
Lecture 3

MIMO&OFDM

- Refer to <<Fundamentals of Wireless Communication>> Chapter 7.1 & 8 & 3.4.4

Part 1: MIMO

➤ Refer to <<Fundamentals of Wireless Communication>> Chapter 7.1 & 8

Capacity of S/ISO AWGN Channel

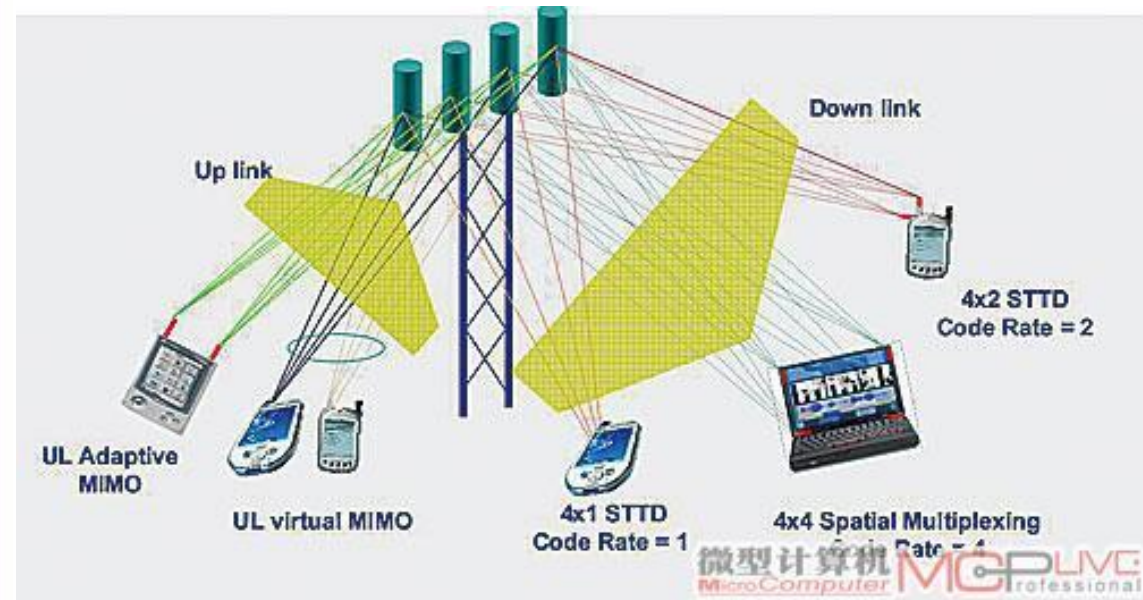
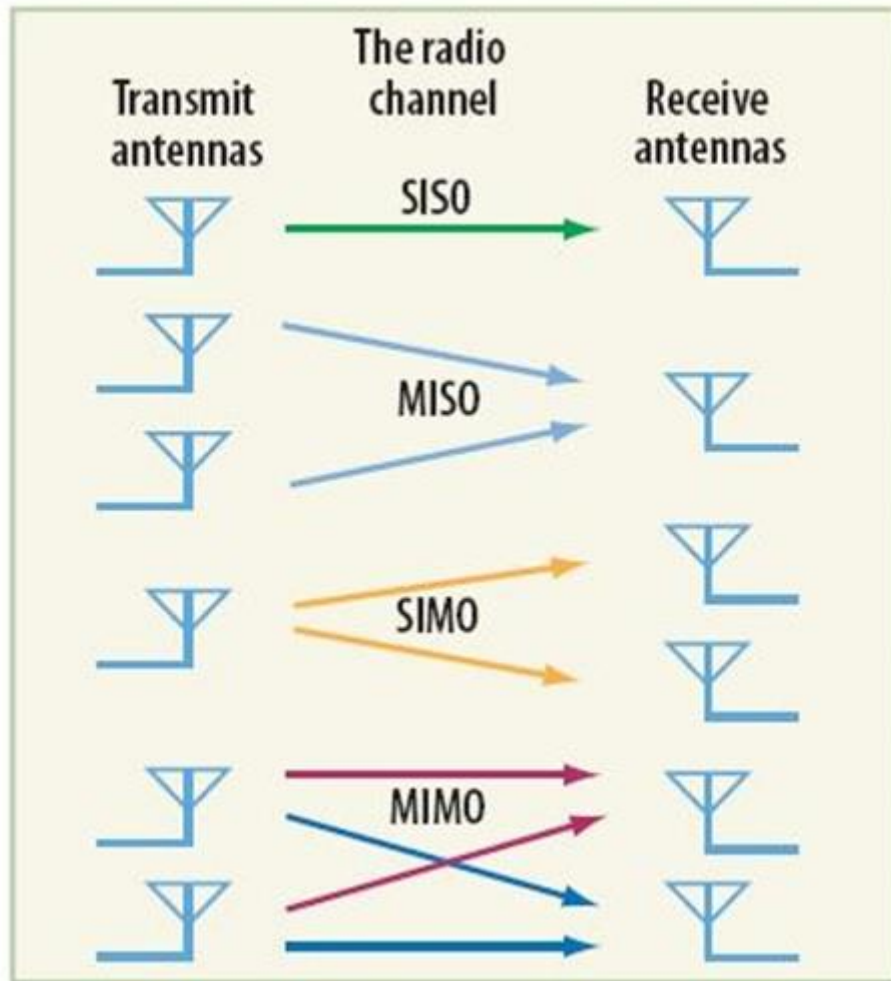
➤ Capacity of AWGN channel

$$\begin{aligned} C_{\text{awgn}} &= \log(1 + \text{SNR}) \quad \text{bits/s/Hz} \\ &= W \log(1 + \text{SNR}) \quad \text{bits/s} \end{aligned}$$

➤ If average transmit power constraint is \bar{P} watts and noise psd is N_0 watts/Hz,

$$C_{\text{awgn}} = W \log \left(1 + \frac{\bar{P}}{N_0 W} \right) \quad \text{bits/s.}$$

