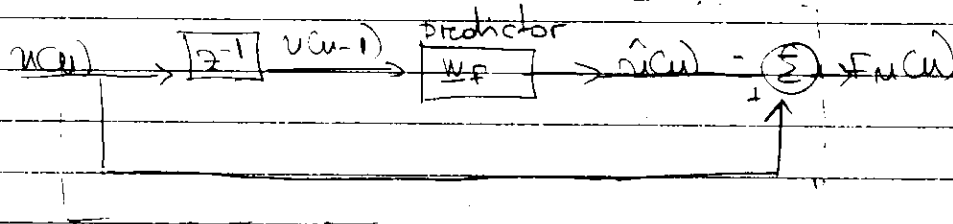


FLP (Forward linear prediction)



↳ predictor error filter

Problem Statement: Given the  $N$  signal values:  $\{x(n-1), x(n-2), \dots, x(n-N)\}$

Find the optimum weights:  $[w_{F1}, \dots, w_{FN}]^T = \underline{w}_F$  such that:

if  $(\hat{x}(n) | \underline{y}_n(n-1)) = \sum_{k=1}^N w_{Fk}^* x(n-k) = \underline{w}_F^H \cdot \underline{y}_n(n-1)$  then

the error  $e(n) = x(n) - \hat{x}(n) | \underline{y}_n(n-1)$  is minimized in the MSE sense, i.e.,

$\min_{\underline{w}_F} E[|e(n)|^2]$   
 $\underline{w}_F \left[ \begin{array}{c} P_u \end{array} \right] \rightarrow$  forward prediction error power

Solution:  $\underline{P}_u \cdot \underline{w}_F = \underline{r}_u = \begin{bmatrix} r^*(1) \\ \vdots \\ r^*(N) \end{bmatrix}$

$P_u = r(0) - \underline{r}_u^H \cdot \underline{w}_F$

Augmented with eqs:  $\begin{bmatrix} r(0) & \underline{r}_u^H \\ \underline{r}_u & \underline{P}_u \end{bmatrix} \begin{bmatrix} 1 \\ -\underline{w}_F \end{bmatrix} = \begin{bmatrix} P_u \\ 0 \end{bmatrix}$

$\left[ \begin{array}{c} \underline{P}_u \\ 0 \end{array} \right] \cdot \left[ \begin{array}{c} \underline{w}_F \\ 1 \end{array} \right] = \begin{bmatrix} P_u \\ 0 \end{bmatrix}$

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

I/O of the FLPEF

### BLP (backward linear predictor)

Given the  $M$  signal values:  $[u(n), \dots, u(n-M+1)] = \underline{u}_M(n)$   
 Find the opt weights  $[w_{b,1}, \dots, w_{b,M}]^T = \underline{w}_b$   
 s.t.  $b_M(n) = u(n-M) - \underline{w}_b^H \underline{u}_M(n)$  is minimized in the MSE sense

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

Solution:  $\underline{R}_M \cdot \underline{w}_b = \underline{r}_M^{BT}$

$$P(w) = r(0) - \underline{r}_M^{BT} \cdot \underline{w}_b$$

Augmented Wt Eqs:  $\underline{R}_{M+1} \underline{c}_M = \begin{bmatrix} r \\ \underline{p}_M \end{bmatrix}$

$$\underline{b}_M(w) = \underline{c}_M^H \underline{u}_{M+1}(w) : \text{I/O BLPEF}$$

We showed that:  $\underline{w}_p = \underline{w}_b^{BT}$

note: As  $M$  increases, the computational complexity increases as  $M^3$  (Gauss-elimination)

### LEVINSON-DURBIN Algorithm

Basic Idea: Solve Wt Eqs for optimum weights for the first 1<sup>st</sup>, 2<sup>nd</sup>, 3<sup>rd</sup>, ...  $M^{\text{th}}$  order recursively  $\Rightarrow$  reduce the computational complexity induced by Gaussian elimination (matrix inversion) method.

Derivation:  $M^{\text{th}}$  order FP:  $\underline{R}_M \cdot \underline{w}_M = \underline{r}_M \Rightarrow$



$$\Rightarrow \underline{L} \underline{w}_u + \underline{w}_{u-1}^{B^*} \cdot \underline{w}_{u,u} = \underline{w}_{u-1} \Rightarrow$$

$$\Rightarrow \underline{L} \cdot \underline{w}_u = \underline{w}_{u-1} - \underline{w}_{u-1}^{B^*} \underline{w}_{u,u} \quad (5)$$

Then, using (5) & (2), to get:

$$\underline{\Gamma}_{u-1}^{B^*} (\underline{w}_{u-1} - \underline{w}_{u-1}^{B^*} \underline{w}_{u,u}) + \underline{\Gamma}(0) \underline{w}_{u,u} = \underline{r}^*(u) \Rightarrow$$

$$\underline{w}_{u,u} = \frac{\underline{r}^*(u) - \underline{\Gamma}_{u-1}^{B^*} \underline{w}_{u-1}}{\underline{\Gamma}(0) - \underline{\Gamma}_{u-1}^{B^*} \underline{w}_{u-1}^{B^*}} \quad (6)$$

