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# ACHIEVABLE RATE FOR GAUSSIAN MULTIPLE RELAY CHANNELS WITH LINEAR RELAYING FUNCTIONS

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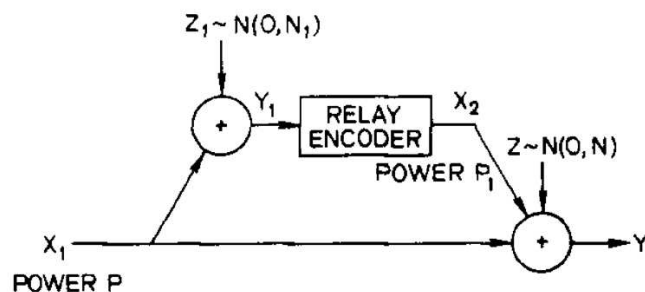
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# Motivation (I)

- Relaying increases capacity in wireless networks.
  - [CovGam79] cooperative source-relay transmission.
  - [WangGian06] opportunistic relaying.



T. Cover and A. El Gamal, "Capacity theorems for the relay channel," *IEEE Trans. on Information Theory*.

- How to operate? decode, compress or amplify?
  - It depends on topology, number of relays and SNR regime.
  - [DanaHass03] amplification achieves capacity scaling law with  $N \rightarrow \infty$ .

# Motivation (II)

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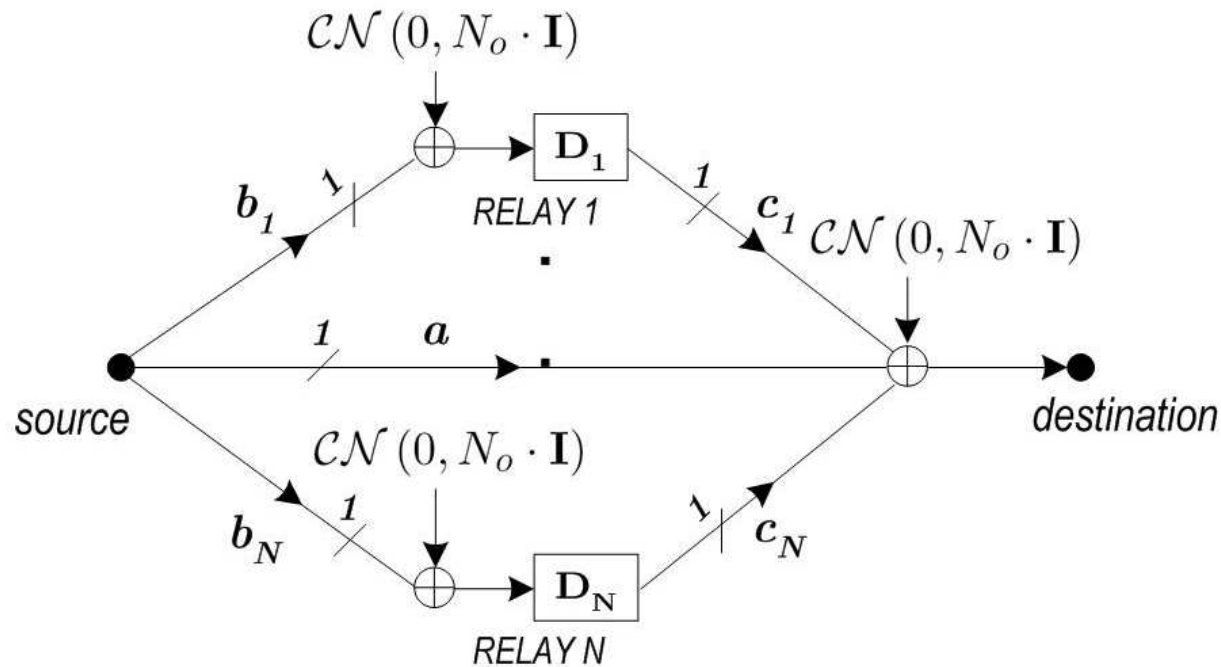
- We study the potentially better amplification scheme: *linear relaying* [Gam et al.06].

$$X_i^t = \sum_{j=0}^{t-1} d_i(t, j) Y_i^j, \quad i = 1, \dots, N.$$

- Resemblance with AWGN channels with memory and colored noise [Gall68] [Hirt88].
  - Strategy: parallel channel decomposition and optimal power allocation.
  - It is not a channel with *time-invariant* ISI: no spectral *waterfilling* can be performed.

