

# MIMO-OFDM Wireless Communications with MATLAB®

$$\min_{\mathbf{H}} (N_R, N_T)$$

Chapter 11. Uzamsal Çoklamalı MIMO  
Sistemleri

Spatial Multiplexing

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# Chapter 11. Signal Detection for Spatially Multiplexed MIMO Systems

## ■ 11.1 LINEAR SIGNAL DETECTION

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- 11.1.2 MMSE Signal Detection

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Bu konudaki varsayımlar

CSIR

Sadece alıcıda

kanal bilgisini  
olduğu varsayımlar

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
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  - 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

$$\begin{aligned} & \text{vector } [x_1, x_2, \dots, x_{N_T}]^T \\ & \text{noise } [z_1, z_2, \dots, z_{N_R}]^T \sim \mathcal{N}(0, \sigma_z^2) \\ & \text{rx } r_x \\ & \mathbf{y} = \mathbf{Hx} + \mathbf{z} \\ & = \mathbf{h}_1 x_1 + \mathbf{h}_2 x_2 + \dots + \mathbf{h}_{N_T} x_{N_T} + \mathbf{z} \\ & \text{N}_{R \times 1} \quad \text{N}_{R \times 1} \quad \text{skaler} \\ & \left[ y_1, y_2, \dots, y_{N_R} \right]^T \end{aligned} \quad (11.1)$$

