

## Chapter 2

- White noises

- 不一定相互独立 +  $Var < +\infty$

$$E\varepsilon_t = 0 \text{ and } \text{cov}(\varepsilon_t, \varepsilon_s) = \begin{cases} \sigma^2, & t = s, \\ \underbrace{0}_{\text{red}}, & t \neq s, \end{cases}$$

- measure of dependence

- Autocovariance

$$\gamma_{s,t} = \text{Cov}(r_s, r_t) = E[(r_s - \mu_s)(r_t - \mu_t)]$$

- Autocorrelation

$$\rho_{s,t} = \text{corr}(r_s, r_t) = \frac{\gamma_{s,t}}{\sqrt{\gamma_{s,s}\gamma_{t,t}}}.$$

- weak stationary

- 前两个moment是constant
- 只与k有关, 与t无关

. Lag-k autocovariance:

$$\gamma_k = \text{Cov}(r_t, r_{t-k}) = E[(r_t - \mu)(r_{t-k} - \mu)].$$

与t无关, 只与k有关, 即时间差

- ACF不等于0, 意味着cov不等于0, 他们是related的, 所以我可以根据之前的信息predict, inefficiency

- strict stationary

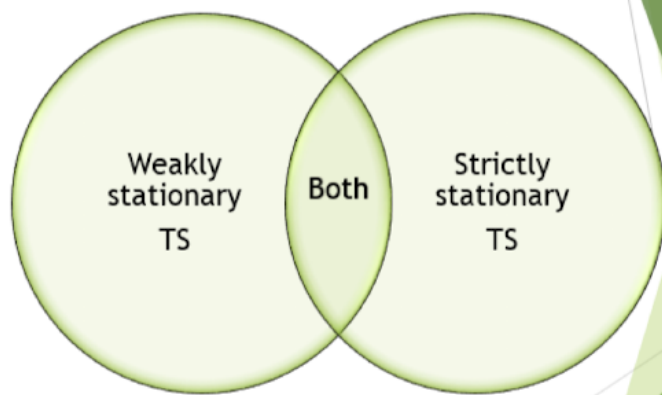
- definition

Strict: distributions are time-invariant, i.e.,

$$\underbrace{P(r_{t_1} \leq z_1, \dots, r_{t_n} \leq z_n)}_{\text{joint distribution}} = P(r_{t_1+k} \leq z_1, \dots, r_{t_n+k} \leq z_n),$$

for  $\forall t_1, \dots, t_n, k$  and  $(z_1, \dots, z_n)$  and  $n$ .

- 没有办法判断到底是不是station的, 只能assump, 然后判断assumption是否正确? close enough
- relationship: strict + finite variance --> weak



- property of WS

- $\rho_0 = 1, \gamma_0 = \text{var}(r_t)$

$$\gamma_0 = \text{cov}(r_t, r_{t-0}) = \text{cov}(r_t, r_t) = \text{var}(r_t)$$

$$\rho_0 = \frac{\text{var}(r_t)}{\sqrt{\text{var}(r_t)} \sqrt{\text{var}(r_t)}} = 1$$

- $|\gamma_k| \leq \gamma_0, |\rho_k| \leq 1$

$$\begin{aligned} |\gamma_k| &= |E[(r_t - \mu)(r_{t-k} - \mu)]| \leq \sqrt{E[(r_t - \mu)^2] E[(r_{t-k} - \mu)^2]} \stackrel{\text{Schwarz}}{\leq} \sqrt{E[X^2] E[Y^2]} \\ &= \sqrt{(\text{var}(r_t))^2} \\ &= \text{var}(r_t) = \gamma_0 \end{aligned}$$

$$\rho_k \leq \frac{\gamma_0}{\gamma_0} = 1$$

- $\gamma_k = \gamma_{-k}, \rho_k = \rho_{-k}$

- $\rho_k = 0$  for all  $k \neq 0 \rightarrow \text{cov} = 0 \rightarrow \text{white noise}$  (应该有WS的条件)

- estimate:

- $\mu = 0$

Test  $H_0 : \mu = 0$  vs  $H_a : \mu \neq 0$ . Compute

$$t = \frac{\sqrt{T} \bar{r}}{\hat{\sigma}_r} \quad \frac{\bar{r}}{\frac{\hat{\sigma}_r}{\sqrt{T}}}$$

Compare  $t$  ratio with  $N(0, 1)$  dist.