# Signal Model and Capacity (I)

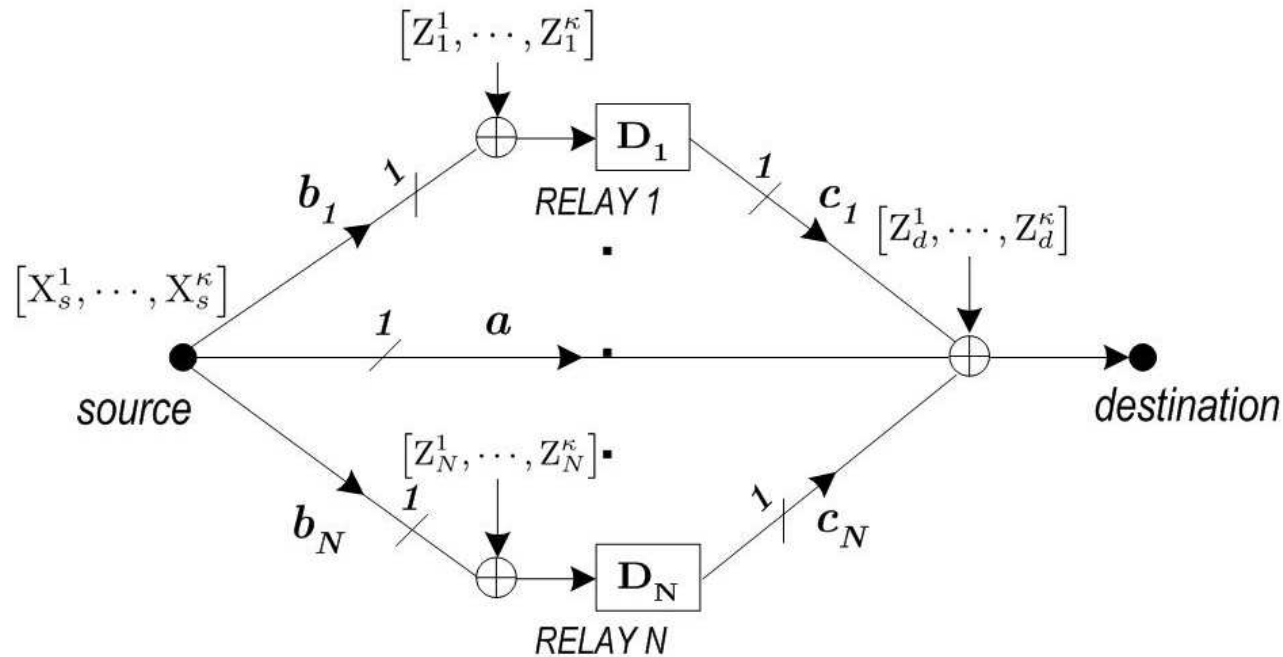
- AWGN multiple relay channel (MRC)



- Remark: parallel relays.