Multiple-Antenna System



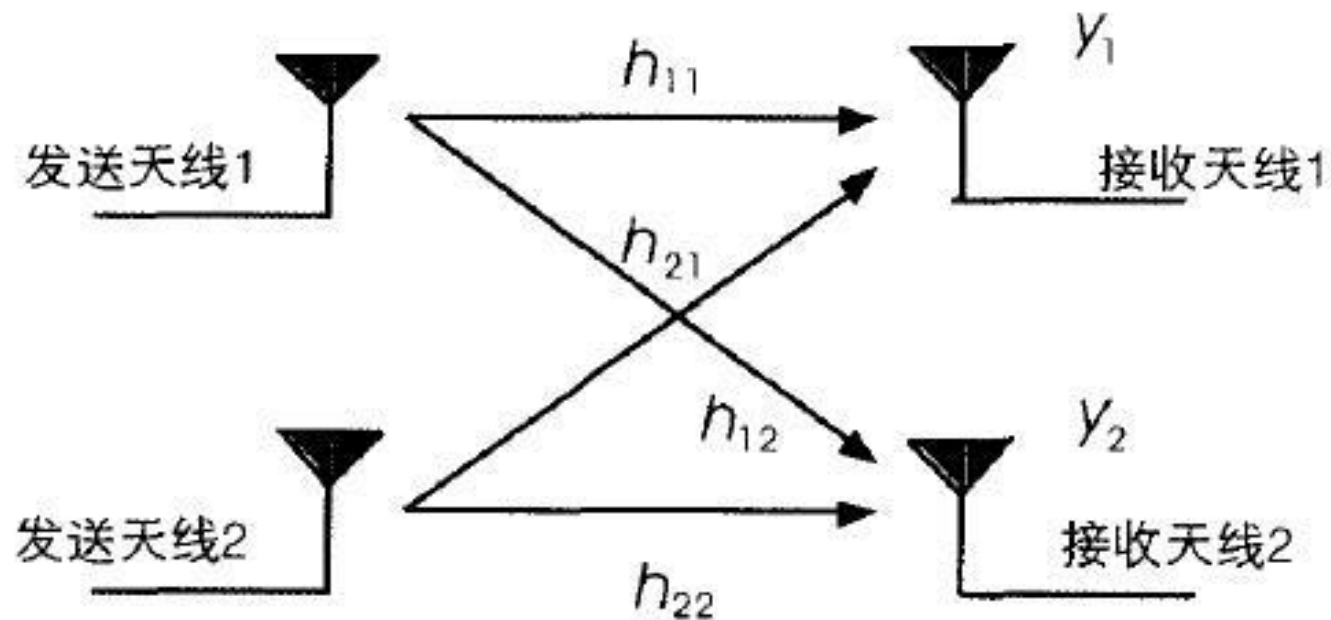
MIMO Channel

Right



MIMO Channel

2x2 Example

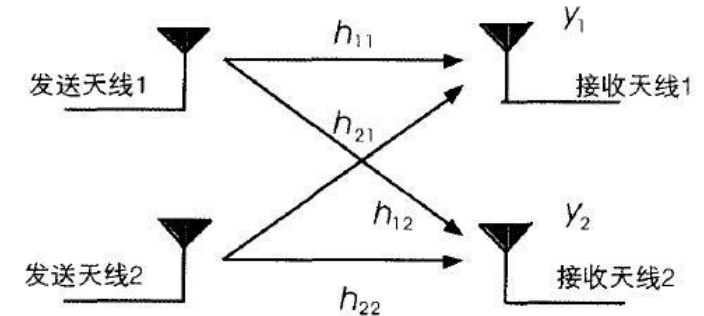


MIMO Capacity via SVD

Narrowband MIMO channel:

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

\mathbf{H} is n_r by n_t , fixed channel matrix.



Singular value decomposition (SVD):

$$\mathbf{H} = \mathbf{U}\mathbf{\Lambda}\mathbf{V}^*$$

\mathbf{U} , \mathbf{V} are complex orthogonal matrices and $\mathbf{\Lambda}$ real diagonal (singular values).

Singular Value Decomposition: SVD

Suppose \mathbf{M} is a $m \times n$ matrix whose entries come from the field K , which is either the field of real numbers or the field of complex numbers. Then there exists a factorization, called a 'singular value decomposition' of \mathbf{M} , of the form

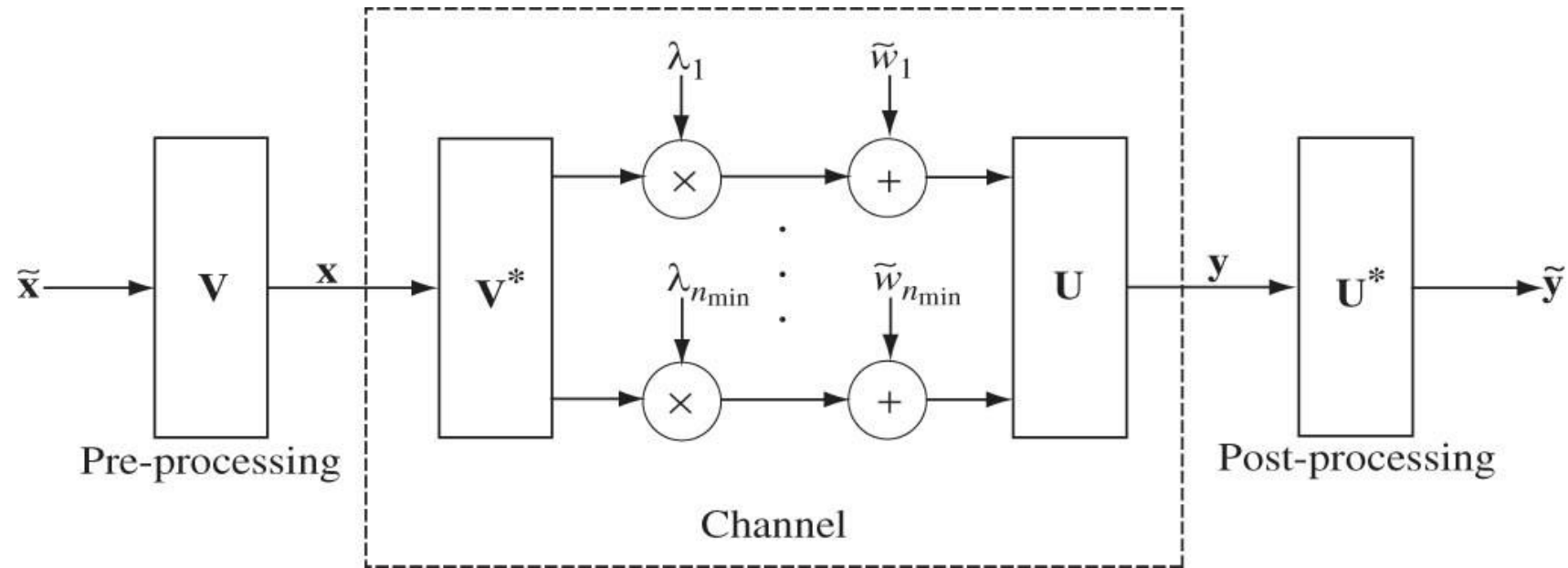
$$\mathbf{M} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^*$$

where

- \mathbf{U} is an $m \times m$ unitary matrix over K (if $K = \mathbb{R}$, unitary matrices are orthogonal matrices),
- $\mathbf{\Sigma}$ is a diagonal $m \times n$ matrix with non-negative real numbers on the diagonal,
- \mathbf{V} is an $n \times n$ unitary matrix over K , and \mathbf{V}^* is the conjugate transpose of \mathbf{V} .

The diagonal entries σ_i of $\mathbf{\Sigma}$ are known as the **singular values** of \mathbf{M} . A common convention is to list the singular values in descending order. In this case, the diagonal matrix, $\mathbf{\Sigma}$, is uniquely determined by \mathbf{M} (though not the matrices \mathbf{U} and \mathbf{V} if \mathbf{M} is not square, see below).

Spatial Parallel Channel



$$\tilde{\mathbf{y}} = \mathbf{\Lambda} \tilde{\mathbf{x}} + \tilde{\mathbf{w}} \quad \tilde{y}_i = \lambda_i \tilde{x}_i + \tilde{w}_i, \quad i = 1, 2, \dots, n_{\min}$$

- Capacity is achieved by **waterfilling** over the eigenmodes of \mathbf{H} . (Analogy to frequency-selective channels.)

