Initial latent states:
$$\underbrace{\begin{bmatrix} \text{eta1} \\ \text{eta2} \end{bmatrix}}_{\boldsymbol{\eta}(t_0)} (t_0) \sim \text{N} \underbrace{\begin{bmatrix} 3.354 \\ 4.33 \end{bmatrix}}_{\text{TOMEANS}}, \underbrace{\begin{bmatrix} 1.913 & -0.054 \\ -0.054 & 0.002 \end{bmatrix}}_{\text{TOVAR}}$$
Deterministic change:
$$\underbrace{d \begin{bmatrix} \text{eta1} \\ \text{eta2} \end{bmatrix}}_{\mathbf{d}\boldsymbol{\eta}(t)} (t) = \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}}_{\mathbf{DRIFT}} \underbrace{\begin{bmatrix} \text{eta1} \\ \text{eta2} \end{bmatrix}}_{\boldsymbol{\eta}(t)} (t) + \underbrace{\begin{bmatrix} 0.467 \\ -0.011 \end{bmatrix}}_{\mathbf{CINT}} dt + \underbrace{\begin{bmatrix} 0.467 \\ -0.011 \end{bmatrix}}_{\mathbf{DRIFT}}$$

 $\begin{bmatrix} 3.354 \\ 4.33 \\ 0.467 \\ -0.011 \end{bmatrix}, \begin{bmatrix} 1.913 & -0.054 & 1.22 & -0.036 \\ -0.054 & 0.002 & -0.033 & 0.001 \\ 1.22 & -0.033 & 1.17 & -0.034 \\ -0.036 & 0.001 & -0.034 & 0.001 \end{bmatrix}$

$$\mathbf{G}_{\text{DIFFUSION}} \text{ } \mathbf{dW}(t)$$
Observations:
$$\underbrace{\begin{bmatrix} \mathbf{y}^1 \\ \mathbf{y}^2 \\ \mathbf{y}^3 \end{bmatrix}}_{\mathbf{Y}(t)} (t) = \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & -34.971 \end{bmatrix}}_{\mathbf{\eta}(t)} \underbrace{\begin{bmatrix} \text{eta1} \\ \text{eta2} \end{bmatrix}}_{\mathbf{\eta}(t)} (t) + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 157.498 \end{bmatrix}}_{\mathbf{T}} + \underbrace{\begin{bmatrix} 0.409 & 0 & 0 \\ 0 & 5.309 & 0 \\ 0 & 0 & 0.536 \end{bmatrix}}_{\mathbf{\Theta}} \underbrace{\begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \end{bmatrix}}_{\mathbf{\epsilon}(t)} (t)$$

MANIFESTVAR

Subject parameter distribution:

Random change:

 $\phi(i)$

Latent noise per time step : $\Delta \left[W_{j \in [1,2]} \right] (t-u) \sim \mathcal{N}(0,t-u)$ Observation noise: $\left[\epsilon_{j \in [1,2]} \right] (t) \sim \mathcal{N}(0,1)$

Linearised approximation of subject parameter distribution shown.

Indivividual specific notation (subscript i) only shown for subject parameter distribution – pop. means shown elsewhere. cholsdcor converts lower tri matrix of std dev and unconstrained correlation to Cholesky factor covariance.

covsdcor = transposed cross product of cholsdcor, to give covariance. See Driver & Voelkle (2018) p11.