# Signal Model and Capacity (I)

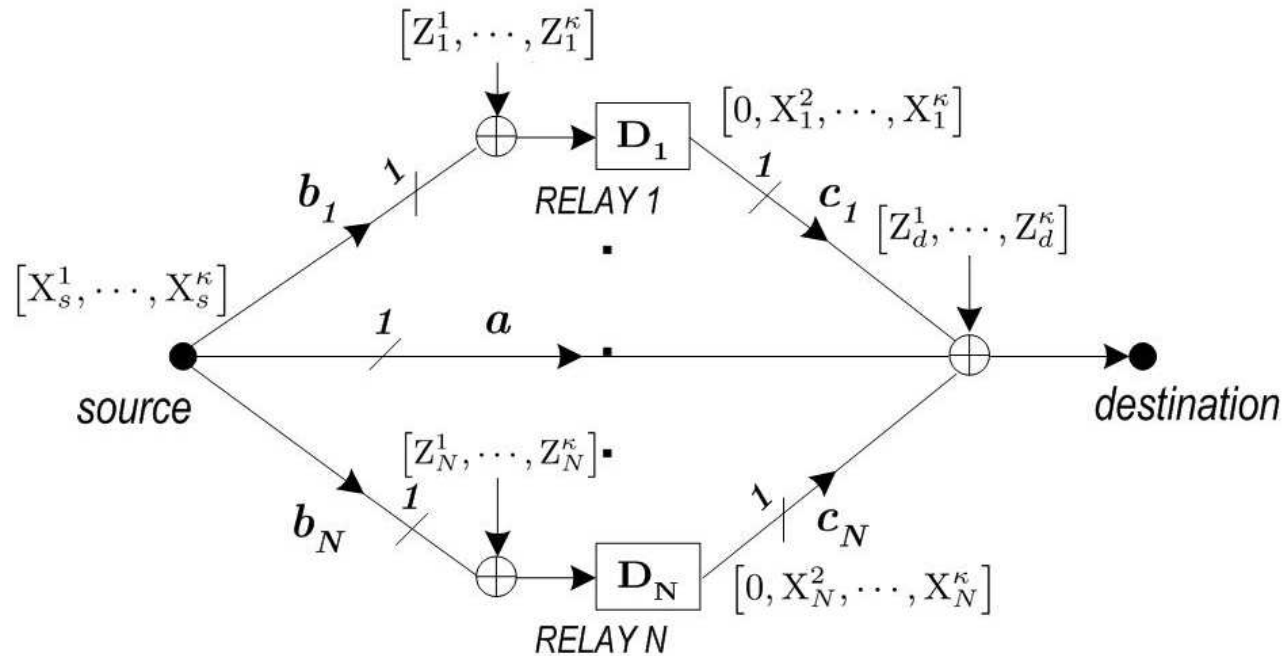
- AWGN multiple relay channel (MRC)



- The source transmits a sequence of  $\kappa$  symbols.

# Signal Model and Capacity (I)

- AWGN multiple relay channel (MRC)

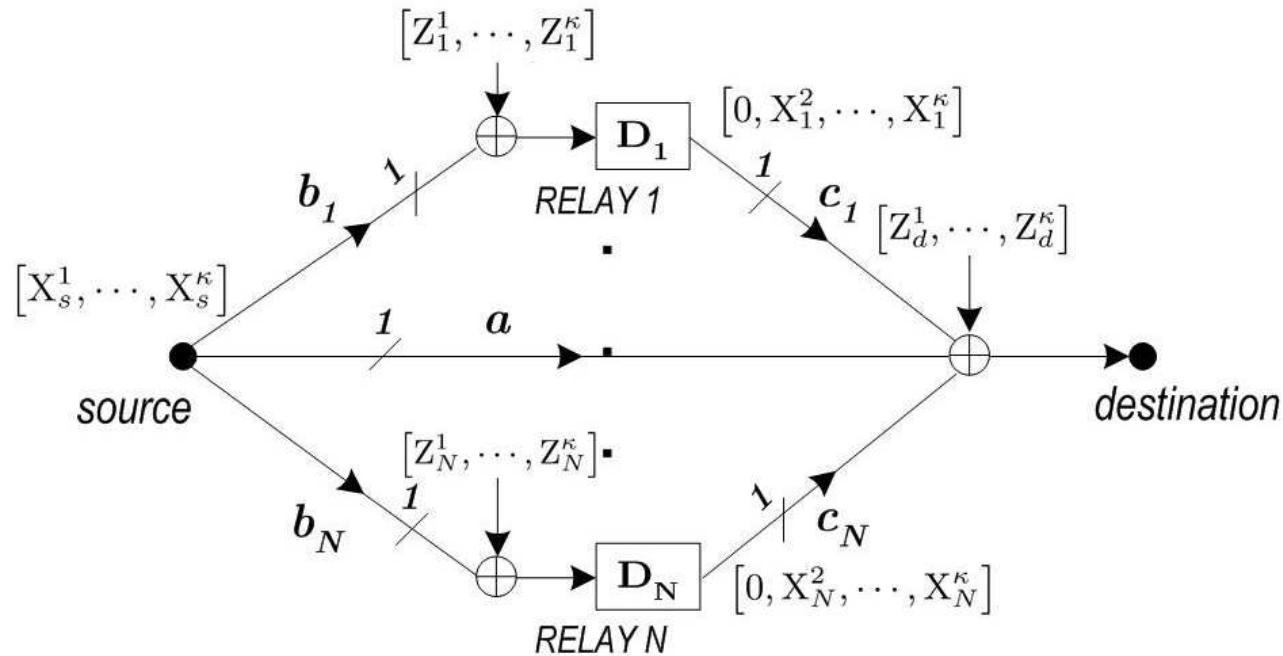


- Relays assist in the communication

$$\mathbf{X}_i^\kappa = \mathbf{D}_i \cdot \mathbf{Y}_i^\kappa = \mathbf{D}_i \cdot (b_i \cdot \mathbf{X}_s^\kappa + \mathbf{Z}_i^\kappa)$$

