



Lecture 11  
MIMO

M. Xiao  
CommTh/EES/KTH

Overview

Space-Time Channel  
Modeling

Information Theoretic  
Limits

Spatial Multiplexing

Space-Time Coding

Beamforming

## Lecture 11: Space-Time Communication / MIMO Advanced Digital Communications (EQ2410)<sup>1</sup>

M. Xiao  
CommTh/EES/KTH

Wednesday, Mar. 2, 2014  
10:00-12:00, B22

<sup>1</sup>Textbook: U. Madhow, *Fundamentals of Digital Communications*, 2008

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### Overview

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## Overview

### Space-Time Communication

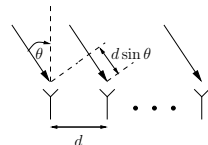
- Communication systems with multiple transmit and receive antennas
- Multiple-input/multiple-output systems (MIMO)
- Multiple transmit/receive antennas  $\rightarrow$  diversity, increased capacity
- Multiple antennas in communication systems
  - Base stations: many antennas
  - Mobiles: 1-2 antennas
  - WIFI (IEEE 802.11n): 2-4 antennas

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### Space-Time Channel Modeling

- Linear antenna array,  $m$  antennas, spacing  $d$



- Phase difference  $\phi$  and delay  $\tau$  between neighboring antennas

$$\phi(\theta) = \frac{2\pi d \sin \theta}{\lambda}, \quad \text{with } \lambda = c/f_c$$

$$\tau = \frac{d \sin \theta}{c}$$

- Received complex baseband signal at antenna  $i$ :  $y(t - i\tau)e^{j(i-1)\phi}$   
( $y(t)$ : received signal at the first antenna)
- Narrowband assumption: bandwidth of  $y \ll f_c \Rightarrow y(t - i\tau) \approx y(t)$ 
  - $\rightarrow$  Gain vector  $\mathbf{a}(\theta) = (1, e^{j\phi}, \dots, e^{j(m-1)\phi})^T$
  - $\rightarrow$  Array manifold:  $\{\mathbf{a}(\theta)\}$ , for  $\theta \in [-\pi/2, \pi/2]$

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## Space-Time Channel Modeling

- Multipath propagation ( $M$  paths,  $T_m \ll T_s$ ):

$$\mathbf{h} = (h_1, \dots, h_m)^T = \sum_{i=1}^M g_i \mathbf{a}(\theta_i)$$

- Gain of the  $i$ -th multipath component:  $g_i$
- Angle of the  $i$ -th multipath component:  $\theta_i$

→ Central limit theorem:  $\mathbf{h}$  is zero-mean, proper complex Gaussian with covariance matrix

$$\mathbf{C}_h = \sum_{i=1}^M |g_i|^2 \mathbf{a}(\theta_i) \mathbf{a}(\theta_i)^H$$

- Power-angle profile:  $P(\theta)$ 
  - Power density for a given angle  $\theta$  ( $\int P(\theta) d\theta = 1$ )
  - For a large number of multipath components, we get

$$\mathbf{C}_h = \int \mathbf{a}(\theta) \mathbf{a}(\theta)^H P(\theta) d\theta$$

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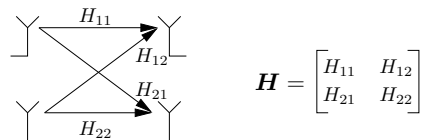
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## Space-Time Channel Modeling

- MIMO
  - $N_T$  transmit antennas, transmit manifold  $\mathbf{a}_T(\theta)$ , departure angle  $\theta$
  - $N_R$  receive antennas, receive manifold  $\mathbf{a}_R(\gamma)$ , arrival angle  $\gamma$
- Channel characterization for narrowband signaling

$$\mathbf{H} = \sum_l g_l \mathbf{a}_R(\gamma_l) \mathbf{a}_T(\theta_l)^T$$

- $\mathbf{H}$ : ( $N_R \times N_T$ ) matrix
- $j$ -th column gives the receive array response to the  $j$ -th transmit antenna



- Line-of-sight (LOS) link:  $\mathbf{H} = \mathbf{a}_R(\gamma) \mathbf{a}_T(\theta)^T$
- Rich scattering:  $H_{ij}$  i.i.d., zero-mean, proper complex Gaussian

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## Information Theoretic Limits – Channel Capacity

- Channel model:  $N_R \times N_T$  MIMO channel

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w} \quad \text{with} \quad \mathbf{w} \sim \mathcal{CN}(\mathbf{0}, 2\sigma^2 \mathbf{I})$$

- Channel capacity (channel unknown at the transmitter)

$$C = \max_{\mathbf{C}_x} I(\mathbf{x}; \mathbf{y}) = \max_{\mathbf{C}_x} \log \det \left( \mathbf{I} + \frac{1}{2\sigma^2} \mathbf{H} \mathbf{C}_x \mathbf{H}^H \right)$$

with the transmit covariance matrix  $\mathbf{C}_x$ .

