

Note that if

$$A(L) = 1 + a_1 L + a_2 L^2, \text{ then: } \begin{aligned} a_1 + a_2 &> -1 \\ a_1 - a_2 &< 1 \\ a_2 &< 1 \end{aligned}$$

## IN GENERAL

CONSIDER AN AR(p) PROCESS

$$A(L)X_t = \varepsilon_t$$

$$A(L) = 1 + a_1 L + a_2 L^2 + \dots + a_p L^p$$

THE AR(p) IS STATIONARY IF THE ROOTS OF  $A(L) = 0$  ALL LIE OUTSIDE THE UNIT CIRCLE.

We can write:

$$A(L) = \prod_{i=1}^p (1 - \mu_i L)$$

↑ roots of the reciprocal

$$\text{if } A(L) = 1 + a_1 L + a_2 L^2$$

$\mu$ 's are roots of

$$A(z) = z^2 + a_1 z + a_2$$

## Variance and Autocovariance Function of AR(2)

let's

$$\tilde{X}_t = \bar{\mu} + \sum_{j=1}^p a_j \tilde{X}_{t-j} + \varepsilon_t$$

ASSUME

1.  $\tilde{X}_t$  is STATIONARY WITH  $\sum_{j=1}^p a_j \neq 1$
2.  $E(\tilde{X}_{t-j} \varepsilon_t) = 0, \forall j \geq 0$

BY STATIONARITY:

$$E(\tilde{X}_t) = \mu \quad \forall t \Rightarrow$$

$$\mu = \bar{\mu} + \sum_{j=1}^p a_j \bar{\mu}$$

$$E(\tilde{X}_t) = \mu = \frac{\bar{\mu}}{1 - \sum_{j=1}^p a_j}$$

the demeaned process:  $X_t$

$$X_t = \sum_{j=1}^p a_j X_{t-j} + \varepsilon_t$$

$$X_{t+k} = \sum_{j=1}^p a_j X_{t+k-j} + \varepsilon_{t+k}$$

/  $X_t$

$$E(X_{t+k} X_t) = \sum_{j=1}^p a_j E(X_{t+k-j} X_t) + E(\varepsilon_{t+k} X_t)$$



$$\gamma(k) = \sum_{j=1}^p a_j \gamma(k-j) + E(\varepsilon_{t+k} X_t)$$

$$\text{where } E(\varepsilon_{t+k} X_t) = \sigma_\varepsilon^2, \text{ if } k=0 \\ = 0, \text{ if } k \geq 1$$

$$\rho(k) = \sum_{j=1}^p a_j \rho(k-j), \quad k \geq 1$$

WE CALL THESE FORMULAS: YULE-WALKER EQUATIONS.  
IF WE KNOW  $\rho(1), \dots, \rho(p)$  WE CAN COMPUTE  $\rho(k)$  FOR  $k \geq p+1$ .

**THE AUTOCORRELATIONS OF AR(p) ARE GRADUALLY  
EXPONENTIALLY DELAYING WITH k.**

YULE-WALKER:

$$A(L) \rho(k) = 0, \quad \forall k \geq 1$$

WHERE  $L^j \rho(k) = \rho(k-j)$ .

TO OBTAIN  $\rho(1), \dots, \rho(p)$  WE SOLVE:

$$\underline{\rho(1)} = a_1 + a_2 \underline{\rho(1)} + \dots + a_p \rho(p-1)$$

$$\underline{\rho(2)} = a_1 \underline{\rho(1)} + a_2 + \dots + a_p \rho(p-2)$$

$$\vdots$$

$$\rho(p) = a_1 \rho(p-1) + a_2 \rho(p-2) + \dots + a_p$$

WITH:  $\rho(-j) = \rho(j)$

- AUTOCOV. OF ORDER  $k > p$  ARE OBTAINED FROM:

$$\rho(k) = \sum_{j=1}^p a_j \rho(k-j), \quad k \geq p+1$$

TO COMPUTE  $\gamma(0) = \text{var}(x_t)$  WE SOLVE:

$$\gamma(0) = \sum_{j=1}^p a_j \gamma(-j) + E(\varepsilon_t X_t)$$

$$\gamma(0) = \sum_{j=1}^p a_j \gamma(j) + \sigma_\varepsilon^2$$

> NCE  $\gamma(j) = \rho(j) \gamma(0) \Rightarrow$

$$\gamma(0) \left[ 1 - \sum_{j=1}^p a_j \rho(j) \right] = \sigma_\varepsilon^2$$

$$\gamma(0) = \frac{\sigma_\varepsilon^2}{1 - \sum_{j=1}^p a_j \rho(j)}$$

1). AR(1):

$$X_t = a_1 X_{t-1} + \varepsilon_t$$

$$\rho(1) = a_1$$

$$\rho(k) = a_1 \rho(k-1), \quad k \geq 1$$

$$\Rightarrow \rho(2) = a_1 \rho(1) = a_1^2$$

$$\rho(k) = a_1^k, \quad k \geq 1$$

$$\gamma(0) = \text{var}(X_t) = \frac{\sigma_\varepsilon^2}{1 - a_1^2}$$

COMPARE WITH...

