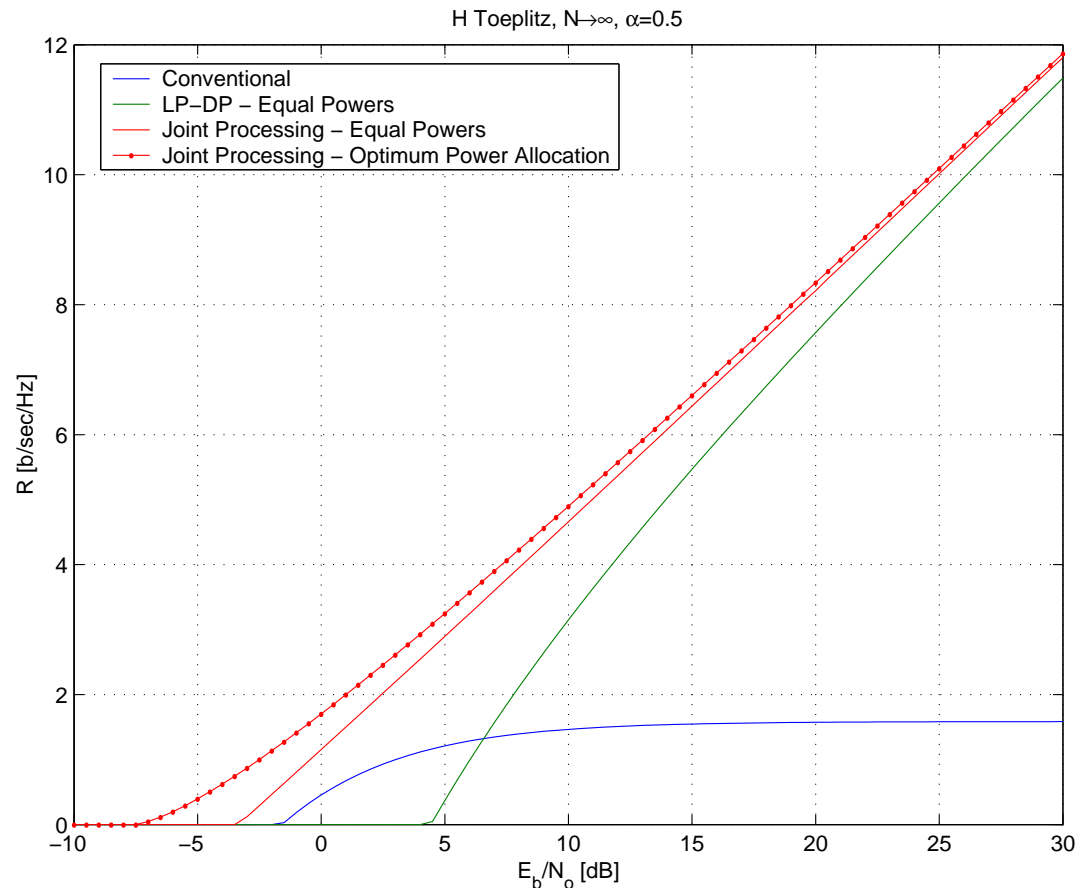
# Shamai-Zaidel '01 (*Cont'd*)

- For non-fading channels equal power allocation is optimal, and the average attainable rate (per user) is given by

$$\bar{R}^{opt} = \begin{cases} \log \left( 1 + \left( \frac{1 + \sqrt{1 - 4\alpha^2}}{2} \right)^2 \bar{P} \right) & 0 \leq \alpha \leq \frac{1}{2} \\ \log (1 + \alpha^2 \bar{P}) & \alpha \geq \frac{1}{2} \end{cases}$$

- The average attainable rate of the LP-DP scheme approaches the rate of the optimum joint decoder in the high-SNR region.
- Numerical results for Rayleigh flat-fading show a similar behavior

# Shamai-Zaidel '01 (Cont'd)



- Average attainable rates without fading, for  $\alpha = \frac{1}{2}$  and  $N \rightarrow \infty$
- S. Shamai (Shitz) and B. M. Zaidel, “Enhancing the cellular downlink capacity via co-processing at the transmitting end,” in *Proc. of IEEE VTC 2001 Spring*, vol. 3, (Rhodes, Greece), pp. 1745–1749, May 6–9, 2001.

