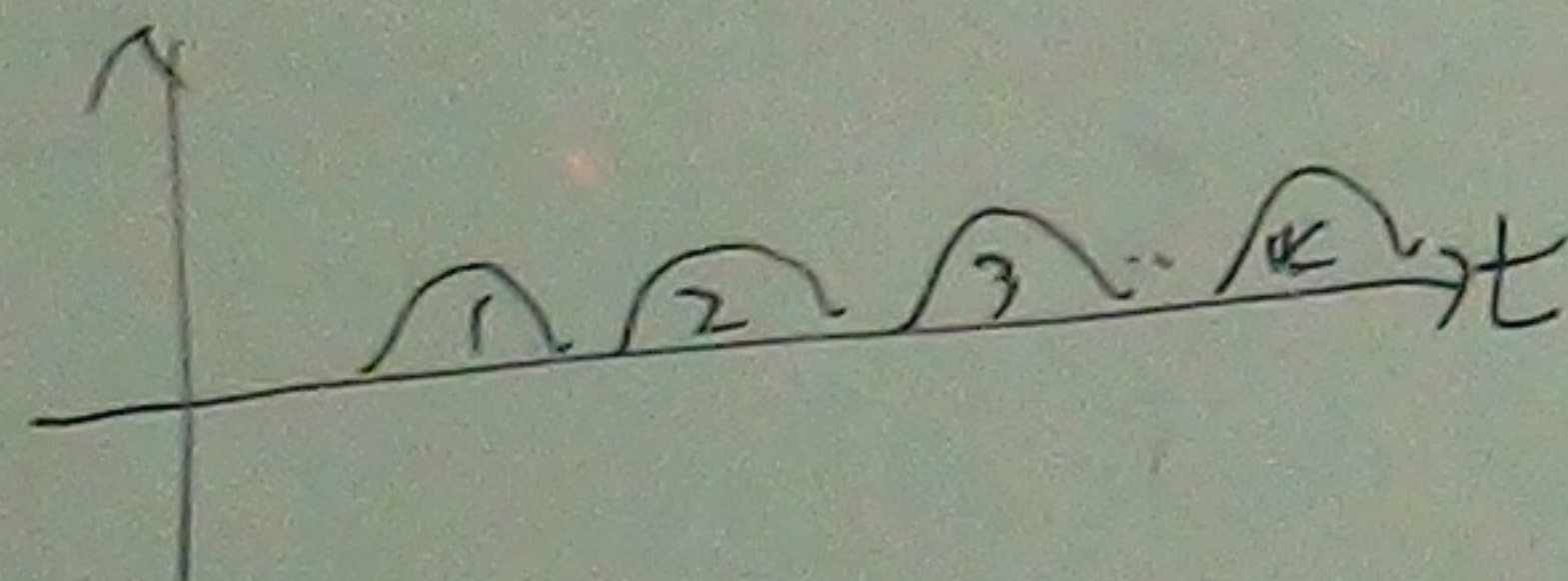
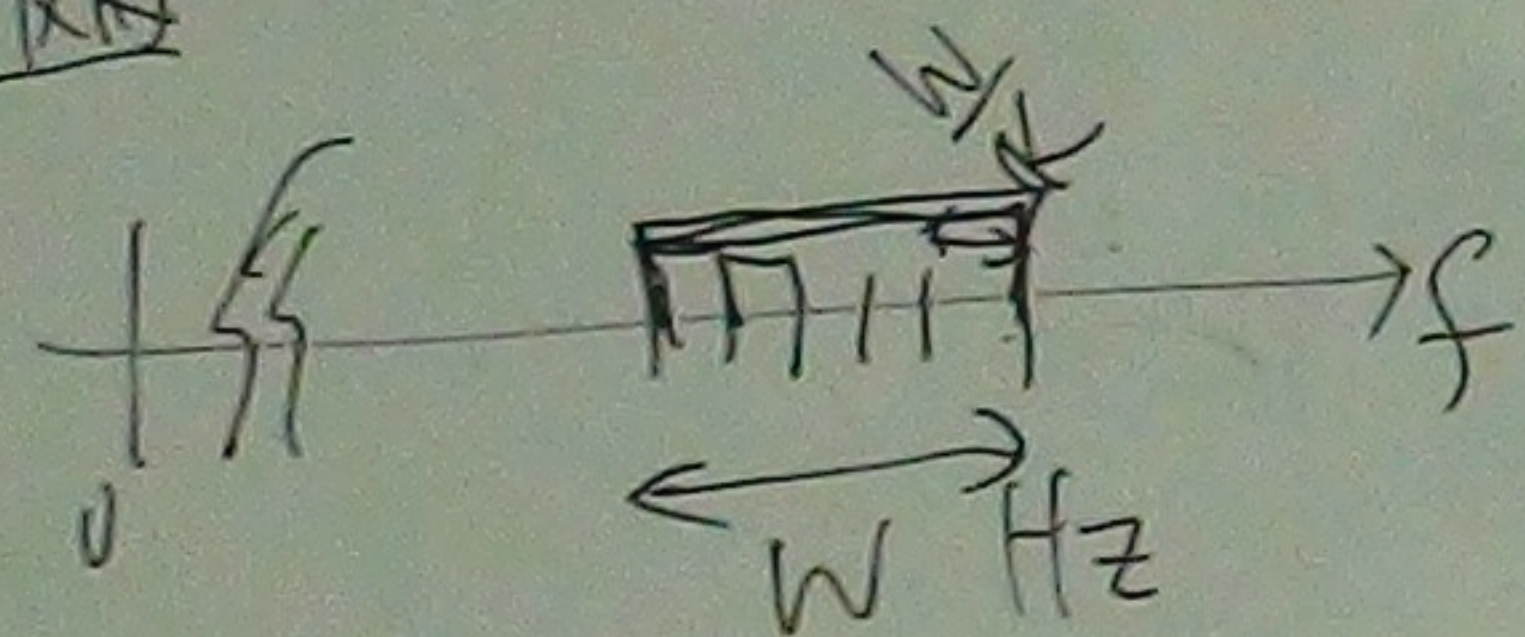
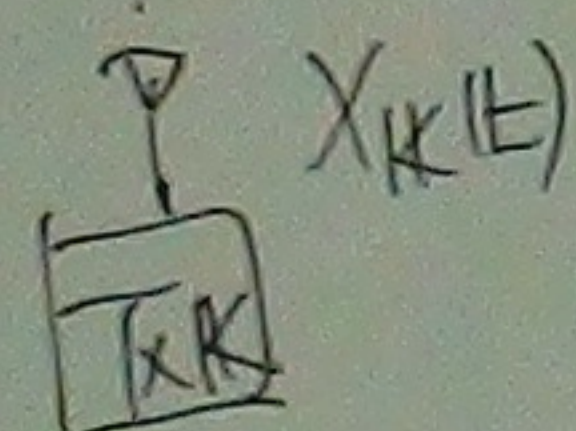
