

MIMO-OFDM Wireless Communications with MATLAB®

$$\min(N_R, N_T)$$

Spatial Multiplexing

Chapter 11. Uzamsal Çoklamalı MIMO
Sistemleri

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Chapter 11. Signal Detection for Spatially Multiplexed MIMO Systems

- 11.1 LINEAR SIGNAL DETECTION

- 11.1.1 ZF Signal Detection

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Bu konudaki varsayım

CSIR

Sadece alıcıda

kanal bilgisi

olduğu varsayım.

Verici her anteninden farklı bilgi basar.

Chapter 11. Signal Detection for Spatially Multiplexed MIMO System

- 11.7 SOFT DECISION FOR MIMO SYSTEMS
 - 11.7.1 Log-likelihood-Ratio (LLR) for SISO Systems
 - 11.7.2 LLR for Linear Detector based MIMO System
 - 11.7.3 LLR for MIMO System with a Candidate Vector Set
 - 11.7.4 LLR for MIMO System using a Limited Candidate Vector Set

Chapter 11. Uzamsal Çoklamalı MIMO Sistemleri

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{z} \quad (11.1)$$

\mathbf{y} (size $N_R \times 1$)
 \mathbf{H} (size $N_R \times N_T$)
 \mathbf{x} (size $N_T \times 1$)
 \mathbf{z} (size $N_R \times 1$)
 $\mathbf{z} = [z_1, z_2, \dots, z_{N_R}]^T \sim \mathcal{N}(0, \sigma_z^2)$
 $\mathbf{y} = \mathbf{h}_1 x_1 + \mathbf{h}_2 x_2 + \dots + \mathbf{h}_{N_T} x_{N_T} + \mathbf{z}$
 \mathbf{h}_i (size $N_R \times 1$)
 x_i (scalar)

11.1 Linear Signal Detection

istenen sinyal dışında her şey
gürültü muamelesi' görür

$$\tilde{\mathbf{x}} = [\tilde{x}_1 \tilde{x}_2 \dots \tilde{x}_{N_T}]^T = \mathbf{W}\mathbf{y} \quad (11.2)$$