# Signal Model and Capacity (I)

- AWGN multiple relay channel (MRC)



- The destination receives

$$\mathbf{Y}_d^\kappa = \left( a \cdot \mathbf{I} + \sum_{i=1}^N b_i c_i \mathbf{D}_i \right) \cdot \mathbf{X}_s^\kappa + \left( \mathbf{Z}_d^\kappa + \sum_{i=1}^N c_i \mathbf{D}_i \mathbf{Z}_i^\kappa \right)$$

# Signal Model and Capacity (II)

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- Achievable rate: theory of discrete-time channels with memory [Hirt88]:

$$\mathcal{C} = \lim_{\kappa \rightarrow \infty} \max_{p_{\mathbf{X}_s^\kappa}, \mathcal{D}} \frac{1}{\kappa} \cdot I(\mathbf{X}_s^\kappa; \mathbf{Y}_d^\kappa)$$

with  $\mathcal{D} = \{\mathbf{D}_1, \dots, \mathbf{D}_N\}$ .

- s.t.  $\text{tr} \left\{ \mathbb{E} \left\{ \mathbf{X}_s^\kappa (\mathbf{X}_s^\kappa)^H \right\} \right\} + \sum_{i=1}^N \text{tr} \left\{ \mathbb{E} \left\{ \mathbf{X}_i^\kappa (\mathbf{X}_i^\kappa)^H \right\} \right\} \leq \kappa P$



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- s.t.  $\text{tr} \{ \mathbf{P}(\boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa}, \mathcal{D}) \} \leq \kappa P$  with

$$\mathbf{P}(\boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa}, \mathcal{D}) = \boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa} \left( \mathbf{I} + \sum_{i=1}^N |b_i|^2 \mathbf{D}_i^H \mathbf{D}_i \right) + \sigma_o^2 \sum_{i=1}^N \mathbf{D}_i^H \mathbf{D}_i.$$

# Signal Model and Capacity (II)

- Achievable rate: theory of discrete-time channels with memory [Hirt88]:

$$\mathcal{C} = \lim_{\kappa \rightarrow \infty} \max_{\Sigma_{X_s^\kappa}, \mathcal{D}} \frac{1}{\kappa} \cdot \log_2 \left( \det \left( \mathbf{I} + \mathbf{H}_e \Sigma_{X_s^\kappa} \mathbf{H}_e^H \right) \right)$$

s.t.  $\text{tr} \{ \mathbf{P} (\Sigma_{X_s^\kappa}, \mathcal{D}) \} \leq \kappa P$  ,  $\Sigma_{X_s^\kappa} \succeq 0$

where

$$\mathbf{H}_e = \frac{1}{\sqrt{\sigma_o^2}} \left( \mathbf{I} + \sum_{i=1}^N |c_i|^2 \mathbf{D}_i \mathbf{D}_i^H \right)^{-\frac{1}{2}} \left( a \cdot \mathbf{I} + \sum_{i=1}^N b_i c_i \mathbf{D}_i \right).$$

- Non-convex optimization for  $\kappa > 2$

# Source Covariance Optimization (I)

**Theorem 1** Consider a given set  $\mathcal{D} = \{\mathbf{D}_1, \dots, \mathbf{D}_N\} \in \mathbb{C}^{\kappa \times \kappa}$  with SVD-Decomposition  $\mathbf{H}_e = \mathbf{U} \mathbf{\Lambda}^{\frac{1}{2}} \mathbf{V}^H$ , where  $\mathbf{\Lambda} = \text{diag}([\lambda_1, \dots, \lambda_\kappa])$ . The MRC achieves the rate

