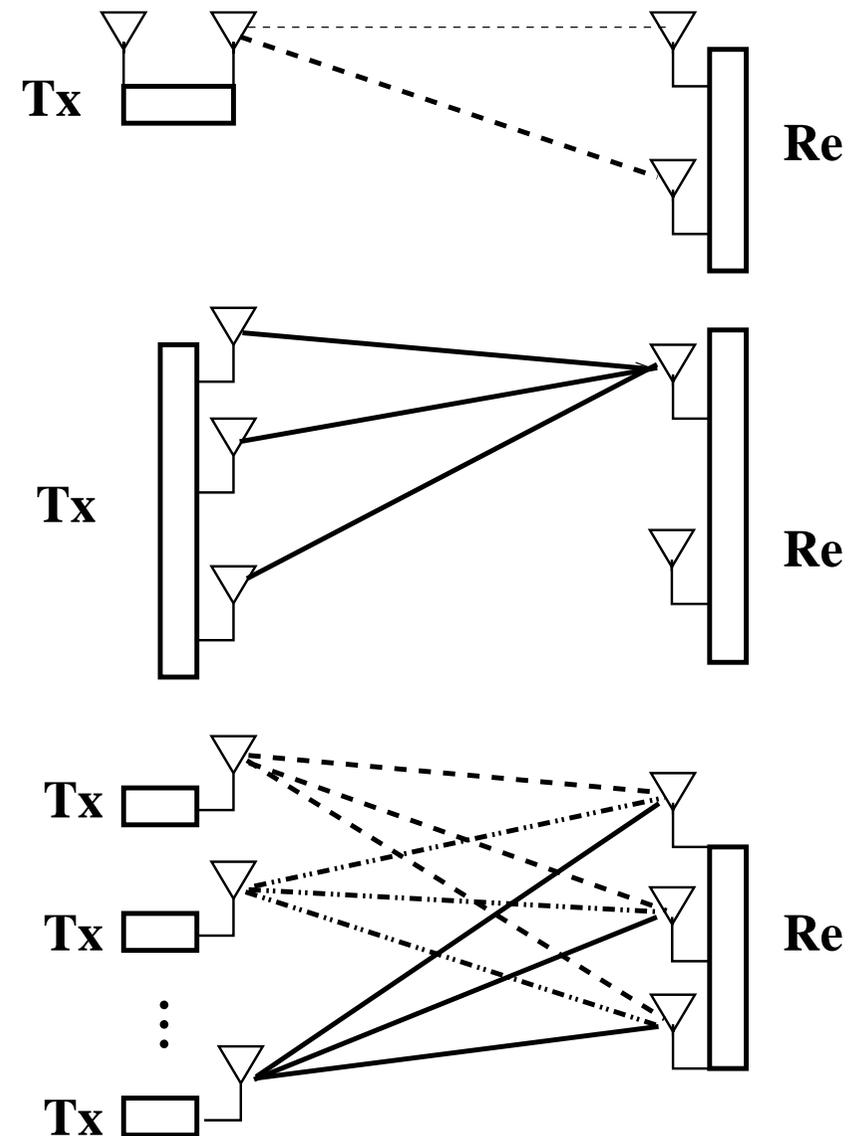


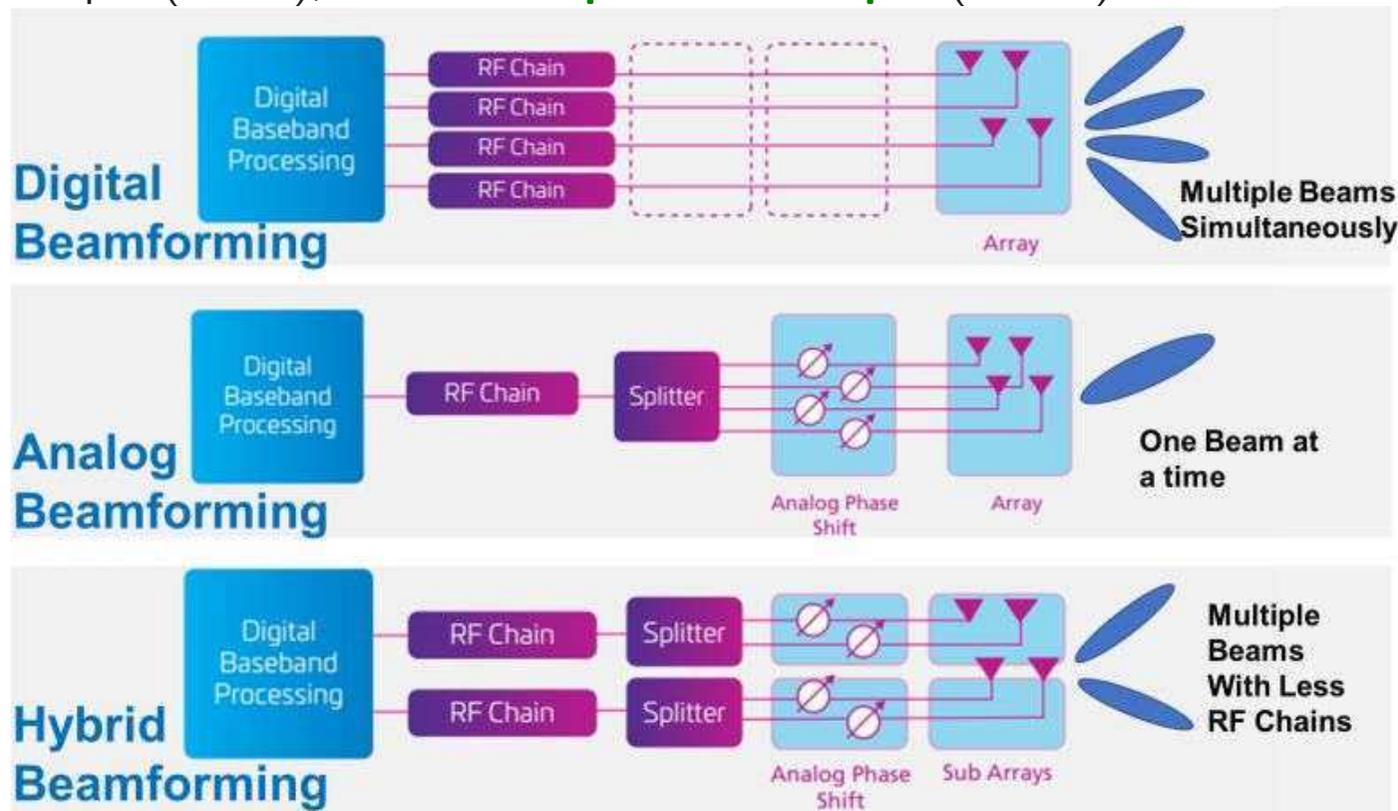
MIMO Introduction

- MIMO are for a) **diversity** gain, and/or b) **multiplexing** gain
1. Create **diversity** for combating fading
 - With sufficient antenna spacing (10 wavelengths), each antenna experiences independent fading → When one signal is in its deep fade, others are unlikely the same
 2. Increase **throughput**
 - Data stream is first S/P, each sub-sequence mapped to an antenna → This creates many “digital pipes” to support higher rate
 3. When multiplexing gain is not used to increase throughput, it can support **multiple users**
 - With multiple receive antennas, each spatially separated user has a unique set of CIRs seen at receiver → This enables SDMA
 4. **Beamforming**, antenna spacing half of wavelength
 - Improve signal quality or support SDMA



MIMO/Beamforming

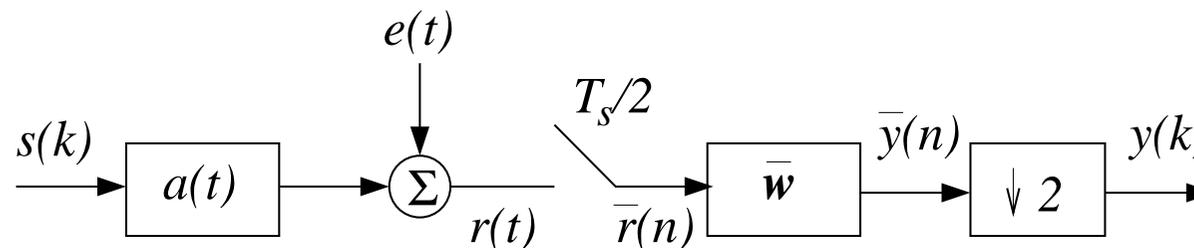
- MIMO classification: single-input single-output (SISO), single-input multi-output (SIMO), multi-input single-output (MISO), and **multi-input multi-output** (MIMO)



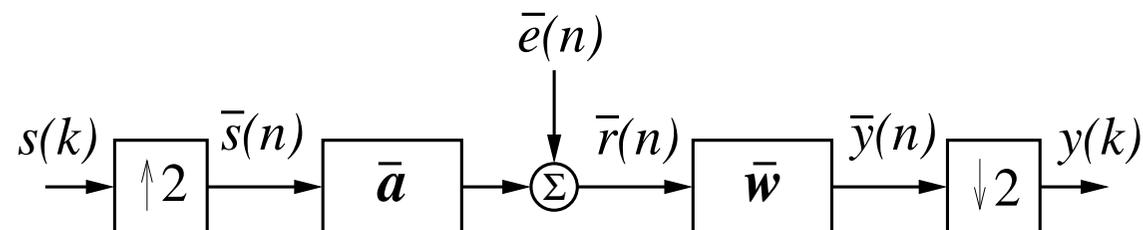
- Beamforming (transmit/receive) classification
 - Digital** beamforming: baseband, one RF chain for each antenna, multi data streams/users
