

(IID) $\rightarrow$ dependence + stationarity

Stock & Watson (SW)

Ruppert & Matteson (SDAFE)

$\hookrightarrow$ one concept of stationarity

$\hookrightarrow$ strict stationarity $\rightarrow$ distribution of blocks of r.v.'s unchanging over time

weak stationarity $\hookrightarrow$ mean, variance, and covariance unchanging over time

moments exist + strict stationarity $\Rightarrow$ weak stationarity

$$\text{cov}(Y_t, Y_s) = \gamma(|t-s|) \quad \gamma(h)$$

$\uparrow$
function of $|t-s|$

$\hookrightarrow$ covariances are function of $|t-s|$

$t-s$ = gap between two time periods.

$\bar{Y}$ Sample average

How are the properties of sample average going to be affected by the change from IID to dependence + stationarity?

$$\bar{Z} = \frac{Z_1 + Z_2 + \dots + Z_n}{n}$$

$Z_1, \dots, Z_n$

dependence $\checkmark$
stationarity
finite moments

$$= \frac{1}{n} \sum_{t=1}^n Z_t$$

$$E(\bar{Z}) = \frac{1}{n} (E(Z_1) + E(Z_2) + \dots + E(Z_n))$$

By stationarity
 $E(Z_1) = E(Z_2) = \dots$

$$= E(Z_n) = E(Z_t) = \mu$$

μ doesn't depend on t

n = # of obs / r.v.'s

$Z_1, \dots, Z_n$ (IID) (μ, σ^2)
E(CONSTA)

$$E(\bar{Z}) = \mu$$

$$\text{Var}(\bar{Z}) = \sigma^2/n$$

$\bar{Z} \xrightarrow{P} \mu$
converges in prob.

$$\text{Var}(\bar{Z}) = \text{Var}\left(\frac{1}{n}(Z_1 + \dots + Z_n)\right)$$

$\leftarrow$ constant

$$= \frac{1}{n^2} \text{Var}(Z_1 + \dots + Z_n)$$

Variance of a sum of r.v.'s?

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$$

dependence $\Rightarrow \frac{1}{4} \left[\begin{aligned} &Var(z_1) + Var(z_2) + \dots + Var(z_n) \\ &+ 2Cov(z_1, z_2) \\ &+ 2Cov(z_1, z_3) \\ &+ 2Cov(z_1, z_4) \\ &\vdots \\ &+ \dots \end{aligned} \right]$

$Var(X+Y+Z) = Var(X) + Var(Y) + Var(Z) + 2Cov(X, Y) + 2Cov(X, Z) + 2Cov(Y, Z)$

organize as matrix

$$Var\left(\frac{z_1 + z_2}{2}\right) = \frac{1}{4} \left[Var(z_1) + Var(z_2) + 2Cov(z_1, z_2) \right]$$

Matrix representation of the covariance terms:

$$\begin{pmatrix} Var(z_1) & Cov(z_1, z_2) \\ Cov(z_2, z_1) & Var(z_2) \end{pmatrix}$$

$$\begin{bmatrix} Var(z_1) & Cov(z_1, z_2) & Cov(z_1, z_3) & Cov(z_1, z_4) & \dots & Cov(z_1, z_n) \\ Cov(z_2, z_1) & Var(z_2) & Cov(z_2, z_3) & Cov(z_2, z_4) & \dots & Cov(z_2, z_n) \\ Cov(z_3, z_1) & Cov(z_3, z_2) & \ddots & \ddots & \ddots & \ddots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \ddots \\ Cov(z_n, z_1) & Cov(z_n, z_2) & \dots & \dots & \dots & Var(z_n) \end{bmatrix}$$

under stationarity:

not equal $\left\{ \begin{aligned} &Cov(z_1, z_2) = Cov(z_2, z_3) = \dots \text{ one period apart} \\ &Cov(z_1, z_3) = Cov(z_2, z_4) = \dots \text{ two periods apart} \end{aligned} \right.$

ELN STA

$$Var(\bar{z}) = \frac{\sigma^2}{n} \Rightarrow \text{Standard error of } \bar{z} = \frac{\sigma}{\sqrt{n}}$$

ELW FMET

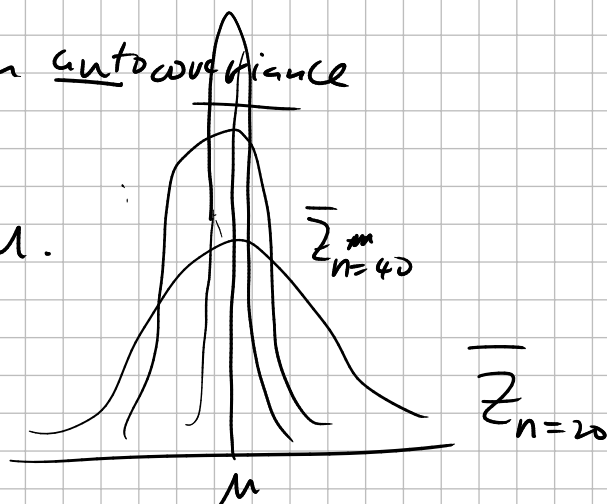
$$Var(\bar{z}) = \frac{\sigma^2}{n} + \text{[uncertainty]}$$