# MIMO BC-MAC Capacity Regions

[Weingarten-Steinberg-Shamai '04]

- DPC capacity region:

$$\mathcal{C}^{\text{DPC}}(P, \mathbf{H}_{1,\dots,m}) = \text{conv} \left\{ \bigcup_{\pi \in \Pi} \bigcup_{\mathbf{B}_1, \dots, \mathbf{B}_m} \left( R_1^{\text{DPC}}(\pi, \mathbf{B}_1, \dots, \mathbf{B}_m, \mathbf{H}_{1,\dots,m}), \dots, R_m^{\text{DPC}}(\pi, \mathbf{B}_1, \dots, \mathbf{B}_m, \mathbf{H}_{1,\dots,m}) \right) \right\}$$

- $\Pi$  is the set of all permutations
- $\pi \in \Pi$  - reverse decoding order ( $\pi_{m+1-j}$  -  $j$ th user to be decoded)
- $R_i^{\text{DPC}}(\pi, \mathbf{B}_1, \dots, \mathbf{B}_m, \mathbf{H}_{1,\dots,m}) = \log \frac{|\mathbf{H}_{\pi_i} (\sum_{j=1}^i \mathbf{B}_{\pi_j}) \mathbf{H}_{\pi_i}^\dagger + \mathbf{I}_{r_{\pi_i} \times r_{\pi_i}}|}{|\mathbf{H}_{\pi_i} (\sum_{j=1}^{i-1} \mathbf{B}_{\pi_j}) \mathbf{H}_{\pi_i}^\dagger + \mathbf{I}_{r_{\pi_i} \times r_{\pi_i}}|}$
- $r_i$  - number of receive antennas of user  $i$

# MIMO BC-MAC Capacity Regions (Cont'd)

[Weingarten-Steinberg-Shamai '04]

- MAC capacity region with individual power constraints:

$$\mathcal{C}^{\text{MAC}}(\mathbf{P}_{1,\dots,m}, \mathbf{H}_{1,\dots,m}^\dagger) = \text{conv} \left\{ \bigcup_{\pi \in \Pi} \left\{ (R_1, \dots, R_m) : R_i = R_i^{\text{MAC}}(\pi, \mathbf{P}_{1,\dots,m}, \mathbf{H}_{1,\dots,m}^\dagger) \forall i \right\} \right\}$$

- $\mathbf{P}_{1,\dots,m}$  are the individual power constraints

- $\pi$  is the decoding order ( $\pi_j$  -  $j$ th user to be decoded)

- $R_i^{\text{MAC}}(\pi, \mathbf{P}_{1,\dots,m}, \mathbf{H}_{1,\dots,m}^\dagger) = \log \left| \mathbf{I} + \frac{\mathbf{H}_{\pi_i}^\dagger \mathbf{P}_{\pi_i} \mathbf{H}_{\pi_i}}{\sum_{j=i+1}^m \mathbf{H}_{\pi_j}^\dagger \mathbf{P}_{\pi_j} \mathbf{H}_{\pi_j} + \mathbf{I}} \right|$

- MAC capacity region with a total power constraint  $P$ :

$$\mathcal{C}^{\text{Union}}(P, \mathbf{H}_{1,\dots,m}^\dagger) = \bigcup_{\text{Tr}(\sum_{i=1}^m \mathbf{P}_i) \leq P} \mathcal{C}^{\text{MAC}}(\mathbf{P}_{1,\dots,m}, \mathbf{H}_{1,\dots,m}^\dagger)$$

# MIMO BC-MAC Capacity Regions (*Cont'd*)

[Weingarten-Steinberg-Shamai '04]

- Main duality result:

$$\begin{aligned} \mathcal{C}^{\text{BC}}(P, \mathbf{H}_{1,\dots,m}) &= \mathcal{C}^{\text{DPC}}(P, \mathbf{H}_{1,\dots,m}) \\ &= \mathcal{C}^{\text{Union}}(P, \mathbf{H}_{1,\dots,m}^\dagger) \end{aligned}$$

- Fundamental Concept: Enhanced Channel  $\Rightarrow$  mitigates the Minkowski inequality penalty in the vector EPI inequality
- Result does not rely on BC-MAC duality

# DPC Achievable Region of the MIMO BC

[G. Caire, SPWAC 2005]