 - Analog** beamforming: RF band, one RF chain for whole antenna array, one data stream
 - Hybrid** beamforming, both base band/RF band, multi data streams/users

Fractional-Spaced Sampling

- Single-user (SISO) system with **fractional-space** sampling
 - At each symbol period take more than one sample \Rightarrow multiple (symbol-rate) subchannel models
 - **Advantage**: robust to carrier/timing recovery errors, perfect reconstruction with finite equalizer
 - **Disadvantage**: noise sample no longer white
- **Baseband** model with $T_s/2$ -spaced receiver:
 - Take two samples during each symbol period
 - $T_s/2$ -spaced equaliser: $\bar{y}(n) = \bar{\mathbf{w}}^H \bar{\mathbf{r}}(n)$
with $\bar{\mathbf{r}}(n) = [\bar{r}(n) \bar{r}(n-1) \cdots \bar{r}(n-2m+1)]^T$ and $\bar{\mathbf{w}} = [\bar{w}_0 \bar{w}_1 \cdots \bar{w}_{2m-1}]^T$
 - $\bar{y}(n)$ is decimated by a factor of 2 to get T_s -spaced output $y(k)$



- **Multirate** model with $T_s/2$ -spaced equaliser
 - k indicates T_s -spaced quantities and n indicates $T_s/2$ -spaced quantities



Multirate/Multichannel Model

- $T_s/2$ -spaced sequence $\{\bar{s}(n)\}$ is zero-filled transmitted symbol sequence $\{s(k)\}$ defined by

$$\bar{s}(n) = \begin{cases} s(n/2), & \text{for even } n \\ 0, & \text{for odd } n \end{cases}$$

