

## ARMA (1,1)

①

- ARMA means autoregressive moving average..
- The time series  $\{X_t\}$  is an ARMA(1,1) process if it is stationary and satisfies

$$X_t - \phi X_{t-1} = Z_t + \theta Z_{t-1}, \quad \text{for all } t,$$

where  $\{Z_t\} \sim WN(0, \sigma^2)$ .

- In terms of back shift or Lag operator  $B$ , we can express  $X_t$  above as,

$$\phi(B) X_t = \theta(B) Z_t, \quad \text{where.}$$

$$\phi(B) = 1 - \phi B \quad \text{and} \quad \theta(B) = 1 + \theta B.$$

## Stationarity, Causality, invertibility ——— .

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- First question is to find if a stationary solution of  $X_t$  exists or not.

- If  $|\phi| < 1$ , we can show that the MA( $\infty$ ) process,

$$X_t = Z_t + (\phi + \theta) \sum_{j=1}^{\infty} \phi^{j-1} Z_{t-j}$$

is the unique stationary solution. This is a causal solution since  $X_t$  can be expressed in terms of  $Z_s$ ,  $s \leq t$ .

- If  $|\phi| > 1$ , similarly it can be shown that,

$$X_t = -\theta/\phi Z_t - (\theta + \phi) \sum_{j=1}^{\infty} \phi^{-(j+1)} Z_{t+j}$$

is the unique stationary solution. This solution is non-causal since  $X_t$  depends on future values. That is  $X_t$  is a function of  $Z_s$  for  $s > t$ .

- There is no stationary solution if  $|\phi| = 1$ .

## Invertibility

(3)

- Just as  $x_t$  can be expressed ~~as~~ in terms of  $z_t$ ,  $z_t$  in some cases  $z_t$  can be expressed in terms of  $x_t$ .
- Invertibility is a concept dual to causality which means that  $z_t$  can be expressed solely in terms of  $x_s$ ,  $s \leq t$ .
- First we show that an ARMA(1,1) process is invertible if  $|\theta| < 1$ .
- Let  $\Xi(z)$  denote the power series expansion of  $\frac{1}{\theta(z)} = \sum_{j=0}^{\infty} (-\theta)^j z^j$ , which exists if  $|\theta| < 1$ .
- Now  $z_t = \Xi(B) \phi(B) x_t = x_t - (\phi + \theta) \sum_{j=1}^{\infty} (-\theta)^{j-1} x_{t-j}$ .  
Thus an ARMA(1,1) process is invertible if  $|\theta| < 1$ .
- Similarly if  $|\theta| > 1$  we can show that  
$$z_t = -\phi/\theta x_t + (\theta + \phi) \sum_{j=1}^{\infty} (-\theta)^{-j-1} x_{t+j}$$
- Note that  $z_t$  now depends on  $x_s$  with  $s \geq t$ . So this ARMA(1,1) process is non-invertible.

- We assume that the ARMA(1,1) process is both causal and invertible. That is  $|\phi| < 1$  and  $|\theta| < 1$ .

- The model is

$$X_t - \phi X_{t-1} = z_t + \theta z_{t-1}$$

or 
$$X_t = \phi X_{t-1} + z_t + \theta z_{t-1}$$

- Now  $X_t X_{t-h} = \phi X_{t-1} X_{t-h} + z_t X_{t-h} + \theta z_{t-1} X_{t-h}$

- Taking Expectation on both sides.

$$\gamma_z(h) = \phi \gamma_z(h-1) + E[z_t X_{t-h}] + \theta E[z_{t-1} X_{t-h}]$$

- For  $h=0$ .

$$\gamma_z(0) = \phi \gamma_z(1) + E[z_t X_t] + \theta E[z_{t-1} X_t]$$

- $E[z_t X_t] = E[\phi X_{t-1} z_t] + E[z_t^2] + \theta E[z_{t-1} z_t]$   
 $= \sigma^2$  (by causality)

