

4.3 Levinson's Recursion (Levinson-Durbin) 4.3.1 Haykin 6.3

- Solution of $R^{-1} \underline{w}_f = \underline{r}$ requires $O(M^3)$ operations.
The structure of the equations lends itself to recursive (or iterative) solutions — $O(M^2)$.
- First look at order recursion, with augmented forward W-TI

$$\begin{bmatrix} r(0) & r(1) & \dots & r(M-1) & r(M) \\ r(1)^* & r(0) & \dots & r(M-2) & r(M-1) \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ r^{*(M-1)} & r^{*(M-2)} & \dots & r(1) & r(0) \\ r^{*(M)} & r^{*(M-1)} & \dots & r^{*(1)} & r(0) \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -w_{f1} \\ \vdots \\ -w_{f,M-2} \\ -w_{f,M-1} \\ -w_{fM} \end{bmatrix} = \begin{bmatrix} P_M \\ 0 \\ \vdots \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Each problem has the one of next lower order embedded in it. Same is true for backward equations.

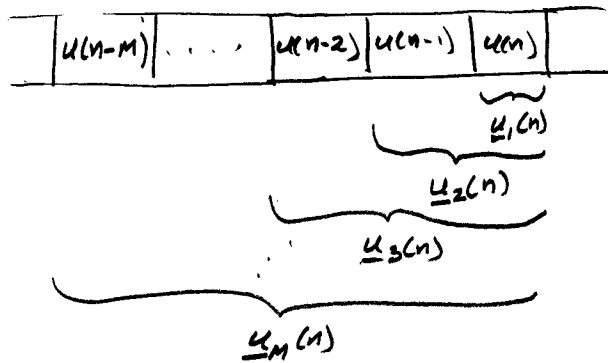
- Next, consider relation between forward and backward partitionings of $R \ ((M+1) \times (M+1))$

$$R_{M+1} = \begin{bmatrix} r(0) & r(1) & \dots & r(M-2) & r(M-1) & r(M) \\ r^{*(1)} & r(0) & \dots & r(M-2) & r(M-1) & r(M-2) \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ r^{*(M-1)} & r^{*(M-2)} & \dots & r(0) & r(1) & r^{*(1)} \\ r^{*(M)} & r^{*(M-1)} & \dots & r^{*(1)} & r(0) & r(0) \end{bmatrix} = \begin{bmatrix} r(0) & \underline{r}_M^T \\ \underline{r}_M & R_M \end{bmatrix} = \begin{bmatrix} R_M & \underline{\tilde{r}}_M^* \\ \underline{\tilde{r}}_M^T & r(0) \end{bmatrix}$$

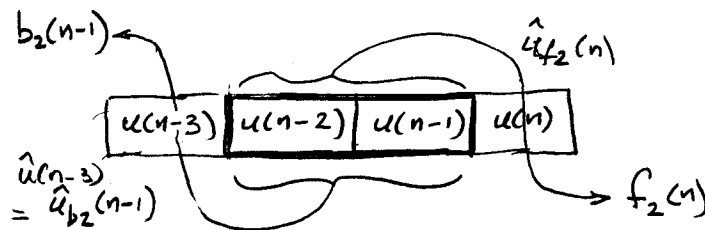
- Can we construct order m solution $\underline{a}_m$ by some calculations on $\underline{a}_{m-1}$?

Recall definitions

4.3.2



and now consider simultaneous forward and backward estimation (order 2 for simplicity)



both estimates based on $\underline{u}_2(n-1)$

The two prediction errors $b_2(n-1)$ and $f_2(n)$ are still correlated (unless $u(n)$ is an AR process of order 2 or less).

Therefore we can estimate one in terms of the other.

