

$$\cdot y(n) = \hat{u}(n|u(n)) = \sum_{k=1}^M w_{p,k}^* u(n-k) = \text{filter output}$$

$$\cdot f_M(n) = u(n) - \hat{u}(n|u(n)) = \text{FLP error}$$

$$\cdot P_M = E[|f_M(n)|^2] = \text{MS prediction error}$$

example: $M=2$ $u(n) = \cos(\omega n)$ $\omega = 0.1$ rad/s

$$\min_{\underline{w}_p} P_M \quad \text{where} \quad \underline{w}_p = [w_{p,1}, w_{p,2}, \dots, w_{p,M}]^T$$

Correlation Matrix

$$d(n) = u(n)$$

$$\underline{u}(n-1) = [u(n-1), \dots, u(n-M)]^T$$

$$\underline{R} = E[\underline{u}(n-1) \underline{u}^H(n-1)] \Rightarrow$$

$$\Rightarrow \underline{R} = E \left[\begin{bmatrix} u(n-1) \\ u(n-2) \\ \vdots \\ u(n-M) \end{bmatrix} \begin{bmatrix} u^*(n-1) & u^*(n-2) & \dots & u^*(n-M) \end{bmatrix} \right] =$$

$$= \begin{bmatrix} r(0) & r(1) & \dots & r(M-1) \\ r^*(1) & r(0) & & r(M-2) \\ \vdots & & \ddots & \vdots \\ r^*(M-1) & r^*(M-2) & & r(0) \end{bmatrix}$$

Cross-correlation vector

$$\underline{p} = E[\underline{u}(n-1) \cdot u^*(n)] = E \left[\begin{bmatrix} u(n-1) \\ u(n-2) \\ \vdots \\ u(n-M) \end{bmatrix} u^*(n) \right] = \begin{bmatrix} r^*(1) \\ r^*(2) \\ \vdots \\ r^*(M) \end{bmatrix} \triangleq \underline{r}$$

Hence the optimal solution of FLP filter

$$\boxed{\underline{R} \cdot \underline{w}_F = \underline{r}} \quad (1)$$

(After final version of the filter equations, division by the AR processes, stochastic)

The resulting Error

$$P_M = \sigma_d^2 - \underline{r}^H \underline{w}_F = r(0) - \underline{r}^H \cdot \underline{w}_F = r(0) - \underline{r}^H \cdot \underline{R}^{-1} \cdot \underline{r} \Rightarrow$$

$$\Rightarrow \boxed{P_M = r(0) - \underline{r}^H \cdot \underline{R}^{-1} \cdot \underline{r}} \quad (2)$$

Combine Eqs (1) & (2) to get the augmented WH Eqs for FP:

$$\begin{bmatrix} r(0) & \underline{r}^H \\ \underline{r} & \underline{R} \end{bmatrix} \begin{bmatrix} \underline{1} \\ -\underline{w}_F \end{bmatrix} = \begin{bmatrix} P_M \\ \underline{0} \end{bmatrix}$$

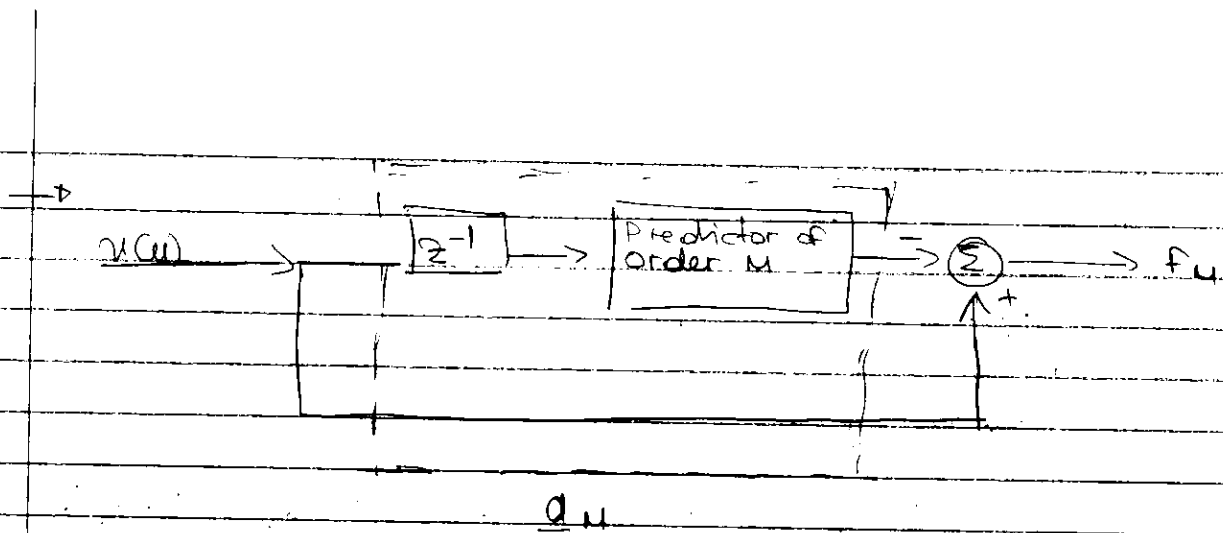
$$\underline{L}_{M+1} \rightarrow \underline{L}_M$$