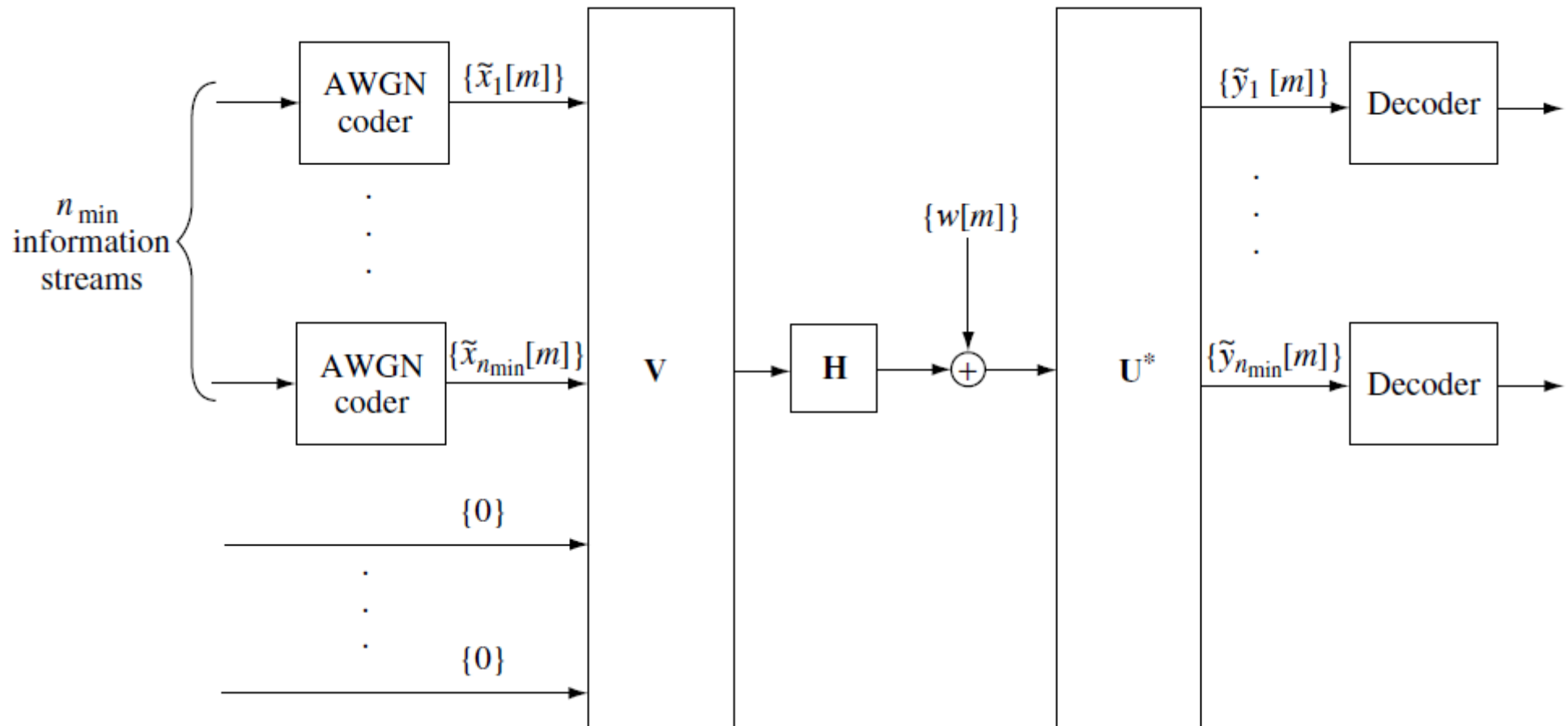
MIMO Capacity by waterfilling

$$C = \sum_{i=1}^{n_{\min}} \log \left(1 + \frac{P_i^* \lambda_i^2}{N_0} \right) \text{ bits/s/Hz},$$

where $P_1^*, \dots, P_{n_{\min}}^*$ are the waterfilling power allocations:

$$P_i^* = \left(\mu - \frac{N_0}{\lambda_i^2} \right)^+,$$

The SVD architecture for MIMO communication



Rank and Condition Number (1)

At high SNR, equal power allocation is near optimal:

$$C \approx \sum_{i=1}^k \log \left(1 + \frac{P \lambda_i^2}{k N_0} \right) \approx k \log \text{SNR} + \sum_{i=1}^k \log \left(\frac{\lambda_i^2}{k} \right)$$

where k is the number of nonzero λ_i^2 's, i.e. the rank of \mathbf{H} .

□ *The parameter k is the number of spatial degrees of freedom per second per hertz.*

Rank and Condition Number (2)

By Jensen's inequality,

$$\frac{1}{k} \sum_{i=1}^k \log \left(1 + \frac{P}{kN_0} \lambda_i^2 \right) \leq \log \left(1 + \frac{P}{kN_0} \left(\frac{1}{k} \sum_{i=1}^k \lambda_i^2 \right) \right)$$

Note

$$\sum_{i=1}^k \lambda_i^2 = \text{Tr}[\mathbf{H}\mathbf{H}^*] = \sum_{i,j} |h_{ij}|^2,$$

The closer the condition number:

$$\frac{\max_i \lambda_i}{\min_i \lambda_i}$$

to 1, the higher the capacity.

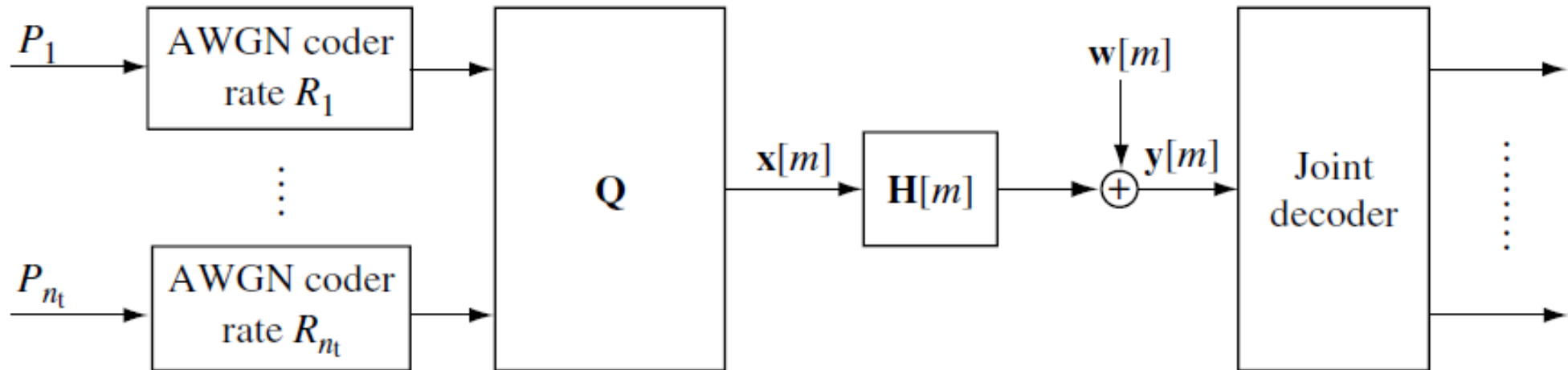
Rank and Condition Number (3)

At **low SNR**, the near optimal policy is to allocate power only to the strongest Eigenmode. The resulting capacity is,

$$C \approx \frac{P}{N_0} \left(\max_i \lambda_i^2 \right) \log_2 e \text{ bits/s/Hz.}$$

- In this regime, the rank or condition number of the channel matrix is less relevant.
- What matters is how much energy gets transferred from the transmitter to the receiver.