\mathbf{W} (size $N_T \times N_R$)
 \mathbf{W} weight matrix

11.1 Linear Signal Detection

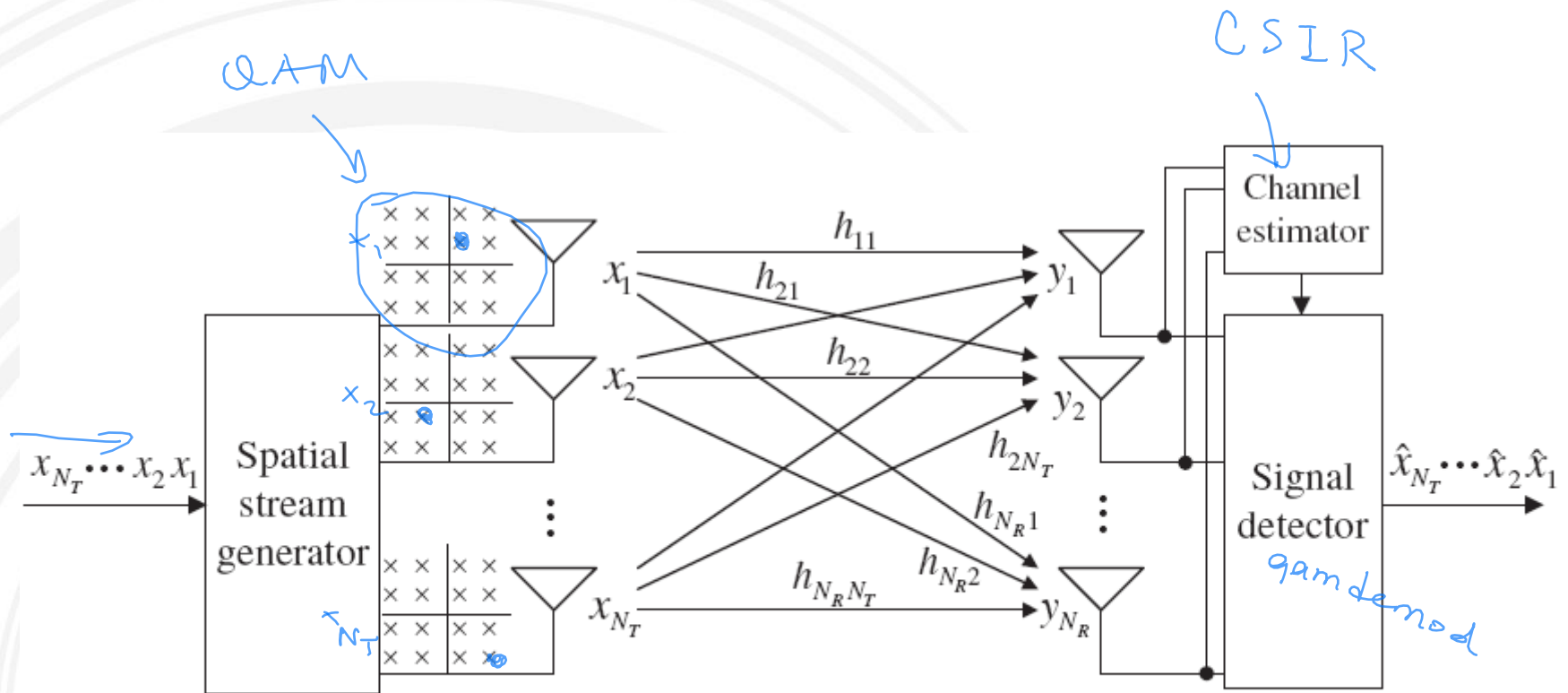


Figure 11.1 Spatially multiplexed MIMO systems.

11.1.1 ZF Signal Detection (Zero Forcing)

weight matrix

$$\mathbf{W}_{ZF} = (\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H$$

$$\bar{\mathbf{y}} = \bar{\mathbf{H}} \bar{\mathbf{x}} + \bar{\mathbf{z}}$$

pseudo inverse

(neden gerecek inv kullanmıyoruz)

(11.3)

$H_{N_R \times N_T}$

$$\bar{\mathbf{W}} \bar{\mathbf{y}} = \bar{\mathbf{W}} \bar{\mathbf{H}} \mathbf{x} + \bar{\mathbf{W}} \bar{\mathbf{z}}$$

$$\tilde{\mathbf{x}}_{ZF} = \mathbf{W}_{ZF} \mathbf{y}$$

$$= \mathbf{x} + (\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H \mathbf{z}$$

$$= \mathbf{x} + \tilde{\mathbf{z}}_{ZF}$$

$$= (\bar{\mathbf{H}}^H \bar{\mathbf{H}})^{-1} \bar{\mathbf{H}}^H \bar{\mathbf{H}} \bar{\mathbf{x}} + (\bar{\mathbf{H}}^H \bar{\mathbf{H}})^{-1} \bar{\mathbf{H}}^H \bar{\mathbf{z}}$$

(11.4)

gürültü gücü

$$\|\tilde{\mathbf{z}}_{ZF}\|_2^2 = \|(\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H \mathbf{z}\|_2^2$$

$$= \|(\mathbf{V} \Sigma^2 \mathbf{V}^H)^{-1} \mathbf{V} \Sigma \mathbf{U}^H \mathbf{z}\|_2^2$$

$$= \|\mathbf{V} \Sigma^{-1} \mathbf{V}^H \mathbf{V} \Sigma \mathbf{U}^H \mathbf{z}\|_2^2$$

$$= \|\mathbf{V} \Sigma^{-1} \mathbf{U}^H \mathbf{z}\|_2^2 = \|\Sigma^{-1} \bar{\mathbf{U}}^H \bar{\mathbf{z}}\|_2^2$$