- Received $T_s/2$ -spaced signal sample is

$$\bar{r}(n) = \sum_{i=0}^{2N_c-1} \bar{a}_i \bar{s}(n-i) + \bar{e}(n)$$

- $T_s/2$ -spaced complex-valued channel impulse response (CIR) is given by

$$\bar{\mathbf{a}} = [\bar{a}_0 \ \bar{a}_1 \ \bar{a}_2 \ \bar{a}_3 \ \cdots \ \bar{a}_{2N_c-1}]^T$$

- Sampling at twice of symbol rate \Rightarrow Two symbol-spaced models: Odd sample model and even sample model, i.e., **multichannel** model
- Take even and odd samples of noise and Rx signal

$$e^e(k) = \bar{e}(2n), \quad e^o(k) = \bar{e}(2n+1), \quad r^e(k) = \bar{r}(2n), \quad r^o(k) = \bar{r}(2n+1)$$

- Define even and odd channels and equalisers (all are symbol-rate sequences) as:

$$\begin{aligned} \bar{\mathbf{a}}^e &= [\bar{a}_0 \ \bar{a}_2 \ \cdots \ \bar{a}_{2N_c-2}]^T, & \bar{\mathbf{a}}^o &= [\bar{a}_1 \ \bar{a}_3 \ \cdots \ \bar{a}_{2N_c-1}]^T, \\ \bar{\mathbf{w}}^e &= [\bar{w}_0 \ \bar{w}_2 \ \cdots \ \bar{w}_{2m-2}]^T, & \bar{\mathbf{w}}^o &= [\bar{w}_1 \ \bar{w}_3 \ \cdots \ \bar{w}_{2m-1}]^T \end{aligned}$$

Multichannel Model (continue)

- Symbol-rate output

$$y(k) = \sum_{i=0}^{2m-1} w_i^* r(k-i) = \mathbf{w}^H \mathbf{r}(k)$$

$$\text{with } \mathbf{w} = \begin{bmatrix} (\bar{\mathbf{w}}^o)^T & (\bar{\mathbf{w}}^e)^T \end{bmatrix}^T \text{ and}$$

$$\mathbf{r}(k) = \begin{bmatrix} (\mathbf{r}^e(k))^T & (\mathbf{r}^o(k))^T \end{bmatrix}^T$$

- In general, T_s/K -spaced sampling will result in K channel models

- As we have symbol-rate model $y(k) = \mathbf{w}^H \mathbf{r}(k)$, all equalisation results apply

- **Blind equalisation example:** consists of a 22-tap $T_s/2$ channel and a 26-tap $T_s/2$ equaliser with 256-QAM and SNR= 60 dB

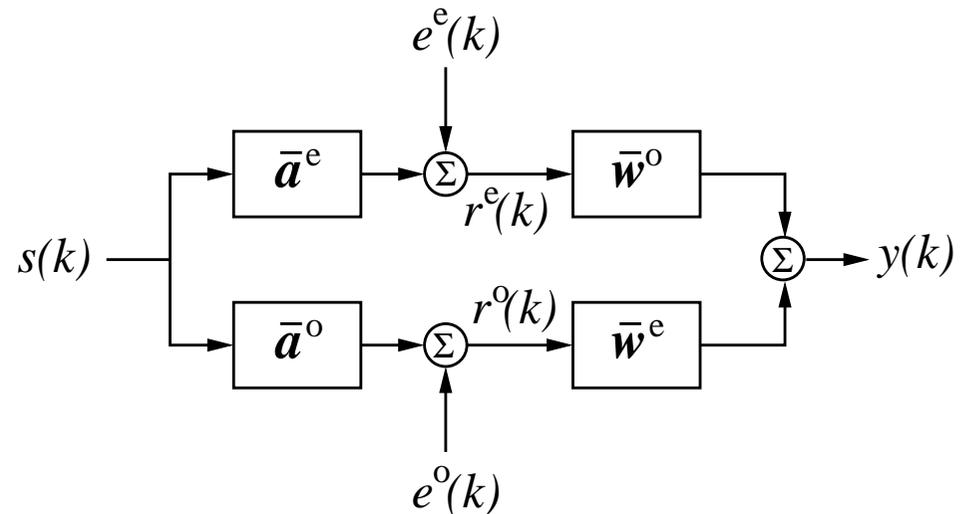
- In simulation, an estimated MSE based on a separate block of data and the maximum distortion measure defined by

$$\text{MD} = \frac{\sum_{i=0}^{n_{\text{tot}}} |f_i| - |f_{i_{\text{max}}}|}{|f_{i_{\text{max}}}|}$$