(Corr. matrix of $\underline{u}_{M+1}(n) = [u(n), \dots, u(n-M)]$)

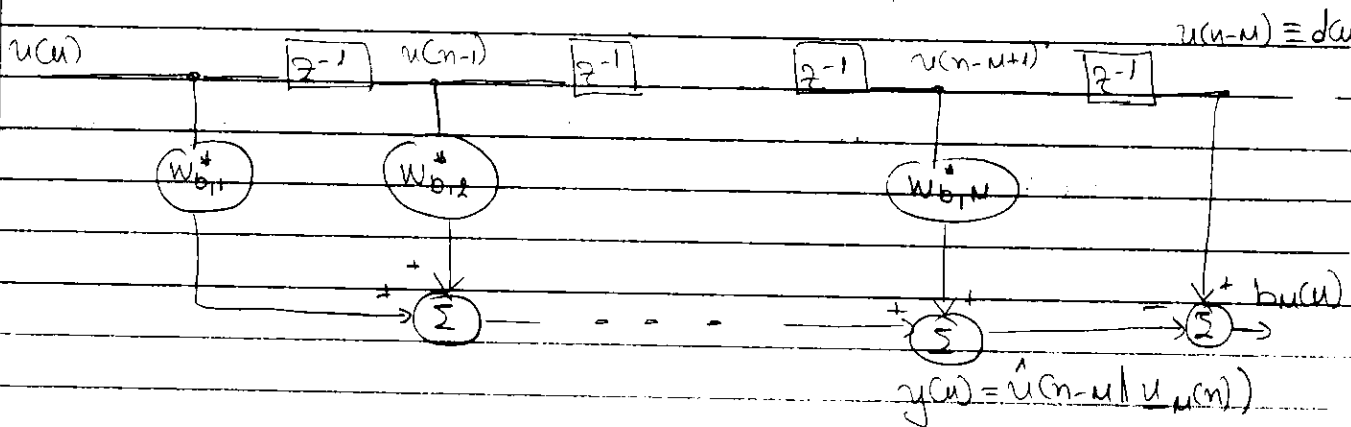
$$\underline{a}_M = \begin{cases} \underline{1} & k=0 \\ -w_{F,k} & k=1, \dots, M \end{cases} \quad \text{is the forward prediction error filter coefficients}$$

$$F_M(n) = u(n) - \underline{w}_F^H \cdot \underline{u}(n-1) = \underline{a}_M^H \cdot \underline{u}_{M+1}(n)$$

LFP & AR processes: The WH Eqs are identical to the Yule-Walker equations, for an M -th order AR process. Hence, for an AR(M) process of known order M , when a FP is optimized in the MS sense, its tap weights take on the same values as the parameters of the process.



