

# Package ‘simStateSpace’

November 9, 2023

**Title** Simulate Data from State Space Models

**Version** 1.0.0

**Description** Offers an efficient and user-friendly framework  
for simulating data in state space models  
where n subjects/units is greater than one  
which is common in social and behavioral sciences.

**URL** <https://github.com/jeksterslab/simStateSpace>,  
<https://jeksterslab.github.io/simStateSpace/>

**BugReports** <https://github.com/jeksterslab/simStateSpace/issues>

**License** GPL (>= 3)

**Encoding** UTF-8

**Roxygen** list(markdown = TRUE)

**Depends** R (>= 3.0.0)

**LinkingTo** Rcpp, RcppArmadillo

**Imports** Rcpp

**Suggests** knitr, rmarkdown, testthat, Matrix

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**NeedsCompilation** yes

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## R topics documented:

OU2SSM	2
Sim2Matrix	3
SimSSM0	5
SimSSM0Fixed	7
SimSSM0Vary	10
SimSSMOU	12

SimSSMOUFixed . . . . .	15
SimSSMOUVary . . . . .	18
SimSSMVAR . . . . .	20
SimSSMVARFixed . . . . .	22
SimSSMVARVary . . . . .	24

<b>Index</b>	<b>27</b>
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OU2SSM	<i>Convert Parameters from the Ornstein–Uhlenbeck Model to State Space Model Parameterization</i>
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---

## Description

This function converts parameters from the Ornstein–Uhlenbeck model to state space model parameterization. See details for more information.

## Usage

```
OU2SSM(mu, phi, sigma_sqrt, delta_t)
```

## Arguments

mu	Numeric vector. The long-term mean or equilibrium level ( $\mu$ ).
phi	Numeric matrix. The rate of mean reversion, determining how quickly the variable returns to its mean ( $\Phi$ ).
sigma_sqrt	Numeric matrix. Cholesky decomposition of the matrix of volatility or randomness in the process ( $\Sigma$ ).
delta_t	Numeric. Time interval ( $\delta_t$ ).

## Details

The state space parameters as a function of the Ornstein–Uhlenbeck model parameters are given by

$$\beta = \exp(-\Phi \Delta_t)$$

$$\alpha = -\Phi^{-1}(\beta - \mathbf{I}_p)$$

$$\text{vec}(\Psi) = \{ [(-\Phi \otimes \mathbf{I}_p) + (\mathbf{I}_p \otimes -\Phi)] [\exp \{ [(-\Phi \otimes \mathbf{I}_p) + (\mathbf{I}_p \otimes -\Phi)] \Delta_t \} - \mathbf{I}_{p \times p}] \text{vec}(\Sigma) \}$$

## Author(s)

Ivan Jacob Agaloos Pesigan

**See Also**

Other Simulation of State Space Models Data Functions: [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSMOUFixed\(\)](#), [SimSSMOUVary\(\)](#), [SimSSMOU\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

**Examples**

```
p <- k <- 2
mu <- c(5.76, 5.18)
phi <- matrix(data = c(0.10, -0.05, -0.05, 0.10), nrow = p)
sigma_sqrt <- chol(
  matrix(data = c(2.79, 0.06, 0.06, 3.27), nrow = p)
)
delta_t <- 0.10

OU2SSM(
  mu = mu,
  phi = phi,
  sigma_sqrt = sigma_sqrt,
  delta_t = delta_t
)
```

---

Sim2Matrix

---

*Simulation Output to Matrix*


---

**Description**

This function converts the output of [SimSSM0\(\)](#), [SimSSMOU\(\)](#), [SimSSMVAR\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSMOUFixed\(\)](#), or [SimSSMVARFixed\(\)](#) to a matrix.

**Usage**

```
Sim2Matrix(x, eta = FALSE)
```

**Arguments**

x	R object. Output of <a href="#">SimSSM0()</a> , <a href="#">SimSSMOU()</a> , <a href="#">SimSSMVAR()</a> , <a href="#">SimSSM0Fixed()</a> , <a href="#">SimSSMOUFixed()</a> , or <a href="#">SimSSMVARFixed()</a> .
eta	Logical. If eta = TRUE, include eta. If eta = FALSE, exclude eta.

**Author(s)**

Ivan Jacob Agaloos Pesigan

**See Also**

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSMOUFixed\(\)](#), [SimSSMOUVary\(\)](#), [SimSSMOU\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

**Examples**

```
# prepare parameters
set.seed(42)
k <- p <- 3
I <- diag(k)
I_sqrt <- chol(I)
null_vec <- rep(x = 0, times = k)
n <- 5
mu0 <- null_vec
sigma0_sqrt <- I_sqrt
alpha <- null_vec
beta <- diag(x = 0.50, nrow = k)
psi_sqrt <- I_sqrt
nu <- null_vec
lambda <- I
theta_sqrt <- chol(diag(x = 0.50, nrow = k))
time <- 50
burn_in <- 0

# generate data
ssm <- SimSSM0(
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  time = time,
  burn_in = burn_in
)

# list to matrix
mat <- Sim2Matrix(ssm)
str(mat)
head(mat)

# generate data
ssm <- SimSSM0Fixed(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
```

```

    nu = nu,
    lambda = lambda,
    theta_sqrt = theta_sqrt,
    time = time,
    burn_in = burn_in
)

# list to matrix
mat <- Sim2Matrix(ssm)
str(mat)
head(mat)

```

SimSSM0

*Simulate Data from a State Space Model (n = 1)***Description**

This function simulates data from a state space model. See details for more information.