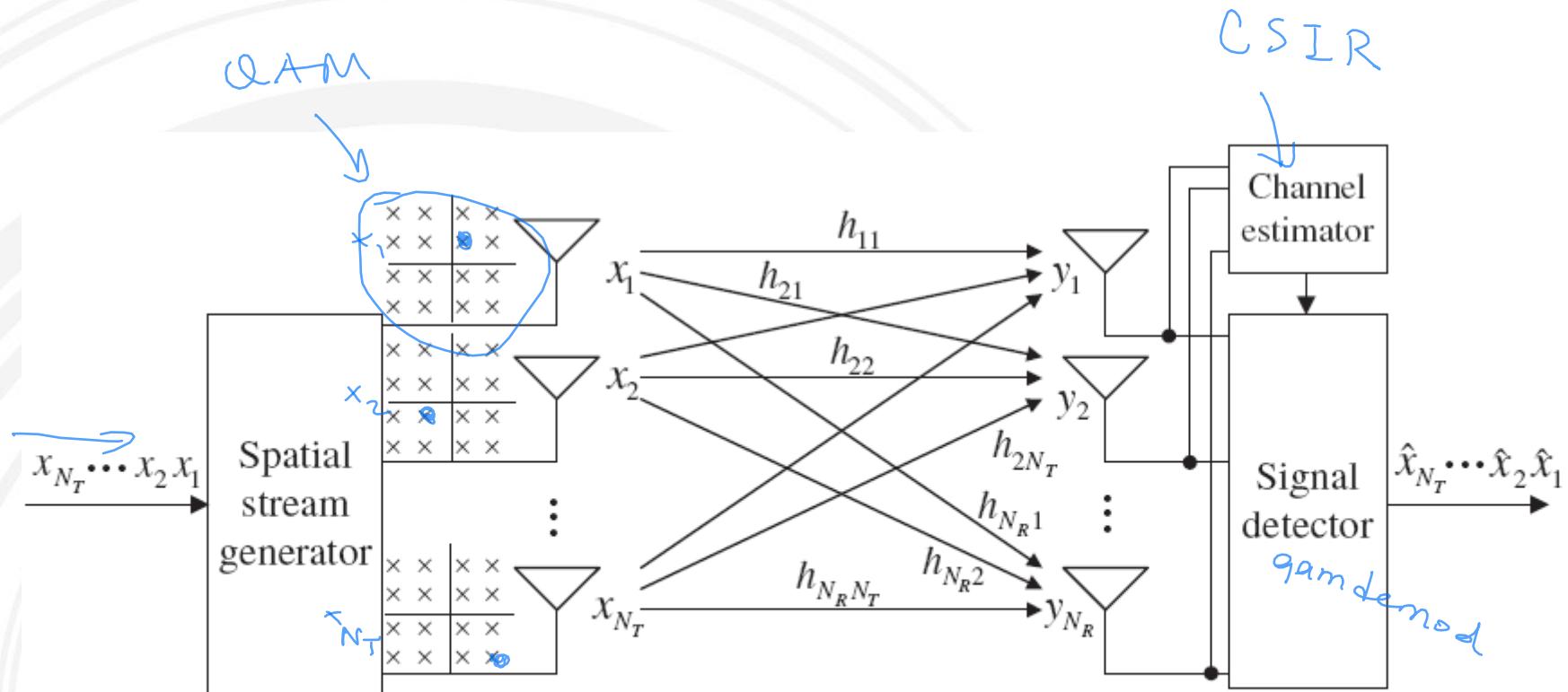
## 11.1 Linear Signal Detection

istenen sinyal dışında her şey  
gürültü muamelesi görünir

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

$N_{R \times 1}$   
 $N_T \times N_R$  weight matrix

# 11.1 Linear Signal Detection



**Figure 11.1** Spatially multiplexed MIMO systems.

# 11.1.1 ZF Signal Detection (Zero Forcing)

Weight matrix

$$\mathbf{W}_{ZF} = (\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H \quad \xrightarrow{\text{pseudo inverse}} \quad \mathbf{H}_{N_{RX} \times N_T} \quad (11.3)$$

$\bar{y} = \bar{H} \bar{x} + \bar{z}$  (neden gerekli inv kullanmıyoruz)

$$\begin{aligned} \tilde{\mathbf{x}}_{ZF} &= \mathbf{W}_{ZF} \mathbf{y} \\ &= \mathbf{x} + (\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H \mathbf{z} \\ &= \mathbf{x} + \tilde{\mathbf{z}}_{ZF} \end{aligned} \quad (11.4)$$

gürültü gücü

$$\begin{aligned} \|\tilde{\mathbf{z}}_{ZF}\|_2^2 &= \|(\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H \mathbf{z}\|^2 \\ &= \left\| \left( \mathbf{V} \Sigma^2 \mathbf{V}^H \right)^{-1} \mathbf{V} \Sigma \mathbf{U}^H \mathbf{z} \right\|^2 \\ &= \left\| \mathbf{V} \Sigma^{-2} \mathbf{V}^H \mathbf{V} \Sigma \mathbf{U}^H \mathbf{z} \right\|^2 \\ &= \left\| \mathbf{V} \Sigma^{-1} \mathbf{U}^H \mathbf{z} \right\|^2 = \left\| \sum \mathbf{U}^{-H} \bar{z} \right\|^2 \end{aligned} \quad (11.5)$$

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

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

tx symbols

$\|\mathbf{Q} \mathbf{x}\| = \|\mathbf{x}\|$   
eger Q unitary ise

## 11.1.1 ZF Signal Detection

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

## 11.1.2 MMSE Signal Detection

Minimum Mean Square Error

(ZF'den Laha iyi)

weight  
matrix

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

bunu maksimize  
eder.

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

gürültünün varyansını  
(gücünü) bilmeyi  
gerektirir.

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

$$= (\mathbf{H}^H \mathbf{H} + \sigma_z^2 \mathbf{I})^{-1} \mathbf{H}^H \mathbf{y}$$

$$= \tilde{\mathbf{x}} + (\mathbf{H}^H \mathbf{H} + \sigma_z^2 \mathbf{I})^{-1} \mathbf{H}^H \mathbf{z}$$

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

$\tilde{\mathbf{x}}$  değil.

(11.9)

## 11.1.2 MMSE Signal Detection

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

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

$$(\bar{V} \bar{\Sigma}^2 \bar{V}^H + \sigma_z^2 \bar{I})^{-1} \bar{V} \bar{\Sigma} = (\bar{V} \bar{\Sigma}^2 \bar{V}^H + \sigma_z^2 \bar{I})^{-1} (\bar{\Sigma}^{-1} \bar{V}^H)^{-1} (\bar{V} \bar{\Sigma}^{-1}) (\bar{V} \bar{\Sigma}^2 \bar{V}^H + \sigma_z^2 \bar{I})$$