BLP (backward linear prediction)



Εφαρμ
$$b_u(n) = u(n-M) - \hat{u}(n-M | u_M(n)) = u(n-M) - \sum_{k=1}^M w_{b,k}^* u(n-k)$$

Τι θέλουμε να ελαχιστοποιήσουμε την ισχύ του σφάλματος εκτίμησης $b_u(n)$

από

$$\min_{\underline{w}_b} P_u = E[|b_u(n)|^2]$$

- $\underline{w}_b = [w_{b,1}^*, \dots, w_{b,M}^*]^T$
- $\underline{u}(n) = [u(n), u(n-1), \dots, u(n-M+1)]^T$
- $d(n) = u(n-M)$
- $\underline{R} = E[\underline{u}(n) \underline{u}^H(n)]$
- $\underline{p} = E[u(n) \cdot d^*(n)] = E \begin{bmatrix} f(n) \\ u(n-1) \\ \vdots \\ u(n-M+1) \end{bmatrix} u^*(n-M) = \begin{bmatrix} f(n) \\ f(n-1) \\ \vdots \\ f(n) \end{bmatrix} = \underline{f}^B$

or more WH Eqs : $\underline{R} \cdot \underline{w}_b = \underline{r}^{B^*}$

BLP ^{error} power : $P_M = t(0) - \underline{r}^{B^T} \cdot \underline{w}_b$

Augmented WH Eqs

$$\begin{bmatrix} \underline{R} & \underline{r}^{B^*} \\ \underline{r}^{B^T} & t(0) \end{bmatrix} \begin{bmatrix} -\underline{w}_b \\ 1 \end{bmatrix} = \begin{bmatrix} \underline{0} \\ P_M \end{bmatrix}$$

$\underbrace{\hspace{10em}}_{\underline{R}_{M+1}}$
 $\underbrace{\hspace{10em}}_{\underline{C}_M}$

where $C_{M+k} = \begin{cases} -w_{b,k+1} & k=0, \dots, M-1 \\ 1 & k=M \end{cases}$

and $b_M(\omega) = \sum_{k=0}^M C_{M+k}^* u(n-k)$

Relation between FLP & BLP

FLP : $\underline{R} \cdot \underline{w}_f = \underline{r}$

BLP : $\underline{R} \cdot \underline{w}_b = \underline{r}^{B^*} \iff E[\underline{u}(\omega) \cdot \underline{u}^H(\omega) \cdot \underline{w}_b] = \underline{r}^{B^*} \iff$

$\iff E[\underline{u}(\omega) \cdot \underline{u}^{BH}(\omega) \cdot \underline{w}_b^B] = \underline{r}^{B^*} \iff (E[\underline{u}(\omega) \cdot \underline{u}^{BH}(\omega) \cdot \underline{w}_b^B])^B = (\underline{r}^{B^*})^B \iff$

$\iff E[\underline{u}^B(\omega) \cdot \underline{u}^{BH}(\omega) \cdot \underline{w}_b^B] = \underline{r}^* \iff E[\underline{u}^B(\omega) \underline{u}^{BH}(\omega)] \cdot \underline{w}_b^B = \underline{r}^* \iff$

$\iff \underline{R}^T \cdot \underline{w}_b^B = \underline{r}^* \iff \underline{R}^{T^*} \underline{w}_b^{B^*} = \underline{r} \iff$

$\iff \underline{R}^H \cdot \underline{w}_b^{B^*} = \underline{r} \iff \underline{R} \cdot \underline{w}_b^{B^*} = \underline{r} \iff \underline{w}_f = \underline{w}_b^{B^*}$

$\underbrace{\hspace{10em}}_{\text{and } \alpha} \quad \underline{R} \cdot \underline{w}_f = \underline{r}$

ΟΡΕΘΗ ΕΙΣΑΓΩΓΩΝ ΛΟΓΩΝ ΕΠΙΔΕΙΞΕΩΝ

FPE
$$P_M = r(0) - \underline{r}^{\#} \cdot \underline{w}_F$$

BPE
$$P_M = r(0) - \underline{r}^{B^T} \underline{w}_B = r(0) - \underline{r}^T \underline{w}_B =$$

$$= r(0) - \underline{r}^{T^*} \underline{w}_B^* = r(0) - \underline{r}^{\#} \cdot \underline{w}_F$$

(επειδή ορίζεται ο υποψήφιος χωρίς να υπάρχει καν
γιατί επιπλέον είναι λογικό το $\underline{r}^T \underline{w}_B^B$ είναι από τα ίδια
αποτελέσματα ο υποψήφιος είναι ο ίδιος)