**Usage**

```

SimSSM0(
  mu0,
  sigma0_sqrt,
  alpha,
  beta,
  psi_sqrt,
  nu,
  lambda,
  theta_sqrt,
  time,
  burn_in
)

```

**Arguments**

mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
alpha	Numeric vector. Vector of intercepts for the dynamic model ( $\alpha$ ).
beta	Numeric matrix. Transition matrix relating the values of the latent variables at time $t - 1$ to those at time $t$ ( $\beta$ ).
psi_sqrt	Numeric matrix. Cholesky decomposition of the process noise covariance matrix ( $\Psi$ ).
nu	Numeric vector. Vector of intercepts for the measurement model ( $\nu$ ).

lambda	Numeric matrix. Factor loading matrix linking the latent variables to the observed variables ( $\Lambda$ ).
theta_sqrt	Numeric matrix. Cholesky decomposition of the measurement error covariance matrix ( $\Theta$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Details

The measurement model is given by

$$\mathbf{y}_t = \boldsymbol{\nu} + \Lambda \boldsymbol{\eta}_t + \boldsymbol{\varepsilon}_t \quad \text{with} \quad \boldsymbol{\varepsilon}_t \sim \mathcal{N}(\mathbf{0}, \Theta)$$

where  $\mathbf{y}_t$ ,  $\boldsymbol{\eta}_t$ , and  $\boldsymbol{\varepsilon}_t$  are random variables and  $\boldsymbol{\nu}$ ,  $\Lambda$ , and  $\Theta$  are model parameters.  $\mathbf{y}_t$  is a vector of observed random variables at time  $t$ ,  $\boldsymbol{\eta}_t$  is a vector of latent random variables at time  $t$ , and  $\boldsymbol{\varepsilon}_t$  is a vector of random measurement errors at time  $t$ , while  $\boldsymbol{\nu}$  is a vector of intercept,  $\Lambda$  is a matrix of factor loadings, and  $\Theta$  is the covariance matrix of  $\boldsymbol{\varepsilon}$ .

The dynamic structure is given by

$$\boldsymbol{\eta}_t = \boldsymbol{\alpha} + \beta \boldsymbol{\eta}_{t-1} + \boldsymbol{\zeta}_t \quad \text{with} \quad \boldsymbol{\zeta}_t \sim \mathcal{N}(\mathbf{0}, \Psi)$$

where  $\boldsymbol{\eta}_t$ ,  $\boldsymbol{\eta}_{t-1}$ , and  $\boldsymbol{\zeta}_t$  are random variables and  $\boldsymbol{\alpha}$ ,  $\beta$ , and  $\Psi$  are model parameters.  $\boldsymbol{\eta}_t$  is a vector of latent variables at time  $t$ ,  $\boldsymbol{\eta}_{t-1}$  is a vector of latent variables at time  $t - 1$ , and  $\boldsymbol{\zeta}_t$  is a vector of dynamic noise at time  $t$  while  $\boldsymbol{\alpha}$  is a vector of intercepts,  $\beta$  is a matrix of autoregression and cross regression coefficients, and  $\Psi$  is the covariance matrix of  $\boldsymbol{\zeta}_t$ .

### Value

Returns a list with the following elements:

- y: A t by k matrix of values for the manifest variables.
- eta: A t by p matrix of values for the latent variables.
- time: A vector of discrete time points from 1 to t.
- n: Number of individuals.

### Author(s)

Ivan Jacob Agaloos Pesigan

### References

Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications: With R examples*. Springer International Publishing. doi:10.1007/9783319524528

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0UFixed\(\)](#), [SimSSM0UVary\(\)](#), [SimSSMOU\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

**Examples**

```

# prepare parameters
set.seed(42)
k <- p <- 3
I <- diag(k)
I_sqrt <- chol(I)
null_vec <- rep(x = 0, times = k)
mu0 <- null_vec
sigma0_sqrt <- I_sqrt
alpha <- null_vec
beta <- diag(x = 0.50, nrow = k)
psi_sqrt <- I_sqrt
nu <- null_vec
lambda <- I
theta_sqrt <- chol(diag(x = 0.50, nrow = k))
time <- 50
burn_in <- 0

# generate data
ssm <- SimSSM0(
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  time = time,
  burn_in = burn_in
)

str(ssm)

```

---

SimSSM0Fixed

---

*Simulate Data using a State Space Model Parameterization for  $n > 1$  Individuals (Fixed Parameters)*


---

**Description**

This function simulates data using a state space model parameterization for  $n > 1$  individuals. In this model, the parameters are invariant across individuals.

