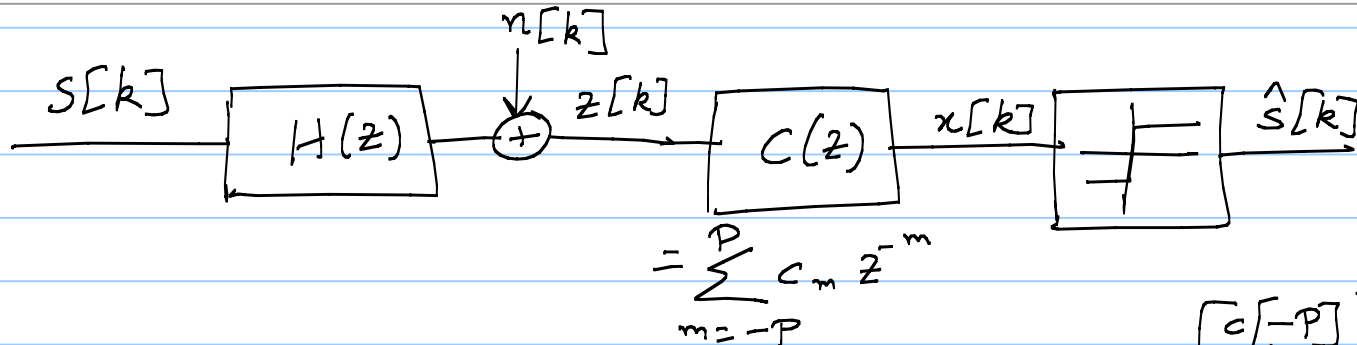


# Lecture 32

Note Title

9/30/2008



Order:  $N = 2P + 1$

$$e[k] = x[k] - s[k]$$

$$MSE = E[|e[k]|^2]$$

$$x[k] = \underline{c}^T \underline{z}_k$$

$$\underline{c} = \begin{bmatrix} c[-P] \\ \vdots \\ c[0] \\ \vdots \\ c[P] \end{bmatrix}$$

$$\underline{z}_k = \begin{bmatrix} z[k+P] \\ \vdots \\ z[k] \\ \vdots \\ z[k-P] \end{bmatrix}$$

$$MSE = E[|e[k]|^2] = E[|s[k]|^2] + \underline{c} \underbrace{E[\underline{z}_k^* \underline{z}_k^T]}_{N \times N} \underline{c} - 2 \operatorname{Re} \left\{ \underline{c}^{*T} \underbrace{E[s[k] \underline{z}_k^*]}_{N \times 1} \right\}$$

$\swarrow$   $E_s$

$$\underline{\alpha} = E \begin{bmatrix} s[k] z^*[k+P] \\ \vdots \\ s[k] z^*[k] \\ \vdots \\ s[k] z^*[k-P] \end{bmatrix} = ? \text{ independent of } k.$$

$$\phi = E[\underline{z}_k^* \underline{z}_k^T] = E \left[ \begin{array}{c} z^*[k+p] \\ \vdots \\ z^*[k] \\ \vdots \\ z^*[k-p] \end{array} \begin{array}{c} [z[k+p] \dots z[k] \dots z[k-p]] \end{array} \right]$$

$$R_z[m] = E[z[k+m] z^*[k]] \xleftrightarrow{\text{DTFT}} S(e^{j\omega})$$

real, non-negative

$$R_z^*[m] = R_z[-m]$$

$$\phi = \begin{bmatrix} R_z[0] & R_z[-1] & \dots & \dots & R_z[-2P] \\ R_z[1] & R_z[0] & \dots & \dots & R_z[-(2P-1)] \\ \vdots & & \ddots & & \vdots \\ R_z[2P] & R_z[2P-1] & \dots & \dots & R_z[0] \end{bmatrix}$$

$$\rightarrow [\phi]_{ij} = R_z[i-j] \quad \text{"Toeplitz"}$$

$$0 \leq i, j \leq 2P$$

$$\rightarrow \phi : \text{Hermitian symmetric } \phi^{*T} = \phi.$$

$$\rightarrow \phi : \text{Positive definite}$$

$$\underline{x}^{*T} \phi \underline{x} \geq 0 \quad \forall \underline{x}$$

$$MSE = E[|s[k]|^2] - 2 \operatorname{Re}\{ \underline{c}^{*T} \underline{\alpha} \} + \underline{c}^{*T} \underline{\phi} \underline{c}$$

