Structure of CKMV

Colombi, Kumbhakar, Martini, Vittadini (2014)

$$y_{it} = \beta_0 + x'_{it}\beta + \underbrace{b_i}_{effect} - \underbrace{u_{it}}_{short-run} - \underbrace{u_{i0}}_{long-run} + \underbrace{e_{it}}_{noise}$$
(1)

CKMV

$$y_{it} = \beta_0 + x'_{it}\beta + \eta_i + \epsilon_{it}$$

Filippini & Greene (2014)

$$y_{it} = \alpha + \beta' x_{it} + \delta_i + \epsilon_{it}$$

Compromise

$$y_{it} = \alpha + \beta' x_{it} + \underbrace{\eta_i}_{w_i-?} + \underbrace{\epsilon_{it}}_{v_{it}-u_{it}}$$

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Three letter identifiers, (T = TRUE), (F = FALSE)
T/F
      random effect
T/F short-run inefficiency (time-varying)
T/F long-run inefficiency (persistent)
(Table 2)
TTT 4-components (CKMV)
TFT TRE (Greene 2005)
FTT
      Kumbhakar and Heshmati (1995)
TTF
       4-comp. without u_{it} \rightarrow 3 comp.
FFT
       Pitt&Lee1981. S&S1984. K1987. BC1988
FTF
       pooled SF
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Assumptions (A1-A4)

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Individuals i=1,2,...,n Periods t=1,2,...,T (A1a)
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"the two random variables"; label: $b_i - u_{i0} = \eta_i$ and $e_{it} - u_{it} = \epsilon_{it}$ are independent in probability.

Dimensions(!): T+1 (per individual), because ϵ has dimension T, and η has dimension 1. See this from the fact that "the full unconditional log-likelihood function for this model [is] based on the joint distribution of" the vector $(\epsilon_{i1},...\epsilon_{iT},\eta_i)'$ (Filippini & Greene 2014: 10)

(A1b)

these random vectors (here: over n; (!) label as N) are independent in probability:

$$egin{array}{ll} m{b}_i & N imes 1 \\ m{u}_{i0} & N imes 1 \\ m{(u}_{i1}, ... m{u}_{iT}) & N imes T \\ m{(e}_{i1}, ... m{e}_{iT}) & N imes T \end{array}$$

$$\eta_i = b_i - u_{i0}$$
 $\epsilon_i = \mathbf{e}_i - \mathbf{u}_i$

(A2)

$$u_{i0} \sim N^{+}(0, \sigma_{1u}^{2})$$

 $b_{i} \sim N(0, \sigma_{b}^{2})$

(A3)

$$u_{it} \sim N^+(0, \sigma_{2u,t}^2)$$

 $e_{it} \sim N(0, \sigma_e^2)$

(A4)

The x'_{it} are vectors of exogenous variables.

Notation

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\mathbf{1}_{T} T \times 1 vector of ones
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$$\mathbf{0}_{T}$$
 $T \times 1$ vector of zeros

$$I_T$$
 $T \times T$ identity matrix

$$y_i$$
 $T \times 1$ vector of output

$$X_i$$
 $T \times p$ matrix of inputs

$$oldsymbol{u}_i \quad (T+1) imes 1$$
 inefficiency vector, i.e. $(oldsymbol{u}_{i0}, oldsymbol{u}_{i1}, ... oldsymbol{u}_{iT})$

 e_i $T \times 1$ vector of noise

Model

$$\mathbf{y}_i = \mathbf{1}_T(\beta_0 + b_i) + \mathbf{X}_i \boldsymbol{\beta} + \mathbf{A} \mathbf{u}_i + \mathbf{e}_i$$

with
$$A = -[\mathbf{1}_T \ \mathbf{I}_T]$$

Matrix **A** is $T \times (T+1)$; e.g. T=3

$$m{A} = - \left[egin{array}{cccc} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{array}
ight]_{3 imes 4} = \left[egin{array}{cccc} -1 & -1 & -0 & -0 \\ -1 & -0 & -1 & -0 \\ -1 & -0 & -0 & -1 \end{array}
ight]$$

Section 3: joint density function of the random components

$$\mathbf{1}_T b_i + \mathbf{A} \mathbf{u}_i + \mathbf{e}_i$$

Statistical properties

 $\phi_{q}\left(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\Omega}
ight)$ pdf of q-dimensional normal RV

 $\bar{\Phi}_q(\mu,\Omega)$ is the probability that a q-variate normal RV belongs to the positive orthant(?)