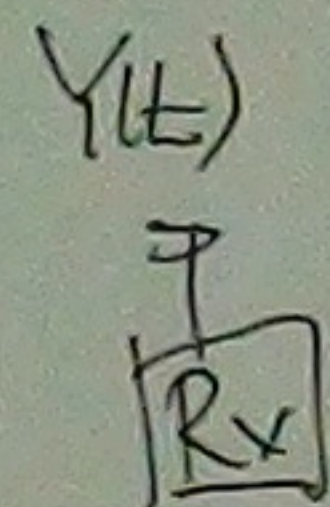
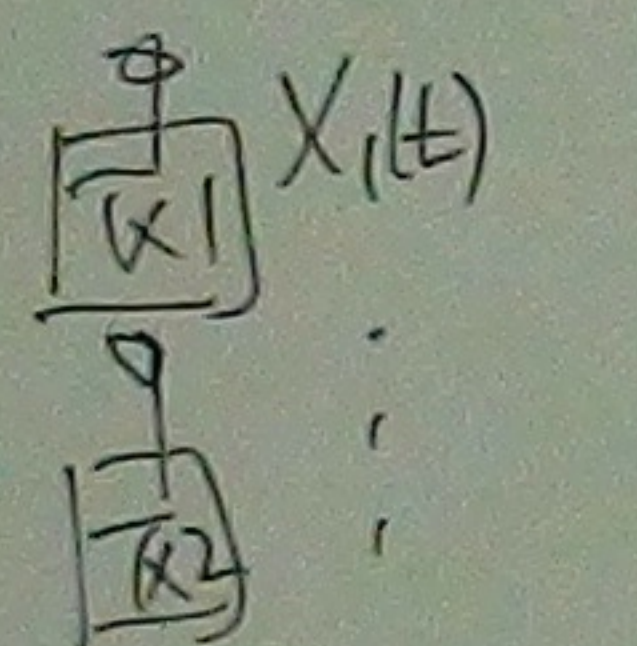