- Helpful tool: use a singular value decomposition to reduce the MIMO channel into a number of  $k$  parallel scalar channels.

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## Information Theoretic Limits – Singular value decomposition

- Define

$$\mathbf{W} = \begin{cases} \mathbf{H}^H \mathbf{H} & \text{if } N_T \leq N_R \\ \mathbf{H} \mathbf{H}^H & \text{if } N_R \leq N_T \end{cases} \quad \text{and} \quad \begin{cases} m = \min(N_T, N_R) \\ M = \max(N_T, N_R) \end{cases}$$

$\mathbf{W}$  is nonnegative definite (i.e., all eigenvalues are nonnegative), and  $\mathbf{W}$  has dimension  $(m \times m)$ .

- Consider the case  $N_T \leq N_R$ ; i.e.,  $m = N_T$  and  $M = N_R$ .
- Let  $\mathbf{v}_i$  be the length- $m$  eigenvector to the eigenvalue  $\lambda_i \geq 0$  of  $\mathbf{W}$  (i.e.,  $\mathbf{W}\mathbf{v}_i = \lambda_i \mathbf{v}_i$  for  $\lambda_i > 0$  and  $\mathbf{W}\mathbf{v}_i = \mathbf{0}$  for  $\lambda_i = 0$ ).
- Matrix of eigenvectors  $\mathbf{V} = (\mathbf{v}_1, \dots, \mathbf{v}_m)$ , normalization  $\mathbf{V}^H \mathbf{V} = \mathbf{I}$  → eigenvectors are an orthonormal basis of the input space.
- Assumption:  $k$  nonzero eigenvalues  $\lambda_i > 0$  for  $i \in \{1, \dots, k\}$ , and  $m - k$  eigenvalues  $\lambda_i = 0$  for  $i \in \{k + 1, \dots, m\}$ .

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## Information Theoretic Limits – Singular value decomposition

- Define the length- $M$  (receive) vectors  $\mathbf{u}_i = \lambda_i^{-\frac{1}{2}} \mathbf{H} \mathbf{v}_i$ ,  $i \in \{1, \dots, k\}$  (i.e., vectors  $\mathbf{u}_i$  are orthonormal,  $\mathbf{u}_i^H \mathbf{u}_j = \delta_{ij}$ ).
- Define  $\mathbf{U} = (\mathbf{u}_1, \dots, \mathbf{u}_k, \mathbf{u}'_1, \dots, \mathbf{u}'_{m-k})$  with  $\mathbf{u}'_i$  such that  $\mathbf{U}^H \mathbf{U} = \mathbf{I}_m$ .  
→ the vectors in  $\mathbf{U}$  are an orthonormal basis for the output space.
- We can show that

$$\mathbf{H} \mathbf{x} = \sum_{i=1}^m \mathbf{u}_i \sqrt{\lambda_i} \mathbf{v}_i^H \mathbf{x} = \mathbf{U} \mathbf{D} \mathbf{V}^H \mathbf{x}$$

$$\mathbf{H} = \mathbf{U} \mathbf{D} \mathbf{V}^H$$

with  $\mathbf{D} = \text{diag}(\sqrt{\lambda_1}, \dots, \sqrt{\lambda_m}) \rightarrow$  singular values  $\sqrt{\lambda_i}$

- Proof: Use  $\mathbf{x} = \sum_{i=1}^m \mathbf{v}_i \mathbf{v}_i^H \mathbf{x}$  and solve  $\mathbf{H} \mathbf{x}$ .
- If we now choose  $\hat{\mathbf{y}} = \mathbf{U}^H \mathbf{y}$  and  $\hat{\mathbf{x}} = \mathbf{V}^H \mathbf{x}$ , we get

$$\hat{\mathbf{y}} = \mathbf{D} \hat{\mathbf{x}} + \hat{\mathbf{w}}$$

→ the channel is decomposed into  $k$  parallel channels!

