

**S3:** Using the result of step **S2**, update coefficients for the  $i$ th pole model as

$$\begin{aligned}\alpha_i^i &= k_i \\ \alpha_j^i &= \alpha_j^{i-1} - k_i \alpha_{i-j}^{i-1}, \quad 1 \leq j \leq i-1\end{aligned}$$

where we have set the highest-order coefficient to  $k_i$  and where  $k_i$  has also been used as a weighting factor on the coefficients of order  $i-1$  to obtain the coefficients of order  $i$ . We refer to the constants  $k_i$  as the *partial correlation coefficients* for which the motivation will shortly become clear. These coefficients are also called the *PARCOR* coefficients.

**S4:** Update the minimum mean-squared prediction error for the  $i$ th pole model as

$$E^i = (1 - k_i^2)E^{i-1}.$$

$E^i$ , being a mean-squared prediction error, provides a way of monitoring the accuracy of prediction when the predictor order is unknown.

**S5:** Repeat steps **S2** to **S4** for  $i = 1, 2, \dots, p$ . At the  $p$ th step we have the desired  $p$ th order predictor coefficients

$$\alpha_j^* = \alpha_j^p, \quad 1 \leq j \leq p$$

where “\*” denotes the (optimal) coefficients that minimize the mean-squared  $p$ th-order prediction error.

The derivation of the Levinson recursion can be found in [15],[16]. From step **S2**, we can show on each iteration that the predictor coefficients  $\alpha_k$  can be written as solely functions of the autocorrelation coefficients (Exercise 5.11). Finally, the desired transfer function is given by

$$H(z) = \frac{A}{1 - \sum_{k=1}^p \alpha_k^* z^{-k}}$$

where the gain  $A$  has yet to be determined. Calculation of the gain depends on certain properties of the estimated prediction coefficients and will be described later in this section.

**Properties** — Properties of the autocorrelation method motivated by the Levinson recursion are given below.

**P1:** The magnitude of each of the partial correlation coefficients is *less than unity*, i.e.,  $|k_i| < 1$ . To show this, we first recall that the mean-squared prediction error is always greater than zero because we cannot attain perfect prediction even when the sequence follows an all-pole model. Therefore, from the Levinson recursion, we have

$$E^i = (1 - k_i^2)E^{i-1} > 0$$

from which

$$(1 - k_i^2) = \frac{E^i}{E^{i-1}} > 0$$

and so it follows that

$$k_i^2 < 1 \quad \text{or} \quad |k_i| < 1.$$