$N_T \times N_R$

$N_R \times 1$

$\bar{\mathbf{y}} = \bar{\mathbf{x}} + (\bar{\mathbf{H}}^H \bar{\mathbf{H}})^{-1} \bar{\mathbf{H}}^H \bar{\mathbf{z}}$
 $\bar{\mathbf{x}}$ symbols $N_T \times 1$

$\|Qx\| = \|x\|$
 eğer Q unitary ise

$$(\mathbf{H} = \mathbf{U} \Sigma \mathbf{V}^H \text{ S.V.D.})$$

unitary unitary
 $\Sigma \in \mathbb{R}^{N_R}$

11.1.1 ZF Signal Detection

$$\begin{aligned}
 E\left\{\|\tilde{\mathbf{z}}_{ZF}\|_2^2\right\} &= E\left\{\left\|\mathbf{\Sigma}^{-1}\mathbf{U}^H\mathbf{z}\right\|_2^2\right\} = \text{tr}\left(\left(\mathbf{\Sigma}^{-1}\bar{\mathbf{U}}^H\bar{\mathbf{z}}\right)\left(\mathbf{\Sigma}^{-1}\bar{\mathbf{U}}^H\bar{\mathbf{z}}\right)^H\right) \\
 &= E\left\{\text{tr}\left(\mathbf{\Sigma}^{-1}\mathbf{U}^H\mathbf{z}\mathbf{z}^H\mathbf{U}\mathbf{\Sigma}^{-1}\right)\right\} \\
 &= \text{tr}\left(\mathbf{\Sigma}^{-1}\mathbf{U}^H E\{\mathbf{z}\mathbf{z}^H\}\mathbf{U}\mathbf{\Sigma}^{-1}\right) \\
 &= \text{tr}\left(\sigma_z^2\mathbf{\Sigma}^{-1}\mathbf{U}^H\mathbf{U}\mathbf{\Sigma}^{-1}\right) \rightarrow \sigma_z^2\bar{\mathbf{I}} \quad (11.6) \\
 &= \sigma_z^2\text{tr}\left(\mathbf{\Sigma}^{-2}\right) \rightarrow \bar{\mathbf{I}} \\
 &= \sum_{i=1}^{N_T} \frac{\sigma_z^2}{\sigma_i^2} \rightarrow \text{diag} \rightarrow \begin{bmatrix} \sigma_1 & & & \\ & \sigma_2 & & \\ & & \ddots & \\ & & & \sigma_{N_T} \end{bmatrix}
 \end{aligned}$$

$\frac{\sigma_z^2}{\min_i(\sigma_i^2)}$
 \rightarrow singular value düşse ise gürültü büyüyor.

11.1.2 MMSE Signal Detection (ZF'den daha iyi)

Minimum Mean Square Error

weight matrix

$$\mathbf{W}_{MMSE} = \underbrace{(\mathbf{H}^H \mathbf{H} + \underbrace{\sigma_z^2 \mathbf{I}}_{N_T \times N_T})^{-1}}_{N_T \times N_T} \underbrace{\mathbf{H}^H}_{N_T \times N_R} \quad (11.7)$$

bunu maksimize eder.

gürültünün Varyansını (gücünü) bilmeyi gerektirir.

1xNR i'inci satırı.

$$\mathbf{w}_{i,MMSE} = \arg \max_{\mathbf{w}=(w_1, w_2, \dots, w_{N_T})} \frac{|\mathbf{w} \mathbf{h}_i|^2 E_x}{E_x \sum_{j=1, j \neq i}^{N_T} |\mathbf{w} \mathbf{h}_j|^2 + \|\mathbf{w}\|^2 \sigma_z^2} \quad (11.8)$$

$$\begin{aligned} \tilde{\mathbf{x}}_{MMSE} &= \mathbf{W}_{MMSE} \mathbf{y} \\ &= (\mathbf{H}^H \mathbf{H} + \sigma_z^2 \mathbf{I})^{-1} \mathbf{H}^H \mathbf{y} \\ &= \underbrace{(\tilde{\mathbf{x}})}_{\neq \bar{\mathbf{x}}} + (\mathbf{H}^H \mathbf{H} + \sigma_z^2 \mathbf{I})^{-1} \mathbf{H}^H \mathbf{z} \\ &= \tilde{\mathbf{x}} + \tilde{\mathbf{z}}_{MMSE} \end{aligned} \quad (11.9)$$

→ $\bar{\mathbf{x}}$ değil.

