

AWGN Real

$$Y = X + \eta \quad ; \quad \text{where } \eta \sim (0, \sigma^2)$$

For Real signals and noise [One Dimensional]

$$C = \frac{1}{2} \log_2 (1 + \text{SNR}) = \frac{1}{2} \log_2 (1 + \rho)$$

$$\text{where } \text{SNR} = \frac{E\{X^2\}}{E\{\eta^2\}} = \frac{P}{\sigma^2} = \rho$$

AWGN Complex

$Y = X + \eta$  ; where  $\eta$  is a complex Gaussian noise with mean zero and variance  $\frac{\sigma^2}{2}$  per dimension.

$$C = \log_2 (1 + \text{SNR}) = \log_2 (1 + \rho)$$

Flat Fading Rayleigh~~Y =~~

$$Y = hX + \eta$$

where,  $\eta$  is a complex Gaussian noise with mean zero and variance  $\frac{\sigma^2}{2}$  per dimension.

$h$  is a complex Gaussian noise with mean zero and variance 0.5 per dimension.

Let  $h = \alpha + j\beta$

The envelope of  $h = \sqrt{\alpha^2 + \beta^2}$  is Rayleigh distributed and the phase  $\theta = \tan^{-1} \frac{\beta}{\alpha}$  is uniformly distributed between  $[0, 2\pi]$ .

$h$  is called the Rayleigh fading coefficient.

Def: Instantaneous SNR:

~~SNR =~~

$$\text{SNR}_i = \frac{|h|^2 E\{X^2\}}{E\{\eta^2\}}$$

$$= \rho \cdot |h|^2$$

where  $\rho = \frac{E\{X^2\}}{E\{\eta^2\}} = \text{SNR for AWGN}$ .

Capacity of Flat Fading Rayleigh channels

$$C = E_h \left\{ \log_2 (1 + \rho \cdot |h|^2) \right\}$$

Ergodic (or Average) Capacity

Def: Instantaneous capacity

$$C_i = \log_2 (1 + \rho \cdot |h|^2)$$

## Fading and MIMO Capacity

P.2

Notice that the instantaneous capacity of fading channels is a random variable which depends on the distribution of the fading coefficient  $h$ .

## MIMO System Model

$$\vec{Y} = H \vec{X} + \vec{N}$$

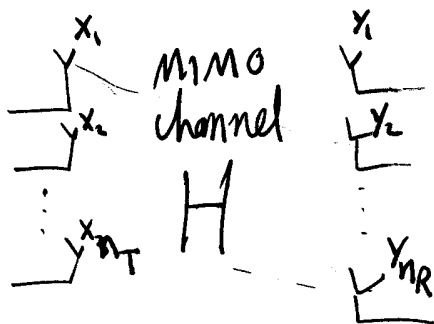
where:

$\vec{X}$  is  $n_T \times 1$  transmit vector

$\vec{Y}$  is  $n_R \times 1$  receive vector

$H$  is the  $(n_R \times n_T)$  channel matrix.

$\vec{N}$  is the  $(n_R \times 1)$  AWGN



$$H = \begin{bmatrix} h_{11} & h_{12} & \dots & h_{1n_T} \\ h_{21} & & & \\ \vdots & & & \\ h_{n_R 1} & & & h_{n_R n_T} \end{bmatrix}$$

where each coefficient is

$$\begin{aligned} h_{ij} &= \alpha + j\beta \\ &= \sqrt{\alpha^2 + \beta^2} \cdot e^{-j \tan^{-1} \frac{\beta}{\alpha}} \\ &= |h_{ij}| e^{j\theta_{ij}} \end{aligned}$$

If  $\alpha$  and  $\beta$  are independent and normally distributed random variables

$\Rightarrow |h_{ij}|$  is a Rayleigh distributed random variable.

MIMO Channel Capacity.

$$C = E_H \left\{ \max_{P(\vec{x}) : P(\Phi) \leq P_T} I(x;y) \right\}$$

where  $\Phi = E\{XX^H\}$  is the covariance matrix of the transmit signal vector  $\vec{x}$ . And  $A^H$  is the Hermitian transpose. For a given  $H$ ,

$$\begin{aligned} I(x;y) &= h(y) - h(y|x) \\ &= h(y) - h(Hx+n|x) \\ &= h(y) - h(n|x) \\ &= h(y) - h(n) \end{aligned}$$

since  $n$  and  $x$  are independent.