$\text{Cov}(\bar{Z}_t, \bar{Z}_{t-j})$ j th-order autocovariance

ELUNSTA Under IID, $\bar{Z} \rightarrow N$.

$$\text{Var}(\bar{Z}) \left(\frac{\sigma^2}{n} \right) \rightarrow 0 \quad \text{As } n \rightarrow \infty$$

As $n \rightarrow \infty$



$1 + 2 + 3 + 4 + \dots = \text{not finite sum}$

$$1 + \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots = 2$$

$$1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \dots = \text{not finite sum}$$

$Y_t = 1 \cdot Y_{t-1} + u_t$ random walk

$$Y_t = \beta_1 Y_{t-1} + u_t \quad 0 < \beta_1 < 1$$

parallels to ECOMETR

$$Y_t = \beta_0 + \beta_1 X_t + \varepsilon_t$$

$$E(\varepsilon_t | X_t) = 0$$

IID,

~~$\text{Var}(\varepsilon_t | X_t) = \sigma^2$~~ cond. homoskedasticity

dependence, stationarity $\Rightarrow$ effects on sample average.

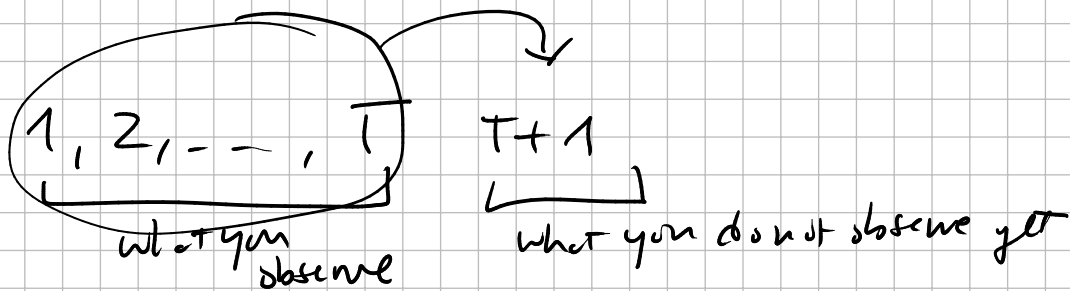
Autocovariances may not necessarily be equal to zero
formulas learned in ELUNSTA / ECOMETR may not apply.

What kind of time series models can we entertain?
and how can they be useful?

SW: fitting autoregressions in order to produce forecasts ^{best hypo...}

SDAPE: figuring out what type of time series model fits the data, forecasting is more of an afterthought

given time series $\rightarrow$ AR? MA? ARMA?



parallels to ECOMETR

What you see

$Y_1, \dots, Y_T$

What you want to predict

Y_{T+1}

best prediction in terms of the smallest mean squared error

$E(Y_{T+1} | Y_1, \dots, Y_T)$

function of past data

ECOMETR

What you see

regressors/features $X_1, X_2, \dots, X_k$

What you want to predict

Y

$E(Y | X_1, \dots, X_k)$ function of features/regressors.

general functions (they are completely unspecified)

SW Section 15.3

$$E(Y_{T+1} | Y_1, \dots, Y_T) = (\beta_0 + \beta_1 Y_T)$$

$$* E(Y_t | Y_{t-1}, Y_{t-2}, \dots) = \beta_0 + \beta_1 Y_{t-1} *$$

Similar to ECOMETR $E(Y | X_1, \dots, X_k) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_k X_k$

Cond. exp = best prediction!

$$\mathbb{E}(u_t | Y_{t-1}, Y_{t-2}, \dots) = 0$$

↓

past things you observe as of time t

$$Y_t = (\beta_0 + \beta_1 Y_{t-1}) + u_t$$

$$\mathbb{E}(Y_t | Y_{t-1}, \dots) = \beta_0 + \beta_1 Y_{t-1} + \underbrace{\mathbb{E}(u_t | Y_{t-1}, \dots)}_{=0}$$

$$\mathbb{E}(u_t | Y_{t-1}, Y_{t-2}, \dots) = 0$$

$$* \mathbb{E}(u_t | Y_{t-1}, Y_{t-2}, \dots) > 0$$

$$* \mathbb{E}(u_t | Y_{t-1}, Y_{t-2}, \dots) < 0$$