The V-BLAST architecture



- If $\mathbf{Q} = \mathbf{V}$ and the powers are given by the waterfilling allocations.
- If $\mathbf{Q} = \mathbf{I}_{n_r}$, then independent data streams are sent on the different transmit antennas

Capacity with precoding

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

The rate achieved in a given channel state \mathbf{H} is

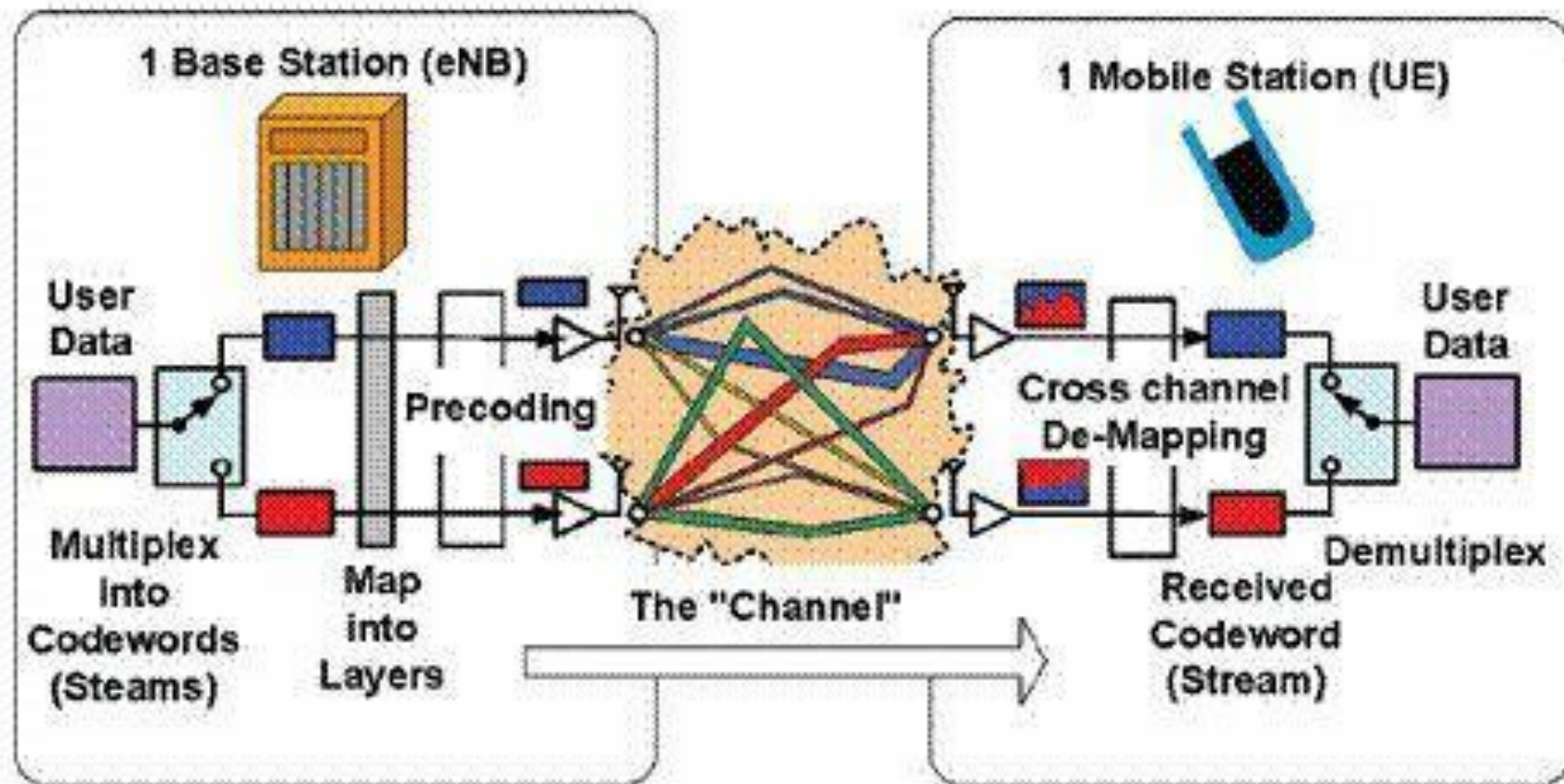
$$\log \det \left(\mathbf{I}_{n_r} + \frac{1}{N_0} \mathbf{H} \mathbf{K}_x \mathbf{H}^* \right)$$

$\mathbf{K}_x = E\{\mathbf{x}\mathbf{x}^*\}$ is the covariance matrix of the transmitted signal \mathbf{x} :

$$\mathbf{K}_x = \mathbf{P} \mathbf{P}^*$$

We call the matrix \mathbf{P} as **precoder**.

Precoding



Capacity with non-AWGN

If w is non-AWGN, and its covariance matrix is $R = E\{w^*w'\}$

$$y = Hx + w$$

The rate achieved in a given channel state H is

$$C = \log \det \left(I + H K_x H^* R^{-1} \right)$$

Proof:

Fast fading MIMO channel with Receiver CSI Only

$$\mathbf{y}[m] = \mathbf{H}[m]\mathbf{x}[m] + \mathbf{w}[m], \quad m = 1, 2, \dots,$$

where $\{\mathbf{H}[m]\}$ is a random fading process.

A long-term rate of reliable communication equal to

$$\mathbb{E}_{\mathbf{H}} \left[\log \det \left(\mathbf{I}_{n_r} + \frac{1}{N_0} \mathbf{H} \mathbf{K}_x \mathbf{H}^* \right) \right]$$

We can now choose the covariance \mathbf{K}_x as a function of the channel *statistics* to achieve a reliable communication rate of

$$C = \max_{\mathbf{K}_x: \text{Tr}[\mathbf{K}_x] \leq P} \mathbb{E} \left[\log \det \left(\mathbf{I}_{n_r} + \frac{1}{N_0} \mathbf{H} \mathbf{K}_x \mathbf{H}^* \right) \right]$$

i.i.d. Rayleigh fading model

For i.i.d. Rayleigh fading model, equal powers are the optimal:

$$C = \mathbb{E} \left[\log \det \left(\mathbf{I}_{n_r} + \frac{\text{SNR}}{n_t} \mathbf{H} \mathbf{H}^* \right) \right]$$

where $\text{SNR} = P/N_0$ is the common SNR at each receive antenna.

$$\begin{aligned} C &= \mathbb{E} \left[\sum_{i=1}^{n_{\min}} \log \left(1 + \frac{\text{SNR}}{n_t} \lambda_i^2 \right) \right] \\ &= \sum_{i=1}^{n_{\min}} \mathbb{E} \left[\log \left(1 + \frac{\text{SNR}}{n_t} \lambda_i^2 \right) \right]. \end{aligned}$$