Theorem: For a complex Gaussian vector  $y \in C^n$ , the differential entropy is

$$h(y) \leq \log_2 \det(\pi e K)$$

with equality if  $y$  is circularly symmetric complex gaussian with

$$E\{yy^H\} = K.$$

$$\begin{aligned} E\{YY^H\} &= E\{(Hx+n)(Hx+n)^H\} \\ &= E\{Hxx^H H^H\} + E\{nn^H\} \\ &= H\Phi H^H + K^n \\ &= K^d + K^n \end{aligned}$$

$$\begin{aligned} I &= h(y) - h(n) \\ &= \log_2 [\det(\pi e (K^d + K^n))] \\ &\quad - \log_2 [\det(\pi e K^n)] \\ &= \log_2 [\det(K^d + K^n)] - \log_2 [\det(K^n)] \\ &= \log_2 [\det((K^d + K^n)(K^n)^{-1})] \\ &= \log_2 [\det(K^d (K^n)^{-1} + I_{nR})] \\ &= \log_2 [\det(H\Phi H^H (K^n)^{-1} + I_{nR})] \end{aligned}$$

When the transmitter has no knowledge about the channel, it is optimal to use a uniform power distribution.

$$\Rightarrow \Phi = \frac{P_T}{n_T} I_{n_T}$$

- Assuming also uncorrelated noise

$$\Rightarrow K^n = \sigma^2 I_{nR}$$

⇒ Ergodic capacity for a complex AWGN MIMO channel is:

$$C = E_H \left\{ \log_2 \left[ \det \left( I_{n_R} + \frac{P_T}{\sigma^2 n_T} H H^H \right) \right] \right\}$$
$$C = E_H \left\{ \log_2 \left[ \det \left( I_{n_R} + \frac{P}{n_T} H H^H \right) \right] \right\}$$

$$\Rightarrow C = E_H \left\{ \log_2 \left[ \sum_{i=1}^k \log_2 \left( 1 + \frac{P}{n_T} \lambda_i \right) \right] \right\}$$

where  $k = \text{rank}(H) \leq \min(n_T, n_R)$

- Using Eigenvalue Decomposition

$$H H^H = E \Lambda E^H$$

where  $E$  is the eigenvector matrix with orthonormal columns and  $\Lambda$  is a diagonal matrix with the eigenvalues.

$$\Rightarrow C = E_H \left\{ \log_2 \left[ \det \left( I_{n_R} + \frac{P}{n_T} E \Lambda E^H \right) \right] \right\}$$
$$= E_H \left\{ \log_2 \left[ \det E \left[ \det \left( E^H I_{n_R} E + \frac{P}{n_T} \Lambda \right) \right] \det E^H \right] \right\}$$

since  $\det E = 1$  and  $E^H E = I$

$$\Rightarrow C = E_H \left\{ \log_2 \left[ \det \left[ I_{n_R} + \frac{P}{n_T} \Lambda \right] \right] \right\}$$

Diagonal Matrix

## - Water Filling

The capacity maximization problem can be represented as

$$\begin{aligned} \max_{\Phi} \quad & \log \det \left( I_{n_R} + \frac{1}{\sigma^2} H \Phi H^H \right) \\ \text{subject to} \quad & \text{tr}(\Phi) \leq P \end{aligned}$$

where:

$$\Phi \equiv E\{x x^H\}$$

$P$  is the power budget

- Using Eigenvalue Decomposition

$$H H^H = E \Lambda E^H$$

where  $\Lambda = \text{diag}\{\lambda_1, \dots, \lambda_k\}$

where  $k = \text{rank}(H) \leq \min(n_T, n_R)$

- Let  $\Phi$  be a diagonal matrix such that  $\Phi = \text{diag}\{q_1, q_2, \dots, q_{n_T}\}$

$\Rightarrow$  The optimal value for  $q_i$  is

$$q_i = \left( v - \frac{\sigma^2}{\lambda_i} \right)^+$$

where  $v$  is the water level such that  $\sum_{i=1}^k q_i = P$