- Let  $\mathbf{S} \in \mathbb{S}_+$  be an input covariance constraint. The region  $\mathcal{R}_{dpc}(\mathbf{S}; \mathbf{H}_{1,\dots,K}, \mathbf{N}_{1,\dots,K})$  defined by

$$\text{coh} \bigcup_{\pi} \bigcup_{\sum_k \mathbf{B}_k \leq \mathbf{S}} \left\{ \mathbf{R} : R_{\pi_k} \leq \log \frac{\det \left( \mathbf{N}_{\pi_k} + \mathbf{H}_{\pi_k} \left( \sum_{i \leq k} \mathbf{B}_{\pi_i} \right) \mathbf{H}_{\pi_k}^\dagger \right)}{\det \left( \mathbf{N}_{\pi_k} + \mathbf{H}_{\pi_k} \left( \sum_{i < k} \mathbf{B}_{\pi_i} \right) \mathbf{H}_{\pi_k}^\dagger \right)} \right\}$$

is achievable by DPC

- We can think of the above region as achieved by individual Gaussian coding with input covariance matrices  $\mathbf{B}_k$ , where while coding for user  $\pi_k$  we precode to take into account all users  $\pi_i$  with  $i > k$
- Successive precoding order:**  $\pi_K, \pi_{K-1}, \dots, \pi_1$
- $\mathcal{R}_{dpc}(\mathcal{E}; \mathbf{H}_{1,\dots,K}, \mathbf{N}_{1,\dots,K}) = \bigcup_{\text{Tr } \mathbf{S} \leq \mathcal{E}} \mathcal{R}_{dpc}(\mathbf{S}; \mathbf{H}_{1,\dots,K}, \mathbf{N}_{1,\dots,K})$

# Weingarten-Steinberg-Shamai: Main Ideas of a Brilliant Proof

[G. Caire, SPWAC 2005]

- Proving optimality for a given covariance constraint  $\mathbf{S}$  is enough
- Proving optimality for squared invertible  $\mathbf{H}_k$  is enough
- In this case, the general MIMO BC is equivalent to an **aligned MIMO BC** given by:

$$\mathbf{y}_k = \mathbf{x} + \mathbf{w}_k$$

with fixed noise marginals  $\mathbf{w}_k \sim \mathcal{N}_c(\mathbf{0}, \mathbf{N}_k)$

- **Step 1:** For every point  $\mathbf{R} \notin \mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,\dots,K})$ , there exists an **enhanced aligned degraded MIMO BC** whose DPC region outerbounds the original capacity region and does not contain  $\mathbf{R}$
- **Step 2:** The capacity region of an aligned degraded MIMO BC coincides with its DPC region

# Step 1

[G. Caire, SPWAC 2005]

- Since  $\mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,\dots,K})$  is convex, for any  $\mathbf{R} \notin \mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,\dots,K})$  there exist a supporting hyperplane that separates  $\mathbf{R}$  and  $\mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,\dots,K})$
- Given a supporting hyperplane touching  $\mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,\dots,K})$  in  $\mathbf{R}^*$ , achieved by DPC in encoding order  $\pi$ , there exist noise covariance matrices

$$\mathbf{N}'_{\pi_1} \leq \mathbf{N}'_{\pi_2} \leq \dots \leq \mathbf{N}'_{\pi_K}$$

(degraded condition) and

$$\mathbf{N}'_k \leq \mathbf{N}_k, \quad k = 1, \dots, K$$

(enhanced condition) with the same supporting hyperplane at  $\mathbf{R}^*$



# Shortcoming of Direct Bergman's Proof in the Vector Case

[G. Caire, SPWAC 2005]

- Let  $(R_1, R_2) \notin \mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,2})$ . Then there exist  $\mathbf{B}_1, \mathbf{B}_2 = \mathbf{S} - \mathbf{B}_1$  such that

$$R_1 \geq R_{1,dpc} = \log \det(\tilde{\mathbf{N}}_1 + \mathbf{B}_1) - \log \det(\tilde{\mathbf{N}}_1),$$

$$R_2 \geq R_{2,dpc} = \log \det(\tilde{\mathbf{N}}_1 + \tilde{\mathbf{N}}_2 + \mathbf{S}) - \log \det(\tilde{\mathbf{N}}_1 + \tilde{\mathbf{N}}_2 + \mathbf{B}_1)$$

- From Fano's inequality (neglecting vanishing terms), two-user case

$$R_1 \leq \frac{1}{n} I(W_1; \mathbf{Y}_1 | W_2), \quad R_2 \leq \frac{1}{n} I(W_2; \mathbf{Y}_2)$$

which yields

$$\frac{1}{n} h(\mathbf{Y}_1 | W_2) \geq \log \det(\pi e(\tilde{\mathbf{N}}_1 + \mathbf{B}_1))$$