$$\begin{aligned} \mathcal{C}(\kappa, \mathcal{D}) &= \max_{\substack{\boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa} : \text{tr}\{\mathbf{P}(\boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa}, \mathcal{D})\} \leq \kappa \mathbf{P} \\ \boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa} \succeq 0}} \frac{1}{\kappa} \cdot \log_2 \left( \det \left( \mathbf{I} + \mathbf{H}_e \boldsymbol{\Sigma}_{\mathbf{X}_s^\kappa} \mathbf{H}_e^H \right) \right) \\ &= \frac{1}{\kappa} \cdot \sum_{n=1}^{\kappa} \log_2 (1 + \lambda_n \psi_n) \end{aligned}$$

where

$$\begin{aligned} \psi_n &= \left[ \frac{1}{\mu \cdot (1 + \phi_n)} - \frac{1}{\lambda_n} \right]^+, \quad \phi_n = \left[ \sum_{i=1}^N \mathbf{V}^H |b_i|^2 \mathbf{D}_i^H \mathbf{D}_i \mathbf{V} \right]_{n,n} \\ \sum_{n=1}^{\kappa} \left[ \frac{1}{\mu} - \frac{1 + \phi_n}{\lambda_n} \right]^+ &= \kappa \mathbf{P} - \text{tr} \left\{ \sigma_o^2 \sum_{i=1}^N \mathbf{D}_i^H \mathbf{D}_i \right\}. \end{aligned}$$

# Source Covariance Optimization (II)

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**Proof:** The Lagrangian of the problem reads

$$\begin{aligned} \mathcal{L}(\boldsymbol{\Sigma}_{X_s^\kappa}, \boldsymbol{\Omega}, \mu) &= \log(\det(\mathbf{I} + \mathbf{H}_e \boldsymbol{\Sigma}_{X_s^\kappa} \mathbf{H}_e^H)) \\ &+ \text{tr}\{\boldsymbol{\Omega} \boldsymbol{\Sigma}_{X_s^\kappa}\} - \mu(\text{tr}\{\mathbf{P}(\boldsymbol{\Sigma}_{X_s^\kappa}, \mathcal{D})\} - \kappa P). \end{aligned}$$

The Karush-Kuhn Tucker conditions are

$$\begin{aligned} i) \quad & \mu \left( \mathbf{I} + \sum_{i=1}^N |b_i|^2 \mathbf{D}_i^H \mathbf{D}_i \right) - \boldsymbol{\Omega} = \mathbf{H}_e^H (\mathbf{I} + \mathbf{H}_e \boldsymbol{\Sigma}_{X_s^\kappa} \mathbf{H}_e^H)^{-1} \mathbf{H}_e \\ ii) \quad & \mu (\text{tr}\{\mathbf{P}(\boldsymbol{\Sigma}_{X_s^\kappa}, \mathcal{D})\} - \kappa P) = 0 \\ iii) \quad & \text{tr}\{\boldsymbol{\Omega} \boldsymbol{\Sigma}_{X_s^\kappa}\} = 0 \end{aligned}$$

For  $\boldsymbol{\Sigma}_{X_s^\kappa} = \mathbf{V} \boldsymbol{\Psi} \mathbf{V}^H$  the KKT conditions (necessary and sufficient for optimality) hold.

# Linear Functions Design

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The optimum linear relaying matrices must satisfy

$$\mathcal{D}^* = \lim_{\kappa \rightarrow \infty} \arg \max_{\mathcal{D} \in \mathcal{C}_{\text{SLT}}^{\kappa \times \kappa}} \mathcal{C}(\kappa, \mathcal{D}).$$

- Non-convex optimization: untractable exhaustive search

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- Suboptimum design:  $\mathcal{D}_{so} \triangleq \{\mathbf{D}_i = \eta_i \mathbf{D}_o(\beta), i = 1, \dots, N\}$ , with

$$\eta_i \triangleq \frac{b_i^* \cdot c_i^*}{\sqrt{\sum_{i=1}^N |b_i \cdot c_i|^2}}$$

$$[\mathbf{D}_o(\beta)]_{p,q} \triangleq \begin{cases} \sqrt{\frac{\beta \kappa}{\kappa-1} \frac{P}{N_o}} & p = q + 1; \quad 1 \leq q \leq \kappa - 1 \\ 0 & \text{elsewhere.} \end{cases}$$

