

Second order Autoregressive Process, AR(2)

The second order Autoregressive Process AR(2) is given by

$$y_t = \delta + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t \quad \text{--- (1)}$$

It is regression of y_t on y_{t-1} and y_{t-2} .

Now, we write eqn (1) in terms of backshift operator as

$$y_t = \delta + \phi_1 B y_t + \phi_2 B^2 y_t + \varepsilon_t$$

$$[\because B^i y_t = y_{t-i}]$$

$$y_t - \phi_1 B y_t - \phi_2 B^2 y_t = \delta + \varepsilon_t$$

$$(1 - \phi_1 B - \phi_2 B^2) y_t = \delta + \varepsilon_t \quad \text{--- (2)}$$

or

$$\Phi(B) y_t = \delta + \varepsilon_t \quad \text{--- (3)}$$

apply $\Phi(B)^{-1}$ to both sides, we obtain

$$Y_t = \Phi(B)^{-1} \delta + \Phi(B)^{-1} \varepsilon_t$$

or

$$Y_t = \mu + \Psi(B) \varepsilon_t \quad - (4)$$

where $\mu = \Phi(B)^{-1} \delta$

$$\Psi(B) = \Phi(B)^{-1}$$

$$Y_t = \mu + \sum_{i=0}^{\infty} \psi_i \varepsilon_{t-i}$$

$$Y_t = \mu + \sum_{i=0}^{\infty} \psi_i B^i \varepsilon_t$$

where

$$\mu = \Phi(B)^{-1} \delta \quad - (5)$$

and

$$\Phi(B)^{-1} = \sum_{i=0}^{\infty} \psi_i B^i = \Psi(B) \quad - (6)$$

Now, we use eqn (6) to obtain the weights in Eqn (4) in terms of ϕ_1 and ϕ_2 .

For that, we use

$$\Phi(B) \Psi(B) = 1 \quad \text{--- (7)}$$

i.e.

$$\Phi(B) \Psi(B)$$

$$(1 - \phi_1 B - \phi_2 B^2)(\psi_0 + \psi_1 B + \psi_2 B^2 + \dots) = 1$$

or

$$\begin{aligned} & \psi_0 + (\psi_1 - \phi_1 \psi_0)B + (\psi_2 - \phi_1 \psi_1 - \phi_2 \psi_0)B^2 \\ & + \dots + (\psi_j - \phi_1 \psi_{j-1} - \phi_2 \psi_{j-2})B^j + \dots = 1 \end{aligned} \quad \text{--- (8)}$$

on RHS of eqⁿ (8) there are no backshift operators, for $\Phi(B)\Psi(B)=1$, we need

$$\left. \begin{aligned} \psi_0 &= 1 \\ \psi_1 - \phi_1 \psi_0 &= 0 \\ (\psi_2 - \phi_1 \psi_1 - \phi_2 \psi_0) &= 0 \\ (\psi_j - \phi_1 \psi_{j-1} - \phi_2 \psi_{j-2}) &= 0 \end{aligned} \right\} \text{ for all } j=2,3, \dots \quad (9)$$

we can solve above eqⁿ (9) to estimate infinitely many parameters.

But ~~eqⁿ (8)~~ ψ_j in eqⁿ (9) satisfy the second-order linear difference equation.

Now, we can express it in terms of two roots m_1 and m_2 of the associated polynomial

$$(\psi_2 - \phi_1 \psi_1 - \phi_2 \psi_0) = 0$$

$$\downarrow \quad \downarrow$$

$$\boxed{m^2 - \phi_1 m - \phi_2 = 0} \quad \text{--- (10)}$$

If the roots obtained by

$$m_1, m_2 = \frac{\phi_1 \pm \sqrt{\phi_1^2 + 4\phi_2}}{2}$$

Satisfy $|m_1|, |m_2| < 1$ then

we have $\sum_{i=0}^{\infty} |\psi_i| < \infty$.

Hence if the roots m_1 and m_2 both less than 1 in absolute value, then the AR(2) model is Causal and stationary.

DATE _____
PAGE _____

If roots of eqn (10) are complex of the form $a \pm ib$, the

condition for {stationary} is that $\sqrt{a^2 + b^2} < 1$.

Under the condition that $|m_1|, |m_2| < 1$, the $AR(2)$ time series, $\{y_t\}$, has an infinite MA representation,

\therefore for the second order autoregressive process to be {stationary}, the parameters ϕ_1 and ϕ_2 must satisfy

$$\phi_1 + \phi_2 < 1$$

$$\phi_2 - \phi_1 < 1$$

$$|\phi_2| < 1$$

Conditions for the stationary of an $AR(2)$ process.

⇒ The mean of a stationary AR(2) is

from eqn (1)

$$Y_t = \delta + \phi_1 Y_{t-1} + \phi_2 Y_{t-2} + \varepsilon_t$$

Taking expectations on both sides of above eqn

$$E(Y_t) = E(\delta) + \phi_1 E(Y_{t-1}) + \phi_2 E(Y_{t-2}) + E(\varepsilon_t)$$

$$\mu = \delta + \phi_1 \mu + \phi_2 \mu + 0$$

$$\mu - \phi_1 \mu - \phi_2 \mu = \delta$$

$$\mu(1 - \phi_1 - \phi_2) = \delta$$

$$\mu = \frac{\delta}{1 - \phi_1 - \phi_2}$$

— (11)

mean of AR(2) process.