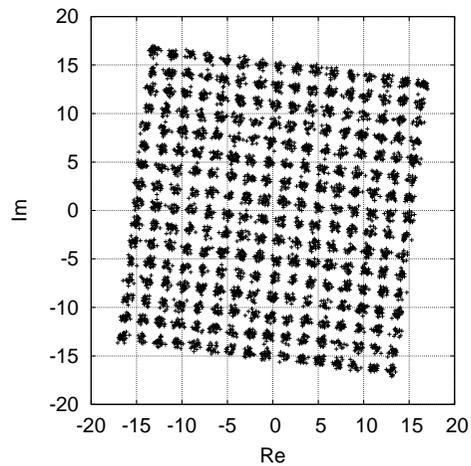
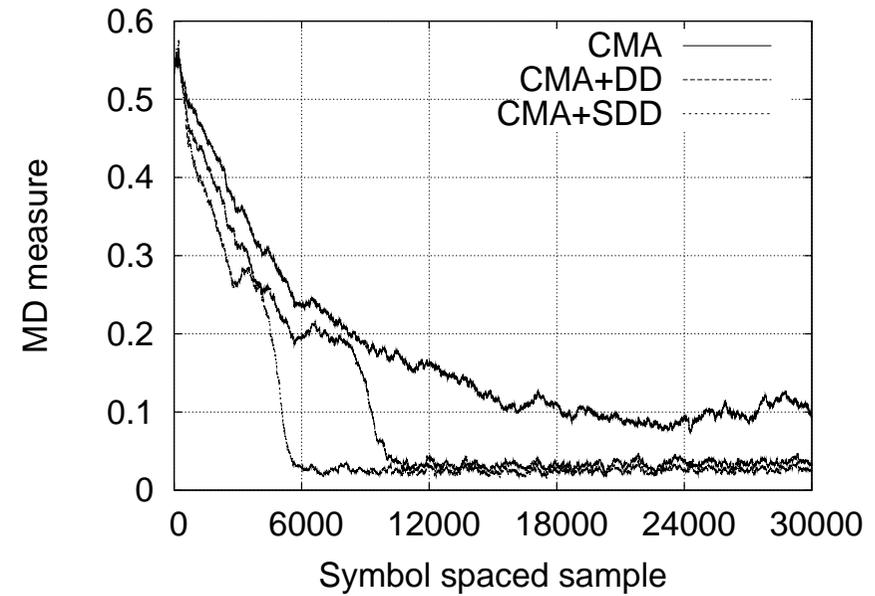
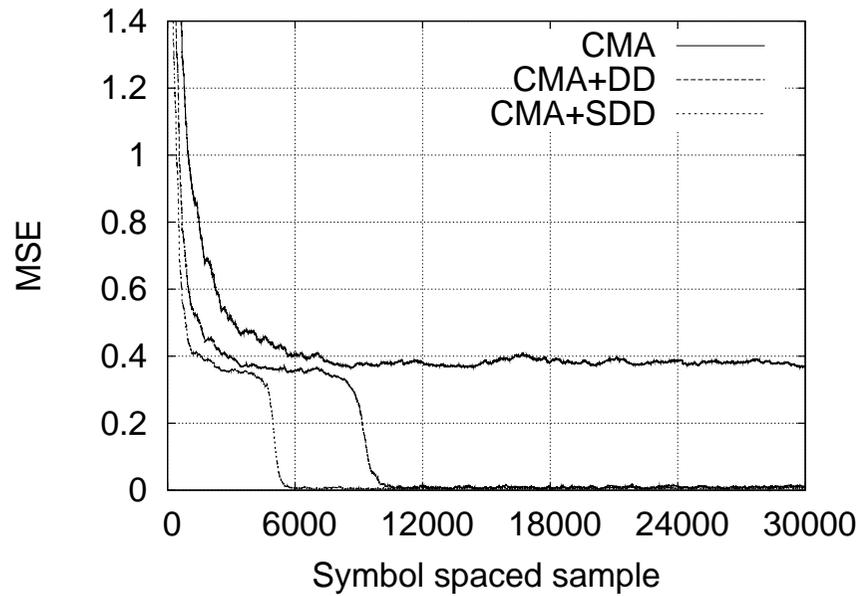
are used to assess convergence rate, where

$$\mathbf{f} = [f_0 \ f_1 \ \dots \ f_{n_{\text{tot}}}]^T = (\bar{\mathbf{w}}^o)^* \star \bar{\mathbf{a}}^e + (\bar{\mathbf{w}}^e)^* \star \bar{\mathbf{a}}^o$$

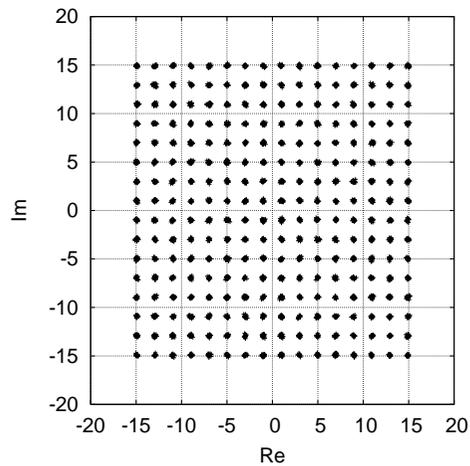
is the combined impulse response of the channel and equaliser, $f_{i_{\text{max}}} = \max\{f_i, 0 \leq i \leq n_{\text{tot}}\}$



