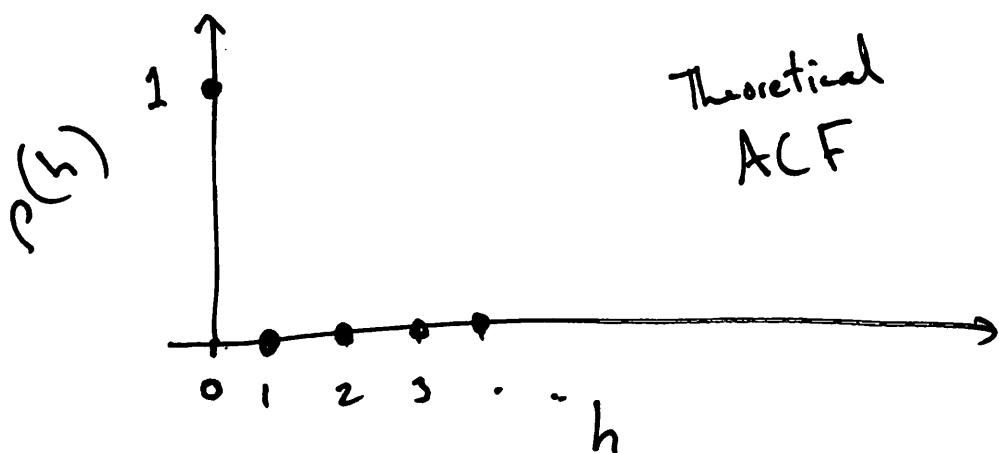


Autocovariance White Noise

①

$$\gamma(h) = \text{Cov}(w_t, w_{t+h}) \longrightarrow E(w_t w_{t+h}) - \underbrace{E(w_t)E(w_h)}_{\text{after disappearance}} \\ = \begin{cases} \text{Var}(w_t) = \sigma^2 & h=0 \\ 0 & h \neq 0 \end{cases}$$

$$\rho(h) = \frac{\gamma(h)}{\gamma(0)} = \begin{cases} \frac{0}{\sigma^2} & h \neq 0 \\ 1 & h = 0 \end{cases}$$



Random Walk with drift

(2)

$$M_t = E(x_t) = E\left(t\delta + \sum_{j=1}^t w_j\right)$$

$$= t\delta$$

linearity of expectation
 $E[w_j] = 0 \quad \forall j$

$$\gamma(t, t+h) = \text{Cov}(x_t, x_{t+h})$$

$$= \text{Cov}\left(t\delta + \sum_{j=1}^t w_j, (t+h)\delta + \sum_{j=1}^{t+h} w_j\right)$$

$$= \text{Cov}\left(t\delta + w_1 + w_2 + \dots + w_t, (t+h)\delta + w_1 + w_2 + \dots + w_t + \dots + w_{t+h}\right)$$

$$\boxed{\begin{aligned} & \text{Cov}\left(\sum_{i=1}^t y_i, \sum_{j=1}^t z_j\right) \\ &= \sum_{i=1}^t \sum_{j=1}^t \text{Cov}(y_i, z_j) \end{aligned}}$$

$$= t \text{Var}(w_1)$$

$$= t \sigma^2$$

depends on t .

Not stationary

$$\delta = 0 \quad \text{Mean function} = 0 + t$$

even here not stationary...

(3)

MA(1)

$$E(x_t) = E(\beta_1 w_{t-1} + w_t) \\ = \neq 0$$

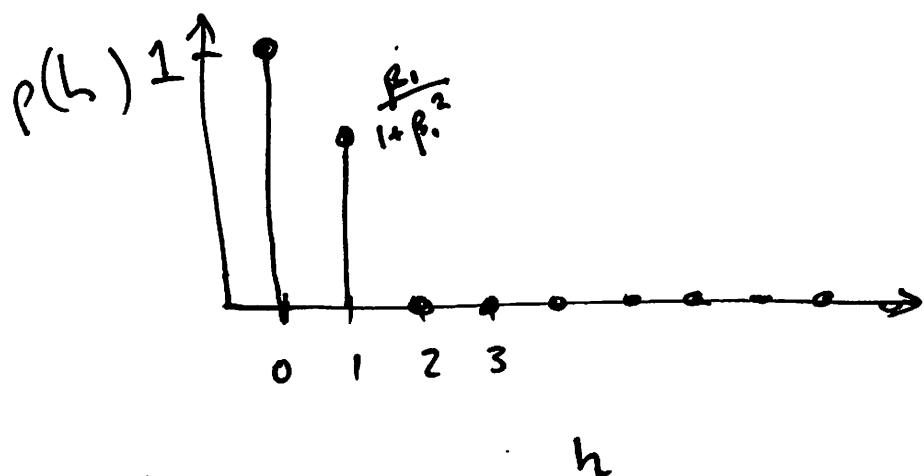
$$\gamma(x_t, x_{t+h}) = \text{Cov}(\beta_1 w_{t-1} + w_t, \beta_1 w_{t+h-1} + w_{t+h})$$

