

# Autocorrelation

## Remedies:

### 1) Specification

true:  $y_t = \alpha + \beta X_t + \underbrace{\rho z_t + \varepsilon_t}_{WN}$

est:  $y_t = \alpha + \beta X_t + u_t$

$z_t$                       else  
 - AR process                      ↓  
   ⇓                                      Not  
 $\varepsilon_t$  - autocor.

### 2) AR(1) Transform (special case of GLS)

Cochran - Orcutt

$y_t = \beta_1 + \beta_2 X_t + u_t$

| • Lp

$$\frac{\sigma_u^2}{\sum (x_i - \bar{x})^2}$$

Assume:  $u_t = \rho u_{t-1} + \varepsilon_t \leftarrow$

$y_t = \beta_1 + \beta_2 X_t + \cancel{\rho u_{t-1}} + \varepsilon_t$

$\rho y_{t-1} = \rho \beta_1 + \rho \beta_2 X_{t-1} + \cancel{\rho u_{t-1}}$

$y_t - \rho y_{t-1} = \beta_1 (1 - \rho) + \beta_2 (X_t - \rho X_{t-1}) + \varepsilon_t$

$y_t^* = \beta_1^* + \beta_2 X_t^* + \varepsilon_t \sim WN$

$$\beta = (X'X)^{-1} X'y$$

$$\begin{bmatrix} 1 & \dots & 1 \\ x_1 & \dots & x_n \end{bmatrix} \begin{bmatrix} 1 \\ x_1 \\ \vdots \\ x_n \end{bmatrix} \hat{\beta}$$

$$\text{Var}(\hat{\beta}_2) = \sigma^2 (X'X)^{-1} = \frac{\sigma_\varepsilon^2}{\sum (x_i - \bar{x})^2}$$

$$\hat{u}_t = \rho \hat{u}_{t-1} + \varepsilon_t$$

$$\hat{\rho} = \frac{\sum \hat{u}_t \cdot \hat{u}_{t-1}}{\sum \hat{u}_{t-1}^2}$$

$$\searrow \quad DW \rightarrow 2(1-\rho)$$

$$\hat{\rho} = 1 - \frac{DW}{2}$$

Price - Winsten correction: (for 1st obs)

$$y_t^* = \beta_1 + \beta_2 x_t^* + \varepsilon_t$$

$$\text{Var}(\varepsilon_t) = \sigma_\varepsilon^2$$

$$\sqrt{1-\rho^2} \mid y_1 = \beta_1 + \beta_2 x_1 + u_1$$

$$\text{Var}(u_t) = \sigma_u^2$$

$$= \text{Var}(\rho u_{t-1} + \varepsilon_t) =$$

$\leftarrow$  RHS

$$= \rho^2 \sigma_u^2 + \sigma_\varepsilon^2 + 2.0$$

$$\Rightarrow \sigma_u^2 = \frac{\sigma_\varepsilon^2}{1-\rho^2}$$

$$\sigma_\varepsilon^2 = (1-\rho^2) \sigma_u^2$$

$$u_t = \rho u_{t-1} + \varepsilon_t$$

③ MA(1)  $\rightarrow$  GLS

$$\hat{\beta}_{GLS} = (X' \Omega^{-1} X)^{-1} \cdot X' \Omega^{-1} y$$

$$\Omega = \sigma_{\varepsilon}^2 \begin{pmatrix} 1 & & & 0 \\ & \ddots & & \\ & & \ddots & \\ 0 & & & 1 \end{pmatrix}$$

$\Downarrow$

$$\hat{\beta} = (X' X)^{-1} \cdot X' y$$

$$\Omega = \sigma_{\varepsilon}^2 \begin{pmatrix} X_1 & 0 & & 0 \\ 0 & \ddots & & \\ \vdots & & \ddots & \\ 0 & \dots & 0 & X_n \end{pmatrix}$$

$\Downarrow$

WLS

AR(1):

$$\Omega = \frac{\sigma_{\varepsilon}^2}{1-\rho^2} \begin{pmatrix} 1 & \rho & \rho^2 & \dots & \rho^{n-1} \\ \rho & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \rho \\ \rho^{n-1} & \dots & \rho & 1 & \end{pmatrix}$$

$$\varepsilon_t = \rho \varepsilon_{t-1} + u_t$$

$\downarrow$

$$\text{Cov}(\varepsilon_t, \varepsilon_{t-k})$$

$=$

$$\frac{\rho^k}{1-\rho^2}$$

MA(1):  $\varepsilon_t = u_t + \rho u_{t-1}$

$$\Omega = \sigma_{\varepsilon}^2 \begin{pmatrix} 1 & \rho & 0 & \dots & 0 \\ \rho & 1 & \rho & \dots & 0 \\ 0 & \rho & 1 & \ddots & \\ \vdots & \vdots & \ddots & \ddots & \rho \\ 0 & 0 & \dots & \rho & 1 \end{pmatrix}$$

$$y_t = \overset{\lambda_1}{\rho} y_{t-1} + \overset{\lambda_2}{\beta_1} (1 - \rho) + \overset{\lambda_3}{\beta_2} x_t - \overset{\lambda_4}{\beta_2 \rho} x_{t-1} + \varepsilon_t$$

ARDL (1, 1) with restrictions

$$\lambda_4 = - \lambda_1 \cdot \lambda_3$$

✓

non-linear restriction

Common Factor test

$$n \cdot \log \left( \frac{RSS_R}{RSS_{UR}} \right) \sim \chi^2_1$$