

## ESTIMATING BOX-JENKINS MODELS

### 1. ARMA(0,0) Model

$$y_t = \phi_0 + a_t$$

The **least squares estimator** of  $\phi_0$  is the sample mean of  $y$ ,  $\hat{\phi}_0 = \sum_{t=1}^T y_t / T = \bar{y}$ .

This estimator is obtained by minimizing the least squares criterion

$$S = \sum_{t=1}^T a_t^2 = \sum_{t=1}^T (y_t - \phi_0)^2$$

with respect to  $\phi_0$ . As it turns out,  $\bar{y}$  is also the **method-of-moments estimator** of  $\phi_0$  since  $E(y_t) = \phi_0$  and the sample mean of  $y$  can be used to estimate it.

### 2. AR(1) Model

$$y_t = \phi_0 + \phi_1 y_{t-1} + a_t$$

The **least squares estimators** of  $\phi_1$  and  $\phi_0$  are, respectively,

$$\hat{\phi}_1 = \frac{\sum_{t=2}^T (y_{t-1} - \bar{y}_{-1})(y_t - \bar{y})}{\sum_{t=2}^T (y_{t-1} - \bar{y}_{-1})}$$

$$\hat{\phi}_0 = \bar{y} - \hat{\phi}_1 \bar{y}_{-1}$$

where  $\bar{y} = \sum_{t=2}^T y_t / (T-1)$  and  $\bar{y}_{-1} = \sum_{t=2}^T y_{t-1} / (T-1)$ . These estimators are obtained by minimizing the least squares criterion

$$S = \sum_{t=1}^T a_t^2 = \sum_{t=1}^T (y_t - \phi_0 - \phi_1 y_{t-1})^2$$

with respect to  $\phi_0$  and  $\phi_1$ .

Alternatively, one could use the **method-of-moments** to estimate the parameters  $\phi_0$  and  $\phi_1$ . Consider the following two moments.

$$E(y_t) = \frac{\phi_0}{1-\phi_1} \quad (1)$$

and

$$\text{Corr}(y_t, y_{t-1}) = \rho_1 = \phi_1 \quad (2)$$

Therefore, a consistent **method-of-moments estimate** of  $\phi_1$  is

$$\hat{\phi}_1 = r_1, \quad (3)$$

where  $r_1$  is the first-order sample autocorrelation coefficient. From (1) we see that the sample mean of  $y$ ,  $\bar{y}$ , can be used to estimate  $\phi_0/(1-\phi_1)$ . That is,

$$\frac{\hat{\phi}_0}{1-\hat{\phi}_1} = \bar{y}. \quad (4)$$

Substituting  $\hat{\phi}_1 = r_1$  into (4) allows us to determine an **method-of-moments estimator** of  $\phi_0$ , namely,

$$\hat{\phi}_0 = \bar{y}(1-r_1). \quad (5)$$

Although the least squares and method-of-moments estimators of  $\phi_0$  and  $\phi_1$  are not the same in finite samples, they equal each other in infinite samples.

### 3.MA(1) Model

$$y_t = \phi_0 + a_t - \theta_1 a_{t-1}$$

Unfortunately, the least squares method cannot be used to estimate  $\phi_0$  and  $\phi_1$  in this model since the “data”  $a_{t-1}$  is not observable. However, we can use the method of moments to estimate these parameters. Consider that

$$E(y_t) = \phi_0 \quad (6)$$

and

$$\rho_1 = \frac{-\theta_1}{1+\theta_1^2}. \quad (7)$$

Replacing these moments with their sample estimates, we have

$$\hat{\phi}_0 = \bar{y} \tag{8}$$

and  $\hat{\theta}_1$  so as to satisfy the moment condition

$$r_1 = \frac{-\hat{\theta}_1}{1 + \hat{\theta}_1^2} \tag{9}$$

and, at the same time, the invertibility condition  $|\hat{\theta}_1| < 1$ . Again,  $r_1$  is the first-order sample autocorrelation coefficient of the time series  $y_t$ . The two roots that will satisfy (9) are

$$\hat{\theta}_1 = \frac{-1 \pm \sqrt{1 - 4r_1^2}}{2r_1} \tag{10}$$

as long as  $r_1 \leq 1/2$ . One then just chooses the root  $\hat{\theta}_1$  that satisfies the invertibility condition.