11.1.2 MMSE Signal Detection

$$H = \overset{\text{SVD}}{\bar{U} \bar{\Sigma} \bar{V}^H}$$

$$\begin{aligned} \|\tilde{\mathbf{z}}_{MMSE}\|_2^2 &= \left\| (\mathbf{H}^H \mathbf{H} + \sigma_z^2 \mathbf{I})^{-1} \mathbf{H}^H \mathbf{z} \right\|^2 \\ &= \left\| (\mathbf{V} \bar{\Sigma}^2 \mathbf{V}^H + \sigma_z^2 \mathbf{I})^{-1} \mathbf{V} \bar{\Sigma} \mathbf{U}^H \mathbf{z} \right\|^2. \end{aligned} \quad (11.10)$$

$$\begin{aligned} (\bar{V} \bar{\Sigma}^2 \bar{V}^H + \sigma_z^2 \mathbf{I})^{-1} \mathbf{V} \bar{\Sigma} &= (\bar{V} \bar{\Sigma}^2 \bar{V}^H + \sigma_z^2 \mathbf{I})^{-1} (\bar{\Sigma}^{-1} \mathbf{V}^H)^T \\ &= (\mathbf{V} \bar{\Sigma}^{-1}) (\bar{V} \bar{\Sigma}^2 \bar{V}^H + \sigma_z^2 \mathbf{I})^{-1} \end{aligned}$$

$$\|\tilde{\mathbf{z}}_{MMSE}\|_2^2 = \left\| (\mathbf{\Sigma} \mathbf{V}^H + \sigma_z^2 \mathbf{\Sigma}^{-1} \mathbf{V}^H)^{-1} \mathbf{U}^H \mathbf{z} \right\|^2 = \left\| \mathbf{V} (\mathbf{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1})^{-1} \mathbf{U}^H \mathbf{z} \right\|^2 \quad (11.11)$$

→ because unitary

11.1.2 MMSE Signal Detection

$$\begin{aligned}
 E\left\{\|\tilde{\mathbf{z}}_{MMSE}\|_2^2\right\} &= E\left\{\left\|\left(\boldsymbol{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1}\right)^{-1} \mathbf{U}^H \mathbf{z}\right\|^2\right\} \\
 &= E\left\{\text{tr}\left(\left(\boldsymbol{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1}\right)^{-1} \mathbf{U}^H \mathbf{z} \mathbf{z}^H \mathbf{U} \left(\boldsymbol{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1}\right)^{-1}\right)\right\} \quad \text{hermitian} \\
 &= \text{tr}\left(\left(\boldsymbol{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1}\right)^{-1} \mathbf{U}^H E\{\mathbf{z} \mathbf{z}^H\} \mathbf{U} \left(\boldsymbol{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1}\right)^{-1}\right) \\
 &\rightarrow \text{trace} \rightarrow \text{diagonal } \sigma_z^2 \mathbf{I}_{N_T \times N_T} \quad \text{diag } \mathbf{U}^H \mathbf{U} = \mathbf{I} \\
 &= \text{tr}\left(\sigma_z^2 \left(\boldsymbol{\Sigma} + \sigma_z^2 \mathbf{\Sigma}^{-1}\right)^{-2}\right) \\
 &= \sum_{i=1}^{N_T} \sigma_z^2 \left(\sigma_i + \frac{\sigma_z^2}{\sigma_i}\right)^{-2} \\
 &\xrightarrow{\text{Sonuç}} = \sum_{i=1}^{N_T} \frac{\sigma_z^2 \sigma_i^2}{(\sigma_i^2 + \sigma_z^2)^2} \quad \text{çok küçük bile olsa sonucu çok etkiler.}
 \end{aligned}
 \tag{11.12}$$