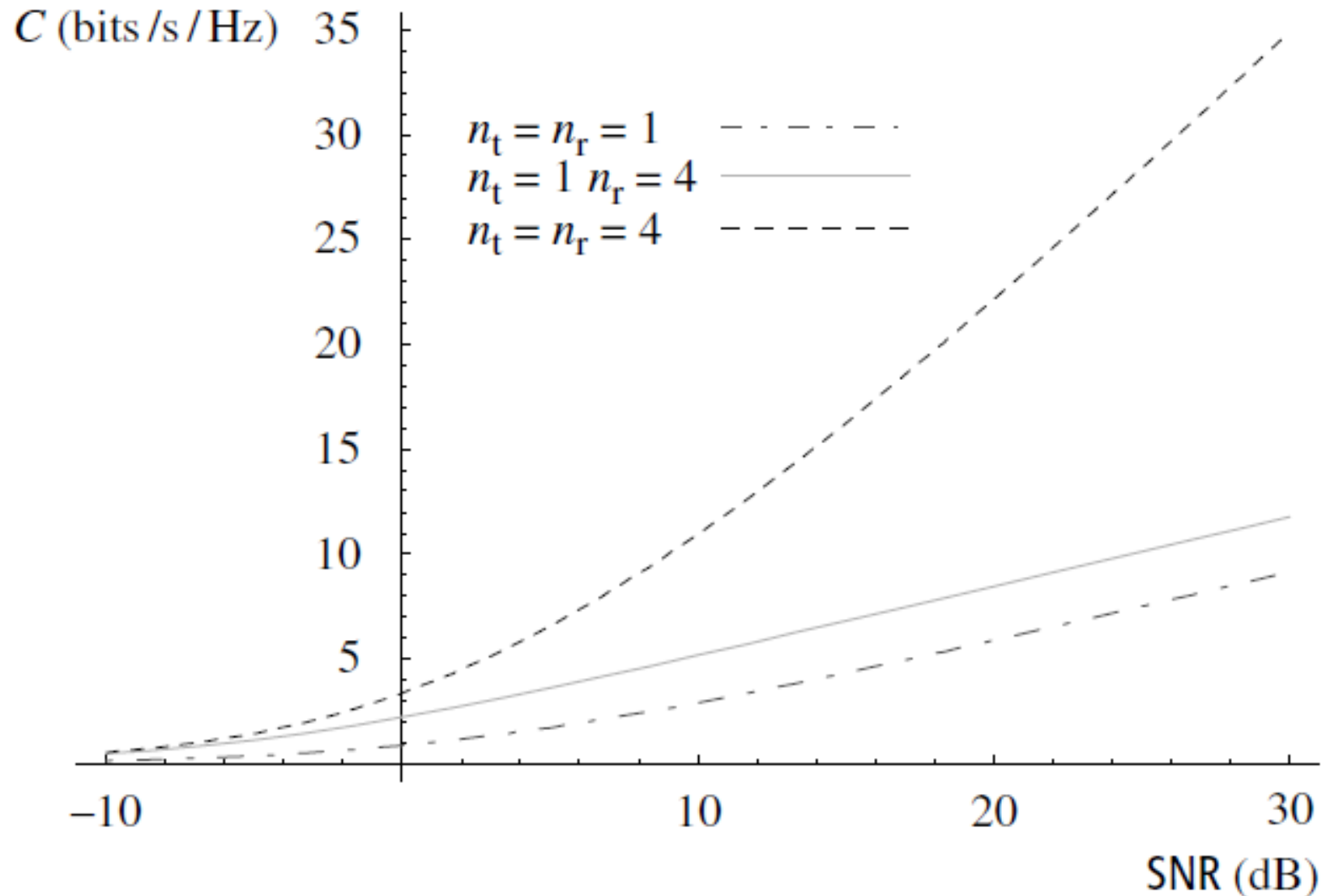
High SNR regime

For i.i.d. Rayleigh fading model, at high SNR, the capacity is

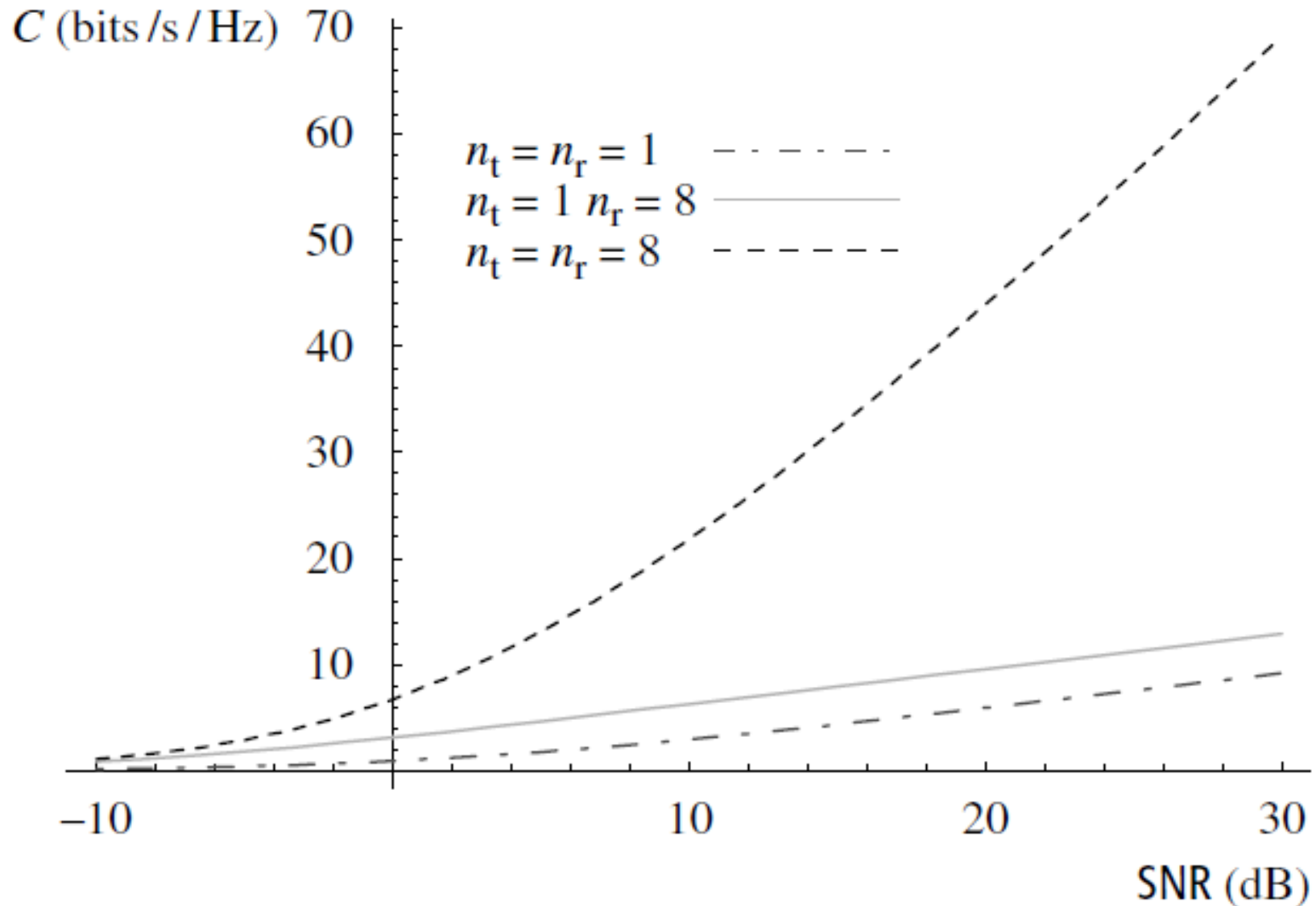
$$C \approx n_{\min} \log \frac{\text{SNR}}{n_t} + \sum_{i=1}^{n_{\min}} \mathbb{E}[\log \lambda_i^2],$$

The asymptotic slope of capacity versus SNR in dB scale is proportional to n , which means that the capacity scales with SNR like $n \log \text{SNR}$.