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## Information Theoretic Limits – MIMO Capacity

- Channel known at the transmitter: use  $\hat{\mathbf{y}} = \mathbf{D} \hat{\mathbf{x}} + \hat{\mathbf{w}}$  and do waterfilling for the symbols  $\hat{\mathbf{x}}$ :

$$C_{\text{CSI-T}} = \sum_{i=1}^k \log\left(1 + \frac{\lambda_i P_i}{2\sigma^2}\right) \quad \text{with} \quad P_i = \left[a - \frac{2\sigma^2}{\lambda_i}\right]^+$$

and  $a$  such that  $\mathbb{E}[\|\mathbf{x}\|^2] \leq P$  is fulfilled ( $P_i = \mathbb{E}[\|\hat{\mathbf{x}}_i\|^2]$ ).

- No channel-state information at the transmitter
  - Choose the input distribution  $P(\mathbf{X})$  to maximize the mutual information

$$I(\mathbf{Y}; \mathbf{X}) = H(\mathbf{Y}) - H(\mathbf{Y}|\mathbf{X})$$

→ Since  $H(\mathbf{Y}|\mathbf{X}) = H(\mathbf{W})$  (i.e., the conditional entropy is independent of the input), it is sufficient to maximize  $H(\mathbf{Y})$ .

→  $H(\mathbf{Y})$  is maximized if  $\mathbf{Y}$  is proper complex Gaussian.

- (Differential) entropy for a complex Gaussian vector  $\mathbf{Z} \sim \mathcal{CN}(\mathbf{m}, \mathbf{C})$

$$H(\mathbf{Z}) = \log \det(\pi e \mathbf{C}) = \sum_{i=1}^n (\log(\lambda_i) + \log(\pi e))$$

with  $\lambda_i$  denoting the eigenvalues of  $\mathbf{C}$

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## Information Theoretic Limits – MIMO Capacity

- With  $\mathbf{C}_w = 2\sigma^2 \mathbf{I}$  and  $\mathbf{C}_y = \mathbf{H}\mathbf{C}_x\mathbf{H}^H + 2\sigma^2 \mathbf{I}$  we get

$$I(\mathbf{Y}; \mathbf{X}) = H(\mathbf{Y}) - H(\mathbf{W}) = \log \det \left( \mathbf{I} + \frac{1}{2\sigma^2} \mathbf{H}\mathbf{C}_x\mathbf{H}^H \right)$$

- Mutual information for spatially white input with  $\mathbf{C}_x = P\mathbf{I}/N_T$

$$C_{white} = \log \det \left( \mathbf{I} + \frac{SNR}{N_T} \mathbf{H}\mathbf{H}^H \right) = \sum_{i=1}^k \log \left( 1 + \frac{SNR}{N_T} \lambda_i \right)$$

with  $SNR = P/2\sigma^2$ .

→ Optimal for rich scattering when entries in  $\mathbf{H}$  are i.i.d., zero-mean complex Gaussian

- Ergodic capacity for rich scattering

$$C_{rich \text{ scattering}} = \sum_{i=1}^k E \left[ \log \left( 1 + \frac{SNR}{N_T} \lambda_i \right) \right]$$

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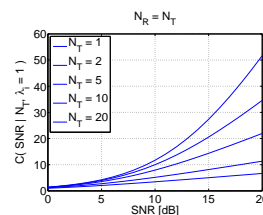


## Spatial Multiplexing

- Motivation: MIMO capacity scales with  $\min(N_T, N_R)$

$$C = \min(N_T, N_R) E \left[ \log \left( 1 + \frac{SNR}{N_T} \lambda \right) \right]$$

→ How can this be achieved?



- Observation: MIMO system can be interpreted as a CDMA system

$$\mathbf{y} = b_1 \mathbf{h}_1 + \dots + b_{N_T} \mathbf{h}_{N_T} + \mathbf{w}$$

- Each user/each stream is mapped to one antenna (or a group of antennas) → *spatial multiplexing*.
- “Spreading codes”  $\mathbf{h}_i$  are generated “by nature”.
- BLAST: Bell-Labs layered space-time architecture
- Linear receiver processing: MMSE or ZF
- Diversity: coding across several antennas.

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## Space-Time Coding

- Motivation
  - Time diversity can only be achieved for (very) mobile users or for a quickly changing environment.
  - For slow or static users (e.g., often in WIFI) frequency and spatial diversity is important.
 → Space-time codes for exploiting spatial diversity.
- Model
  - Narrowband, time-invariant system without time and frequency diversity
  - $N_T$  transmit antennas,  $N_R = 1$  receive antennas ( $N_R > 1$ : maximum ratio combining for antenna outputs)
 
$$y[m] = \mathbf{h}\mathbf{x}[m] + w[m] = h_1x_1[m] + \dots + h_{N_T}x_{N_T}[m] + w[m]$$
 with  $\mathbf{h} = (h_1, \dots, h_{N_T})$ .
- Capacity:  $C(\mathbf{h}) = \log(1 + G \cdot SNR)$  with  $G = \frac{\|\mathbf{h}\|^2}{N_T} = \frac{1}{N_T} \sum_{l=1}^{N_T} |h_l|^2$   
→ fluctuations in  $h_i$  are averaged out; improved outage capacity.
- Problem: without structure in  $\mathbf{x}$  the detection complexity is high (e.g., exponential in  $N_T$  for ML). → Space-time code design!