11.1.2 MMSE Signal Detection

$$\sigma_{\min}^2 = \min \{ \sigma_1^2, \sigma_2^2, \dots, \sigma_{N_T}^2 \}$$

$$E \left\{ \|\tilde{\mathbf{z}}_{ZF}\|_2^2 \right\} = \sum_{i=1}^{N_T} \frac{\sigma_z^2}{\sigma_i^2} \approx \frac{\sigma_z^2}{\sigma_{\min}^2} \quad \text{for ZF} \quad (11.13a)$$

$$E \left\{ \|\tilde{\mathbf{z}}_{MMSE}\|_2^2 \right\} = \sum_{i=1}^{N_T} \frac{\sigma_z^2 \sigma_i^2}{(\sigma_i^2 + \sigma_z^2)^2} \approx \frac{\sigma_z^2 \sigma_{\min}^2}{(\sigma_{\min}^2 + \sigma_z^2)^2} \quad \text{for MMSE} \quad (11.13b)$$

gürültü daha az
ettirir eder.

N_T antenler N_T sembol iletilirse, $N_R \geq N_T$ olmak zorunda.
diversity order $N_R - N_T + 1$
 $N_T = 1$ ise, ZF MRC ile eşdeğer olur. (diversity order N_R)

11.2 OSIC Signal Detection

Ordered Successive Interference Cancellation.
 (i) = i olmak zorunda değil.