# ACF(2)

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$$\begin{aligned} \bullet E[z_{t-1} x_t] &= \phi E[z_{t-1} x_{t-1}] + E[z_{t-1} z_t] + \theta E[z_t^2] \\ &= \phi \sigma^2 + \theta \sigma^2 = (\phi + \theta) \sigma^2 \end{aligned}$$

$$\bullet \text{ So } \gamma_x(0) = \phi \gamma_x(1) + \sigma^2 + \theta (\phi + \theta) \sigma^2$$

• When  $h = 1$ .

$$\begin{aligned} \gamma_x(1) &= \phi \gamma_x(0) + E[z_t x_{t-1}] + \theta E[z_{t-1} x_{t-1}] \\ &= \phi \gamma_x(0) + \theta \sigma^2 \end{aligned}$$

$E[z_t x_{t-1}] = 0$  [by causality]

$$\begin{aligned} \bullet \gamma_x(0) &= \phi (\phi \gamma_x(0) + \theta \sigma^2) + \sigma^2 + \theta (\phi + \theta) \sigma^2 \\ &= \phi^2 \gamma_x(0) + \theta \phi \sigma^2 + \sigma^2 + \theta (\phi + \theta) \sigma^2 \end{aligned}$$

$$\text{So } \gamma_x(0) = \frac{(1 + 2\theta\phi + \theta^2) \sigma^2}{1 - \phi^2}$$

Note that  $\gamma_x(0)$  is not defined if  $\phi^2 = 1$ .

$$\bullet \gamma_x(1) = \frac{\phi (1 + 2\theta\phi + \theta^2)}{1 - \phi^2} \sigma^2 + \theta \sigma^2 = \frac{(\phi + \theta)(1 + \theta\phi)}{1 - \phi^2} \sigma^2$$

## ACF and PACF

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- For  $h = 2$ .

$$\begin{aligned}\gamma_z(2) &= \phi \gamma_z(1) + E[z_t x_{t-2}] + \theta E[z_{t-1} x_{t-2}] \\ &= \phi \gamma_z(1).\end{aligned}$$

- For  $h > 2$ .

$$\begin{aligned}\gamma_z(h) &= \phi \gamma_z(h-1) + E[z_t x_{t-h}] + \theta E[z_{t-1} x_{t-h}] \\ &= \phi \gamma_z(h-1).\end{aligned}$$

- To summarise

$$\gamma_z(h) = \begin{cases} \frac{1 + 2\theta\phi + \theta^2}{1 - \phi^2} \sigma^2 & \text{if } h = 0 \\ \frac{(\phi + \theta)(1 + \theta\phi)}{1 - \phi^2} \sigma^2 & \text{if } h = 1 \\ \phi \gamma_z(h-1) & \text{if } h \geq 2. \end{cases}$$

- Thus the ACF is given by

$$\rho_z(h) = \begin{cases} 1 & \text{if } h = 0 \\ \frac{(\phi + \theta)(1 + \theta\phi)}{1 + 2\theta\phi + \theta^2} & \text{if } h = 1 \\ \phi \rho_z(h-1) & \text{if } h \geq 2 \end{cases}$$

## PACF

- PACF of an  $ARMA(1,1)$  is complicated, It can be calculated from the general formula.
- However since ~~AR(1)~~  $MA(1)$  is a special case of  $ARMA(1,1)$ , the PACF function of the latter will not become zero after any finite lag.
- From the ACF function it is also clear that it won't ~~be~~ become zero for any finite  $h$ .
- In general the nature of ACF and PACF plots are more complicated ~~than~~ for  $ARMA(1,1)$  than  $AR(1)$  or  $MA(1)$ .