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

but  $\mathbf{V}$  is a unitary matrix

## 11.1.2 MMSE Signal Detection

$$\begin{aligned}
 E\left\{\|\tilde{\mathbf{z}}_{MMSE}\|_2^2\right\} &= E\left\{\left\|\left(\Sigma + \sigma_z^2 \Sigma^{-1}\right)^{-1} \mathbf{U}^H \mathbf{z}\right\|^2\right\} \\
 &= E\left\{\operatorname{tr}\left(\left(\Sigma + \sigma_z^2 \Sigma^{-1}\right)^{-1} \mathbf{U}^H \mathbf{z} \mathbf{z}^H \mathbf{U} \left(\Sigma + \sigma_z^2 \Sigma^{-1}\right)^{-1}\right)\right\} \\
 &= \operatorname{tr}\left(\left(\Sigma + \sigma_z^2 \Sigma^{-1}\right)^{-1} \mathbf{U}^H E\{\mathbf{z} \mathbf{z}^H\} \mathbf{U} \left(\Sigma + \sigma_z^2 \Sigma^{-1}\right)^{-1}\right) \\
 &\xrightarrow{\text{hermitian}} \operatorname{tr}\left(\sigma_z^2 \left(\Sigma + \sigma_z^2 \Sigma^{-1}\right)^{-2}\right) \\
 &\xrightarrow{\text{diagonal}} \sigma_z^2 \sum_{i=1}^{N_T} \left(\sigma_i + \frac{\sigma_z^2}{\sigma_i}\right)^{-2} \quad (11.12) \\
 &\xrightarrow{\text{trace}} = \sum_{i=1}^{N_T} \sigma_z^2 \left(\sigma_i + \frac{\sigma_z^2}{\sigma_i}\right)^{-2} \\
 &\xrightarrow{\text{Sonuç}} = \sum_{i=1}^{N_T} \frac{\sigma_z^2 \sigma_i^2}{(\sigma_i^2 + \sigma_z^2)^2}.
 \end{aligned}$$

hermitian  
 diagonal  $\sigma_z^2$  /  $N_T \times N_T$   
 $\mathbf{U}^H \mathbf{U} = \mathbf{I}$

Sonuç  $\rightarrow$   $\sigma_z^2$  çok küçük bile olsa sonucu çok etkilenez.

## 11.1.2 MMSE Signal Detection

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

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

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

gürülti daha az  
etki eder

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

## 11.2 OSIC Signal Detection

$(i) = i$  olmak zorunda değil.