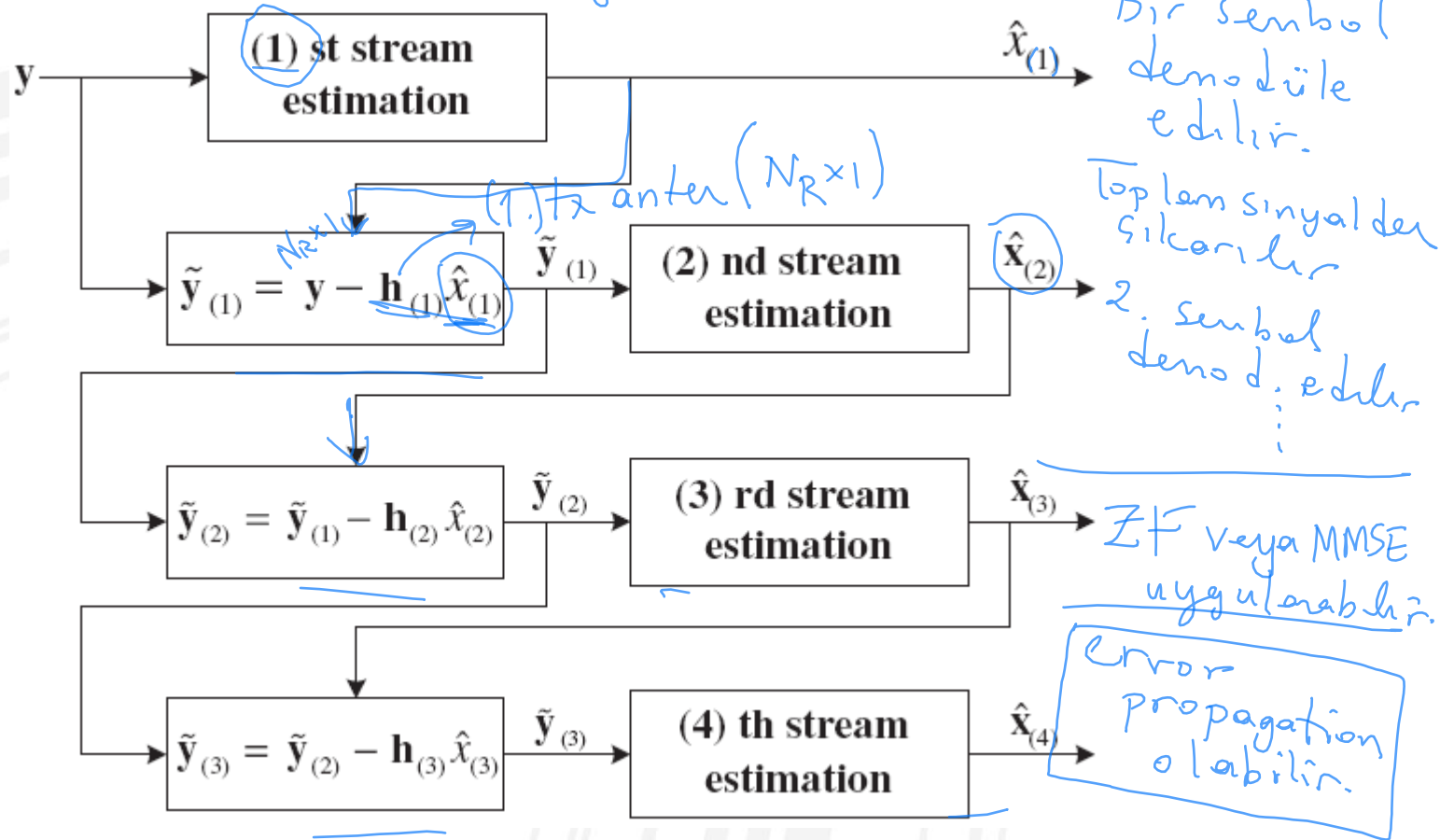


Figure 11.2 Illustration of OSIC signal detection for four spatial streams (i.e., $N_T = 4$).

11.2 OSIC Signal Detection

$$\tilde{\mathbf{y}}_{(1)} = \mathbf{y} - \mathbf{h}_{(1)} \hat{x}_{(1)} \quad (11.14)$$

$$= \mathbf{h}_{(1)} (x_{(1)} - \hat{x}_{(1)}) + \mathbf{h}_{(2)} x_{(2)} + \cdots + \mathbf{h}_{(N_T)} x_{(N_T)} + \mathbf{z}.$$

Method 1
MMSE
least detection
noise hata yayılımı (sıra öncelikli)

$$\text{SINR}_i = \frac{E_x |\mathbf{w}_{i,MMSE} \mathbf{h}_i|^2}{E_x \sum_{l \neq i} |\mathbf{w}_{i,MMSE} \mathbf{h}_l| + \sigma_z^2 \|\mathbf{w}_{i,MMSE}\|^2}, \quad i = 1, 2, \dots, N_T \quad (11.15)$$

→ i'inci sütun.
→ diğer sinyaller → gürültü

$$\mathbf{H}^{(1)} = [\mathbf{h}_1 \mathbf{h}_2 \cdots \mathbf{h}_{l-1} \mathbf{h}_{l+1} \cdots \mathbf{h}_{N_T}] \quad (11.16)$$

Method 2
ZF
SNR based ordering
→ $N_T \times (N_T - 1)$ oldu.
→ \mathbf{h}_l 'yi çıkarttık

$$\text{SNR}_i = \frac{E_x}{\sigma_z^2 \|\mathbf{w}_i\|^2}, \quad i = 1, 2, \dots, N_T. \quad (11.17)$$