CSN definitions

Random vector ${\it z}$ has a (o,q) CSN distribution with parameters $\mu,\Gamma,{\it D},\nu,\Delta$ and pdf:

$$f(\mathbf{z}, \boldsymbol{\mu}, \boldsymbol{\Gamma}, \boldsymbol{D}, \boldsymbol{\nu}, \boldsymbol{\Delta}, o, q) = \frac{\phi_o(\mathbf{z}, \boldsymbol{\mu}, \boldsymbol{\Gamma}) \,\bar{\Phi}_q(\boldsymbol{D}(\mathbf{z} - \boldsymbol{\mu}) - \boldsymbol{\nu}, \boldsymbol{\Delta})}{\bar{\Phi}_q(-\boldsymbol{\nu}, \boldsymbol{D} + \boldsymbol{D}\boldsymbol{\Gamma}\boldsymbol{D'})}$$
(2)

Moment-generating function of z

$$E\left(\exp\{\mathbf{t}'\mathbf{z}\}\right) = \frac{\bar{\Phi}_q\left(\mathbf{D}\Gamma\mathbf{t} - \boldsymbol{\nu}, \boldsymbol{\Delta} + \mathbf{D}\Gamma\mathbf{D}'\right)}{\bar{\Phi}_q\left(-\boldsymbol{\nu}, \boldsymbol{\Delta} + \mathbf{D}\Gamma\mathbf{D}'\right)} \exp\{\mathbf{t}'\boldsymbol{\mu} + \frac{1}{2}\mathbf{t}'\Gamma\mathbf{t}\}$$
(3)

CSN in the TTT model

Relevant matrices

$$oldsymbol{V} = \left[egin{array}{cc} \sigma_{1u}^2 & oldsymbol{0}_{\mathcal{T}} \ oldsymbol{0}_{\mathcal{T}} & oldsymbol{\Psi} \end{array}
ight]_{(\mathcal{T}+1) imes(\mathcal{T}+1)}$$

"...where Ψ is the diagonal matrix with the variances $\sigma_{2u,t}^2$ (t=1,2,...,T) on the main diagonal."

$$\Psi = \begin{bmatrix} \sigma_{2u,1}^2 & 0 & 0 \\ 0 & \sigma_{2u,2}^2 & 0 \\ 0 & 0 & \sigma_{2u,3}^2 \end{bmatrix}_{3\times 3}$$

$$oldsymbol{\Sigma} = \sigma_{\mathrm{e}}^2 oldsymbol{I}_T + \sigma_{b}^2 oldsymbol{1}_T oldsymbol{1}_T'$$

$$egin{aligned} oldsymbol{\Lambda} &= oldsymbol{V} - oldsymbol{V} oldsymbol{A}' \left(oldsymbol{\Sigma} + oldsymbol{A} oldsymbol{V} oldsymbol{A}' oldsymbol{\Delta}'^1 oldsymbol{A}
ight)^{-1} oldsymbol{A} \ &= \left(oldsymbol{V}^{-1} + oldsymbol{A}' oldsymbol{\Sigma}^{-1} oldsymbol{A}
ight)^{-1} \end{aligned}$$

 $\dim (T+1) \times (T+1)$

$$extbf{\emph{R}} = extbf{\emph{VA'}} ig(\Sigma + extbf{\emph{AVA'}} ig)^{-1} \ = \Lambda extbf{\emph{A'}} \Sigma^{-1}$$

$$\dim (T+1) \times (T)$$

Examples, T=3

$$oldsymbol{V} = \left[egin{array}{cccc} \sigma_{1u}^2 & 0 & 0 & 0 \ 0 & \sigma_{2u,1}^2 & 0 & 0 \ 0 & 0 & \sigma_{2u,2}^2 & 0 \ 0 & 0 & 0 & \sigma_{2u,3}^2 \end{array}
ight]_{4 imes4}$$