First, estimate $f_2(n)$ with $b_2(n-1)$.

$$\hat{f}_2(n) = \gamma_2^* b_2(n-1)$$

$$\gamma_2 = \frac{E[f_2^*(n) b_2(n-1)]}{E[|b_2(n-1)|^2]}$$

is the PARCOR coefficient

The error in this estimate is $f_3(n)$, since $b_2(n-1)$ is what's left in $u(n-3)$ after projection onto $\underline{u}_2(n-1)$. So

$$f_3(n) = f_2(n) - \gamma_2^* b_2(n-1)$$

$$\underline{a}_{f_3}^+ \underline{u}_+(n) = [\underline{a}_{f_2}^+ \mid 0] \underline{u}_+(n) - \gamma_2^* [0 \mid \underline{a}_{b_2}^+] \underline{u}_+(n)$$

$$\underline{a}_{f_3} = \begin{bmatrix} \underline{a}_{f_2} \\ 0 \end{bmatrix} - \gamma_2 \begin{bmatrix} 0 \\ \tilde{\underline{a}}_{f_2}^* \end{bmatrix}$$

a recursion - we just need to find γ_2

since $\underline{a}_{b_2} = \tilde{\underline{a}}_{f_2}^*$

We could also have estimated $b_2(n-1)$ with $f_2(n)$:

$$\hat{b}_2(n-1) = h_2^* f_2(n) \quad h_2 = \frac{E[b_2^*(n-1) f_2(n)]}{E[|f_2(n)|^2]}$$

But $E[|f_2(n)|^2] = E[|b_2(n-1)|^2]$ (why?) so $h_2 = \gamma_2^*$ and

$$\hat{b}_2(n-1) = \gamma_2 f_2(n)$$

The error $b_2(n-1) - \hat{b}_2(n-1)$ is $b_3(n)$, since $\hat{b}_2(n-1) + \hat{u}_{b_2}^{(n-1)}$ is effectively an estimate of $u(n-3)$ based on $u(n-2), u(n-1), u(n)$.

$$b_3(n) = b_2(n-1) - \gamma_2 f_2(n)$$

$$\underline{a}_{b_3}^+ \underline{u}_4(n) = [0 \mid \underline{a}_{b_2}^+] \underline{u}_4(n) - \gamma_2 [\underline{a}_{f_2}^+ \mid 0] \underline{u}_4(n)$$

$$\underline{a}_{b_3} = \begin{bmatrix} 0 \\ \underline{a}_{b_2} \end{bmatrix} - \gamma_2^* \begin{bmatrix} \underline{a}_{f_2} \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ \underline{a}_{b_2} \end{bmatrix} - \gamma_2^* \begin{bmatrix} \tilde{a}_{b_2}^* \\ 0 \end{bmatrix}$$

This is a recursion similar to the one for forward filters. In fact, they are the same: apply $\sim^*$ to this equation, and the fact that $\tilde{a}_{b_2}^* = a_{f_2}$ and get

$$\underline{a}_{f_3} = \begin{bmatrix} \underline{a}_{f_2} \\ 0 \end{bmatrix} - \gamma_2 \begin{bmatrix} 0 \\ \tilde{a}_{f_2}^* \end{bmatrix}$$

the forward filter recursion

- Now that we have motivation, we'll follow formal derivation in Haykin. Use $\underline{a}_m = \underline{a}_{fm}$.

$$\underline{a}_m = \begin{bmatrix} \underline{a}_{m-1} \\ 0 \end{bmatrix} + K_m \begin{bmatrix} 0 \\ \tilde{r}^* \\ \underline{a}_{m-1} \end{bmatrix} \quad K_m = -\delta_{m-1}$$