Method 3
Column norm based ordering

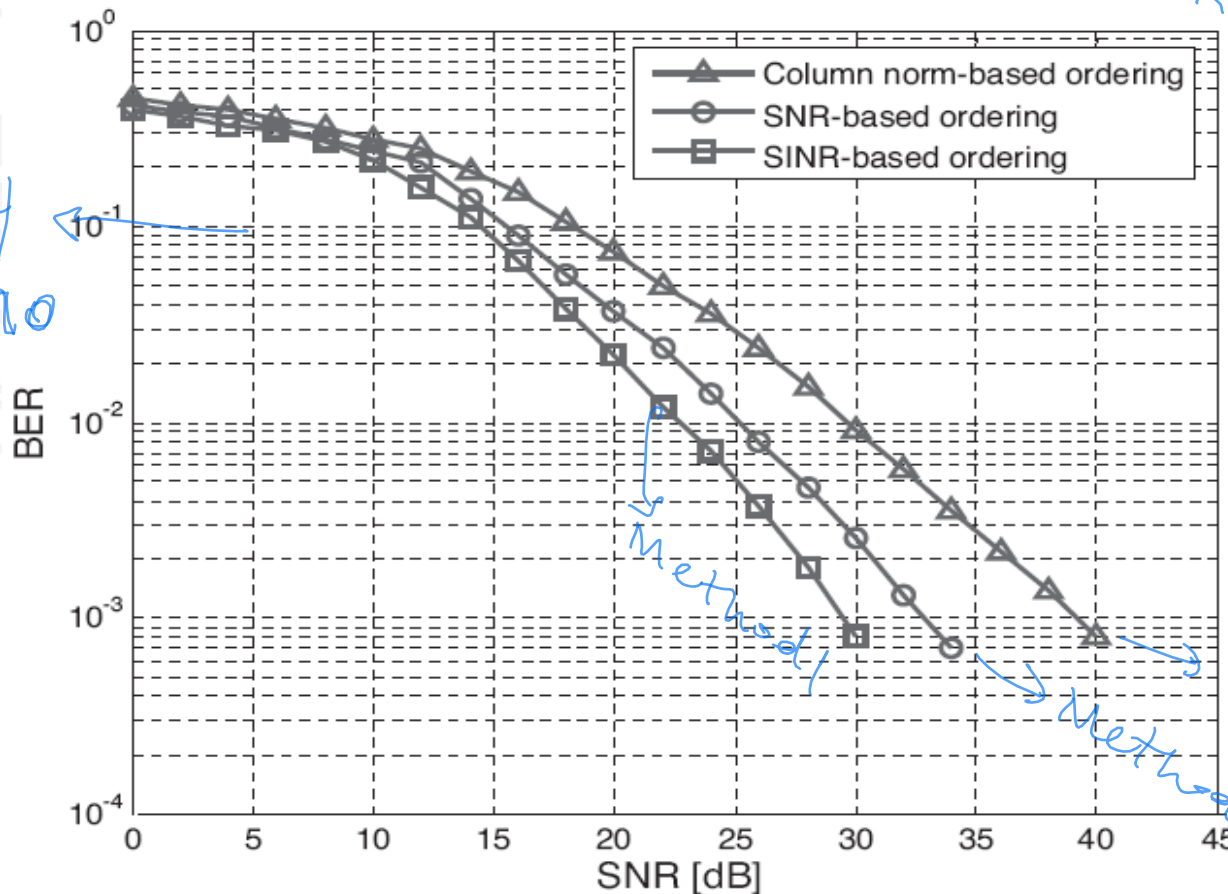
$$\|\mathbf{h}_i\|^2 \quad \mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{z} = \mathbf{h}_1 x_1 + \mathbf{h}_2 x_2 + \cdots + \mathbf{h}_{N_T} x_{N_T} + \mathbf{z} \quad (11.18)$$

11.2 OSIC Signal Detection

OSIC diversity order

daha fazla $N_R - N_T + 1$

$N_T = N_R = 4$
4x4 MIMO



iinci detect edilen sonb'dan sesitlilik derecesi

Method 1

Method 2

Method 3

Figure 11.3 Performance of OSIC methods with different detection ordering.

Programlar

- Program 11.1 “OSIC_detector” implementing the various OSIC signal detection methods
- Program 11.2 “QAM16_slicer”
- Program 11.3 “ML_detector” for ML signal detection

11.3 ML Signal Detection

- Alınan sinyal vektörü (\mathbf{y})

üstel karmaşıklığa sahip

- Gönderilmesi olası bütün sinyal vektörleri $\mathbf{x} \in \mathcal{C}^{N_T}$ için kanal matrisi ile çarpım. $\mathbf{H}\mathbf{x}$

16QAM için 16^{N_T} tane şey denenir.

