SSJ User's Guide

Package `randvarmulti`

Generating Random Vectors

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This package is a multivariate version of the package `randvar`. It implements random number generators for some (nonuniform) multivariate distributions.

Contents

Overview

This package provides a collection of classes for non-uniform random variate generation, very similar to `randvar`, but for multivariate distributions.

RandomMultivariateGen

This class is the multivariate counterpart of `RandomVariateGen`. It is the base class for general random variate generators over the d-dimensional real space $\mathbb{R}^d$. It specifies the signature of the `nextPoint` method, which is normally called to generate a random vector from a given distribution. Contrary to univariate distributions and generators, here the inversion method is not well defined, so we cannot construct a multivariate generator simply by passing a multivariate distribution and a stream; we must specify a generating method as well. For this reason, this class is abstract. Generators can be constructed only by invoking the constructor of a subclass. This is an important difference with `RandomVariateGen`.

```
package umontreal.iro.lecuyer.randvarmulti;

public abstract class RandomMultivariateGen
```

Methods

`abstract public void nextPoint (double[] p);`

Generates a random point p using the the stream contained in this object.

`public void nextArrayOfPoints (double[][] v, int start, int n)`

Generates n random points. These points are stored in the array `v`, starting at index `start`. Thus `v[start][i]` contains coordinate i of the first generated point. By default, this method calls `nextPoint` n times, but one can override it in subclasses for better efficiency. The array argument `v[][d]` must have d elements reserved for each generated point before calling this method.

`public int getDimension()`

Returns the dimension of this multivariate generator (the dimension of the random points).

`public RandomStream getStream()`

Returns the `RandomStream` used by this object.

`public void setStream (RandomStream stream)`

Sets the `RandomStream` used by this object to `stream`.

IIDMultivariateGen

Extends `RandomMultivariateGen` for a vector of independent identically distributed (i.i.d.) random variables.

```
package umontreal.iro.lecuyer.randvarmulti;

public class IIDMultivariateGen extends RandomMultivariateGen
```

Constructor

`public IIDMultivariateGen (RandomVariateGen gen1, int d)`

Constructs a generator for a `d`-dimensional vector of i.i.d. variates with a common one-dimensional generator `gen1`.

Methods

`public void setDimension (int d)`

Changes the dimension of the vector to `d`.

`public void nextPoint (double[] p)`

Generates a vector of i.i.d. variates.

`public void setGen1 (RandomVariateGen gen1)`

Sets the common one-dimensional generator to `gen1`.

`public RandomVariateGen getGen1()`

Returns the common one-dimensional generator used in this class.

`public String toString()`

Returns a string representation of the generator.

MultinormalGen

Extends `RandomMultivariateGen` for a *multivariate normal* (or *multinormal*) distribution [1]. The d-dimensional multivariate normal distribution with mean vector $\boldsymbol{\mu} \in \mathbb{R}^d$ and (symmetric positive-definite) covariance matrix $\boldsymbol{\Sigma}$, denoted $N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$, has density

$$f(\mathbf{X}) = \frac{1}{\sqrt{(2\pi)^d \det(\boldsymbol{\Sigma})}} \exp\left(-(\mathbf{X}-\boldsymbol{\mu})^{\mathsf{t}} \boldsymbol{\Sigma}^{-1} (\mathbf{X}-\boldsymbol{\mu})/2\right),$$

for all $\mathbf{X} \in \mathbb{R}^d$, and $\mathbf{X}^{\mathsf{t}}$ is the transpose vector of $\mathbf{X}$. If $\mathbf{Z} \sim N(\mathbf{0}, \mathbf{I})$ where $\mathbf{I}$ is the identity matrix, $\mathbf{Z}$ is said to have the *standard multinormal* distribution.

For the special case $d = 2$, if the random vector $\mathbf{X} = (X_1, X_2)^{\mathsf{t}}$ has a bivariate normal distribution, then it has mean $\boldsymbol{\mu} = (\mu_1, \mu_2)^{\mathsf{t}}$, and covariance matrix

$$\boldsymbol{\Sigma} = \begin{bmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{bmatrix}$$

if and only if $\mathrm{Var}[X_1] = \sigma_1^2$, $\mathrm{Var}[X_2] = \sigma_2^2$, and the linear correlation between X_1 and X_2 is ρ, where $\sigma_1 > 0$, $\sigma_2 > 0$, and $-1 \le \rho \le 1$.

```
package umontreal.iro.lecuyer.randvarmulti;

public class MultinormalGen extends RandomMultivariateGen
```

Constructors

`public MultinormalGen (NormalGen gen1, int d)`

Constructs a generator with the standard multinormal distribution (with $\boldsymbol{\mu} = \mathbf{0}$ and $\boldsymbol{\Sigma} = \mathbf{I}$) in d dimensions. Each vector $\mathbf{Z}$ will be generated via d successive calls to `gen1`, which must be a *standard normal* generator.

```
protected MultinormalGen (NormalGen gen1, double[] mu,
                          DoubleMatrix2D sigma)
```

Constructs a multinormal generator with mean vector `mu` and covariance matrix `sigma`. The mean vector must have the same length as the dimensions of the covariance matrix, which must be symmetric and positive-definite. If any of the above conditions is violated, an exception is thrown. The vector $\mathbf{Z}$ is generated by calling d times the generator `gen1`, which must be *standard normal*.