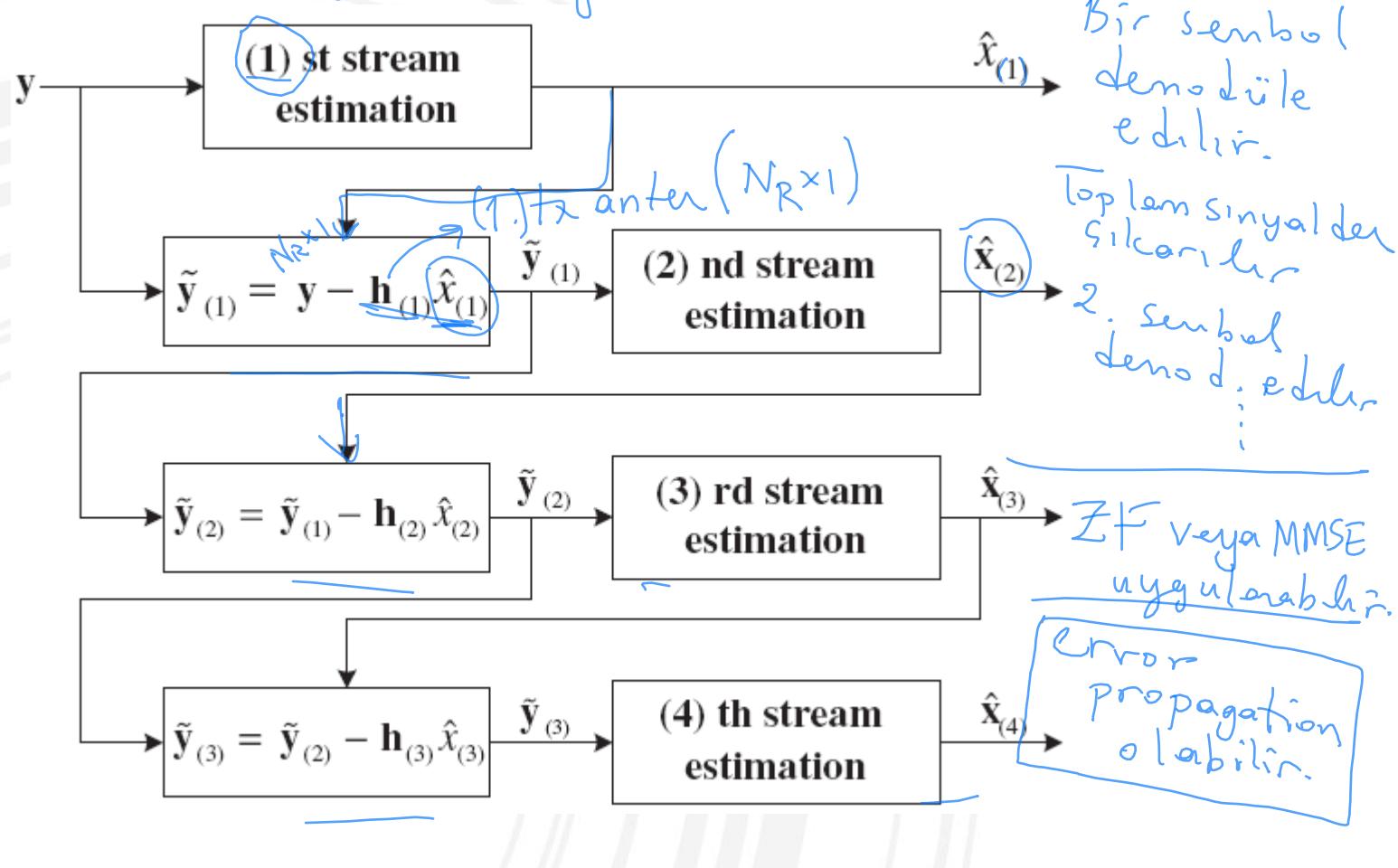


Figure 11.2 Illustration of OSIC signal detection for four spatial streams (i.e.,  $N_T = 4$ ).

# 11.2 OSIC Signal Detection

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

Method 1  
MMSE  
SINR detection  
 $\mathbf{E}_{\mathbf{x}} \left| \mathbf{w}_{i,MMSE} \mathbf{h}_i \right|^2$   $\rightarrow$  ilk sütun,  
 $\mathbf{E}_{\mathbf{x}} \sum_{l \neq i} \left| \mathbf{w}_{i,MMSE} \mathbf{h}_l \right|^2 + \sigma_z^2 \left\| \mathbf{w}_{i,MMSE} \right\|^2$   $\rightarrow$  geri kalan sinyalleri  
 $\rightarrow$  güç nüfus  
 $\rightarrow$   $N_T \times (N_T - 1)$  olur.

$$SINR_i = \frac{\mathbf{E}_{\mathbf{x}} \left| \mathbf{w}_{i,MMSE} \mathbf{h}_i \right|^2}{\mathbf{E}_{\mathbf{x}} \sum_{l \neq i} \left| \mathbf{w}_{i,MMSE} \mathbf{h}_l \right|^2 + \sigma_z^2 \left\| \mathbf{w}_{i,MMSE} \right\|^2}, \quad i = 1, 2, \dots, N_T \quad (11.15)$$

Method 2  $\mathbf{H}^{(1)} = [\mathbf{h}_1 \ \mathbf{h}_2 \ \dots \ \mathbf{h}_{i-1} \ \mathbf{h}_{i+1} \ \dots \ \mathbf{h}_{N_T}]$   $\rightarrow$  hepsi silindi  $\rightarrow$   $N_T \times (N_T - 1)$  olur.  $\rightarrow$   $\mathbf{H}^{(2)} = [\mathbf{h}_1 \ \mathbf{h}_2 \ \dots \ \mathbf{h}_i \ \mathbf{h}_{i+1} \ \dots \ \mathbf{h}_{N_T}]$   $\rightarrow$   $N_T \times N_T$  olur.

ZF based ordering  $\rightarrow$   $\mathbf{H}^{(1)} = [\mathbf{h}_1 \ \mathbf{h}_2 \ \dots \ \mathbf{h}_{i-1} \ \mathbf{h}_{i+1} \ \dots \ \mathbf{h}_{N_T}]$   $\rightarrow$  hepsi silindi  $\rightarrow$   $N_T \times (N_T - 1)$  olur.