2) AR(2)

$$X_t = a_1 X_{t-1} + a_2 X_{t-2} + \varepsilon_t$$

$$\rho(1) = a_1 + a_2 \rho(2)$$

$$\rho(2) = a_1 \rho(1) + a_2 \Rightarrow \rho(1) = \frac{a_1}{1 - a_2}$$

$$\rho(2) = \frac{a_1^2}{1 - a_2} + a_2 = \frac{a_1^2 + a_2(1 - a_2)}{1 - a_2}$$

$$\gamma(0) = \frac{(1 - a_2) \sigma_\varepsilon^2}{(1 + a_2)(1 - a_1 - a_2)(1 + a_1 - a_2)}$$

FOR  $k > p = 2$ , i.e. FOR  $k \geq 3$ 

$$\rho(k) = a_1 \rho(k-1) + a_2 \rho(k-2)$$

HOMOGENOUS, 2<sup>nd</sup> ORDER DIFF EQ:  $\rho(k) = A_1 \mu_1^k + A_2 \mu_2^k$ ,  
 $\mu_1, \mu_2$  are roots of  $A(z) = z^2 + a_1 z + a_2$ .

## PARTIAL AUTOCORRELATIONS

LINEAR FORECASTS:

CONSIDER A STATIONARY PROCESS  $X_t$ , FOR WHICH THE AUTOCORRELATION MATRICES  $R(m)$  ARE ALL NONSINGULAR FOR INTEGER  $m$ .

WE ARE INTERESTED IN THE BEST LINEAR PREDICTION OF  $X_t$  BASED ON THE  $k$  PREVIOUS VALUES:  $X_{t-1}, X_{t-2}, \dots, X_{t-k}$

ASSUME  $X_t$  IS DEMEANED.

THE PREDICTION:

$$E(X_t | X_{t-1}, \dots, X_{t-k}) = a_1(k) X_{t-1} + \dots + a_k(k) X_{t-k}$$

WHERE THE VECTOR  $a(k) = (a_1(k), \dots, a_k(k))'$  OF THE REGRESSION COEFF IS:

$$a(k) = R(k)^{-1} \begin{bmatrix} \rho(1) \\ \vdots \\ \rho(k) \end{bmatrix}$$

FOR AR(p):

$$a(k) = \sum_{j=1}^p a_j \rho(k-j), \quad k = \dots$$

GIVEN THAT:

$$\rho(0) = 1 \quad \text{AND} \quad \rho(-k) = \rho(k)$$

AND  $a_j$  DENOTE HERE THE AUTOREGRESSIVE COEFF.

$$\begin{bmatrix} 1 & \rho(1) & \rho(2) & \dots & \rho(p-1) \\ \rho(1) & 1 & \rho(1) & \dots & \rho(p-2) \\ \vdots & & & & \\ \rho(p-1) & \rho(p-2) & \rho(p-3) & \dots & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_p \end{bmatrix} = \begin{bmatrix} \rho(1) \\ \rho(2) \\ \vdots \\ \rho(p) \end{bmatrix}$$

$$R \cdot a = \rho$$

$$a = R^{-1} \rho$$

INTRODUCING DIMENSIONS, WE HAVE SEQUENCE OF EQUATIONS:

$$R_k a_k = \rho_k \quad k=1, 2, 3, \dots$$

WHERE

$$a_k = [a_{k1}, a_{k2}, \dots, a_{kk}]'$$

$$a_k = R_k^{-1} \rho_k.$$

THE COEF  $a_{kk}$  IS EQUAL TO THE COEFF OF CORRELATION

$$\text{BTW: } X_t - E(X_t | X_{t-1}, \dots, X_{t-k+1})$$

$$\text{AND } X_{t-k} - E(X_{t-k} | X_{t-1}, \dots, X_{t-k+1})$$

IT MEASURES THE LINEAR LINK BTW  $X_t$  AND  $X_{t-k}$  ONCE THE INFLUENCE OF INTERVENING VARIABLES  $X_{t-1}, \dots, X_{t-k+1}$  HAS BEEN REMOVED

FOR AN AR(p) WE HAVE

$$a_{kk} = 0 \quad \forall k \geq p+1$$



SOME  $a_{kk}$ :

$$a_{11} = \rho_1$$

$$a_{22} = \frac{\begin{vmatrix} 1 & \rho_1 \\ \rho_1 & \rho_2 \end{vmatrix}}{\begin{vmatrix} 1 & \rho_1 \\ \rho_1 & 1 \end{vmatrix}} = \frac{\rho_2 - \rho_1^2}{1 - \rho_1^2}$$

$$a_{33} = \frac{\begin{vmatrix} 1 & \rho_1 & \rho_1 \\ \rho_1 & 1 & \rho_2 \\ \rho_2 & \rho_1 & \rho_3 \end{vmatrix}}{\begin{vmatrix} 1 & \rho_1 & \rho_2 \\ \rho_1 & 1 & \rho_1 \\ \rho_2 & \rho_1 & 1 \end{vmatrix}} =$$

DURBIN - LEVINSON

$$a_{k+1, k+1} = \frac{\rho(k+1) - \sum_{j=1}^k a_{kj} \rho(k+1-j)}{1 - \sum_{j=1}^k a_{kj} \rho(j)}$$

$$a_{k+1, j} = a_{kj} - a_{k+1, k+1} a_{k, k-j+1}, \quad j = 1, 2, \dots, k$$