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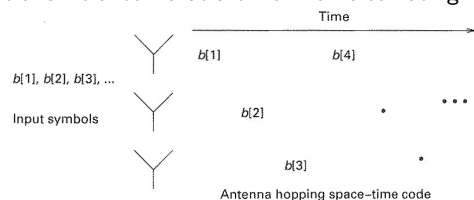
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## Space-Time Coding – Antenna Hopping

- Use one transmit antenna at a time in an alternating fashion



[U. Madhow, *Fundamentals of Dig. Comm.*, 2008]

- Achievable rate: average of the rates of the sub-channels

$$C_{\text{hopping}}(\mathbf{h}) = \frac{1}{N_T} \sum_{l=1}^{N_T} \log(1 + |h_l|^2 SNR) < C(\mathbf{h})$$

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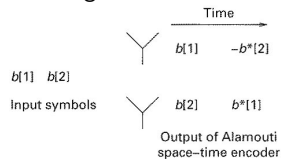
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## Space-Time Coding – Alamouti Code

- Space-time code which achieves the capacity  $C(\mathbf{h})$  for  $N_T = 2$
- Two symbols  $b[1]$  and  $b[2]$  are transmitted using two channel uses

$$\mathbf{x}[1] = \begin{bmatrix} x_1[1] \\ x_2[1] \end{bmatrix} = \begin{bmatrix} b[1] \\ b[2] \end{bmatrix}$$

$$\mathbf{x}[2] = \begin{bmatrix} x_1[2] \\ x_2[2] \end{bmatrix} = \begin{bmatrix} -b^*[2] \\ b^*[1] \end{bmatrix}$$



[U. Madhow, *Fundamentals of Dig. Comm.*, 2008]

- The corresponding received signals

$$y[1] = h_1 x_1[1] + h_2 x_2[1] + w[1] = h_1 b[1] + h_2 b[2] + w[1]$$

and

$$y[2] = h_1 x_1[2] + h_2 x_2[2] + w[2] = -h_1 b^*[2] + h_2 b^*[1] + w[2]$$

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## Space-Time Coding – Alamouti Code

- Consider now

$$\mathbf{y} = \begin{bmatrix} y[1] \\ y^*[2] \end{bmatrix} = \underbrace{\begin{bmatrix} h_1 \\ h_2^* \end{bmatrix}}_{=\mathbf{v}_0} b[1] + \underbrace{\begin{bmatrix} h_2 \\ -h_1^* \end{bmatrix}}_{=\mathbf{v}_1} b[2] + \begin{bmatrix} w_1 \\ w_2^* \end{bmatrix}$$

- Correlate  $\mathbf{y}$  with  $\mathbf{v}_0$  and  $\mathbf{v}_1$  to get the decision variables

$$Z_1 = \mathbf{v}_0^H \mathbf{y}, \text{ for } b[1],$$

$$Z_2 = \mathbf{v}_1^H \mathbf{y}, \text{ for } b[2]$$

- Observation:  $\mathbf{v}_0^H \mathbf{v}_1 = \mathbf{v}_1^H \mathbf{v}_0 = 0$ , i.e., two parallel AWGN channels with the effective SNR  $\|\mathbf{v}_0\|^2 / 2\sigma^2 = \|\mathbf{v}_1\|^2 / 2\sigma^2 = \|\mathbf{h}\|^2 / 2\sigma^2$ .
- With  $P_1 = P_2 = P/2$  the Alamouti code achieves the capacity  $C(\mathbf{h})$ .
- Optimal strategy if the transmitter does not know the channel  $\mathbf{h}$ .

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## Beamforming

- If the channel vector  $\mathbf{h}$  is known at the transmitter, we can use the spatial matched filter at the transmitter  $\mathbf{h}^H$  by transmitting

$$\mathbf{x}[n] = \mathbf{h}^H b[n]$$

- Received signal at time  $n$

$$y[n] = b[n] \cdot \sum_{k=1}^{N_T} |h_k|^2 + w[n] = b[n] \cdot \|\mathbf{h}\|^2 + w[n]$$

- Optimal strategy if the transmitter knows the channel  $\mathbf{h}$ .

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