Method 3  $\rightarrow$   $SINR_i = \frac{\mathbf{E}_{\mathbf{x}}}{\sigma_z^2 \left\| \mathbf{w}_i \right\|^2}, \quad i = 1, 2, \dots, N_T. \quad (11.17)$

Column norm based ordering  $\rightarrow$   $\left\| \mathbf{h}_i \right\|^2$   $\mathbf{y} = \mathbf{Hx} + \mathbf{z} = \mathbf{h}_1 x_1 + \mathbf{h}_2 x_2 + \dots + \mathbf{h}_{N_T} x_{N_T} + \mathbf{z} \quad (11.18)$

## 11.2 OSIC Signal Detection

$N_T = N_R = 4$   
4x4 MIMO

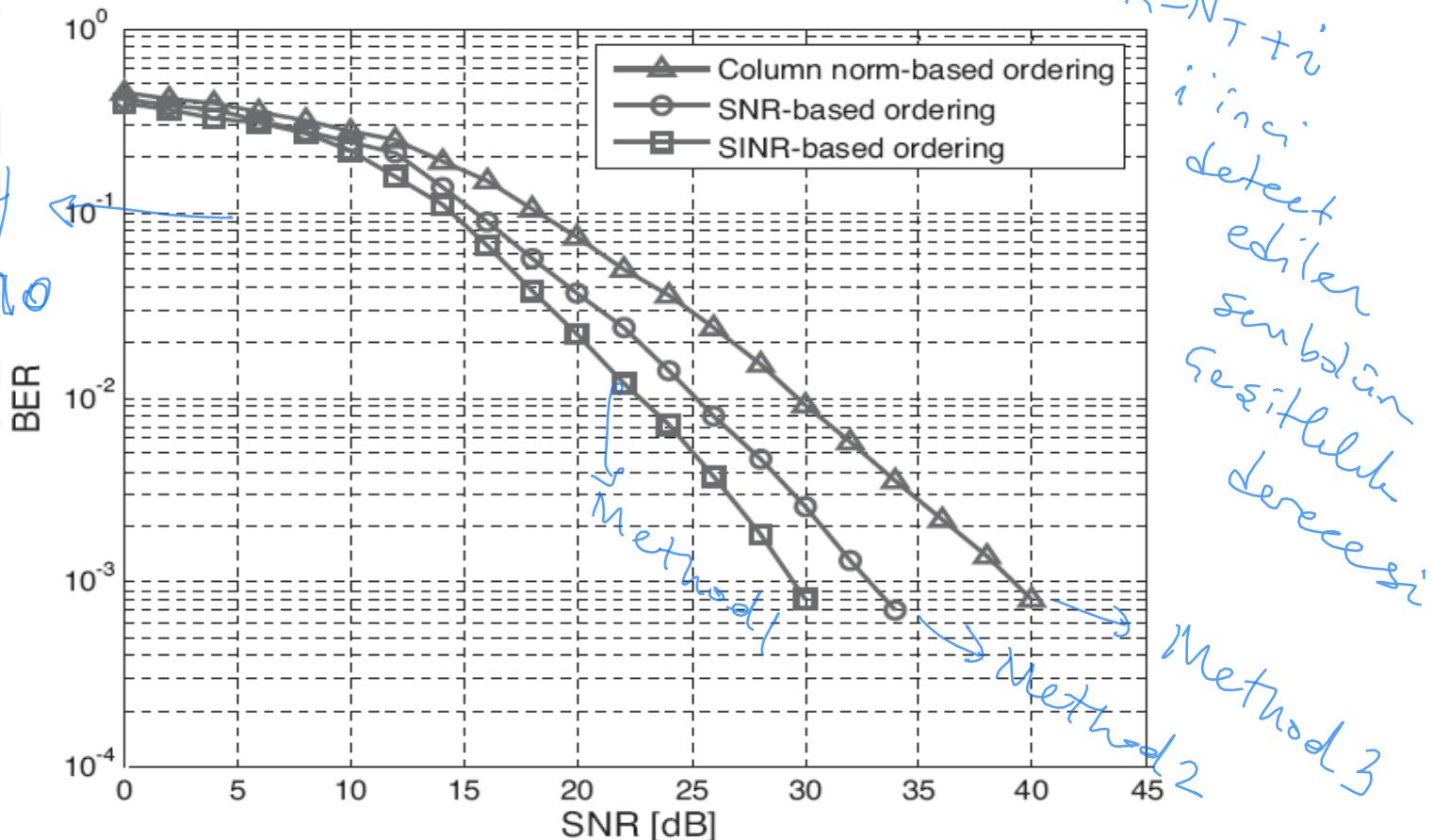


Figure 11.3 Performance of OSIC methods with different detection ordering.