Step m: premult by $R_{m+1} = E[\underline{u}_{m+1}(n) \underline{u}_{m+1}(n)^T]$

$$R_{m+1} \underline{a}_m = \begin{bmatrix} R_m & \tilde{r}^+ \\ \tilde{r}^T & r(0) \end{bmatrix} \begin{bmatrix} \underline{a}_{m-1} \\ 0 \end{bmatrix} + K_m \begin{bmatrix} r(0) & \tilde{r}^+ \\ \tilde{r} & R_m \end{bmatrix} \begin{bmatrix} 0 \\ \tilde{r}^* \\ \underline{a}_{m-1} \end{bmatrix} = \begin{bmatrix} P_m \\ \underline{0}_m \end{bmatrix}$$

$$\begin{bmatrix} R_m \underline{a}_{m-1} \\ \tilde{r}^T \underline{a}_{m-1} \end{bmatrix} + K_m \begin{bmatrix} \tilde{r}^+ \tilde{r}^* \\ R_m \tilde{r}^* \end{bmatrix} = \begin{bmatrix} P_m \\ \underline{0}_m \end{bmatrix}$$

$$\begin{bmatrix} P_{m-1} \\ \underline{0}_{m-1} \\ \Delta_{m-1} \end{bmatrix} + K_m \begin{bmatrix} \Delta_{m-1}^* \\ \underline{0}_{m-1} \\ P_{m-1} \end{bmatrix} = \begin{bmatrix} P_m \\ \underline{0}_m \end{bmatrix}$$

hence

$$\Delta_{m-1} + K_m P_{m-1} = 0 \Rightarrow \boxed{K_m = -\frac{\Delta_{m-1}}{P_{m-1}}}$$

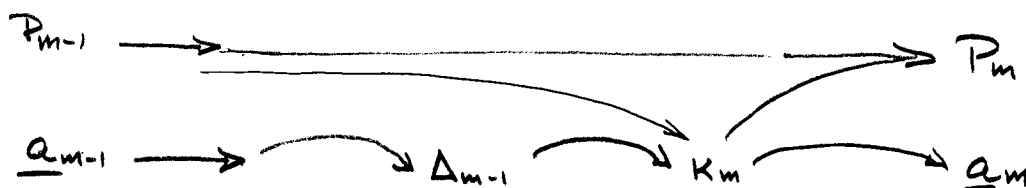
$$P_m = P_{m-1} + K_m \Delta_{m-1}^*$$

$$= P_{m-1} - \frac{|\Delta_{m-1}|^2}{P_{m-1}}$$

$$= P_{m-1} - |K_m|^2 P_{m-1}$$

$$\boxed{P_m = (1 - |K_m|^2) P_{m-1}}$$

— that's all there is to it.



• Some observations:

→ K_m , the reflection coeff (related to PARCOR coeff)
decreases to 0 as $m \rightarrow \infty$ (consider est of $f_m(n)$
from $b_m(n-1)$)

→ Zero order has

$$\underline{a}_0 = [1] \quad P_0 = r(0) \quad (\text{trivial case})$$

$$\text{hence } P_M = P_0 \prod_{m=1}^M (1 - |K_m|^2)$$

→ from the recursion

$$\underline{a}_m = \begin{bmatrix} \underline{a}_{m-1} \\ 0 \end{bmatrix} + K_m \begin{bmatrix} 0 \\ \underline{a}_{m-1}^* \end{bmatrix} = \begin{bmatrix} 1 \\ a_{m,1} \\ a_{m,m-1} \\ K_m \end{bmatrix} = \begin{bmatrix} a_{m,0} \\ a_{m,1} \\ a_{m,m-1} \\ a_{m,m} \end{bmatrix} \quad \begin{array}{l} \text{note} \\ \text{last} \\ \text{component} \\ \text{is } K_m \end{array}$$

→ from the PARCOR coefficient

$$\gamma_m = \frac{E[f_m^*(n) b_m(n-1)]}{E[|b_m(n-1)|^2]} = -K_{m+1}$$

$$E[f_m^*(n) b_m(n-1)] = -K_{m+1} P_m = \Delta_m$$

hence the intermediate quantities

$$\Delta_{m-1} = E[f_{m-1}^*(n) b_{m-1}(n-1)] \quad \text{the covariance}$$

$$K_m = - \frac{E[f_{m-1}^*(n) b_{m-1}(n-1)]}{E[|b_{m-1}(n-1)|^2]}$$

→ $O(m)$ operations in step m ,

so $O(M^2/2)$ operations cumulative

→ Levinson-Durbin and conventional solution of Yule Walker give same result $\underline{a}_m, P_m$ but L-D generates $\underline{a}_m, P_m$ for all intermediate orders, too.