**Usage**

```

SimSSM0Fixed(
  n,
  mu0,

```

```

    sigma0_sqrt,
    alpha,
    beta,
    psi_sqrt,
    nu,
    lambda,
    theta_sqrt,
    time,
    burn_in
)

```

### Arguments

n	Positive integer. Number of individuals.
mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
alpha	Numeric vector. Vector of intercepts for the dynamic model ( $\alpha$ ).
beta	Numeric matrix. Transition matrix relating the values of the latent variables at time $t - 1$ to those at time $t$ ( $\beta$ ).
psi_sqrt	Numeric matrix. Cholesky decomposition of the process noise covariance matrix ( $\Psi$ ).
nu	Numeric vector. Vector of intercepts for the measurement model ( $\nu$ ).
lambda	Numeric matrix. Factor loading matrix linking the latent variables to the observed variables ( $\Lambda$ ).
theta_sqrt	Numeric matrix. Cholesky decomposition of the measurement error covariance matrix ( $\Theta$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Details

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda}\boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t} \quad \text{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}(\mathbf{0}, \boldsymbol{\Theta})$$

where  $\mathbf{y}_{i,t}$ ,  $\boldsymbol{\eta}_{i,t}$ , and  $\boldsymbol{\varepsilon}_{i,t}$  are random variables and  $\boldsymbol{\nu}$ ,  $\boldsymbol{\Lambda}$ , and  $\boldsymbol{\Theta}$  are model parameters.  $\mathbf{y}_{i,t}$  is a vector of observed random variables at time  $t$  and individual  $i$ ,  $\boldsymbol{\eta}_{i,t}$  is a vector of latent random variables at time  $t$  and individual  $i$ , and  $\boldsymbol{\varepsilon}_{i,t}$  is a vector of random measurement errors at time  $t$  and individual  $i$ , while  $\boldsymbol{\nu}$  is a vector of intercept,  $\boldsymbol{\Lambda}$  is a matrix of factor loadings, and  $\boldsymbol{\Theta}$  is the covariance matrix of  $\boldsymbol{\varepsilon}$ .

The dynamic structure is given by

$$\boldsymbol{\eta}_{i,t} = \boldsymbol{\alpha} + \boldsymbol{\beta}\boldsymbol{\eta}_{i,t-1} + \boldsymbol{\zeta}_{i,t} \quad \text{with} \quad \boldsymbol{\zeta}_{i,t} \sim \mathcal{N}(\mathbf{0}, \boldsymbol{\Psi})$$



where  $\eta_{i,t}$ ,  $\eta_{i,t-1}$ , and  $\zeta_{i,t}$  are random variables and  $\alpha$ ,  $\beta$ , and  $\Psi$  are model parameters.  $\eta_{i,t}$  is a vector of latent variables at time  $t$  and individual  $i$ ,  $\eta_{i,t-1}$  is a vector of latent variables at time  $t - 1$  and individual  $i$ , and  $\zeta_{i,t}$  is a vector of dynamic noise at time  $t$  and individual  $i$  while  $\alpha$  is a vector of intercepts,  $\beta$  is a matrix of autoregression and cross regression coefficients, and  $\Psi$  is the covariance matrix of  $\zeta_{i,t}$ .

### Value

Returns a list of length n. Each element is a list with the following elements:

- y: A t by k matrix of values for the manifest variables.
- eta: A t by p matrix of values for the latent variables.
- time: A vector of discrete time points from 1 to t.
- id: A vector of ID numbers of length t.
- n: Number of individuals.

### Author(s)

Ivan Jacob Agaloos Pesigan

### References

Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications: With R examples*. Springer International Publishing. doi:10.1007/9783319524528

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSM0UFixed\(\)](#), [SimSSM0UVary\(\)](#), [SimSSM0U\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

### Examples

```
# prepare parameters
set.seed(42)
k <- p <- 3
I <- diag(k)
I_sqrt <- chol(I)
null_vec <- rep(x = 0, times = k)
n <- 5
mu0 <- null_vec
sigma0_sqrt <- I_sqrt
alpha <- null_vec
beta <- diag(x = 0.50, nrow = k)
psi_sqrt <- I_sqrt
nu <- null_vec
lambda <- I
theta_sqrt <- chol(diag(x = 0.50, nrow = k))
time <- 50
burn_in <- 0
```

```
# generate data
ssm <- SimSSM0Fixed(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  time = time,
  burn_in = burn_in
)

str(ssm)
```

---

SimSSM0Vary

---

*Simulate Data using a State Space Model Parameterization for  $n > 1$  Individuals (Varying Parameters)*


---

### Description

This function simulates data using a state space model parameterization for  $n > 1$  individuals. In this model, the parameters can vary across individuals.

### Usage

```
SimSSM0Vary(
  n,
  mu0,
  sigma0_sqrt,
  alpha,
  beta,
  psi_sqrt,
  nu,
  lambda,
  theta_sqrt,
  time,
  burn_in
)
```

### Arguments

n	Positive integer. Number of individuals.
mu0	List of numeric vectors. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).

sigma0_sqrt	List of numeric matrices. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
alpha	List of numeric vectors. Vector of intercepts for the dynamic model ( $\alpha$ ).
beta	List of numeric matrices. Transition matrix relating the values of the latent variables at time $t - 1$ to those at time $t$ ( $\beta$ ).
psi_sqrt	List of numeric matrices. Cholesky decomposition of the process noise covariance matrix ( $\Psi$ ).
nu	List of numeric vectors. Vector of intercepts for the measurement model ( $\nu$ ).
lambda	List of numeric matrices. Factor loading matrix linking the latent variables to the observed variables ( $\Lambda$ ).
theta_sqrt	List of numeric matrices. Cholesky decomposition of the measurement error covariance matrix ( $\Theta$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Details

Parameters can vary across individuals by providing a list of parameter values. If the length of any of the parameters (mu0, sigma0\_sqrt, alpha, beta, psi\_sqrt, nu, lambda, and theta\_sqrt) is less than  $n$ , the function will cycle through the available values.