$$= \begin{cases} \beta_1 \sigma^2, & h = 1, -1 \quad |h| = 1 \\ \beta_1^2 \sigma^2 + \sigma^2, & h = 0 \\ 0, & h > 1 \quad |h| \geq 2 \\ & h < -1 \end{cases}$$

$$\rho(h) = \frac{\gamma(h)}{\gamma(0)} \begin{cases} 1 & h = 0 \\ \frac{\beta_1 \sigma^2}{\beta_1^2 \sigma^2 + \sigma^2} = \frac{\beta_1}{1 + \beta_1^2} & |h| = 1 \\ 0 & |h| \geq 2 \end{cases}$$

Stationary!

(4)



(5)

AR(1)

$$x_t = \alpha_1 x_{t-1} + \omega_t$$

$$E(x_t) = E(\alpha_1 x_{t-1} + \omega_t)$$

$$= E(\alpha_1 (\alpha_1 x_{t-2} + \omega_{t-1}) + \omega_t)$$

$$= E(\alpha_1^2 x_{t-2} + \alpha_1 \omega_{t-1} + \omega_t)$$

$$= E(\alpha_1^2 (\alpha_1 x_{t-3} + \omega_{t-2}) + \alpha_1 \omega_{t-1} + \omega_t)$$

$$= E(\omega_t + \alpha_1 \omega_{t-1} + \alpha_1^2 \omega_{t-2} + \alpha_1^3 \omega_{t-3} + \dots)$$

$$= 0 \quad \begin{matrix} \text{intuitively pleasant} \\ \text{TRUE} \end{matrix}$$

$$E\left(\sum_{i=1}^{\infty} \alpha^{i-1} \omega_{t-i}\right)$$

↑

$$\hookrightarrow \sum_{i=0}^{\infty} E[\alpha^i \omega_{t-i}]$$

this isn't
always true

(6)

$$\text{Cov}(x_t, x_{t+h}) = \text{Cov} \left(w_t + \alpha_1 w_{t-1} + \alpha_1^2 w_{t-2} + \dots, \right.$$

3

$$w_{t+h} + \alpha_1 w_{t+h-1} + \alpha_1^2 w_{t+h-2} + \dots$$

$$\left. + \alpha_1^h w_t + \alpha_1^{h+1} w_{t-1} + \dots \right).$$

$$= \alpha_1^h \sigma^2 + \alpha_1^{h+2} \sigma^2 + \alpha_1^{h+4} \sigma^2 + \dots$$

$$= \alpha_1^h \sigma^2 \sum_{i=0}^{\infty} \alpha^{2i}$$

"geometric series"
doesn't always
converge

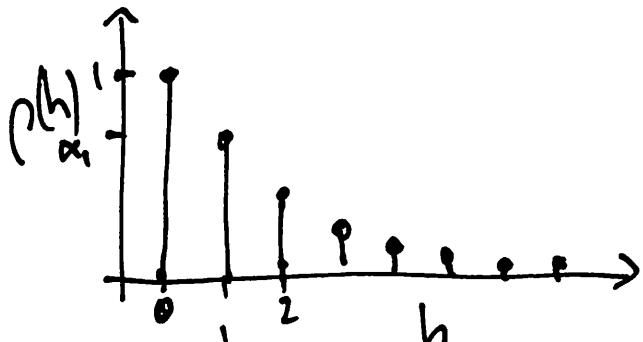
will if $|\alpha^2| < 1$

converges to $\frac{1}{1-\alpha^2}$

$$= \frac{\alpha_1^h \sigma^2}{1-\alpha_1^2} \quad \text{if } |\alpha_1^2| < 1$$

$\Leftrightarrow |\alpha_1| < 1$ weakly stationary

$$\rho(h) \approx \rho(h) = \frac{\gamma(h)}{\gamma(0)} = \begin{cases} 1 & h=0 \\ \alpha_1^{|h|} & |h| > 0 \end{cases}$$



$$|\alpha_1| < 1$$