$$\Sigma = \begin{bmatrix} \sigma_e^2 & 0 & 0 \\ 0 & \sigma_e^2 & 0 \\ 0 & 0 & \sigma_e^2 \end{bmatrix} + \begin{bmatrix} \sigma_b^2 & \sigma_b^2 & \sigma_b^2 \\ \sigma_b^2 & \sigma_b^2 & \sigma_b^2 \end{bmatrix}$$
$$= \begin{bmatrix} \sigma_e^2 + \sigma_b^2 & \sigma_e^2 + \sigma_b^2 & \sigma_b^2 \\ \sigma_b^2 & \sigma_e^2 + \sigma_b^2 & \sigma_b^2 \end{bmatrix}_{3\times3}$$

$$\begin{split} & \pmb{\Lambda} = \Big(\left[\begin{array}{cccc} \sigma_{1u}^2 & 0 & 0 & 0 & 0 \\ 0 & \sigma_{2u,1}^2 & 0 & 0 & 0 \\ 0 & 0 & \sigma_{2u,2}^2 & 0 & 0 \\ 0 & 0 & 0 & \sigma_{2u,3}^2 & 0 \\ \end{array} \right]^{-1} \\ & + \left[\begin{array}{ccccc} -1 & -1 & -1 & -1 & \sigma_{2u,2} & \sigma_{2u,3} \\ -1 & -0 & -0 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 \\ -1 & -0 & -0 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 \\ -1 & -1 & -1 & -1 & -0 & -0 \\ -1 & -0 & -1 & -0 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 & \sigma_{2u,3}^2 \\ -1 & -1 & -1 & -1 & -0 & -0 \\ -1 & -1 & -0 & -1 & -0 \\ -1 & -0 & -0 & -1 & 1 \\ \end{bmatrix} \Big)^{-1} \end{split}$$

$$\mathbf{R} = \mathbf{\Lambda} \begin{bmatrix} -1 & -1 & -1 \\ -1 & -0 & -0 \\ -0 & -1 & -0 \\ -0 & -0 & -1 \end{bmatrix} \begin{bmatrix} \sigma_e^2 + \sigma_b^2 & \sigma_b^2 & \sigma_b^2 \\ \sigma_b^2 & \sigma_e^2 + \sigma_b^2 & \sigma_b^2 \\ \sigma_b^2 & \sigma_b^2 & \sigma_e^2 + \sigma_b^2 \end{bmatrix}^{-1}$$

Proposition 1

"... y_i has a (T, T+1) CSN distribution with the parameters:

Location
$$\mu = \mathbf{1}_T \beta_0 + \mathbf{X}_i \boldsymbol{\beta}$$

Scale
$$\Gamma = \Sigma + extit{AVA'}$$

Skewness
$$D = R$$

Mean in
$$\bar{\Phi}$$
 $\nu=0$

Covariance in
$$ar{\Phi}$$
 $oldsymbol{\Delta}=oldsymbol{\Lambda}$

Examples?

Conditional density of y_i

$$f(\mathbf{y}_{i}) = \phi_{T} \left(\mathbf{y}_{i}; \, \mathbf{1}_{T} \beta_{0} + \mathbf{X}_{i} \boldsymbol{\beta}, \, \boldsymbol{\Sigma} + \mathbf{AVA'} \right)$$

$$\frac{\bar{\Phi}_{T+1} \left(\mathbf{R} \left(\mathbf{y}_{i} - \mathbf{X}_{i} \boldsymbol{\beta} - \mathbf{1}_{T} \beta_{0} \right), \, \boldsymbol{\Lambda} \right)}{2^{-(T+1)}}$$

Note that

$$\mathbf{y}_i - \mathbf{X}_i \boldsymbol{\beta} - \mathbf{1}_T \beta_0 = \mathbf{1}_T b_i + \mathbf{A} \mathbf{u}_i + \mathbf{e}_i$$

= $\eta_i + \boldsymbol{\epsilon}_i$

Moment-generating function (dimensions t and k?)