### Value

Returns a list of length  $n$ . Each element is a list with the following elements:

- $y$ : A  $t$  by  $k$  matrix of values for the manifest variables.
- $\eta$ : A  $t$  by  $p$  matrix of values for the latent variables.
- $time$ : A vector of discrete time points from 1 to  $t$ .
- $id$ : A vector of ID numbers of length  $t$ .
- $n$ : Number of individuals.

### Author(s)

Ivan Jacob Agaloos Pesigan

### References

Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications: With R examples*. Springer International Publishing. doi:10.1007/9783319524528

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0\(\)](#), [SimSSM0UFixed\(\)](#), [SimSSM0UVary\(\)](#), [SimSSM0U\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

**Examples**

```

# prepare parameters
# In this example, beta varies across individuals
set.seed(42)
k <- p <- 3
iden <- diag(k)
iden_sqrt <- chol(iden)
null_vec <- rep(x = 0, times = k)
n <- 5
mu0 <- list(null_vec)
sigma0_sqrt <- list(iden_sqrt)
alpha <- list(null_vec)
beta <- list(
  diag(x = 0.1, nrow = k),
  diag(x = 0.2, nrow = k),
  diag(x = 0.3, nrow = k),
  diag(x = 0.4, nrow = k),
  diag(x = 0.5, nrow = k)
)
psi_sqrt <- list(iden_sqrt)
nu <- list(null_vec)
lambda <- list(iden)
theta_sqrt <- list(chol(diag(x = 0.50, nrow = k)))
time <- 50
burn_in <- 0

ssm <- SimSSM0Vary(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  time = time,
  burn_in = burn_in
)

str(ssm)

```

---

SimSSMOU

---

*Simulate Data from the Ornstein–Uhlenbeck Model using a State Space Model Parameterization ( $n = 1$ )*


---

**Description**

This function simulates data from the Ornstein–Uhlenbeck model using a state space model parameterization. See details for more information.

**Usage**

```

SimSSMOU(
  mu0,
  sigma0_sqrt,
  mu,
  phi,
  sigma_sqrt,
  nu,
  lambda,
  theta_sqrt,
  delta_t,
  time,
  burn_in
)

```

**Arguments**

mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
mu	Numeric vector. The long-term mean or equilibrium level ( $\mu$ ).
phi	Numeric matrix. The rate of mean reversion, determining how quickly the variable returns to its mean ( $\Phi$ ).
sigma_sqrt	Numeric matrix. Cholesky decomposition of the matrix of volatility or randomness in the process ( $\Sigma$ ).
nu	Numeric vector. Vector of intercepts for the measurement model ( $\nu$ ).
lambda	Numeric matrix. Factor loading matrix linking the latent variables to the observed variables ( $\Lambda$ ).
theta_sqrt	Numeric matrix. Cholesky decomposition of the measurement error covariance matrix ( $\Theta$ ).
delta_t	Numeric. Time interval ( $\delta_t$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

**Details**

The measurement model is given by

$$\mathbf{y}_t = \boldsymbol{\nu} + \boldsymbol{\Lambda}\boldsymbol{\eta}_t + \boldsymbol{\varepsilon}_t \quad \text{with} \quad \boldsymbol{\varepsilon}_t \sim \mathcal{N}(\mathbf{0}, \boldsymbol{\Theta})$$

where  $\mathbf{y}_t$ ,  $\boldsymbol{\eta}_t$ , and  $\boldsymbol{\varepsilon}_t$  are random variables and  $\boldsymbol{\nu}$ ,  $\boldsymbol{\Lambda}$ , and  $\boldsymbol{\Theta}$  are model parameters.  $\mathbf{y}_t$  is a vector of observed random variables at time  $t$ ,  $\boldsymbol{\eta}_t$  is a vector of latent random variables at time  $t$ , and  $\boldsymbol{\varepsilon}_t$  is a vector of random measurement errors at time  $t$ , while  $\boldsymbol{\nu}$  is a vector of intercept,  $\boldsymbol{\Lambda}$  is a matrix of factor loadings, and  $\boldsymbol{\Theta}$  is the covariance matrix of  $\boldsymbol{\varepsilon}$ .

The dynamic structure is given by

$$d\eta_t = \Phi (\mu - \eta_t) dt + \Sigma^{\frac{1}{2}} dW_t$$

where  $\mu$  is the long-term mean or equilibrium level,  $\Phi$  is the rate of mean reversion, determining how quickly the variable returns to its mean,  $\Sigma$  is the matrix of volatility or randomness in the process, and  $dW$  is a Wiener process or Brownian motion, which represents random fluctuations.

### Value

Returns a list with the following elements:

- y: A t by k matrix of values for the manifest variables.
- eta: A t by p matrix of values for the latent variables.
- time: A vector of continuous time points of length t starting from 0 with delta\_t increments.
- n: Number of individuals.