# Linear Functions Design

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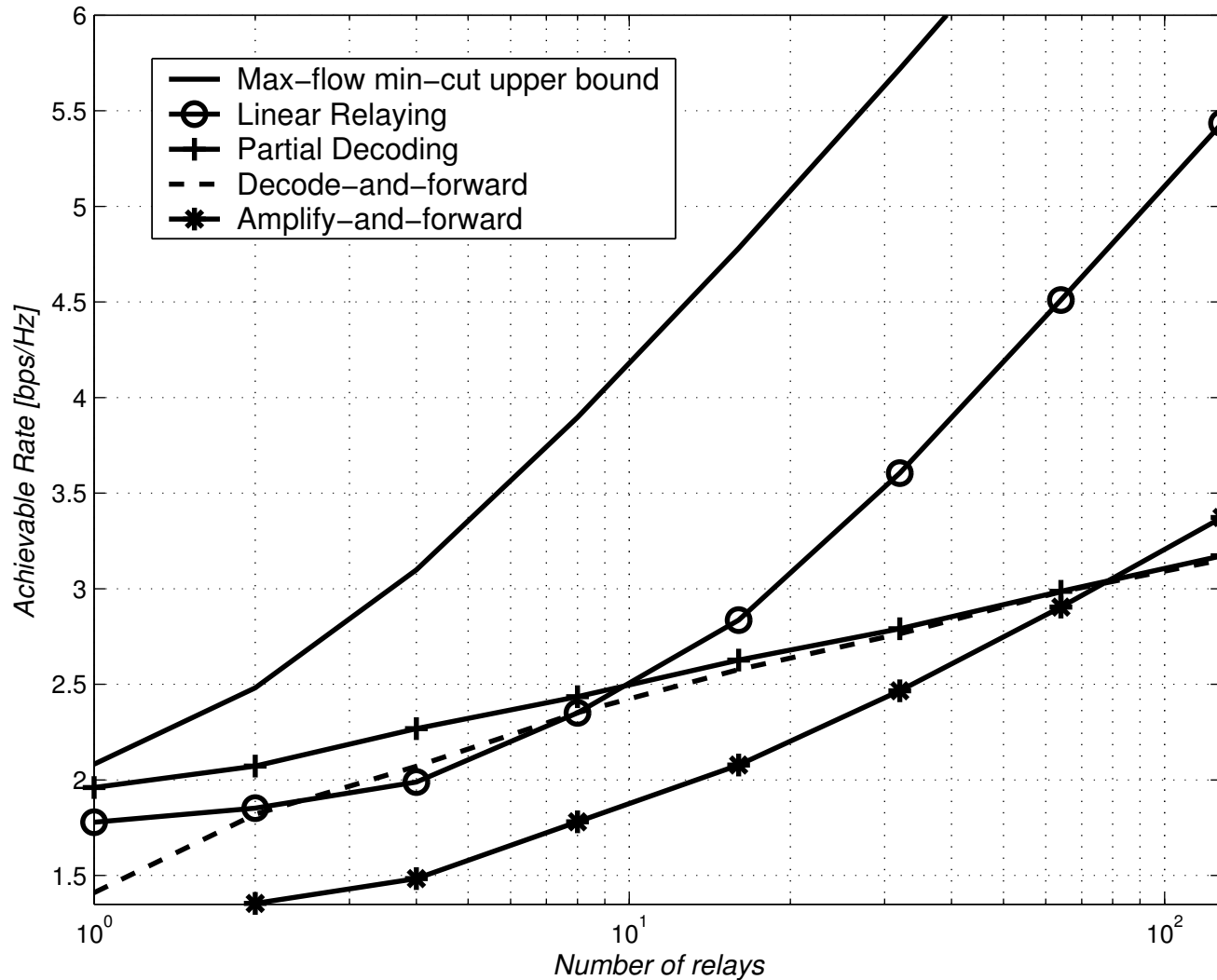
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$$\mathbf{D}_o(\beta) \triangleq \begin{pmatrix} 0 & \dots & 0 \\ \sqrt{\frac{\beta \kappa}{\kappa-1} \frac{P}{N_o}} & \ddots & 0 \\ 0 & \sqrt{\frac{\beta \kappa}{\kappa-1} \frac{P}{N_o}} & 0 \end{pmatrix}$$