# 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 ( $y$ ) *üstel karmaşılığa sahip*
- Gönderilmesi olası bütün sinyal vektörleri  $x \in \mathbb{C}^{N_T}$  için kanal matrisi ile çarpım.  $Hx$  *16 QAM için 16<sup>N\_T</sup> tane sey denerir.*
- Bunlar arasındaki uzaklığı minimize eden  $x$  değeri. *teker telker*

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

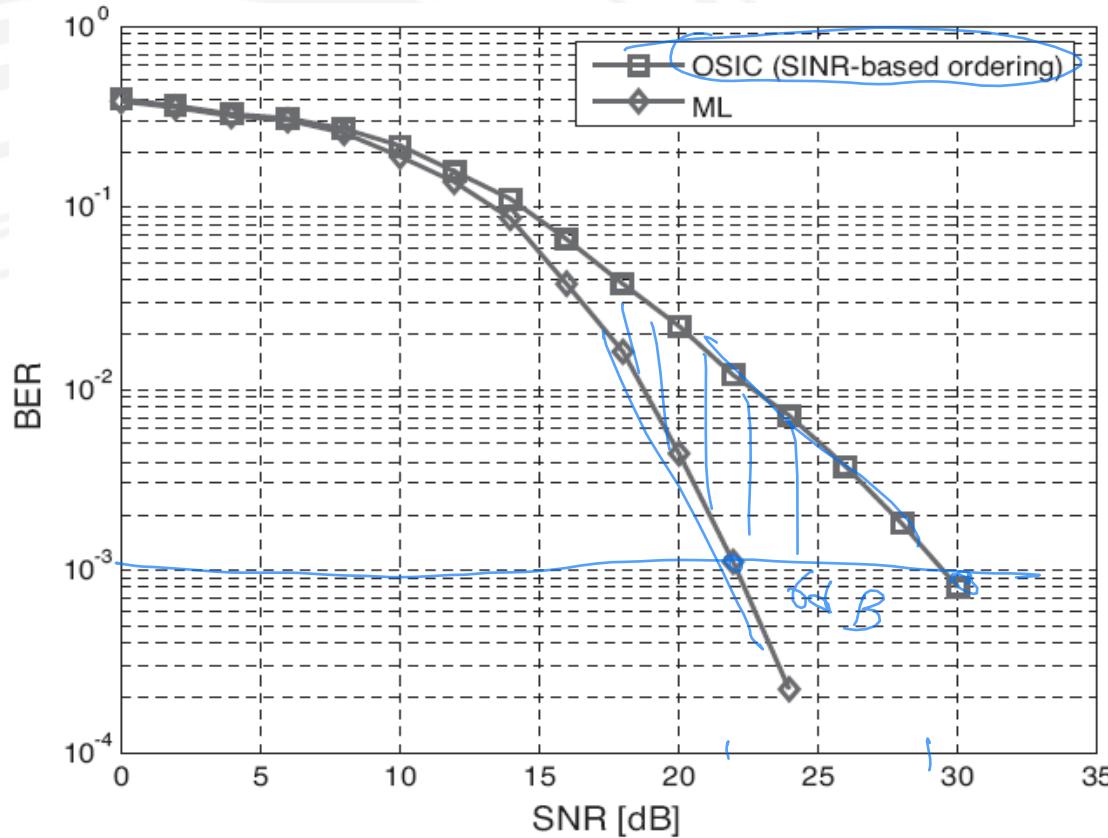
- Bütün vektörler eşit olasılıklı ise bu yöntem optimaldir.
- Bit/sembol ( $k$ ) ve verici anten sayısı arttıkça karmaşılık üstel artar  $O(2^{kN_T})$  *ütun suner*

$$\log_2 M$$

↓  
benchmark.

# 11.3 ML Signal Detection

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



Sanior  
4T4MIMO  
16QAM, 8in

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