# Shortcoming of Direct Bergman's Proof in the Vector Case (*Cont'd*)

[G. Caire, SPWAC 2005]

- In order to lowerbound  $\frac{1}{n}h(\mathbf{Y}_2) = \frac{1}{n}I(\mathbf{Y}_2; W_2) + \frac{1}{n}h(\mathbf{Y}|W_2) \geq R_2 + \frac{1}{n}h(\mathbf{Y}_2|W_2)$ , we seek a lowerbound to  $\frac{1}{n}h(\mathbf{Y}_2|W_2)$  via the EPI:

$$\frac{1}{Mn}h(\mathbf{Y}_2|W_2) = \frac{1}{Mn}h(\mathbf{Y}_1 + \tilde{\mathbf{Z}}|W_2) \geq \log \left( \det \left( \pi e(\mathbf{B}_1 + \tilde{\mathbf{N}}_1) \right)^{1/M} + \det(\pi e\tilde{\mathbf{N}}_2)^{1/M} \right)$$

- Unfortunately, Minkowski's inequality goes in the wrong direction:

$$\det(\mathbf{K}_1)^{1/M} + \det(\mathbf{K}_2)^{1/M} \leq \det(\mathbf{K}_1 + \mathbf{K}_2)^{1/M}$$

- This is not a problem for  $M = 1$  (scalar case), but it prevents from going on in the vector case !
- If there exists  $\alpha > 0$  such that  $\alpha(\mathbf{B}_1 + \tilde{\mathbf{N}}_1) = \tilde{\mathbf{N}}_2$  (proportionality condition), then Minkowski's inequality holds with equality, and we have

$$\frac{1}{n}h(\mathbf{Y}_2|W_2) \geq \log \det(\pi e(\mathbf{B}_1 + \tilde{\mathbf{N}}_1 + \tilde{\mathbf{N}}_2))$$

# Step 1 (Cont'd)

[G. Caire, SPWAC 2005]

- Under the parallel condition (always true when scalar) we have

$$\begin{aligned}\frac{1}{n}h(\mathbf{Y}_2) &\geq R_2 + \frac{1}{n}h(\mathbf{Y}_2|W_2) \\ &> R_{2,dpc} + \frac{1}{n}h(\mathbf{Y}_2|W_2) \\ &= \log \det \left( \pi e(\tilde{\mathbf{N}}_1 + \tilde{\mathbf{N}}_2 + \mathbf{S}) \right) \\ &\quad - \log \det \left( \pi e(\tilde{\mathbf{N}}_1 + \tilde{\mathbf{N}}_2 + \mathbf{B}_1) \right) + \frac{1}{n}h(\mathbf{Y}_2|W_2) \\ &\geq \log \det \left( \pi e(\tilde{\mathbf{N}}_1 + \tilde{\mathbf{N}}_2 + \mathbf{S}) \right)\end{aligned}$$

contradiction !!!

- This shows that any achievable rate must lie inside the Gaussian DPC region

# Step 2

[G. Caire, SPWAC 2005]

- Consider an aligned degraded MIMO BC, defined by the noise increment covariances  $\tilde{\mathbf{N}}_k = \mathbf{N}_k - \mathbf{N}_{k-1} \geq \mathbf{0}$ ,  $k = 1, \dots, K$ ,  $\mathbf{N}_0 = \mathbf{0}$
- Let  $\mathbf{R} \in \partial\mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}_{1,\dots,K})$  and let  $\mathbf{B}_1, \dots, \mathbf{B}_K$  denote the input covariances achieving  $\mathbf{R}$  (they exist and satisfy  $\sum_k \mathbf{B}_k = \mathbf{S}$ )
- There exist  $\tilde{\mathbf{N}}'_k$  such that  $\sum_{i=1}^k \tilde{\mathbf{N}}'_i \leq \tilde{\mathbf{N}}_k$  (**enhanced channel**),  
 $\alpha_k \left( \sum_{i=1}^k \mathbf{B}_i + \sum_{i=1}^k \tilde{\mathbf{N}}'_i \right) = \tilde{\mathbf{N}}'_{k+1}$  (**proportionally**) and  
 $\mathbf{R} \in \partial\mathcal{R}_{dpc}(\mathbf{S}; \mathbf{N}'_{1,\dots,K})$  achieved by the same input covariances (**rate preservation**)
- We can apply Bergman's converse proof to the enhanced channel and show optimality of Gaussian DPC coding