### Author(s)

Ivan Jacob Agaloos Pesigan

### References

Uhlenbeck, G. E., & Ornstein, L. S. (1930). On the theory of the brownian motion. *Physical Review*, 36(5), 823–841. doi:10.1103/physrev.36.823

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSM0UFixed\(\)](#), [SimSSM0UVary\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

### Examples

```
# prepare parameters
set.seed(42)
p <- k <- 2
I <- diag(p)
I_sqrt <- chol(I)
mu0 <- c(-3.0, 1.5)
sigma0_sqrt <- I_sqrt
mu <- c(5.76, 5.18)
phi <- matrix(data = c(0.10, -0.05, -0.05, 0.10), nrow = p)
sigma_sqrt <- chol(
  matrix(data = c(2.79, 0.06, 0.06, 3.27), nrow = p)
)
nu <- rep(x = 0, times = k)
lambda <- diag(k)
theta_sqrt <- chol(diag(x = 0.50, nrow = k))
delta_t <- 0.10
time <- 50
```

```

burn_in <- 0

# generate data
ssm <- SimSSMOU(
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  mu = mu,
  phi = phi,
  sigma_sqrt = sigma_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  delta_t = delta_t,
  time = time,
  burn_in = burn_in
)

str(ssm)

```

---

SimSSMOUFixed

*Simulate Data from an Ornstein–Uhlenbeck Model using a State Space Model Parameterization for  $n > 1$  Individuals (Fixed Parameters)*

---

## Description

This function simulates data from an Ornstein–Uhlenbeck model using a state space model parameterization for  $n > 1$  individuals. In this model, the parameters are invariant across individuals. See details for more information.

## Usage

```

SimSSMOUFixed(
  n,
  mu0,
  sigma0_sqrt,
  mu,
  phi,
  sigma_sqrt,
  nu,
  lambda,
  theta_sqrt,
  delta_t,
  time,
  burn_in
)

```

### Arguments

n	Positive integer. Number of individuals.
mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
mu	Numeric vector. The long-term mean or equilibrium level ( $\mu$ ).
phi	Numeric matrix. The rate of mean reversion, determining how quickly the variable returns to its mean ( $\Phi$ ).
sigma_sqrt	Numeric matrix. Cholesky decomposition of the matrix of volatility or randomness in the process ( $\Sigma$ ).
nu	Numeric vector. Vector of intercepts for the measurement model ( $\nu$ ).
lambda	Numeric matrix. Factor loading matrix linking the latent variables to the observed variables ( $\Lambda$ ).
theta_sqrt	Numeric matrix. Cholesky decomposition of the measurement error covariance matrix ( $\Theta$ ).
delta_t	Numeric. Time interval ( $\delta_t$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Details

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda}\boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t} \quad \text{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}(\mathbf{0}, \boldsymbol{\Theta})$$

where  $\mathbf{y}_{i,t}$ ,  $\boldsymbol{\eta}_{i,t}$ , and  $\boldsymbol{\varepsilon}_{i,t}$  are random variables and  $\boldsymbol{\nu}$ ,  $\boldsymbol{\Lambda}$ , and  $\boldsymbol{\Theta}$  are model parameters.  $\mathbf{y}_{i,t}$  is a vector of observed random variables at time  $t$  and individual  $i$ ,  $\boldsymbol{\eta}_{i,t}$  is a vector of latent random variables at time  $t$  and individual  $i$ , and  $\boldsymbol{\varepsilon}_{i,t}$  is a vector of random measurement errors at time  $t$  and individual  $i$ , while  $\boldsymbol{\nu}$  is a vector of intercept,  $\boldsymbol{\Lambda}$  is a matrix of factor loadings, and  $\boldsymbol{\Theta}$  is the covariance matrix of  $\boldsymbol{\varepsilon}$ .

The dynamic structure is given by

$$d\boldsymbol{\eta}_{i,t} = \boldsymbol{\Phi}(\boldsymbol{\mu} - \boldsymbol{\eta}_{i,t}) dt + \boldsymbol{\Sigma}^{\frac{1}{2}} d\mathbf{W}_{i,t}$$

where  $\boldsymbol{\mu}$  is the long-term mean or equilibrium level,  $\boldsymbol{\Phi}$  is the rate of mean reversion, determining how quickly the variable returns to its mean,  $\boldsymbol{\Sigma}$  is the matrix of volatility or randomness in the process, and  $d\mathbf{W}$  is a Wiener process or Brownian motion, which represents random fluctuations.

### Value

Returns a list of length n. Each element is a list with the following elements:

- y: A t by k matrix of values for the manifest variables.
- eta: A t by p matrix of values for the latent variables.
- time: A vector of continuous time points of length t starting from 0 with delta\_t increments.
- id: A vector of ID numbers of length t.
- n: Number of individuals.



**Author(s)**

Ivan Jacob Agaloos Pesigan

**References**

Uhlenbeck, G. E., & Ornstein, L. S. (1930). On the theory of the brownian motion. *Physical Review*, 36(5), 823–841. doi:[10.1103/physrev.36.823](https://doi.org/10.1103/physrev.36.823)

**See Also**

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSMOUVary\(\)](#), [SimSSMOU\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

**Examples**

```
# prepare parameters
set.seed(42)
p <- k <- 2
I <- diag(p)
I_sqrt <- chol(I)
n <- 5
mu0 <- c(-3.0, 1.5)
sigma0_sqrt <- I_sqrt
mu <- c(5.76, 5.18)
phi <- matrix(data = c(0.10, -0.05, -0.05, 0.10), nrow = p)
sigma_sqrt <- chol(
  matrix(data = c(2.79, 0.06, 0.06, 3.27), nrow = p)
)
nu <- rep(x = 0, times = k)
lambda <- diag(k)
theta_sqrt <- chol(diag(x = 0.50, nrow = k))
delta_t <- 0.10
time <- 50
burn_in <- 0

# generate data
ssm <- SimSSMOUFixed(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  mu = mu,
  phi = phi,
  sigma_sqrt = sigma_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  delta_t = delta_t,
  time = time,
  burn_in = burn_in
)
```

```
str(ssm)
```

---

SimSSMOUVary

---

*Simulate Data from an Ornstein–Uhlenbeck Model using a State Space Model Parameterization for  $n > 1$  Individuals (Varying Parameters)*


---

## Description

This function simulates data from an Ornstein–Uhlenbeck model using a state space model parameterization for  $n > 1$  individuals. In this model, the parameters can vary across individuals.