□ T-IT 2005 "Continuous-Time Equivalents of Welch Bound Equality Sequences"

MA (multiple access) $\leftarrow \begin{matrix} FD \\ TD \\ CD \end{matrix} \approx DS/SS \text{ MA} \leftarrow \begin{matrix} \text{short-code} \\ \text{long-code} \end{matrix} \leftarrow \text{generalized random spreading}$



$$Y(t) = \sum_{k=1}^K \sum_{m=-\infty}^{\infty} \sqrt{P} d_k[m] S_k(t - mT) + N(t)$$

$$\|S_k(t)\|^2 = T = \int_{-\infty}^{\infty} |S_k(t)|^2 dt = T$$

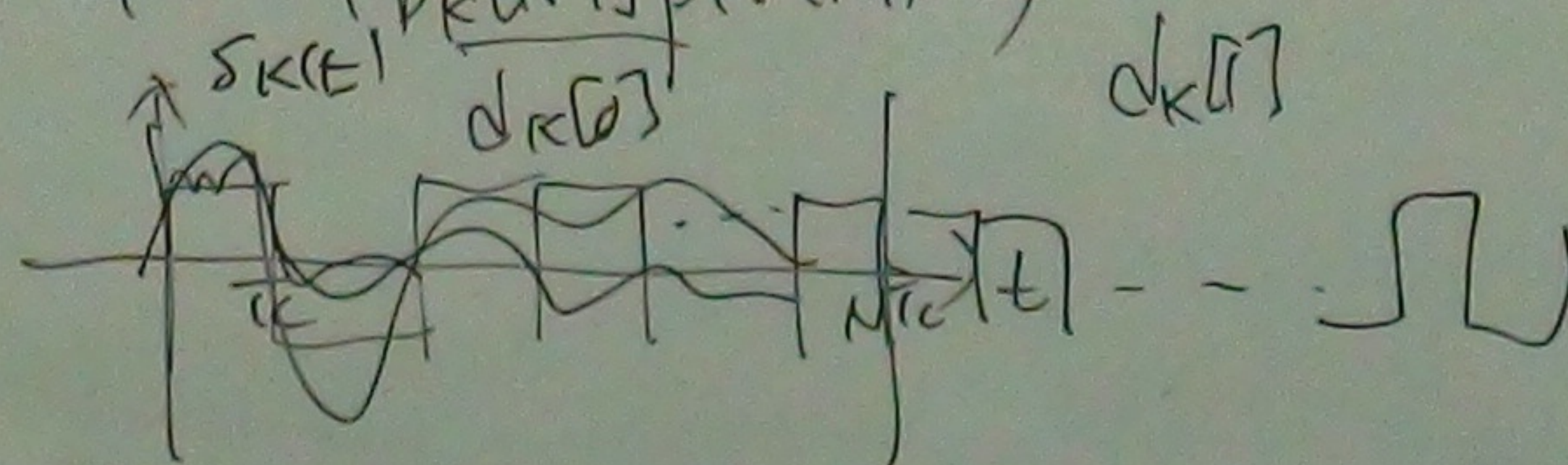
$(d_k[m])_m$ zero-mean, unit-variance white symbol seq