Fast Fading Capacity for I.I.D. Rayleigh Fading



Fast Fading Capacity for I.I.D. Rayleigh Fading



Low SNR regime

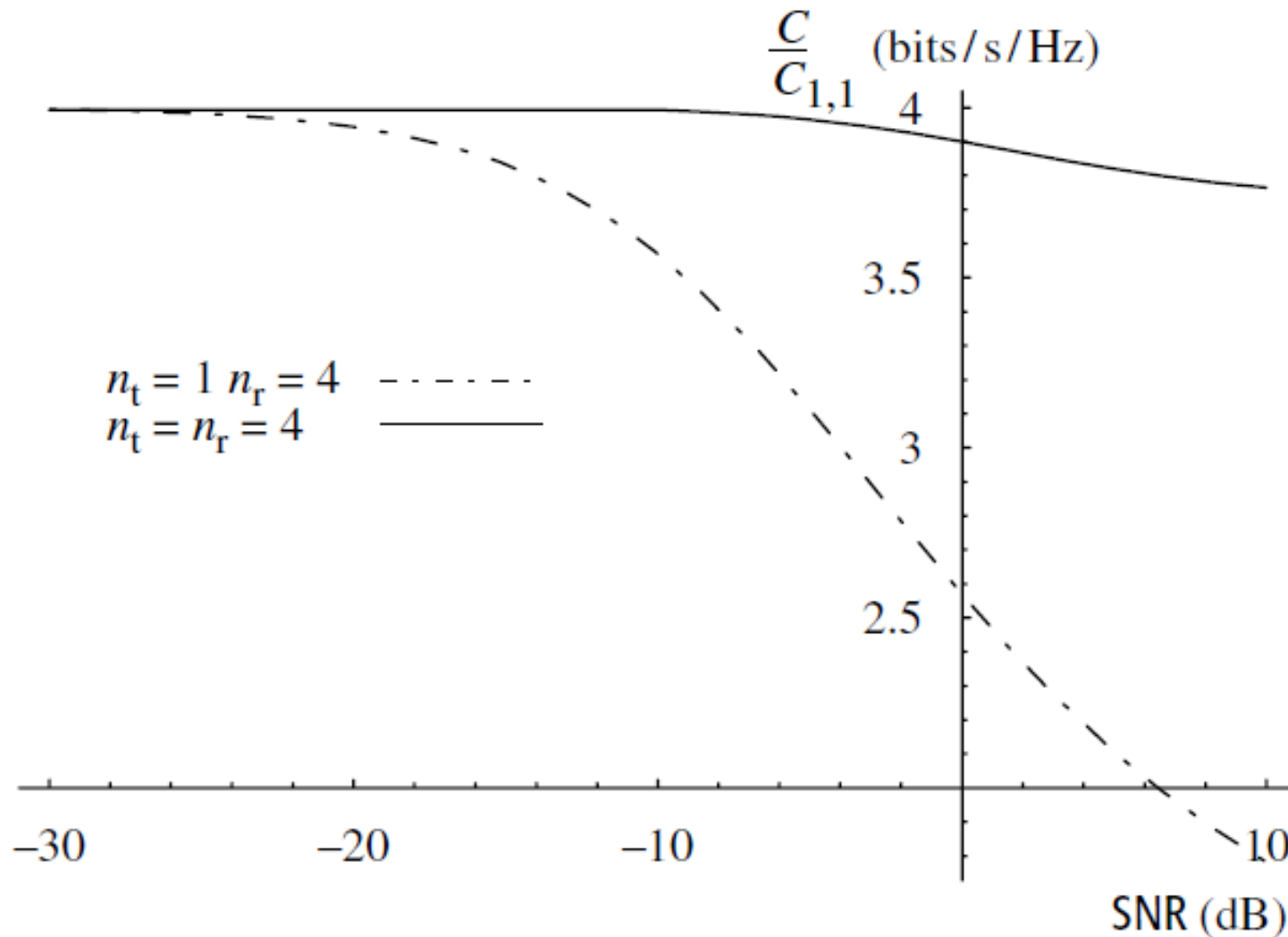
For i.i.d. Rayleigh fading model, at Low SNR, the capacity is

$$\begin{aligned} C &= \sum_{i=1}^{n_{\min}} \mathbb{E} \left[\log \left(1 + \frac{\text{SNR}}{n_t} \lambda_i^2 \right) \right] \\ &\approx \sum_{i=1}^{n_{\min}} \frac{\text{SNR}}{n_t} \mathbb{E} [\lambda_i^2] \log_2 e \\ &= \frac{\text{SNR}}{n_t} \mathbb{E} [\text{Tr}[\mathbf{H}\mathbf{H}^*]] \log_2 e \\ &= \frac{\text{SNR}}{n_t} \mathbb{E} \left[\sum_{i,j} |h_{ij}|^2 \right] \log_2 e \\ &= n_r \text{SNR} \log_2 e \text{ bits/s/Hz.} \end{aligned}$$

Thus, at low SNR, an n_t by n_r system yields a power gain of n_r over a single antenna system.

Fast Fading Capacity: Low SNR

n_r – fold power gain at low SNR



Fast fading MIMO channel with full CSI

Decompose the channel matrix as

$$\mathbf{H}[m] = \mathbf{U}[m]\mathbf{\Lambda}[m]\mathbf{V}[m]^*$$

A parallel channel

$$\tilde{y}_i[m] = \lambda_i[m]\tilde{x}_i[m] + \tilde{w}_i[m], \quad i = 1, \dots, n_{\min}$$

waterfilling policy

$$P^*(\lambda) = \left(\mu - \frac{N_0}{\lambda^2} \right)^+$$

capacity

$$C = \sum_{i=1}^{n_{\min}} \mathbb{E} \left[\log \left(1 + \frac{P^*(\lambda_i)\lambda_i^2}{N_0} \right) \right]$$

Linear Receiver

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

Linear MMSE (minimum mean squared error) receiver \mathbf{D}

$$\tilde{\mathbf{x}} = \mathbf{D}\mathbf{y} = \mathbf{D}(\mathbf{H}\mathbf{x} + \mathbf{w})$$

Mean squared error (MSE)

$$\text{MSE} = \mathbb{E} \left\{ \text{trace} \left[(\mathbf{x} - \tilde{\mathbf{x}})(\mathbf{x} - \tilde{\mathbf{x}})^* \right] \right\}$$