Rewrite,

$$MSE = E[|s[k]|^2] - \underline{\alpha}^{*T} \underline{\phi}^{-1} \underline{\alpha} + \underbrace{(\underline{\phi}^{-1} \underline{\alpha} - \underline{c})^{*T} \underline{\phi} (\underline{\phi}^{-1} \underline{\alpha} - \underline{c})}_{\geq 0}$$

$$\underline{c}_{opt} = \underline{\phi}^{-1} \underline{\alpha}$$

$$MMSE = E_s - \underline{\alpha}^{*T} \underline{c}_{opt}$$

$$\rightarrow MSE = f(\underline{c})$$

$$\nabla_{\underline{c}} MSE = 2 \underline{\phi} \underline{c} - 2 \underline{\alpha}$$

$$\underline{c}_{opt} = \underline{\phi}^{-1} \underline{\alpha}$$

Deriving MMSE-LE of order  $N=2P+1$



Solving linear system of equations

$$\begin{array}{ccc} \phi & \underline{c} & = & \underline{r} \\ \downarrow & \downarrow & & \downarrow \\ N \times N & N \times 1 & & N \times 1 \end{array}$$

$$z[k] = s[k] * h[k] + n[k]$$

$$\underline{r} = \begin{bmatrix} E[s[k]z^*[k+P]] \\ \vdots \\ E[s[k]z^*[k-P]] \end{bmatrix}$$

$$[\phi]_{ij} = R_z[i-j]$$
$$0 \leq i, j \leq 2P$$

## MSE Gradient Algorithm (known: $\phi, \underline{\alpha}$ )

iterative:  $\underline{c}_0, \underline{c}_1, \underline{c}_2, \dots, \underline{c}_j, \underline{c}_{j+1}, \dots \rightarrow \underline{c}_{opt}$

$$\underline{c}_{j+1} = \underline{c}_j - \underbrace{\frac{\beta}{2}}_{\text{step size}} \underbrace{\nabla_{\underline{c}_j} E[e[k]^2]}_{\text{gradient}}$$

$$\begin{aligned}\underline{c}_{j+1} &= \underline{c}_j + \beta(\underline{\alpha} - \phi \underline{c}_j) \\ &= (\mathbf{I} - \beta \phi) \underline{c}_j + \beta \underline{\alpha}\end{aligned}$$

$$\underline{q}_{j+1} = \underline{c}_{j+1} - \underbrace{\underline{c}_{opt}}_{\phi^T \underline{\alpha}} = (\mathbf{I} - \beta \phi) \underline{q}_j$$

$$\underline{q}_{j+1} = \underbrace{(\mathbf{I} - \beta\phi)^{j+1}}_{\substack{\mathbf{O} \\ N \times N}} \underline{q}_0$$

$\phi$ : Hermitian symmetric.

"Spectral decomposition"

$$\phi = \sum_{i=1}^n \lambda_i \underbrace{\underline{v}_i \underline{v}_i^{*T}}_{\substack{\downarrow \\ \text{eigenvalue of } \phi}}$$

$$\underline{v}_i \underline{v}_j^{*T} = \begin{cases} 0, & i \neq j \\ 1, & i = j \end{cases}$$

$$\lambda_{\max} = \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n = \lambda_{\min}$$

$$(\mathbf{I} - \beta\phi)^j = \sum_{i=1}^n (1 - \beta\lambda_i)^j \underline{v}_i \underline{v}_i^{*T} \xrightarrow{?} \mathbf{O}$$



MSE Gradient  $\longrightarrow \underline{C}_{opt}$

if  $|1 - \beta \lambda_i| < 1 \quad i=1, 2, \dots, n$

$$-1 < 1 - \beta \lambda_{max} < 1$$

(m)

$$0 < \beta < \frac{2}{\lambda_{max}}$$

Fastest  
 $\underline{C}_j \longrightarrow \underline{C}_{opt} \Rightarrow \beta_{opt} = \frac{2}{\lambda_{min} + \lambda_{max}}$

Fastest  
 $MSE_j \longrightarrow 0 \Rightarrow \boxed{\beta = \frac{1}{\lambda_{max}}}$