$\{(d_k[m])_m\}_k$ uncorrelated

$$P_k = \frac{P \|S_k(t)\|^2}{T} = P \cdot \frac{1}{K}$$

$$S_k(t) = \sum_{n=0}^{N-1} \underbrace{d_k[n]}_{\text{given}} p(t - nT_c)$$

$$= d_k[0] p(t) + d_k[1] p(t - T_c) + \dots + d_k[N-1] p(t - (N-1)T_c)$$

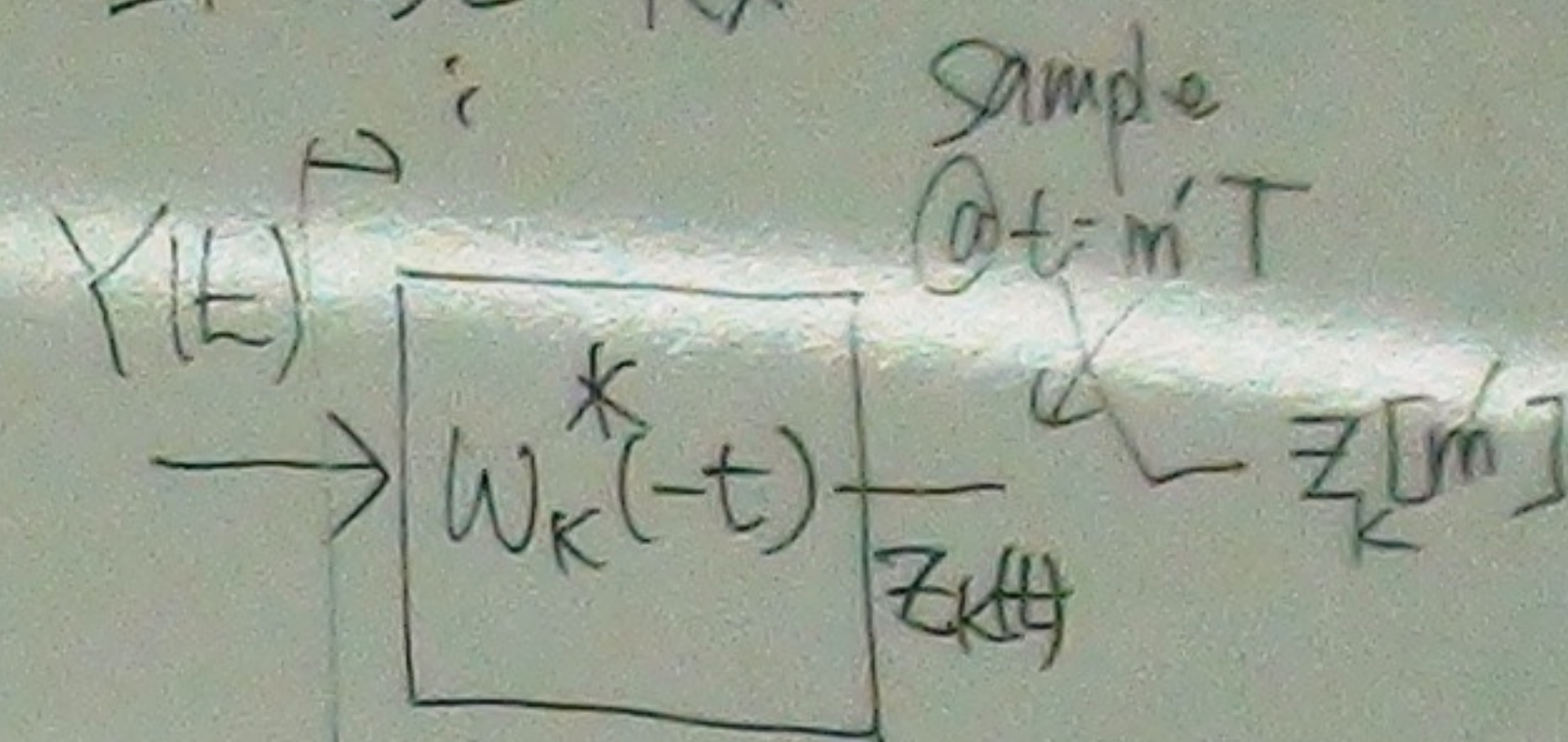


$$KT \gg \frac{1}{W}$$

$$T = NT_c$$

(X)

