Notation from Lutkepohl 2005:

(Section 2.1.1., p.13)

The object of interest in the following is the VAR(p) model (VAR model of order p),

$$y_t = \nu + A_1 y_{t-1} + \dots + A_p y_{t-p} + u_t, \quad t = 0, \pm 1, \pm 2, \dots,$$

where

- $y_t = (y_{1t}, \dots, y_{Kt})'$ is a $(K \times 1)$ random vector,
- the A_i are fixed $(K \times K)$ coefficient matrices,
- $\nu = (\nu_1, \dots, \nu_K)'$ is a fixed $(K \times 1)$ vector of intercept terms allowing for the possibility of a nonzero mean $E(y_t)$,
- $u_t = (u_{1t}, \dots, u_{Kt})'$ is a K-dimensional white noise or innovation process:
 - $-E(u_t)=0,$
 - $E(u_t u_t') = \Sigma_u,$
 - $-E(u_t u_s') = 0 \text{ for } s \neq t,$
 - The covariance matrix Σ_u is assumed to be nonsingular if not otherwise stated.

Multivariate Least Squares Estimation

(Section 3.2.1, p.70)

It is assumed that a time series y_1, \ldots, y_T of the y variables is available, that is, we have a sample of size T for each of the K variables for the same sample period. In addition, p presample values for each variable, y_{-p+1}, \ldots, y_0 , are assumed to be available. Partitioning a multiple time series into sample and presample values is convenient in order to simplify the notation.

$$Y := (y_1, \dots, y_T) \qquad (K \times T)$$

$$B := (v, A_1, \dots, A_p) \qquad (K \times (Kp+1))$$

$$Z_t := \begin{bmatrix} 1 \\ y_t \\ \vdots \\ y_{t-p+1} \end{bmatrix} \qquad ((Kp+1) \times 1)$$

$$Z := (Z_0, \dots, Z_{T-1}) \qquad ((Kp+1) \times T)$$

$$U := (u_1, \dots, u_T) \qquad (K \times T)$$

$$y := \text{vec}(Y) \qquad (KT \times 1)$$

$$\beta := \text{vec}(B) \qquad ((K^2p + K) \times 1)$$

$$u := \text{vec}(U) \qquad (KT \times 1)$$

The OLS estimate of B is given by (eqn.3.2.10, p.72):

$$\widehat{B} = YZ'(ZZ')^{-1}.$$

An unbiased estimator for Σ_u is (eqn.3.2.19, p.75):

$$\widehat{\Sigma}_u = \frac{1}{T - Kp - 1} \widehat{U}\widehat{U}'.$$

Estimated standard errors of the coefficient estimates are the square roots of the diagonal elements of (eqn.3.2.21, p.77):

$$(ZZ')^{-1}\otimes\widehat{\Sigma}_u$$