## Usage

```
SimSSMOUVary(
  n,
  mu0,
  sigma0_sqrt,
  mu,
  phi,
  sigma_sqrt,
  nu,
  lambda,
  theta_sqrt,
  delta_t,
  time,
  burn_in
)
```

## Arguments

n	Positive integer. Number of individuals.
mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
mu	List of numeric vectors. The long-term mean or equilibrium level ( $\mu$ ).
phi	List of numeric matrices. The rate of mean reversion, determining how quickly the variable returns to its mean ( $\Phi$ ).
sigma_sqrt	List of numeric matrices. Cholesky decomposition of the matrix of volatility or randomness in the process ( $\Sigma$ ).
nu	Numeric vector. Vector of intercepts for the measurement model ( $\nu$ ).
lambda	Numeric matrix. Factor loading matrix linking the latent variables to the observed variables ( $\Lambda$ ).
theta_sqrt	Numeric matrix. Cholesky decomposition of the measurement error covariance matrix ( $\Theta$ ).

delta_t	Numeric. Time interval ( $\delta_t$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Details

Parameters can vary across individuals by providing a list of parameter values. If the length of any of the parameters (`mu0`, `sigma0_sqrt`, `mu`, `phi`, `sigma_sqrt`, `nu`, `lambda`, `theta_sqrt`) is less than `n`, the function will cycle through the available values.

### Value

Returns a list of length `n`. Each element is a list with the following elements:

- `y`: A `t` by `k` matrix of values for the manifest variables.
- `eta`: A `t` by `p` matrix of values for the latent variables.
- `time`: A vector of discrete time points from 1 to `t`.
- `id`: A vector of ID numbers of length `t`.
- `n`: Number of individuals.

### Author(s)

Ivan Jacob Agaloos Pesigan

### References

Uhlenbeck, G. E., & Ornstein, L. S. (1930). On the theory of the brownian motion. *Physical Review*, 36(5), 823–841. doi:[10.1103/physrev.36.823](https://doi.org/10.1103/physrev.36.823)

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSMOUFixed\(\)](#), [SimSSMOU\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

### Examples

```
# prepare parameters
# In this example, phi varies across individuals
set.seed(42)
p <- k <- 2
iden <- diag(p)
iden_sqrt <- chol(iden)
n <- 5
mu0 <- list(c(-3.0, 1.5))
sigma0_sqrt <- list(iden_sqrt)
mu <- list(c(5.76, 5.18))
phi <- list(
```

```

as.matrix(Matrix::expm(diag(x = -0.1, nrow = k))),
as.matrix(Matrix::expm(diag(x = -0.2, nrow = k))),
as.matrix(Matrix::expm(diag(x = -0.3, nrow = k))),
as.matrix(Matrix::expm(diag(x = -0.4, nrow = k))),
as.matrix(Matrix::expm(diag(x = -0.5, nrow = k)))
)
sigma_sqrt <- list(
  chol(
    matrix(data = c(2.79, 0.06, 0.06, 3.27), nrow = p)
  )
)
nu <- list(rep(x = 0, times = k))
lambda <- list(diag(k))
theta_sqrt <- list(chol(diag(x = 0.50, nrow = k)))
delta_t <- 0.10
time <- 50
burn_in <- 0

ssm <- SimSSMOUVary(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  mu = mu,
  phi = phi,
  sigma_sqrt = sigma_sqrt,
  nu = nu,
  lambda = lambda,
  theta_sqrt = theta_sqrt,
  delta_t = delta_t,
  time = time,
  burn_in = burn_in
)

str(ssm)

```

---

SimSSMVAR

---

*Simulate Data from the Vector Autoregressive Model using a State Space Model Parameterization ( $n = 1$ )*


---

## Description

This function simulates data from the vector autoregressive model using a state space model parameterization. See details for more information.

## Usage

```
SimSSMVAR(mu0, sigma0_sqrt, alpha, beta, psi_sqrt, time, burn_in)
```

### Arguments

mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
alpha	Numeric vector. Vector of intercepts for the dynamic model ( $\alpha$ ).
beta	Numeric matrix. Transition matrix relating the values of the latent variables at time $t - 1$ to those at time $t$ ( $\beta$ ).
psi_sqrt	Numeric matrix. Cholesky decomposition of the process noise covariance matrix ( $\Psi$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Details

The measurement model is given by

$$\mathbf{y}_t = \boldsymbol{\eta}_t.$$

The dynamic structure is given by

$$\boldsymbol{\eta}_t = \boldsymbol{\alpha} + \boldsymbol{\beta}\boldsymbol{\eta}_{t-1} + \boldsymbol{\zeta}_t \quad \text{with} \quad \boldsymbol{\zeta}_t \sim \mathcal{N}(\mathbf{0}, \boldsymbol{\Psi})$$

where  $\boldsymbol{\eta}_t$ ,  $\boldsymbol{\eta}_{t-1}$ , and  $\boldsymbol{\zeta}_t$  are random variables and  $\boldsymbol{\alpha}$ ,  $\boldsymbol{\beta}$ , and  $\boldsymbol{\Psi}$  are model parameters.  $\boldsymbol{\eta}_t$  is a vector of latent variables at time  $t$ ,  $\boldsymbol{\eta}_{t-1}$  is a vector of latent variables at  $t - 1$ , and  $\boldsymbol{\zeta}_t$  is a vector of dynamic noise at time  $t$  while  $\boldsymbol{\alpha}$  is a vector of intercepts,  $\boldsymbol{\beta}$  is a matrix of autoregression and cross regression coefficients, and  $\boldsymbol{\Psi}$  is the covariance matrix of  $\boldsymbol{\zeta}_t$ .