◦ LMMSE Rx

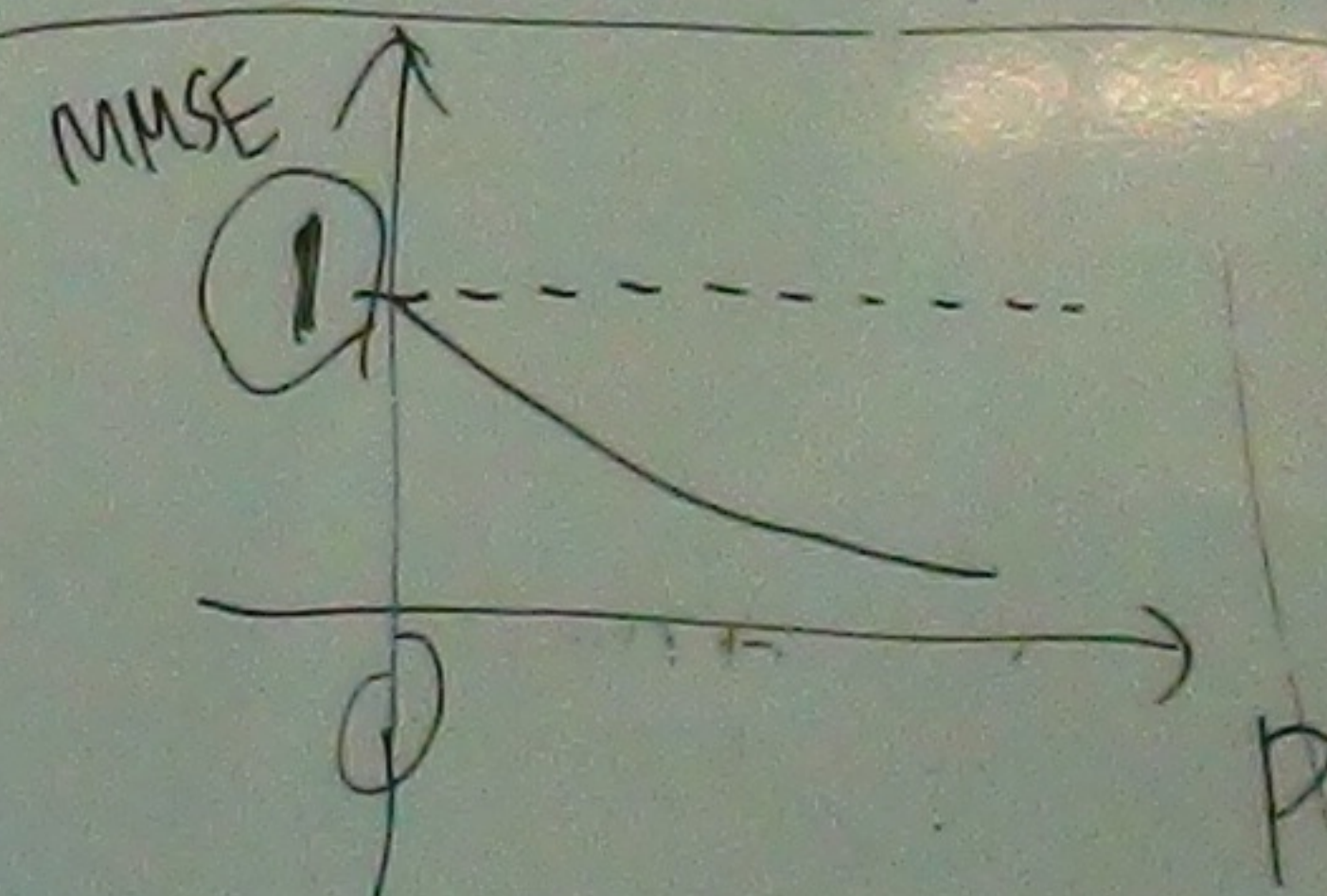


$$TMSE = \sum_{k=1}^K E \{ \|z_k[m] - d_k[m]\|^2 \}$$

min TMSE

$$\equiv \left\{ \min_{W_k(t)} E \{ \|z_k[m] - d_k[m]\|^2 \} \right\}_{k=1}^K$$

$$\boxed{\begin{aligned} MMSE_k &= \frac{1}{MSINR_k + 1} \\ MSINR_k &= \frac{1}{MMSE_k} - 1 \end{aligned}}$$



□ T-IT 2005 "Continuous-Time Equivalents of Welch Bound Equality Sequences" If $\underline{w}_{\text{MSINR}}$ maximizes SINR, then $k \underline{w}_{\text{MSINR}}$ " " (K ≠ 0)

○ MSE vs. SINR

$$\underline{Y} = \underline{d} \underline{s} + \underline{N}$$

$\underline{d}$: random variable

$$\begin{cases} E\{\underline{d}\} = 0 \\ E\{\underline{d}^2\} = 0 \\ E\{|\underline{d}|^2\} = 1 \end{cases}$$

$\underline{s}$: given deterministic

$\underline{N}$: additive noise

$$\begin{cases} E\{\underline{N}\} = 0 \\ E\{\underline{N}\underline{N}^H\} = 0 \\ E\{\underline{N}\underline{N}^H\} = \underline{R}_{\underline{N}} \end{cases}$$

$$\min_{\underline{w}} E\{|\underline{w}^H \underline{Y} - \underline{d}|^2\} = \text{MMSE}$$

$$\text{SINR} \triangleq \frac{|\underline{w}^H \underline{s}|^2}{E\{|\underline{w}^H \underline{N}|^2\}}$$