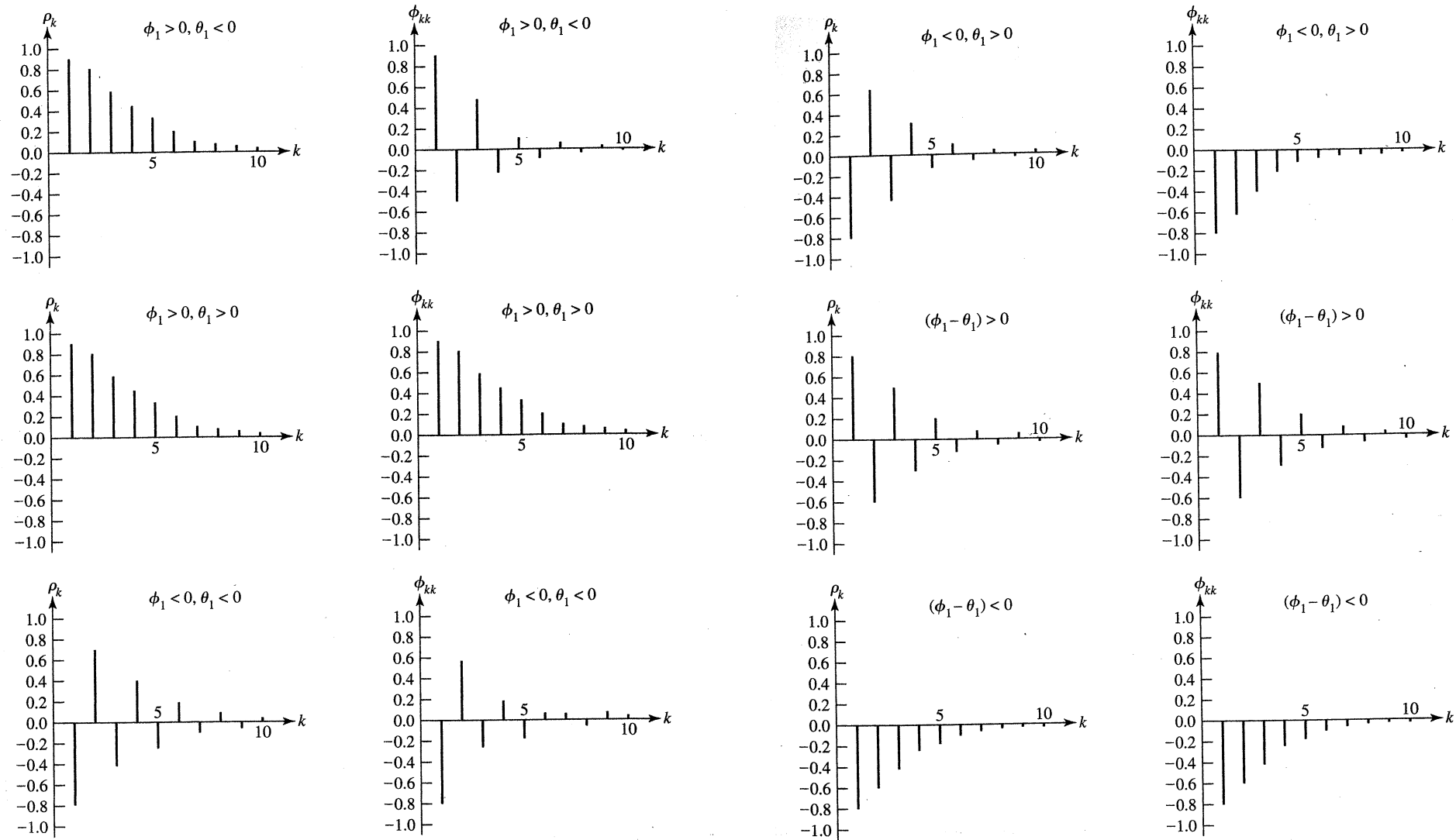


FIGURE 3.14 ACF and PACF of ARMA(1, 1) model  $(1 - \phi_1 B)\hat{Z}_t = (1 - \theta_1 B)a_t$ .

FIGURE 3.14 (Continued)



## ARMA (p,q) process

- $\{x_t\}$  is an ARMA(p,q) process if  $\{x_t\}$  is stationary and if for every  $t$ ,

$$x_t - \phi_1 x_{t-1} - \dots - \phi_p x_{t-p} = z_t + \theta_1 z_{t-1} + \dots + \theta_q z_{t-q},$$

where  $\{z_t\} \sim WN(0, \sigma^2)$ .