- Bunlar arasındaki uzaklığı minimize eden \mathbf{x} değeri.

teker teker

$$\hat{\mathbf{x}}_{ML} = \underset{\mathbf{x} \in \mathcal{C}^{N_T}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{H}\mathbf{x}\|^2 \quad \text{minimum distance} \quad (11.19)$$

- Bütün vektörler eşit olasılıklı ise bu yöntem optimaldir.

- Bit/sembol (k) ve verici anten sayısı arttıkça karmaşıklık üstel artar

$$O(2^{kN_T})$$

$$\rightarrow \log_2 M$$

uzun süre

↓
benchmark.

11.3 ML Signal Detection

$$\hat{\mathbf{x}}_{ML} = \underset{\mathbf{x} \in \mathbb{C}^{N_T}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{H}\mathbf{x}\|^2 \quad (11.19)$$

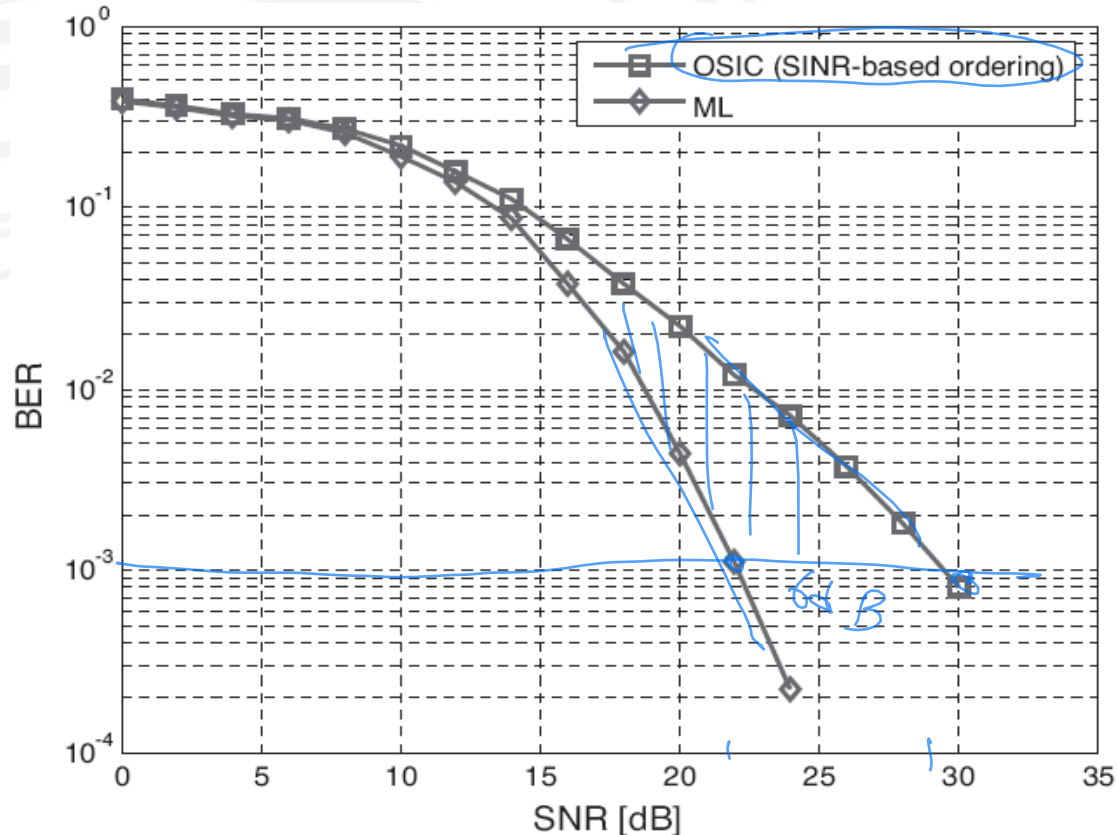


Figure 11.4 Performance comparison: OSIC vs. ML detection methods.