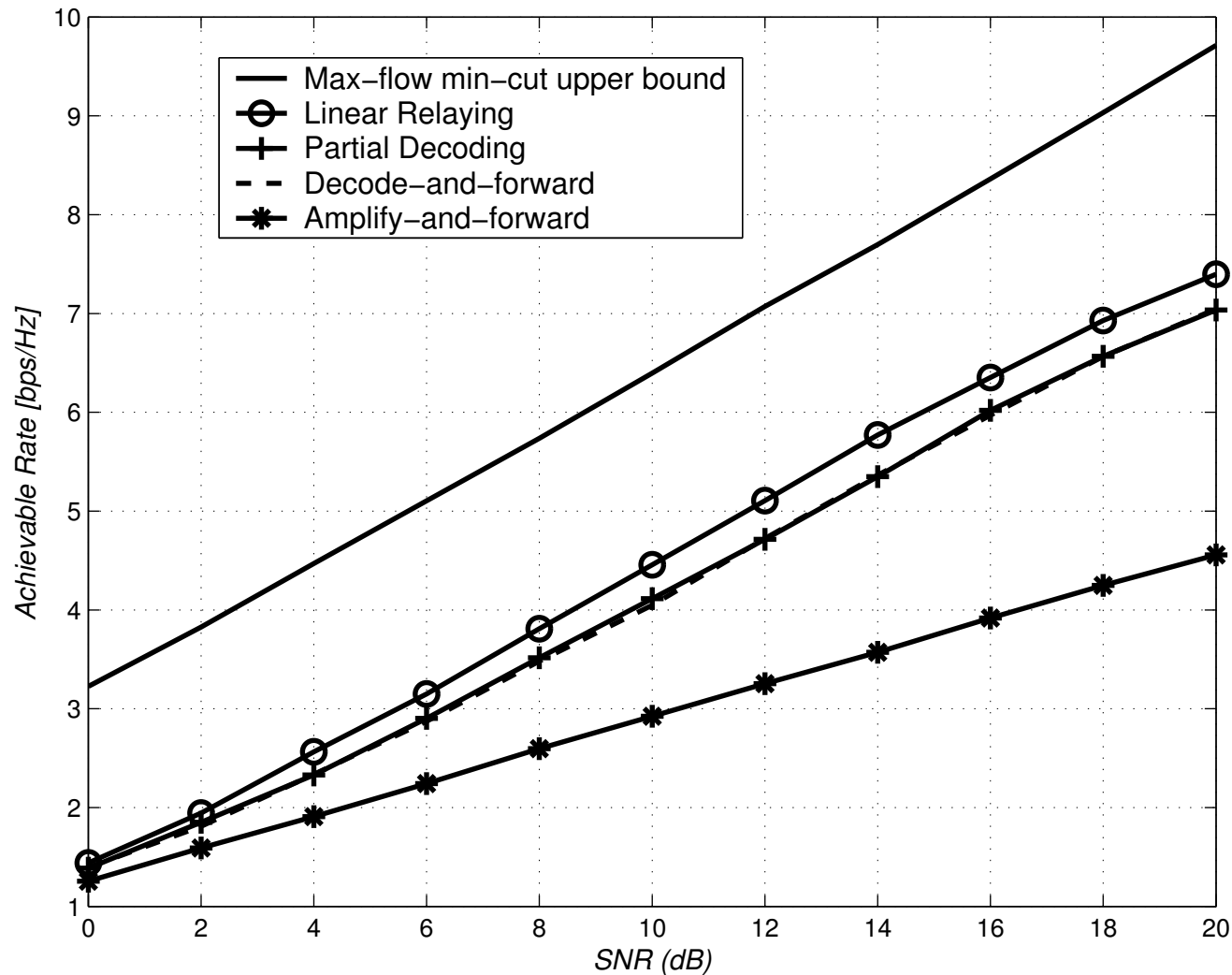
# Numerical Results



Rayleigh-distributed, *time-invariant* channels,  $\mathbb{E}\{\mathcal{C}\}$ .



# Numerical Results



Rayleigh-distributed, *time-invariant* channels,  $\mathbb{E} \{C\}$ .

# Conclusions

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- Amplify-based relaying studied as AWGN channels with block memory and colored noise.
- *Linear relaying* is the most spectrally efficient technique for  $N \rightarrow \infty$ 
  - Capacity doubles with respect to *amplify-and-forward*.
- Shortcoming: worse than *partial decoding* for low number of relays.