$$E\left(\exp\{t'\mathbf{y}_i\}\right) = \frac{\bar{\Phi}_{(T+1)}\left(\mathbf{V}^{\frac{1}{2}}\mathbf{A}'\mathbf{t},\mathbf{I}\right)}{2^{-(T+1)}}\exp\{t'(\mathbf{1}_{T}\beta_0 + \mathbf{X}_i\beta) + \frac{1}{2}\mathbf{t}'(\mathbf{\Sigma} + \mathbf{AVA}')\mathbf{t}\}\tag{4}$$

shows that these parameters can be identified:

$$-m{AV}^{rac{1}{2}}=\left[\sigma_{1u}m{1}_{T}\ m{\Psi}^{rac{1}{2}}
ight]$$
 $m{\Sigma}+m{AVA}'=m{\Gamma} \qquad (ext{scale pdf})_{T imes T}$ $\hat{m{y}}_{i}=m{1}_{T}eta_{0}+m{X}_{i}m{eta}$

Proposition 2

Log-likelihood function of NT observations (following from assumption A1b and Proposition 1) of the N independent CSN RVs $(y_i - X_i\beta)$.

$$ln L = N(T + 1) ln(2)$$

$$+ \sum_{i} \left[ln \phi_{T} (\mathbf{y}_{i} - \mathbf{X}_{i}\beta; \mathbf{1}_{T}\beta_{0}, \Sigma + \mathbf{A}V\mathbf{A}') \right]$$

$$+ \sum_{i} \left[ln \bar{\Phi}_{T+1} (\mathbf{R} (\mathbf{y}_{i} - \mathbf{X}_{i}\beta - \mathbf{1}_{T}\beta_{0}); \mathbf{\Lambda}) \right]$$
(5)

$$\ln L = \textit{N}(\textit{T}+1) \; \ln \left(2\right) + \sum_{i} \; \left[\ln \phi_{\textit{T}} \left(\eta_{i} + \epsilon_{i} + \mathbf{1}_{\textit{T}} \beta_{0}; \; \mathbf{1}_{\textit{T}} \beta_{0}, \boldsymbol{\Sigma} + \textit{AVA}' \right) \right] \\ + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right) \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\textit{R} \left(\eta_{i} + \epsilon_{i} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left(\ln \bar{\Phi}_{\textit{T}+1} \right); \; \boldsymbol{\Lambda} \right] \right] + \sum_{i} \left[\ln \bar{\Phi}_{\textit{T}+1} \left$$

Multivariate normal pdf, p > 1

$$\phi = f(v) = (2\pi)^{-\frac{\rho}{2}} |\Sigma|^{-\frac{1}{2}} \exp\left\{-\left(\frac{1}{2}\right) (v - \mu)' \Sigma^{-1} (v - \mu)\right\}$$

$$\ln \phi = \ln f(v) = p\left(-\frac{1}{2}\right) \ln (2\pi) - \frac{1}{2} \ln |\Sigma| \left\{-\left(\frac{1}{2}\right) (v - \mu)' \Sigma^{-1} (v - \mu)\right\}$$

log pdf of Proposition 2

$$\ln \phi_T (\underbrace{\eta_i + \epsilon_i + \mathbf{1}_T \beta_0}_{label \ as \ w_i}; \ \mathbf{1}_T \beta_0, \ \Sigma + AVA')$$

$$\ln \phi_T(w_i) = T\left(-\frac{1}{2}\right) \ln (2\pi) - \frac{1}{2} \ln |\Gamma| \left\{-\left(\frac{1}{2}\right) (w_i - \mu)' \Gamma^{-1} (w_i - \mu)\right\}$$

Note:
$$w_i - \mu = (\eta_i + \epsilon_i + \mathbf{1}_T \beta_0) - \mathbf{1}_T \beta_0$$

CDF

$$ar{\Phi}_{T+1}\left(m{R}\left(m{y}_i-m{X}_im{eta}-m{1}_Teta_0
ight);m{\Lambda}
ight)$$
 $m{R}(m{y}_i-m{X}_im{eta}-m{1}_Teta_0) \qquad ((T+1) imes T)(T imes 1) o \qquad (T+1) imes 1$
 $ar{\Phi}_{T+1}\left(.
ight) \qquad 1 imes 1$