

Variance and Autocovariance function of a stationary AR(2)

✓ from eqn (1)

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

✓ The autocovariance function is

$$\gamma(k) = \text{cov}(y_t, y_{t-k})$$

$$\gamma(k) = \text{cov}(\delta + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t, y_{t-k})$$

$$\gamma(k) = \phi_1 \text{cov}(y_{t-1}, y_{t-k}) + \phi_2 \text{cov}(y_{t-2}, y_{t-k}) + \text{cov}(\varepsilon_t, y_{t-k})$$

— (2)

$$V(\varepsilon_t) = \sigma^2 \quad \text{lag } k=0$$

$$V(\varepsilon_t) = 0 \quad k \neq 0$$

$$\gamma(k) = \phi_1 \gamma(k-1) + \phi_2 \gamma(k-2) + \sigma^2 \quad \text{if } k=0$$

$$\gamma(k) = \phi_1 \gamma(k-1) + \phi_2 \gamma(k-2) + 0 \quad \text{if } k > 0$$

Auto covariance fn \Downarrow

\Downarrow

Thus for $\{k=0\}$ \Downarrow
we have

$$\gamma(0) = \phi_1 \gamma(1) + \phi_2 \gamma(2) + \sigma^2$$

and

~~$\gamma(0)$~~

$$\gamma(k) = \phi_1 \gamma(k-1) + \phi_2 \gamma(k-2)$$

$k=1, 2, 3, \dots$

— (3)

The eqn in (3) are called the

{Yule-Walker} equations for
 $\gamma(k)$.

$$\text{Cov}(y_t, y_{t-k})$$

$$\gamma(0) \\ \text{Var}$$

$$\gamma(-1) = \gamma(1) \\ \gamma(-2) = \gamma(2)$$

Again from eqn ①

$$V(\text{constant}) = 0$$

$$y_t = \underbrace{\delta}_{\text{Constant}} + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t$$

Taking Variance on both sides

$$\text{Var}(y_t) = \text{Var}(\underbrace{\phi_1 y_{t-1}}_X + \underbrace{\phi_2 y_{t-2}}_Y + \underbrace{\varepsilon_t}_Z)$$

NOTE \Rightarrow $\text{Var}(X+Y+Z) = \text{Var}(X) + \text{Var}(Y) + \text{Var}(Z) + 2\text{Cov}(XY) + 2\text{Cov}(YZ) + 2\text{Cov}(XZ)$

$$\text{Var}(ax) = a^2 \text{Var}(x)$$

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

$$+ \text{Var}(\varepsilon_t) + 2\text{Cov}(\phi_1 \phi_2 y_{t-1} y_{t-2})$$

$$+ 2\text{Cov}(\phi_2 y_{t-2} \varepsilon_t) + 2\text{Cov}(\phi_1 y_{t-1} \varepsilon_t)$$

$= 0 \qquad \qquad \qquad = 0$

$$\text{Var}(y_t) = \phi_1^2 \text{var}(y_{t-1}) + \phi_2^2 \text{var}(y_{t-2}) + \text{var}(\varepsilon_t)$$

$$+ 2\phi_1\phi_2 \text{cov}(y_{t-1}, y_{t-2})$$

Var(

or

$$\gamma(0) = \phi_1^2 \gamma(0) + \phi_2^2 \gamma(0) + \sigma^2 + 2\phi_1\phi_2 \gamma(1) \quad (4)$$

$$[\gamma(0) = \text{var}(y_t) = \text{var}(y_{t-1}) = \text{var}(y_{t-2})]$$

$$\gamma(1) = \text{cov}(y_{t-1}, y_{t-2})]$$

Stationary

From Eqn (3)

for $k=1$

$$\gamma(k) = \phi_1 \gamma(k-1) + \phi_2 \gamma(k-2)$$

$k=1, 2, 3, \dots$

$$\gamma(1) = \phi_1 \gamma(0) + \phi_2 \gamma(-1)$$

$$\gamma(1) = \phi_1 \gamma(0) + \phi_2 \gamma(1)$$

$$[\therefore \gamma(-1) = \gamma(1)]$$

which provides

$$\gamma(1) - \phi_2 \gamma(1) = \phi_1 \gamma(0)$$

$$(1 - \phi_2) \gamma(1) = \phi_1 \gamma(0)$$

$$\gamma(1) = \frac{\phi_1 \gamma(0)}{1 - \phi_2}$$

Put this value of $\gamma(1)$ in eqn (4)
we get

$$\gamma(0) = \frac{(1 - \phi_2) \sigma^2}{(1 + \phi_2) (1 - \phi_1 - \phi_2) (1 + \phi_1 - \phi_2)}$$