Optimal \mathbf{D} to minimize MSE is $\mathbf{D} = \mathbf{H}^* (\mathbf{H}\mathbf{H}^* + \mathbf{N}_0 \mathbf{I})^{-1}$

Part 2: OFDM

➤ Refer to <<Fundamentals of Wireless Communication>> Chapter 3.4.4

Frequency-Selective Channel

Frequency-Selective Channel:

$$y[m] = \sum_{\ell} h_{\ell}[m]x[m - \ell] + w[m]$$

Assuming a finite number of non-zero taps L :

$$y[m] = \sum_{\ell=0}^{L-1} h_{\ell}x[m - \ell] + w[m].$$

We transmit over only a finite duration, say N_c symbols:

$$\mathbf{d} = [d[0], d[1], \dots, d[N_c - 1]]^t.$$

Cyclic Prefix (CP)

A *prefix* of length $L-1$ consisting of data symbols:

$$d[N_c - L + 1], d[N_c - L + 2], \dots, d[N_c - 1]$$

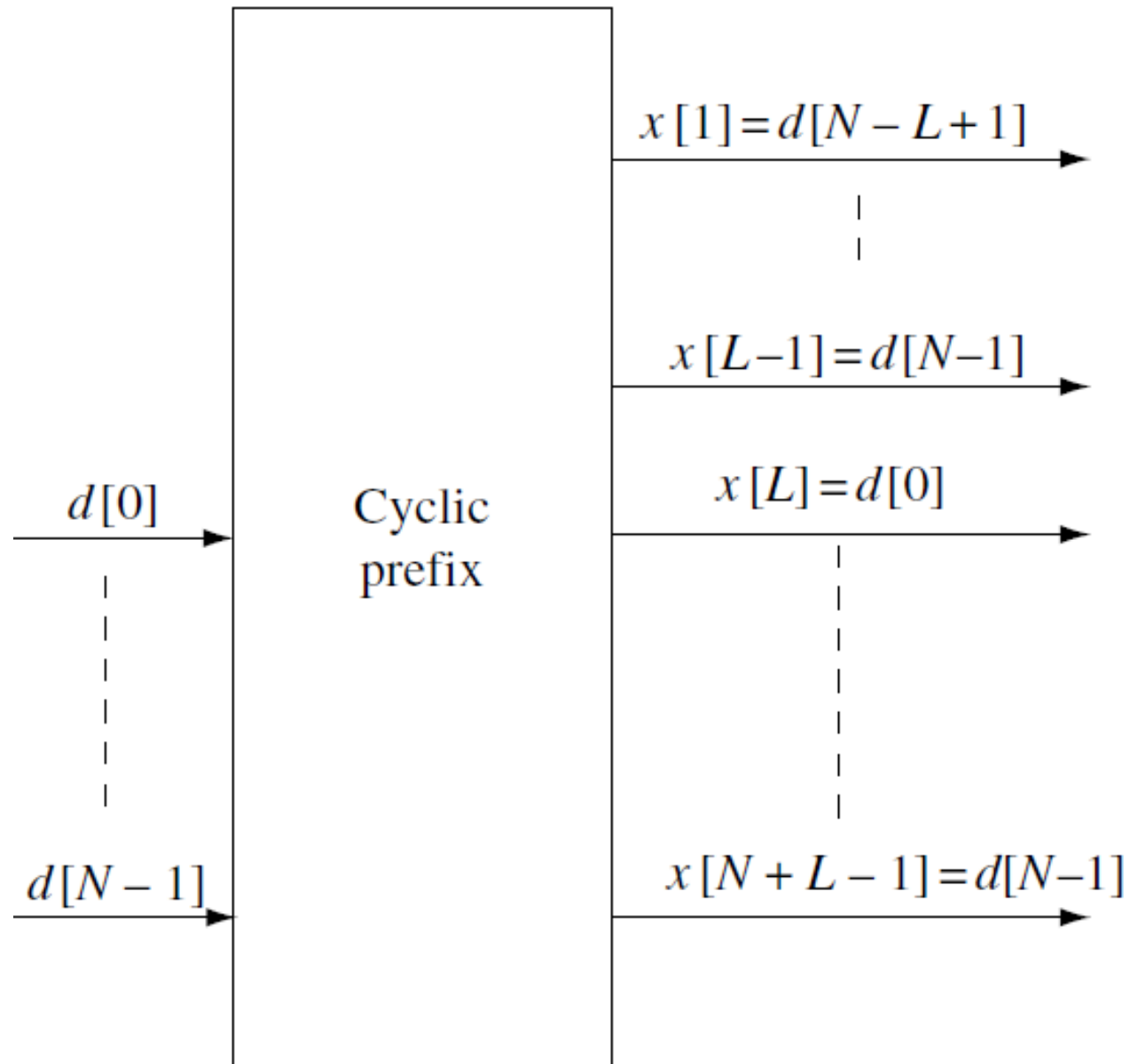
An $N_c + L - 1$ input block:

$$\mathbf{x} = [d[N_c - L + 1], d[N_c - L + 2], \dots, d[N_c - 1], d[0], d[1], \dots, d[N_c - 1]]^t$$

The output to the channel:

$$y[m] = \sum_{\ell=0}^{L-1} h_{\ell} x[m - \ell] + w[m], \quad m = 1, \dots, N_c + L - 1$$

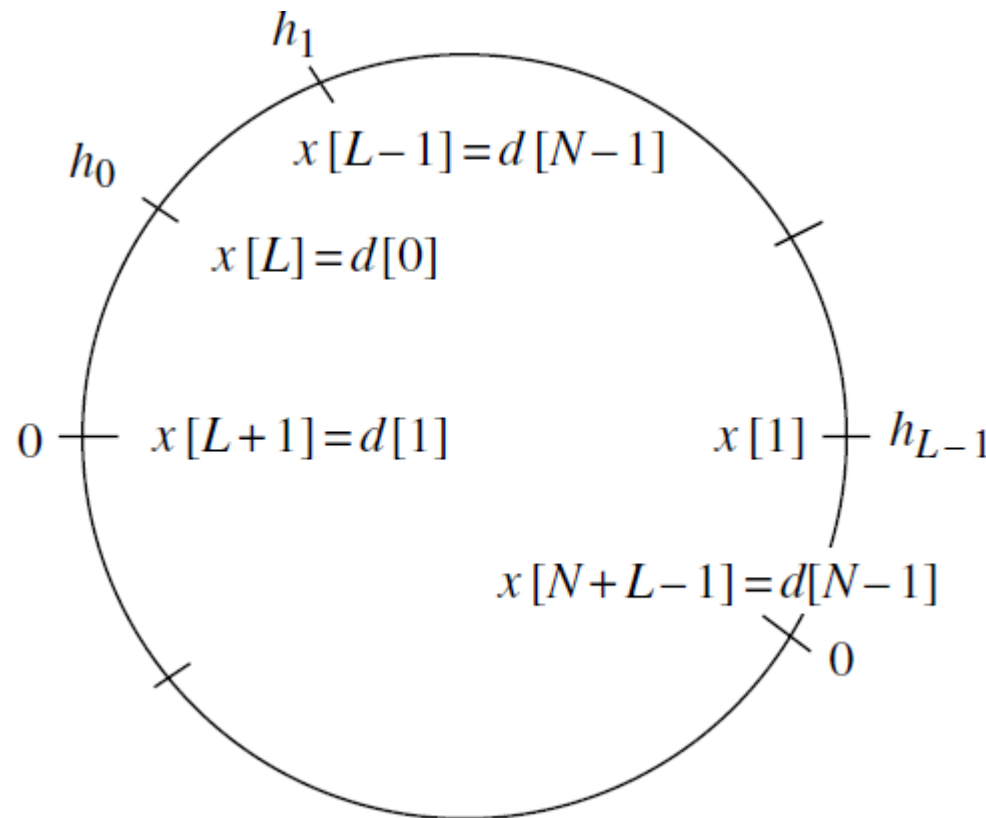
Cyclic Prefix (CP)