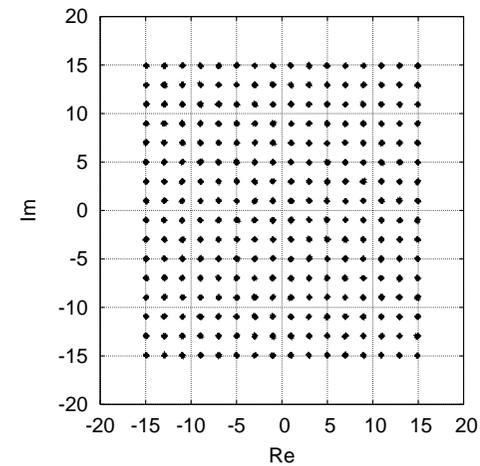
Simulation Results



(a) CMA



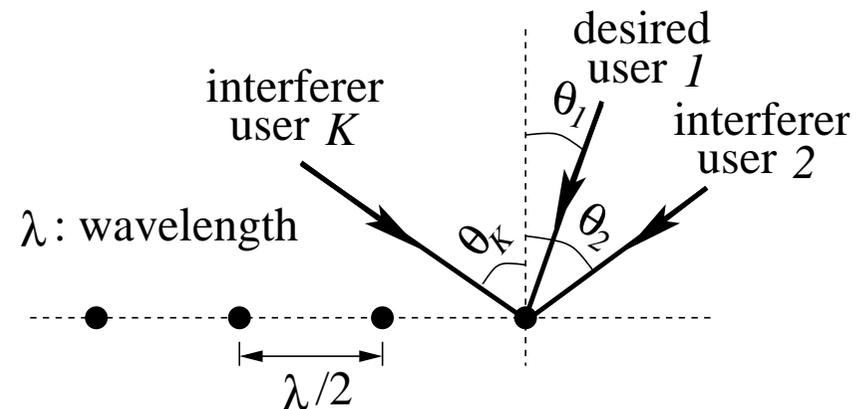
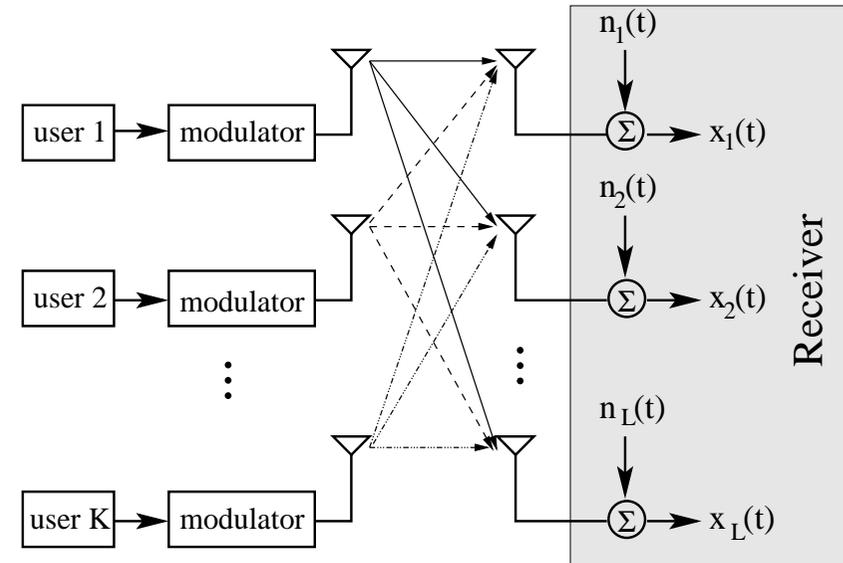
(b) CMA+DD



(c) CMA+SDD

Digital Beamforming Assisted Receiver

- MIMO based **SDMA** system:
 - Uplink, BS **digital beamforming** receiver to implement **multiuser detection**
 - Assume L receiver antennas supporting K single-antenna users
 - Narrowband channels with $m_i(k) = A_i b_i(k)$, A_i : channel coefficient for user i and $b_i(k)$: k th symbol of user i
 - Symbol-rate sampling, BS detects user i data for $1 \leq i \leq K$
- Uniformly spaced linear antenna array:
 - $t_l(\theta_i)$ be relative time delay at array element l for user i
 - θ_i angle of arrival for user i and carrier $\omega = 2\pi f_c$



$$x_l(k) = \sum_{i=1}^K m_i(k) \exp(j\omega t_l(\theta_i)) + n_l(k) = \bar{x}_l(k) + n_l(k), \quad 1 \leq l \leq L$$

Beamforming Assisted Receiver (continue)

- System model: Antenna array output $\mathbf{x}(k) = [x_1(k) \ x_2(k) \ \cdots \ x_L(k)]^T$ is expressed as

$$\mathbf{x}(k) = \bar{\mathbf{x}}(k) + \mathbf{n}(k) = \mathbf{H}\mathbf{b}(k) + \mathbf{n}(k)$$

where $\mathbf{n}(k) = [n_1(k) \ n_2(k) \ \cdots \ n_L(k)]^T$ has a covariance matrix of $E[\mathbf{n}(k)\mathbf{n}^H(k)] = 2\sigma_n^2\mathbf{I}_L$ with \mathbf{I}_L representing the $L \times L$ identity matrix, uplink channel matrix \mathbf{H} is given by

$$\mathbf{H} = [\mathbf{h}_1 \ \mathbf{h}_2 \ \cdots \ \mathbf{h}_K] = [A_1\mathbf{s}_1 \ A_2\mathbf{s}_2 \ \cdots \ A_K\mathbf{s}_K]$$

steering vector for source i is

$$\mathbf{s}_i = [\exp(j\omega t_1(\theta_i)) \ \exp(j\omega t_2(\theta_i)) \ \cdots \ \exp(j\omega t_L(\theta_i))]^T$$

and transmitted symbol vector $\mathbf{b}(k) = [b_1(k) \ b_2(k) \ \cdots \ b_K(k)]^T$

- Given channel matrix \mathbf{H} , **maximum likelihood** MUD

$$\hat{\mathbf{b}}_{\text{ML}}(k) = \arg \min_{\mathbf{b} \in \mathcal{B}} \|\mathbf{x}(k) - \mathbf{H}\mathbf{b}\|$$

Feasible solution set $\mathcal{B} = \{\mathbf{b}_1, \cdots, \mathbf{b}_{N_f}\}$, with $N_f = K^M$ for M -QAM

- Linear** beamforming/MUD: BS beamformer output vector $\mathbf{y}(k) = [y_1(k) \ \cdots \ y_K(k)]^T$

$$\hat{\mathbf{b}}(k) = \mathbf{y}(k) = \mathbf{W}^H \mathbf{x}(k)$$

Beamformer complex-valued weight matrix $\mathbf{W} = [\mathbf{w}_1 \ \cdots \ \mathbf{w}_K]$ with $\mathbf{w}_i = [w_{1,i} \ \cdots \ w_{L,i}]^T$