→ We want $\{r(0), r(1), \dots, r(m)\} \longrightarrow \{\underline{a}_0, \underline{a}_1, \dots, \underline{a}_m\}$
 $\{k_1, \dots, k_m\} \{P_0, P_1, \dots, P_m\}$

but we can also do this:

$$\{r(0), k_1, \dots, k_m\} \longrightarrow \{\underline{a}_0, \underline{a}_1, \dots, \underline{a}_m\}$$

$$\{P_0, \dots, P_m\}$$

since

$$\underline{a}_m = \begin{bmatrix} \underline{a}_{m-1} \\ 0 \end{bmatrix} + k_m \begin{bmatrix} 0 \\ \tilde{\underline{a}}_{m-1}^* \end{bmatrix}, \quad P_m = (1 - |k_m|^2) P_{m-1}$$

• Inverse L-D

$$\underline{a}_m \longrightarrow \{\underline{a}_0, \dots, \underline{a}_m\}, \{k_1, \dots, k_m\}$$

From the highest order PEF, we can obtain those of lower order, as well as $\{k_1, \dots, k_m\}$ (which we obtain as last components of PEFs)

$$\underline{a}_m = \begin{bmatrix} \underline{a}_{m-1} \\ 0 \end{bmatrix} + k_m \begin{bmatrix} 0 \\ \tilde{\underline{a}}_{m-1}^* \end{bmatrix} \xrightarrow{\text{row } k, k=1 \dots m-1} a_{m,k} = a_{m-1,k} + k_m a_{m-1,m-k}^*$$

$$\tilde{\underline{a}}_m^* = \begin{bmatrix} 0 \\ \tilde{\underline{a}}_{m-1}^* \end{bmatrix} + k_m^* \begin{bmatrix} \underline{a}_{m-1} \\ 0 \end{bmatrix} \xrightarrow{\text{row } m-k} a_{m,m-k}^* = a_{m-1,m-k}^* + k_m^* a_{m-1,k}$$

$$\begin{bmatrix} a_{m,k} \\ a_{m,m-k}^* \end{bmatrix} = \begin{bmatrix} 1 & k_m \\ k_m^* & 1 \end{bmatrix} \begin{bmatrix} a_{m-1,k} \\ a_{m-1,m-k}^* \end{bmatrix}$$

invert

$$\frac{1}{1 - |k_m|^2} \begin{bmatrix} 1 & -k_m \\ -k_m^* & 1 \end{bmatrix}$$

solve for lower order coefficients from higher

$$a_{m-1,k} = \frac{a_{m,k} - k_m a_{m,m-k}^*}{1 - |k_m|^2}$$

$$= \frac{a_{m,k} - a_{m,m} a_{m,m-k}^*}{1 - |k_m|^2}$$

- Recovery of autocorrelation function: in Haykin 6.4

From $\{P_0, K_1, \dots, K_m\}$ we can recover the $\{r(0), r(1), \dots, r(m)\}$.