Removing Cyclic Prefix

The ISI extends over the first $L-1$ symbols and the receiver ignores it by considering only the output over the time interval $m \in [L, N_c+L-1]$:

$$y[m] = \sum_{\ell=0}^{L-1} h_{\ell} d[(m - L - \ell) \text{ modulo } N_c] + w[m].$$



Cyclic Convolution

Denoting the output of length N_c by

$$\mathbf{y} = [y[L], \dots, y[N_c + L - 1]]^t$$

The channel by a vector of length N_c

$$\mathbf{h} = [h_0, h_1, \dots, h_{L-1}, 0, \dots, 0]^t$$

The output by *cyclic convolution*:

$$\mathbf{y} = \mathbf{h} \otimes \mathbf{d} + \mathbf{w}$$

Discrete Fourier transform (DFT)

DFT of \mathbf{d} :

$$\tilde{d}_n := \frac{1}{\sqrt{N_c}} \sum_{m=0}^{N_c-1} d[m] \exp\left(\frac{-j2\pi nm}{N_c}\right), \quad n = 0, \dots, N_c - 1$$

DFT of both sides of $\mathbf{y} = \mathbf{h} \otimes \mathbf{d} + \mathbf{w}$

$$\text{DFT}(\mathbf{h} \otimes \mathbf{d})_n = \sqrt{N_c} \text{DFT}(\mathbf{h})_n \cdot \text{DFT}(\mathbf{d})_n, \quad n = 0, \dots, N_c - 1$$

N_c parallel subchannel:

$$\tilde{y}_n = \tilde{h}_n \tilde{d}_n + \tilde{w}_n, \quad n = 0, \dots, N_c - 1$$

Connection between the frequency-selective channel and the MIMO channel (1)

The circular convolution operation $\mathbf{u} = \mathbf{h} \otimes \mathbf{d}$ can be viewed as a linear transformation:

$$\mathbf{u} = \mathbf{C}\mathbf{d}$$

where

$$\mathbf{C} := \begin{bmatrix} h_0 & 0 & \cdot & 0 & h_{L-1} & h_{L-2} & \cdot & h_1 \\ h_1 & h_0 & 0 & \cdot & 0 & h_{L-1} & \cdot & h_2 \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ 0 & \cdot & 0 & h_{L-1} & h_{L-2} & \cdot & h_1 & h_0 \end{bmatrix}$$

\mathbf{C} is a *circulant matrix*.

Connection between the frequency-selective channel and the MIMO channel (2)

The DFT of \mathbf{d} can be represented as an N_c -length vector $\mathbf{U}\mathbf{d}$, where \mathbf{U} is the unitary matrix with its (k,n) th entry equal to

$$\frac{1}{\sqrt{N_c}} \exp\left(\frac{-j2\pi kn}{N_c}\right), \quad k, n = 0, \dots, N_c - 1$$

Then $\text{DFT}(\mathbf{h} \otimes \mathbf{d})_n = \sqrt{N_c} \text{DFT}(\mathbf{h})_n \cdot \text{DFT}(\mathbf{d})_n$ is equivalent to

$$\mathbf{U}\mathbf{u} = \Lambda \mathbf{U}\mathbf{d}$$

where

$$\Lambda_{nn} = \tilde{h}_n := \left(\sqrt{N_c} \mathbf{U}\mathbf{h}\right)_n, \quad n = 0, \dots, N_c - 1$$

Connection between the frequency-selective channel and the MIMO channel (3)

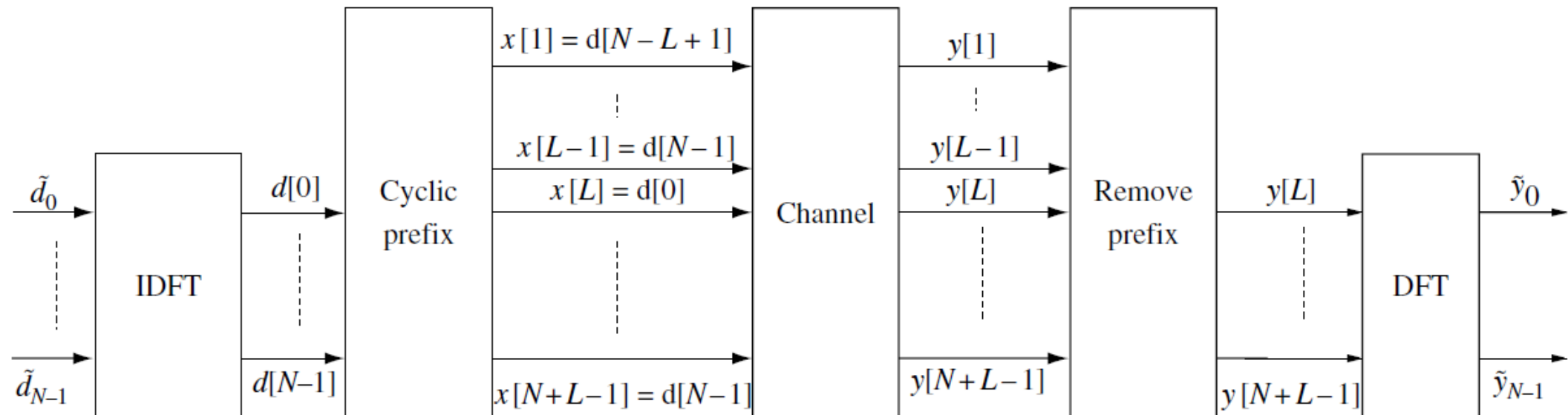
We come to the conclusion that

$$\mathbf{C} = \mathbf{U}^{-1} \mathbf{\Lambda} \mathbf{U}.$$

Then $\mathbf{y} = \mathbf{h} \otimes \mathbf{d} + \mathbf{w}$ is equivalent to

$$\mathbf{y} = \mathbf{C} \mathbf{d} + \mathbf{w} = \mathbf{U}^{-1} \mathbf{\Lambda} \mathbf{U} \mathbf{d} + \mathbf{w}.$$

The OFDM transmission and reception schemes



Homework

- 1. If $\mathbf{H} = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$, $\mathbf{K}_x = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, the noise power $N_0=0\text{dB}$, compute the capacity of the AWGN MIMO channel.
- 2. For the MIMO channel $\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w}$, how to parallelize it?

- Tips: For Q1, see P16; For Q2, see P 6-8
- Requirements: Submitted next week.