Linear Beamforming/MUD

- **Zero-forcing** (ZF) beamforming/MUD

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

- **Minimum mean square error** (MMSE) beamforming/MUD

$$\mathbf{W}_{MMSE}^H = \left(\mathbf{H}^H \mathbf{H} + \frac{2\sigma_n^2}{\sigma_s^2} \mathbf{I}_K \right)^{-1} \mathbf{H}^H \text{ or } \mathbf{W}_{MMSE} = \mathbf{H} \left(\mathbf{H}^H \mathbf{H} + \frac{2\sigma_n^2}{\sigma_s^2} \mathbf{I}_K \right)^{-1}$$

with $E[|b_i(k)|^2] = \sigma_s^2$. For user i ,

$$\mathbf{w}_{i,MMSE} = \left(\mathbf{H}\mathbf{H}^H + \frac{2\sigma_n^2}{\sigma_s^2} \mathbf{I}_L \right)^{-1} \mathbf{h}_i$$

MMSE solution can be implemented adaptively using the LMS or RLS algorithm

- Further consider for user i , beamformer's output is given by

$$y_i(k) = \mathbf{w}_i^H \mathbf{x}(k) = \mathbf{w}_i^H \bar{\mathbf{x}}(k) + \mathbf{w}_i^H \mathbf{n}(k) = \bar{y}_i(k) + e_i(k)$$

where $e_i(k)$ is Gaussian distributed having a zero mean and $E[|e_i(k)|^2] = 2\sigma_n^2 \mathbf{w}_i^H \mathbf{w}_i$

- Assume M -QAM modulation and define combined impulse response of beamformer and system as

$$\mathbf{w}_i^H \mathbf{H} = \mathbf{w}_i^H [\mathbf{h}_1 \ \mathbf{h}_2 \ \cdots \ \mathbf{h}_K] = [c_{i,1} \ c_{i,2} \ \cdots \ c_{i,K}]$$

Linear Beamforming/MUD (continue)

- **Minimum symbol error rate:** beamformer's output can also be expressed as

$$y_i(k) = c_{i,i}b_i(k) + \sum_{1 \leq l \leq K, l \neq i} c_{i,l}b_l(k) + e_i(k)$$

- Define decision variable as $d_i(k) = d_{i_R}(k) + jd_{i_I}(k) = \frac{y_i(k)}{c_{i,i}}$, then symbol decision $\hat{b}_i(k) = \hat{b}_{i_R}(k) + j\hat{b}_{i_I}(k)$ is given by

$$\hat{b}_{i_R}(k) = \begin{cases} u_1, & \text{if } d_{i_R}(k) \leq u_1 + 1 \\ u_l, & \text{if } u_l - 1 < d_{i_R}(k) \leq u_l + 1 \text{ for } 2 \leq l \leq \sqrt{M} - 1 \\ u_{\sqrt{M}}, & \text{if } d_{i_R}(k) > u_{\sqrt{M}} - 1 \end{cases}$$

$$\hat{b}_{i_I}(k) = \begin{cases} u_1, & \text{if } d_{i_I}(k) \leq u_1 + 1 \\ u_q, & \text{if } u_q - 1 < d_{i_I}(k) \leq u_q + 1 \text{ for } 2 \leq q \leq \sqrt{M} - 1 \\ u_{\sqrt{M}}, & \text{if } d_{i_I}(k) > u_{\sqrt{M}} - 1 \end{cases}$$

where M -QAM symbol set is defined as $\{u_l + ju_q, 1 \leq l, q \leq \sqrt{M}\}$

- Given \mathbf{H} , we have set of \bar{y}_i , which is function of \mathbf{w}_i , and symbol error rate is **mixture** of Q functions
- Minimize this symbol error rate leads to **MSER** solution $\mathbf{w}_{i\text{MSER}}$

Adaptive MBER Beamforming Solution

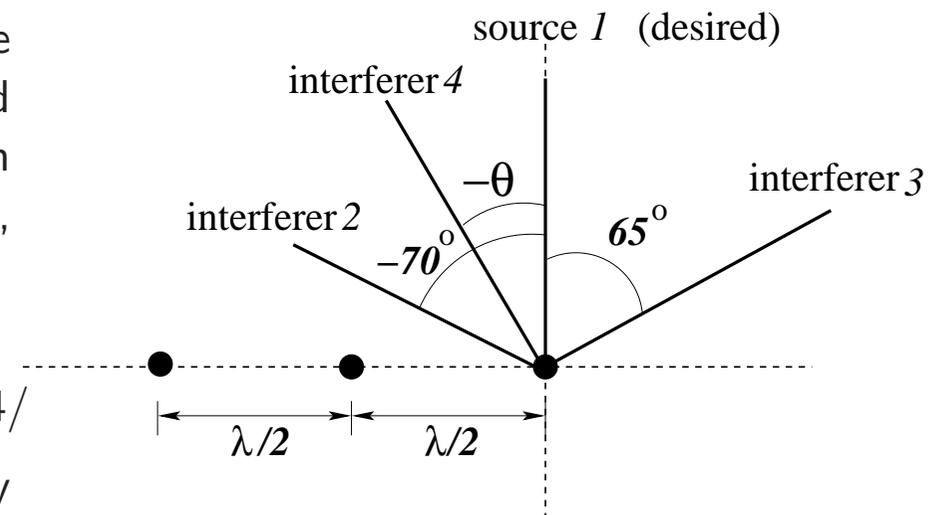
- We can derive minimum symbol error rate solution $\mathbf{w}_{i,\text{MSER}}$ for general QAM
 - Unlike MMSE solution, there is no closed-form solution for $\mathbf{w}_{i,\text{MSER}}$
 - Gradient optimisation must be used
- MSER solution can be implemented adaptively using the LSER algorithm
- For details see:

S. Chen, H.-Q. Du and L. Hanzo, "Adaptive minimum symbol error rate beamforming assisted receiver for quadrature amplitude modulation systems," in *VTC2006-Spring* (Melbourne, Australia), May 7-10, 2006

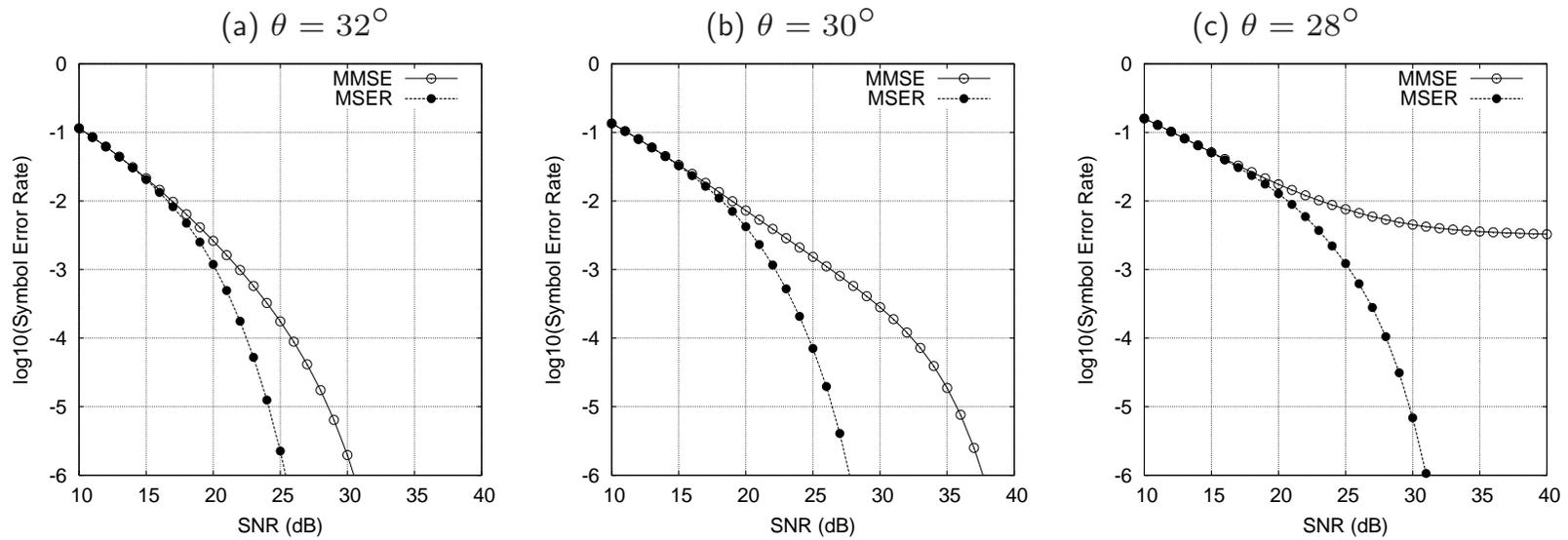
PDF copy can be download from:

<https://www.southampton.ac.uk/~sqc/ELEC6214/>

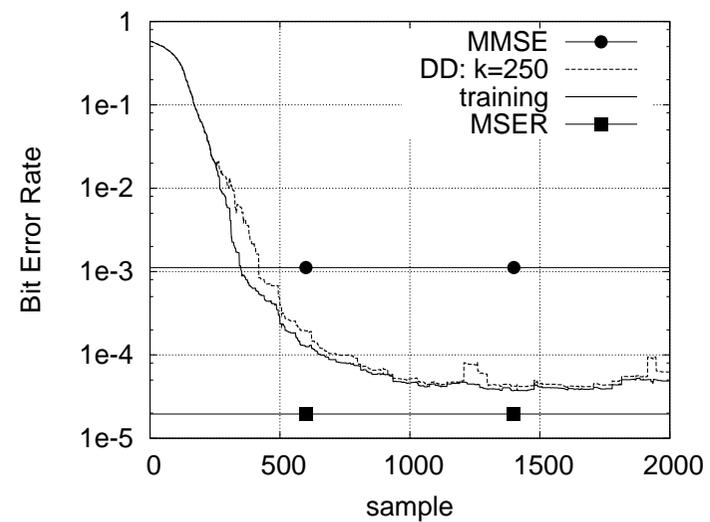
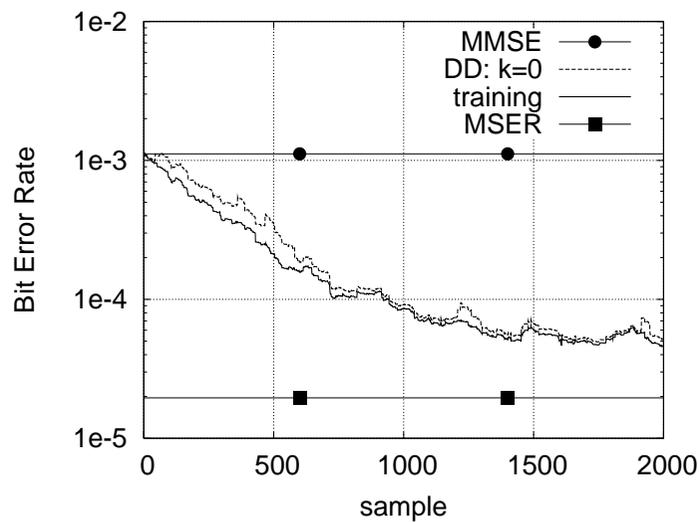
- Example: 16QAM, 4 users, 3-element antenna array