- $ACF = 0$

$$\hat{\rho}_k = \frac{\sum_{t=1}^{T-k} (r_t - \bar{r})(r_{t+k} - \bar{r})}{\sum_{t=1}^T (r_t - \bar{r})^2},$$

- individual

$$H_0 : \rho_1 = 0 \text{ vs } H_a : \rho_1 \neq 0$$

$$t = \frac{\hat{\rho}_1}{\sqrt{1/T}} = \sqrt{T} \hat{\rho}_1 \sim N(0, 1).$$

can't do one

- joint: 要拒绝  $H_0 \leftarrow p$  足够小

Joint test (Ljung-Box statistics): 要拒绝  $H_0 \Rightarrow p\text{-value} <$

$H_0: \rho_1 = \dots = \rho_m = 0$  vs  $H_a: \rho_i \neq 0$  for some  $i$ . 不能太大  $\Rightarrow$  it is not too large

$$Q(m) = T(T+2) \sum_{k=1}^m \frac{\hat{\rho}_k^2}{T-k} \sim \chi_m^2.$$

Asym.  $\chi^2$  dist with  $m$  degrees of freedom.

- 注意:  $m$ 不能太大,  $m$ 太大, ACF用的数据太少

- notation:  $Br_t = r_{t-1}, B2 = 2$
- $r_t = \mu_t + a_t \mu_t$ : predict part, the best predictor of  $r_t$ 
  - 如果我们令  $g_t \in F_t \rightarrow g_t \in f(F_{t-1})$

- $r_t = \mu_t + a_t \mu_t$ : predict part, the best predictor of  $r_t$

- 如果我们令  $g_t \in F_t \rightarrow g_t \in f(F_{t-1})$

$$E(r_t - g_t)^2 > E(r_t - \mu_t)^2 \text{ if } g_t \neq \mu_t.$$

- proof

$$E(y_t - g_t)^2 = E[(y_t - \mu_t)(g_t - \mu_t)] = E[(y_t - \mu_t)^2] + E[(\mu_t - g_t)^2] - 2E[(y_t - \mu_t)(g_t - \mu_t)]$$

$$E[(y_t - \mu_t)(g_t - \mu_t)] = E[E[(y_t - \mu_t)(g_t - \mu_t) | F_{t-1}]] = E[g_t - \mu_t] E[y_t - \mu_t | F_{t-1}] = E[g_t - \mu_t] (\mu_t - \mu_t) = 0$$

- $a_t$  is white noise but no sure it's iid

$$\begin{aligned}\sigma_t^2 &= \text{Var}(r_t | F_{t-1}) \\ &= E[(Y_t - \mu_t)^2 | F_{t-1}] - \underbrace{(E[Y_t - \mu_t | F_{t-1}])^2}_{=0} \quad \mu_t = E[Y_t | F_{t-1}] \\ &= E[a_t^2 | F_{t-1}] = \text{Var}(a_t)\end{aligned}$$

$$\Rightarrow r_t = \mu_t + \frac{\sigma_t}{\sigma_t} \varepsilon_t a_t$$

$$\textcircled{1} E[\varepsilon_t] = E[E[\varepsilon_t | F_{t-1}]] = E[E[\frac{a_t}{\sigma_t} | F_{t-1}]] = E[\frac{1}{\sigma_t} E[a_t | F_{t-1}]] = 0$$

$$\textcircled{2} \text{Var}(\varepsilon_t) = E[\varepsilon_t^2] = E[E[\frac{a_t^2}{\sigma_t^2} | F_{t-1}]] = E[\frac{1}{\sigma_t^2} E[a_t^2 | F_{t-1}]] = 1$$

$$\textcircled{3} \text{for } a_t, \quad r_k = E[a_t a_{t+k}] = E[E[a_t a_{t+k} | F_{t+k}]] \\ = E[a_{t+k} E[a_t | F_{t+k}]] = 0$$

$\Rightarrow a_t$  is unrelated  $\Rightarrow \varepsilon_t$  is unrelated.

$$\Rightarrow \varepsilon_t \sim WN(0,1)$$