### Value

Returns a list with the following elements:

- y: A  $t$  by  $k$  matrix of values for the manifest variables.
- eta: A  $t$  by  $p$  matrix of values for the latent variables.
- time: A vector of discrete time points from 1 to  $t$ .
- n: Number of individuals.

### References

Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications: With R examples*. Springer International Publishing. doi:10.1007/9783319524528

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSM0UFixed\(\)](#), [SimSSM0UVary\(\)](#), [SimSSM0U\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVARVary\(\)](#)

**Examples**

```
# prepare parameters
set.seed(42)
k <- 3
I <- diag(k)
I_sqrt <- chol(I)
null_vec <- rep(x = 0, times = k)
mu0 <- null_vec
sigma0_sqrt <- I_sqrt
alpha <- null_vec
beta <- diag(x = 0.5, nrow = k)
psi_sqrt <- I_sqrt
time <- 50
burn_in <- 0

# generate data
ssm <- SimSSMVAR(
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  time = time,
  burn_in = burn_in
)

str(ssm)
```

---

SimSSMVARFixed

---

*Simulate Data from a Vector Autoregressive Model using a State Space Model Parameterization for  $n > 1$  Individuals (Fixed Parameters)*


---

**Description**

This function simulates data from a vector autoregressive model using a state space model parameterization for  $n > 1$  individuals. In this model, the parameters are invariant across individuals.

**Usage**

```
SimSSMVARFixed(n, mu0, sigma0_sqrt, alpha, beta, psi_sqrt, time, burn_in)
```

**Arguments**

n	Positive integer. Number of individuals.
mu0	Numeric vector. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	Numeric matrix. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).

alpha	Numeric vector. Vector of intercepts for the dynamic model ( $\alpha$ ).
beta	Numeric matrix. Transition matrix relating the values of the latent variables at time $t - 1$ to those at time $t$ ( $\beta$ ).
psi_sqrt	Numeric matrix. Cholesky decomposition of the process noise covariance matrix ( $\Psi$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.

### Value

Returns a list of length  $n$ . Each element is a list with the following elements:

- $y$ : A  $t$  by  $k$  matrix of values for the manifest variables.
- $\eta$ : A  $t$  by  $p$  matrix of values for the latent variables.
- $time$ : A vector of discrete time points from 1 to  $t$ .
- $id$ : A vector of ID numbers of length  $t$ .
- $n$ : Number of individuals.

### Author(s)

Ivan Jacob Agaloos Pesigan

### References

Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications: With R examples*. Springer International Publishing. doi:10.1007/9783319524528

### See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSMOUFixed\(\)](#), [SimSSMOUVary\(\)](#), [SimSSMOU\(\)](#), [SimSSMVARVary\(\)](#), [SimSSMVAR\(\)](#)

### Examples

```
# prepare parameters
set.seed(42)
k <- 3
iden <- diag(k)
iden_sqrt <- chol(iden)
null_vec <- rep(x = 0, times = k)
n <- 5
mu0 <- null_vec
sigma0_sqrt <- iden_sqrt
alpha <- null_vec
beta <- diag(x = 0.5, nrow = k)
psi_sqrt <- iden_sqrt
time <- 50
```

```

burn_in <- 0

ssm <- SimSSMVARFixed(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  time = time,
  burn_in = burn_in
)

str(ssm)

```

---

SimSSMVARVary	<i>Simulate Data from a Vector Autoregressive Model using a State Space Model Parameterization for <math>n &gt; 1</math> Individuals (Varying Parameters)</i>
---------------	---

---

## Description

This function simulates data from a vector autoregressive model using a state space model parameterization for  $n > 1$  individuals. In this model, the parameters can vary across individuals.

## Usage

```
SimSSMVARVary(n, mu0, sigma0_sqrt, alpha, beta, psi_sqrt, time, burn_in)
```

## Arguments

n	Positive integer. Number of individuals.
mu0	List of numeric vectors. Mean of initial latent variable values ( $\mu_{\eta 0}$ ).
sigma0_sqrt	List of numeric matrices. Cholesky decomposition of the covariance matrix of initial latent variable values ( $\Sigma_{\eta 0}$ ).
alpha	List of numeric vectors. Vector of intercepts for the dynamic model ( $\alpha$ ).
beta	List of numeric matrices. Transition matrix relating the values of the latent variables at time $t - 1$ to those at time $t$ ( $\beta$ ).
psi_sqrt	List of numeric matrices. Cholesky decomposition of the process noise covariance matrix ( $\Psi$ ).
time	Positive integer. Number of time points to simulate.
burn_in	Positive integer. Number of burn-in points to exclude before returning the results.