Just consider the Yule Walker equations of order m

$$\begin{bmatrix} r(0) & r(1) & \dots & r(m-1) & r(m) \\ r^*(1) & r(0) & & & r(m-1) \\ & & \ddots & & \\ & & & r(1)^* & r(0) \\ r^*(m) & r^*(m-1) & \dots & r(1)^* & r(0) \end{bmatrix} \begin{bmatrix} 1 \\ a_{m,1} \\ \vdots \\ a_{m,m-1} \\ K_m \end{bmatrix} = \begin{bmatrix} P_m \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Solution 1: Use row 1 to express $r(m)$ in terms of lower:

$$r(0) + a_{m,1} r(1) + \dots + a_{m,m-1} r(m-1) + K_m r(m) = P_m = (1 - |K_m|^2) P_{m-1}$$

$$r(m) = \frac{(1 - |K_m|^2) P_{m-1}}{K_m} - \frac{1}{K_m} \sum_{k=0}^{m-1} a_{m,k} r(k)$$

So given $P_0 (= r(0))$, $\underline{a}_0 (= [1])$, we form

$\underline{a}_1, r(1), P_1$; $\underline{a}_2, r(2), P_2$; etc.

Solution 2: Use row m to express $r(m)$ in terms of lower

$$r^*(m) = -a_{m,1} r^*(m-1) - \dots - a_{m,m-1} r^*(1) - a_{m,m} r(0)$$

$$r(m) = - \sum_{k=1}^m a_{m,k}^* r(m-k) \quad \left(= - \sum_{k=1}^{m-1} a_{m,k}^* r(m-k) - K_m^* P_0 \right)$$

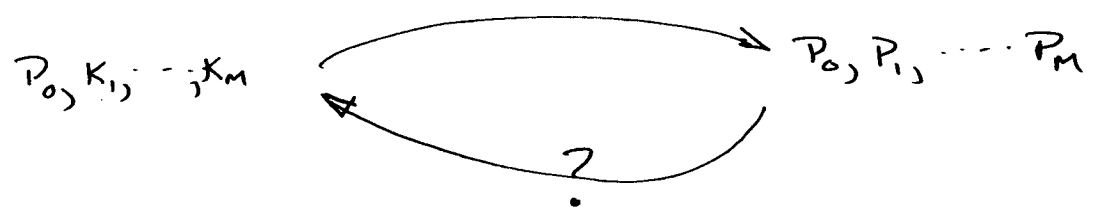
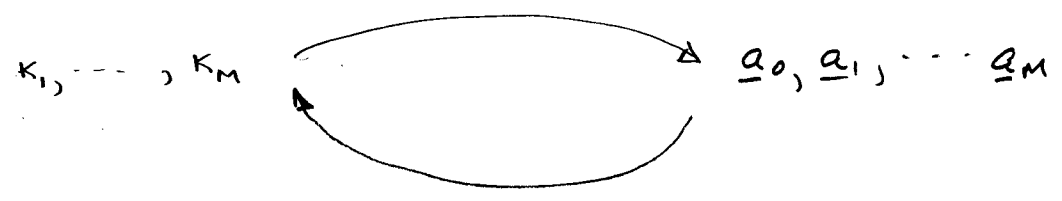
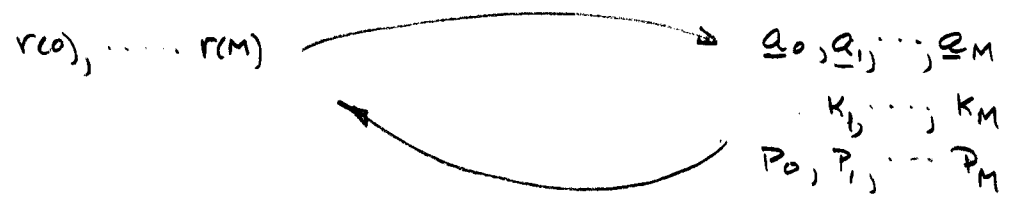
So given $P_0, \underline{a}_0$, do this:

$\underline{a}_1, r(1)$; $\underline{a}_2, r(2)$; $\dots$

Both require the $\{\underline{a}_m\}$, which are generated from the $\{K_m\}$ through the basic recursion.

• Summary of relations:

A process is characterized by the shape and size of its power spectrum ($\underline{a}_M$ and $r(\omega)$, respectively). Only $M+1$ parameters for AR, order M .



→ Can you think of another common set of parameters?