$\underbrace{1 - (\phi_1 + \phi_2)}_{2} \quad \underbrace{1 + \phi_1 - \phi_2}_{4}$

Variance for AR(2) process

$$\phi_2 < \phi_1 < 1$$

$$\phi_1 + \phi_2 < 1$$

$$\phi_2 - \phi_1 < 1$$

$$|\phi_2| < 1$$

Auto correlation function of
a stationary AR(2) process

$$P(k) = \frac{\gamma(k)}{\gamma(0)} = \frac{\text{cov}(y_t, y_{t-k})}{\text{Var } y_t}$$

✓ From eqn (3) Yule-walker eqn

$$\gamma(k) = \phi_1 \gamma(k-1) + \phi_2 \gamma(k-2)$$

$k=1, 2, 3, \dots$

dividing above equation by $\gamma(0)$
(Variance of y_t)

$$\frac{\gamma(k)}{\gamma(0)} = \phi_1 \frac{\gamma(k-1)}{\gamma(0)} + \phi_2 \frac{\gamma(k-2)}{\gamma(0)}$$

$$P_k = \phi_1 P_{k-1} + \phi_2 P_{k-2} \quad \text{--- (1)}$$

$k=1, 2, 3, \dots$

for $k=1$

$$\checkmark P_1 = \phi_1 P_0 + \phi_2 P_{-1}$$

$$P_0 = 1$$

$$\text{or } P_1 = \phi_1 P_0 + \phi_2 P_1 \quad \leftarrow P_{-1} = P_1$$

$$(\because P_{-1} = P_1)$$

we get

~~$$P_1 - \phi_2 P_1 = \phi_1 P_0$$~~

$$P_1 - \phi_2 P_1 = \phi_1 P_0$$

$$(1 - \phi_2) P_1 = \phi_1 P_0$$

$$P_1 = \frac{\phi_1 P_0}{1 - \phi_2} \quad (\because P_0 = 1)$$

$$P_1 = \frac{\phi_1}{1 - \phi_2} \quad - (2)$$

for $k=2$

$$P_2 = \phi_1 P_1 + \phi_2 P_0$$

$$P_2 = \phi_1 P_1 + \phi_2 \quad (\because P_0 = 1)$$

$$P_2 = \phi_1 \times \frac{\phi_1}{1 - \phi_2} + \phi_2 \quad [\text{from } \textcircled{2}]$$

$$P_2 = \frac{\phi_1^2}{1 - \phi_2} + \phi_2 \quad \textcircled{3}$$

✓ For $k \geq 3$

the autocorrelation coefficient
can be obtained ~~from~~
recursively starting from
the difference equation (1).

A general solution can be
obtained through the
roots m_1 and m_2 of the
associated polynomial

C-I

✓ $m^2 - \phi_1 m - \phi_2 = 0$.

as $\left\{ P(k) = C_1 m_1^k + C_2 m_2^k \right\} \quad \text{--- (4)}$

where C_1 and C_2 are
particular constant to be
determined from the
initial conditions $P_0 = 1$
($C_1 + C_2 = 1$) and

$$P_1 = \frac{\phi_1}{1 - \phi_2}$$

According to eqn (4), the coefficients
 $\{ \rho(k) \text{ will be } \leq 1 \}$ if

$|m_1| < 1$ and $|m_2| < 1$, which
 are the conditions of stationarity
 of the process.

and, in this case, the ACF is
 a mixture of two exponential
 decay terms.

CASE-II If m_1 and m_2 are complex conjugates in the form $a \pm ib$, we then have

$$\sqrt{a^2 + b^2} < 1$$

$$P(k) = R^k [C_1 \cos(\lambda k) + C_2 \sin(\lambda k)]$$

$k \geq 3$

where $R = |m_i| = \sqrt{a^2 + b^2}$

and λ is determined by

$$\cos(\lambda) = \frac{a}{R}$$

$$\sin(\lambda) = \frac{b}{R}$$

Hence $a + ib = R [\cos(\lambda) + i \sin(\lambda)]$.

C_1 and C_2 are particular constant.

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The ACF in this case has the form of a sinusoid with damping factor R and frequency λ , i.e., the period is $\frac{2\pi}{\lambda}$.

CASE-III

If there is one real root m_0

$m_1 = m_2 = m_0$, we then have

$$P(k) = (C_1 + C_2 k) m_0^k \quad k \geq 3$$

in this case ACF will exhibit an exponential decay pattern.