## Details

Parameters can vary across individuals by providing a list of parameter values. If the length of any of the parameters ( $\mu_0$ ,  $\sigma_0_{\text{sqrt}}$ ,  $\alpha$ ,  $\beta$ , and  $\psi_{\text{sqrt}}$ ) is less than  $n$ , the function will cycle through the available values.

## Value

Returns a list of length  $n$ . Each element is a list with the following elements:

- $y$ : A  $t$  by  $k$  matrix of values for the manifest variables.
- $\eta$ : A  $t$  by  $p$  matrix of values for the latent variables.
- $\text{time}$ : A vector of discrete time points from 1 to  $t$ .
- $\text{id}$ : A vector of ID numbers of length  $t$ .
- $n$ : Number of individuals.

## Author(s)

Ivan Jacob Agaloos Pesigan

## References

Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications: With R examples*. Springer International Publishing. doi:[10.1007/9783319524528](https://doi.org/10.1007/9783319524528)

## See Also

Other Simulation of State Space Models Data Functions: [OU2SSM\(\)](#), [Sim2Matrix\(\)](#), [SimSSM0Fixed\(\)](#), [SimSSM0Vary\(\)](#), [SimSSM0\(\)](#), [SimSSM0UFixed\(\)](#), [SimSSM0UVary\(\)](#), [SimSSM0U\(\)](#), [SimSSMVARFixed\(\)](#), [SimSSMVAR\(\)](#)

## Examples

```
# prepare parameters
# In this example, beta varies across individuals
set.seed(42)
k <- 3
iden <- diag(k)
iden_sqrt <- chol(iden)
null_vec <- rep(x = 0, times = k)
n <- 5
mu0 <- list(null_vec)
sigma0_sqrt <- list(iden_sqrt)
alpha <- list(null_vec)
beta <- list(
  diag(x = 0.1, nrow = k),
  diag(x = 0.2, nrow = k),
  diag(x = 0.3, nrow = k),
  diag(x = 0.4, nrow = k),
  diag(x = 0.5, nrow = k)
)
```

```
psi_sqrt <- list(iden_sqrt)
time <- 50
burn_in <- 0

ssm <- SimSSMVARVary(
  n = n,
  mu0 = mu0,
  sigma0_sqrt = sigma0_sqrt,
  alpha = alpha,
  beta = beta,
  psi_sqrt = psi_sqrt,
  time = time,
  burn_in = burn_in
)

str(ssm)
```

# Index

## \* Simulation of State Space Models Data

### Functions

OU2SSM, [2](#)  
Sim2Matrix, [3](#)  
SimSSM0, [5](#)  
SimSSM0Fixed, [7](#)  
SimSSM0Vary, [10](#)  
SimSSMOU, [12](#)  
SimSSMOUFixed, [15](#)  
SimSSMOUVary, [18](#)  
SimSSMVAR, [20](#)  
SimSSMVARFixed, [22](#)  
SimSSMVARVary, [24](#)

### \* misc

Sim2Matrix, [3](#)

### \* simStateSpace

OU2SSM, [2](#)  
Sim2Matrix, [3](#)  
SimSSM0, [5](#)  
SimSSM0Fixed, [7](#)  
SimSSM0Vary, [10](#)  
SimSSMOU, [12](#)  
SimSSMOUFixed, [15](#)  
SimSSMOUVary, [18](#)  
SimSSMVAR, [20](#)  
SimSSMVARFixed, [22](#)  
SimSSMVARVary, [24](#)

### \* sim

OU2SSM, [2](#)  
SimSSM0, [5](#)  
SimSSM0Fixed, [7](#)  
SimSSM0Vary, [10](#)  
SimSSMOU, [12](#)  
SimSSMOUFixed, [15](#)  
SimSSMOUVary, [18](#)  
SimSSMVAR, [20](#)  
SimSSMVARFixed, [22](#)  
SimSSMVARVary, [24](#)

Sim2Matrix, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [17](#), [19](#), [21](#), [23](#),  
[25](#)

SimSSM0, [2](#), [3](#), [5](#), [9](#), [11](#), [14](#), [17](#), [19](#), [21](#), [23](#), [25](#)

SimSSM0(), [3](#)

SimSSM0Fixed, [2](#), [3](#), [6](#), [7](#), [11](#), [14](#), [17](#), [19](#), [21](#),  
[23](#), [25](#)

SimSSM0Fixed(), [3](#)

SimSSM0Vary, [2](#), [3](#), [6](#), [9](#), [10](#), [14](#), [17](#), [19](#), [21](#), [23](#),  
[25](#)

SimSSMOU, [2](#), [3](#), [6](#), [9](#), [11](#), [12](#), [17](#), [19](#), [21](#), [23](#), [25](#)

SimSSMOU(), [3](#)

SimSSMOUFixed, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [15](#), [19](#), [21](#),  
[23](#), [25](#)

SimSSMOUFixed(), [3](#)

SimSSMOUVary, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [17](#), [18](#), [21](#),  
[23](#), [25](#)

SimSSMVAR, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [17](#), [19](#), [20](#), [23](#), [25](#)

SimSSMVAR(), [3](#)

SimSSMVARFixed, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [17](#), [19](#), [21](#),  
[22](#), [25](#)

SimSSMVARFixed(), [3](#)

SimSSMVARVary, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [17](#), [19](#), [21](#),  
[23](#), [24](#)

OU2SSM, [2](#), [3](#), [6](#), [9](#), [11](#), [14](#), [17](#), [19](#), [21](#), [23](#), [25](#)