$|\underline{w}^H \underline{s}|^2 \leftarrow$ average signal power

$$\underline{w}^H \underline{Y} = \underbrace{\underline{w}^H \underline{s}}_{\text{signal}} + \underbrace{\underline{w}^H \underline{N}}_{\text{IN}}$$

$$= \frac{|\underline{w}^H \underline{s}|^2}{\text{Var}\{\underline{w}^H \underline{Y} | \underline{d}\}}$$

$$|\underline{d}| = 1, \text{ MPSK}$$

$$E\{|\underline{d}|^2\} = 1$$

$$\text{MMSE} = \frac{\text{MSINR} + 1}{\text{MSINR}}$$

Not unique!

$$|\underline{w}^H \underline{s}|^2 = C_i$$

$$\max_{\underline{w}} \frac{|\underline{w}^H \underline{s}|^2}{\underline{w}^H \underline{R}_{\underline{N}} \underline{w}}$$

$$\min_{\underline{w}} \underline{w}^H (\underline{s} \underline{s}^H + \underline{R}_{\underline{N}}) \underline{w} - 2 \text{Re}\{\underline{w}^H \underline{s}\} + 1$$

$$\left(1 - \frac{\underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}}{1 + \underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}}\right) \underline{R}_{\underline{N}}^{-1} \underline{s}$$

$$\min_{\underline{w}} \underline{w}^H \underline{R}_{\underline{N}} \underline{w} \text{ s.t. } \underline{w}^H \underline{s} = 1$$

$$\mathcal{L} = \underline{w}^H \underline{R}_{\underline{N}} \underline{w} - \lambda (\underline{w}^H \underline{s} - 1)$$

$$\begin{cases} \frac{\partial \mathcal{L}}{\partial \underline{w}^H} = 0 \Rightarrow 2 \underline{R}_{\underline{N}} \underline{w}_{\text{opt}} - \lambda \underline{s} = 0 \\ \frac{\partial \mathcal{L}}{\partial \lambda} = 0 \Rightarrow \underline{w}_{\text{opt}}^H \underline{s} = 1 \end{cases}$$

$$\underline{w}_{\text{opt}} = \frac{1}{\underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}} \underline{R}_{\underline{N}}^{-1} \underline{s}$$

$$\underline{w}_{\text{MSINR}} \propto \underline{R}_{\underline{N}}^{-1} \underline{s}$$

$$\text{MMSE} = 1 - \frac{\underline{s}^H (\underline{R}_{\underline{N}} + \underline{s} \underline{s}^H) \underline{s}}{\underline{s}^H \underline{R}_{\underline{N}} \underline{s}} = 1 - \frac{\underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}}{\underline{s}^H \underline{R}_{\underline{N}} \underline{s}}$$

$$\frac{\partial \mathcal{L}}{\partial \underline{w}^H} = 0 \Leftrightarrow 2(\underline{s} \underline{s}^H + \underline{R}_{\underline{N}}) \underline{w}_{\text{opt}} - 2 \underline{s} = 0$$

$$\underline{w}_{\text{opt}} = (\underline{s} \underline{s}^H + \underline{R}_{\underline{N}})^{-1} \underline{s} = \left(\underline{R}_{\underline{N}}^{-1} + \frac{\underline{R}_{\underline{N}}^{-1} \underline{s} \underline{s}^H \underline{R}_{\underline{N}}^{-1}}{1 + \underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}} \right) \underline{s}$$

$$= 1 - \frac{\underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}}{1 + \underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}} = \frac{1}{1 + \frac{\underline{s}^H \underline{R}_{\underline{N}}^{-1} \underline{s}}{\underline{s}^H \underline{R}_{\underline{N}} \underline{s}}} = \frac{1}{\text{MSINR} + 1}$$

$$\max_{\underline{w}} \frac{\underline{w}^H (\underline{s} \underline{s}^H) \underline{w}}{\underline{w}^H (\underline{R}_{\underline{N}}) \underline{w}}$$

$$= \frac{C - (C+ab)}{(C+ab)C} = \frac{-ab}{(C+ab)C} = -\frac{\frac{a}{C} \frac{b}{C}}{(1+\frac{ab}{C})}$$