Beamforming Results

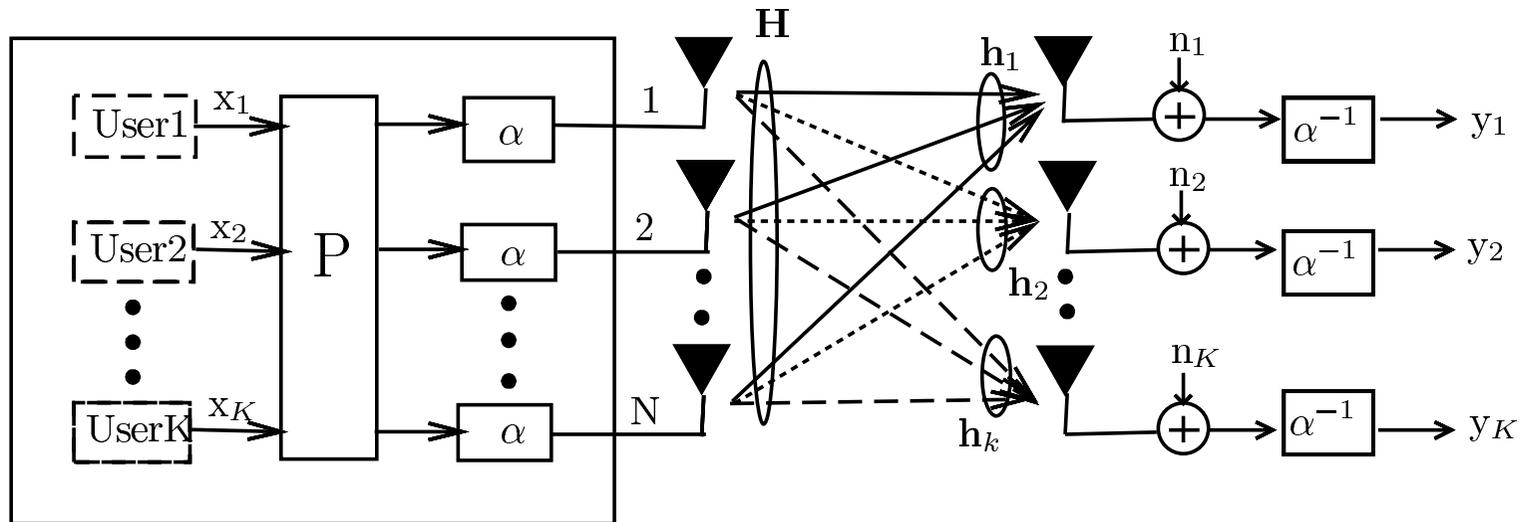


$\theta = 30^\circ$ and SNR = 26 dB, LSER



Transmit Beamforming/Precoding

- BS employs N transmit antennas to communicate with K single-receive-antenna MSs, i.e. downlink



- MSs unable to perform multiuser detection or receive beamforming
 - BS does **multiuser transmission/transmit beamforming/precoding** to combat MUI
- Received signal vector of K MSs, $\mathbf{y}(k) = [y_1(k) \ y_2(k) \ \cdots \ y_K(k)]^T$, is given by

$$\mathbf{y}(k) = \mathbf{H}^T \mathbf{P} \mathbf{x}(k) + \alpha^{-1} \mathbf{n}(k)$$

- Transmit vector $\mathbf{x}(k) = [x_1(k) \ x_2(k) \ \cdots \ x_K(k)]^T$
- Downlink channel matrix $\mathbf{H} = [\mathbf{h}_1 \ \mathbf{h}_2 \ \cdots \ \mathbf{h}_K] \in \mathbb{C}^{N \times K}$
- Precoding matrix $\mathbf{P} = [\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_K] \in \mathbb{C}^{N \times K}$
- $\mathbf{n}(k)$ is AWGN vector, and α a transmit power normalization factor

Digital Precoding Design

- The task is to design the precoding matrix \mathbf{P} so that
 - i th MS's receive signal $y_i(k)$ is **sufficient** statistic to detect $x_i(k)$
 - Zero-forcing precoding: $\mathbf{P}_{ZF} = \mathbf{H}^* (\mathbf{H}^T \mathbf{H}^*)^{-1}$
 - Similarly one can have MMSE precoding, and even MBER precoding
 - BS cannot estimate downlink channel matrix, and MSs have to feed back corresponding downlink channel estimates to BS
 - For TDD, uplink channel and downlink channel are reciprocal
 - BS can estimate uplink channel, which is needed in uplink MUD, and it can use estimated uplink channel as downlink channel in precoding design
 - Fundamentally, for TDD, uplink receive beamforming or MUD is exactly equivalent to downlink transmit beamforming or MUT
1. L.-L. Yang, "Design of linear multiuser transmitters from linear multiuser receivers," in *Proc. ICC 2007* (Glasgow, Scotland), June 24-28, 2007, pp. 5258–5263
 2. S. Chen and L.-L. Yang, "Downlink MBER beamforming design based on uplink MBER receive beamforming for TDD-SDMA induced MIMO systems," *Communications and Networks*, vol.2, pp. 145–151, Aug. 2010



Summary

- MIMO system introduction: diversity gain and multiplexing gain
- Single-user fractional-spaced receiver
 - Baseband continuous-time model, discrete-time multirate model, discrete-time multichannel model
- Digital beamforming assisted receiver for QAM modulation
 - Also known as multiuser detection
- Digital transmit beamforming or precoding
 - Also known as multiuser transmission
- For TDD, equivalency of uplink receive beamforming or MUD and downlink transmit beamforming or MUT