- In terms of backward shift operator  $B$  it can be expressed as

$$\phi(B) x_t = \theta(B) z_t,$$

where  $\phi(z)$  and  $\theta(z)$  are the  $p$ th and  $q$ th degree polynomials,

$$\phi(z) = 1 - \phi_1 z - \dots - \phi_p z^p.$$

and

$$\theta(z) = 1 + \theta_1 z + \dots + \theta_q z^q.$$

- A stationary solution of  $\{x_t\}$  exists and is unique if and only if

$$\phi(z) \neq 0 \quad \text{for all } |z| = 1.$$

- The solution may be complex.

## Causality and Invertibility

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- An ARMA( $p, q$ ) process  $\{x_t\}$  is causal, or a causal function of  $\{z_t\}$ , if there exist constants  $\{\psi_j\}$  such that  $\sum_{j=0}^{\infty} |\psi_j| < \infty$  and 
$$x_t = \sum_{j=0}^{\infty} \psi_j z_{t-j} \text{ for all } t.$$

Causality is equivalent to the condition

$$\phi(z) = 1 - \phi_1 z - \dots - \phi_p z^p \neq 0 \text{ for all } |z| \leq 1.$$

- An ARMA( $p, q$ ) process  $\{x_t\}$  is invertible if there exist constants  $\{\pi_j\}$  such that  $\sum_{j=0}^{\infty} |\pi_j| < \infty$  and 
$$z_t = \sum_{j=0}^{\infty} \pi_j x_{t-j} \text{ for all } t.$$

Invertibility is equivalent to the condition

$$\theta(z) = 1 + \theta_1 z + \dots + \theta_q z^q \neq 0 \text{ for all } |z| \leq 1.$$

- It can be shown that  $\pi_j + \sum_{k=1}^q \theta_k \pi_{j-k} = -\phi_j$ ,  $j=0, 1, \dots$   
where  $\phi_0 := -1$ ,  $\phi_j := 0 \quad \forall j > p$  and  $\pi_j := 0$  for  $j < 0$ .

## Example

(11)

- Consider the ARMA(1,1) process  $\{x_t\}$  satisfying the equations
$$x_t - .5x_{t-1} = z_t + .4z_{t-1}, \quad \{z_t\} \sim WN(0, \sigma^2).$$

Here  $\phi(z) = 1 - .5z$ . Now  $\phi(z) = 0$  has solution  $z = 2$ .

There is no solution in the unit interval. So the process is stationary and causal.

The corresponding MA( $\infty$ ) representation of  $\{x_t\}$  are ~~found~~ as follows:

$$\psi_0 = 1, \quad \psi_1 = .4 + .5, \quad \psi_2 = .5(.4 + .5), \quad \psi_j = .5^{j-1}(.4 + .5), \quad j=1, 2, \dots$$

Also  $\theta(z) = 1 + .4z$ ,  $\theta(z) = 0$  has a solution  $z = -1/4 = -2.5$ , which is located outside the unit interval.

The corresponding  $\{\pi_j\}$  is given by.

$$\pi_0 = 1, \quad \pi_1 = -(.4 + .5), \quad \pi_2 = -(.4 + .5)(-.4), \quad \pi_j = -(.4 + .5)(-.4)^{j-1}, \\ j = 1, 2, \dots$$

## Example 2

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- ARMA(2,1) process defined by .

$$X_t - .75 X_{t-1} + .5625 X_{t-2} = Z_t + 1.25 Z_{t-1}, \quad \{Z_t\} \sim \text{WN}(0, \sigma^2)$$

$\phi(z) = 1 - .75z + .5625z^2$  has zeroes at

$z = 2(1 \pm i\sqrt{3})/3$ , which lie outside the unit circle. So

the process is causal.

$\theta(z) = 1 + 1.25z$  has a zero  $z = -.8$  which is inside the unit interval, so this process is not invertible.