□ T-IT 2005 "Continuous-Time Equivalents of Welch Bound Equality Sequences"

$$\textcircled{1} \quad Y(t) = \sum_{k=1}^K \sum_{m=-\infty}^{\infty} d_k[m] s_k(t-mT) + \underbrace{N(t)}_{\text{AWGN}}$$

$$\equiv Y(t) = \sum_{k=1}^K \sum_{m=-\infty}^{\infty} d_k[m] s_k(t-mT) + N(t)$$

$$\underline{w}_k(t) \leftrightarrow \underline{w}_k(f) = \underline{s}_k(f)^H \left(\sigma^2 \mathbf{I} + \sum_{k=1}^K \underline{s}_k(f) \underline{s}_k(f)^H \right)^{-1} \underline{s}_k(f)$$

$$\text{MMSE}_k = 1 - \int_{-\frac{1}{2T}}^{\frac{1}{2T}} \underline{s}_k(f)^H \left(\sum_{k=1}^K \underline{s}_k(f) \underline{s}_k(f)^H \right)^{-1} \underline{s}_k(f) df \quad R_{xx}(f)$$

$$\text{TMSE} = \sum_{k=1}^K \text{MMSE}_k = K - \int_{-\frac{1}{2T}}^{\frac{1}{2T}} \left[\sum_{k=1}^K \underline{s}_k(f)^H \left(\sum_{k=1}^K \underline{s}_k(f) \underline{s}_k(f)^H \right)^{-1} \underline{s}_k(f) \right] df$$

$$\left(\sigma^2 \mathbf{I} + \Lambda \right)^{-1} \Lambda$$

$$= \frac{\lambda_k}{\sigma^2 + \lambda_k}$$

$$\textcircled{2} \quad \min_{\substack{\underline{s}_k(t) \\ \underline{s}_k(f)}} \text{TMSE} = K - T \int_{-\frac{1}{2T}}^{\frac{1}{2T}} \sum_{k=1}^K \frac{\lambda_k(f)}{\lambda_k(f) + \sigma^2} df$$

$$\Rightarrow \underline{s}(f) \underline{s}(f)^H = \left(\frac{K}{N} \right) T \mathbf{I}$$

$\underbrace{\hspace{10em}}_{K=N} \uparrow \forall f$

s.t. $\|\underline{s}_k(t)\|^2 = T, \forall k$

$$\int_{-\frac{1}{2T}}^{\frac{1}{2T}} \|\underline{s}_k(f)\|^2 df = T$$

$\downarrow (ABC) = \text{tr}(BCA)$

$\boxed{K < N}$
 $\underline{s}(f)^H \underline{s}(f) = \mathbf{I}$

$$\textcircled{3} \quad \sum_{k=1}^K \left[\underline{s}_k^H \left(\sigma^2 \mathbf{I} + \sum_{k=1}^K \underline{s}_k \underline{s}_k^H \right)^{-1} \underline{s}_k \right] \text{ scalar}$$

$$= \sum_{k=1}^K \text{tr} \left(\left(\sigma^2 \mathbf{I} + \sum_{k=1}^K \underline{s}_k \underline{s}_k^H \right)^{-1} \underline{s}_k \underline{s}_k^H \right)$$

$$= \text{tr} \left\{ \left(\sigma^2 \mathbf{I} + \sum_{k=1}^K \underline{s}_k \underline{s}_k^H \right)^{-1} \left(\sum_{k=1}^K \underline{s}_k \underline{s}_k^H \right) \right\} \quad L \times K$$

where $\sum_{k=1}^K \underline{s}_k \underline{s}_k^H = \begin{bmatrix} \underline{s}_1 & \underline{s}_2 & \dots & \underline{s}_K \end{bmatrix} \begin{bmatrix} \underline{s}_1^H \\ \underline{s}_2^H \\ \vdots \\ \underline{s}_K^H \end{bmatrix} = \underline{S} \underline{S}^H$

$\underline{S} = \begin{bmatrix} \underline{s}_1 & \underline{s}_2 & \dots & \underline{s}_K \end{bmatrix}$

$$= \text{tr} \left\{ \left(\sigma^2 \mathbf{I} + \underline{U} \underline{\Lambda} \underline{U}^H \right)^{-1} \underline{U} \underline{\Lambda} \underline{U}^H \right\} = \sum_{k=1}^K \frac{\lambda_k}{\sigma^2 + \lambda_k}$$