Define  $k_u \triangleq -\underline{w}_{u,u}$ . From (5) & (6)

$$\underline{w}_u = \begin{bmatrix} \underline{L} \cdot \underline{w}_u \\ \underline{w}_{u,u} \end{bmatrix} = \begin{bmatrix} \underline{w}_{u-1} \\ 0 \end{bmatrix} + k_u \begin{bmatrix} \underline{w}_{u-1}^* \\ -I \end{bmatrix} \quad (7)$$

$$\underline{\alpha}_u = \begin{bmatrix} 1 \\ -\underline{w}_u \end{bmatrix}, \quad \underline{\alpha}_{u-1} = \begin{bmatrix} 1 \\ -\underline{w}_{u-1} \end{bmatrix}$$

$$(7) \Rightarrow \begin{bmatrix} 1 \\ -\underline{w}_u \end{bmatrix} = \begin{bmatrix} 1 \\ -\underline{w}_{u-1} \\ 0 \end{bmatrix} + k_u \begin{bmatrix} 0 \\ -\underline{w}_{u-1}^* \\ 1 \end{bmatrix} \Rightarrow$$

$$\Rightarrow \underline{\alpha}_u = \begin{bmatrix} \underline{\alpha}_{u-1} \\ 0 \end{bmatrix} + k_u \begin{bmatrix} 0 \\ \underline{\alpha}_{u-1}^{B^*} \\ 1 \end{bmatrix} \quad (8)$$

In scalar form:  $\alpha_{u,k} = \alpha_{u-1,k} + k_u \alpha_{u-1,u-k}^*$ ,  $k=0,1,\dots,u-1$

$$\alpha_{u,0} = 1$$

$$\alpha_{u-1,0} = 1$$

(9)

Define:  $\Delta_{u-1} \stackrel{\Delta}{=} r^{\Delta}(u) - r_{u-1}^{BT} \cdot \underline{w}_{u-1} = r_{u-1}^{BT} \cdot \underline{a}_{u-1} =$   
 $= \sum_{l=0}^{u-1} \alpha_{u-1,l} r(l-u) \stackrel{\text{HW4}}{=} E \left[ b_{u-1}^{(u-1)} \cdot f_{u-1}^*(\omega) \right]$

↑ also da  $\delta_{ij} = \delta_{ji}$

Also,  $r(u) - r_{u-1}^{BT} \cdot \underline{w}_{u-1} = P_{u-1}$

Hence, from (6):

$$k_u \stackrel{\Delta}{=} -w_{u,u} = -\frac{\Delta_{u-1}}{P_{u-1}} = -\frac{\sum_{l=0}^{u-1} \alpha_{u-1,l} r(l-u)}{P_{u-1}} =$$

$$= -\frac{E \left[ b_{u-1}^{(u-1)} \cdot f_{u-1}^*(\omega) \right]}{E \left[ |f_{u-1}^*(\omega)|^2 \right]} \quad (10)$$

$$P_u = r(u) - r_u^H \cdot \underline{w}_u \stackrel{(7)}{=} r(u) - r_u^H \left( \begin{bmatrix} \underline{w}_{u-1} \\ 0 \end{bmatrix} + k_u \begin{bmatrix} \underline{w}_{u-1}^{BT} \\ -1 \end{bmatrix} \right) =$$

$$= r(u) - r_{u-1}^H \cdot \underline{w}_{u-1} - k_u \left( r_{u-1}^H \cdot \underline{w}_{u-1}^{BT} - r(u) \cdot 1 \right) =$$

$$= P_{u-1} + k_u \underbrace{\left( r^{\Delta}(u) - r_{u-1}^{BT} \cdot \underline{w}_{u-1} \right)}_{\Delta_{u-1}} \Rightarrow$$

$$\Rightarrow P_u = P_{u-1} + k_u \Delta_{u-1} = P_{u-1} - |k_u|^2 P_{u-1} \Rightarrow$$

$$P_u = (1 - |k_u|^2) P_{u-1} \stackrel{(10)}{=} P_0 \prod_{l=1}^u (1 - |k_l|^2)$$

Levinson-Durbin

$$P_0 = r(0)$$

$$m = 1, 2, 3, \dots, M$$

$$k_m = \alpha_{m,m} = \frac{-\Delta_{m-1}}{P_{m-1}} = \frac{-\sum_{l=0}^{m-1} \alpha_{m-1,l} r(l-m)}{P_{m-1}}$$

(5)

$$(\Delta_0 = r^{\Delta}(1))$$

$$k = 0, 1, \dots, w-1$$

$$\alpha_{m,k} = \alpha_{w-1,k} + k_m \alpha_{w-1, w-k}$$

$$\text{end } P_m = \frac{P_{m-1}}{m-1} (1 - |k_m|^2) \quad (\alpha_{k,m} = 0 \quad m > 0)$$

end

AC Sequence

$r(0) \quad r(1) \quad r(2) \quad \dots \quad r(N)$

2<sup>nd</sup> order statistics

LD /  $\uparrow$  ILD

ILD  $\uparrow$  LD

$P_m, \alpha_{m,1}, \alpha_{m,2}, \dots, \alpha_{m,m}$

$(k_1, \dots, k_m)$   
 $r(0), \alpha_{1,1}, \dots, \alpha_{m,m}$

Reflection Coefficients

нормированная погрешность  $\sim$   $O(N^{-2})$  Levinson-Darboux error

note 3: 1 LD (inverse Levinson-Darboux)

LD (Levinson-Darboux)