`protected MultinormalGen (NormalGen gen1, double[] mu, double[][] sigma)`

Equivalent to `MultinormalGen (gen1, mu, new DenseDoubleMatrix2D (sigma))`.

Methods

`public double[] getMu()`

Returns the mean vector used by this generator.

`public double getMu (int i)`

Returns the i-th component of the mean vector for this generator.

`public void setMu (double[] mu)`

Sets the mean vector to `mu`.

`public void setMu (int i, double mui)`

Sets the i-th component of the mean vector to `mui`.

`public DoubleMatrix2D getSigma()`

Returns the covariance matrix $\boldsymbol{\Sigma}$ used by this generator.

`public void nextPoint (double[] p)`

Generates a point from this multinormal distribution.

MultinormalCholeskyGen

Extends `MultinormalGen` for a *multivariate normal* distribution [1], generated via a Cholesky decomposition of the covariance matrix. The covariance matrix $\mathbf{\Sigma}$ is decomposed (by the constructor) as $\mathbf{\Sigma} = \mathbf{A}\mathbf{A}^t$ where $\mathbf{A}$ is a lower-triangular matrix (this is the Cholesky decomposition), and $\mathbf{X}$ is generated via

$$\mathbf{X} = \boldsymbol{\mu} + \mathbf{A}\mathbf{Z},$$

where $\mathbf{Z}$ is a d-dimensional vector of independent standard normal random variates, and $\mathbf{A}^t$ is the transpose of $\mathbf{A}$. The covariance matrix $\mathbf{\Sigma}$ must be positive-definite, otherwise the Cholesky decomposition will fail. The decomposition method uses the `CholeskyDecomposition` class in `colt`.

```
package umontreal.iro.lecuyer.randvarmulti;

   import cern.colt.matrix.DoubleMatrix2D;
   import cern.colt.matrix.impl.DenseDoubleMatrix2D;
   import cern.colt.matrix.linalg.CholeskyDecomposition;


public class MultinormalCholeskyGen extends MultinormalGen
```

Constructors

```
   public MultinormalCholeskyGen (NormalGen gen1, double[] mu,
                                  double[][] sigma)
```

Equivalent to `MultinormalCholeskyGen(gen1, mu, new DenseDoubleMatrix2D(sigma))`.

```
   public MultinormalCholeskyGen (NormalGen gen1, double[] mu,
                                  DoubleMatrix2D sigma)
```

Constructs a multinormal generator with mean vector `mu` and covariance matrix `sigma`. The mean vector must have the same length as the dimensions of the covariance matrix, which must be symmetric and positive-definite. If any of the above conditions is violated, an exception is thrown. The vector $\mathbf{Z}$ is generated by calling d times the generator `gen1`, which must be a *standard normal* 1-dimensional generator.

Methods

```
   public DoubleMatrix2D getCholeskyDecompSigma()
```

Returns the lower-triangular matrix $\mathbf{A}$ in the Cholesky decomposition of $\mathbf{\Sigma}$.

```
   public void setSigma (DoubleMatrix2D sigma)
```

Sets the covariance matrix $\mathbf{\Sigma}$ of this multinormal generator to `sigma` (and recomputes $\mathbf{A}$).

```
   public static void nextPoint (NormalGen gen1, double[] mu,
                                 double[][] sigma, double[] p)
```

Equivalent to `nextPoint(gen1, mu, new DenseDoubleMatrix2D(sigma), p)`.

```
public static void nextPoint (NormalGen gen1, double[] mu,
                              DoubleMatrix2D sigma, double[] p)
```

Generates a d-dimensional vector from the multinormal distribution with mean vector `mu` and covariance matrix `sigma`, using the one-dimensional normal generator `gen1` to generate the coordinates of $\mathbf{Z}$, and using the Cholesky decomposition of $\mathbf{\Sigma}$. The resulting vector is put into `p`. Note that this static method will be very slow for large dimensions, since it computes the Cholesky decomposition at every call. It is therefore recommended to use a `MultinormalCholeskyGen` object instead, if the method is to be called more than once.

```
public void nextPoint (double[] p)
```

Generates a point from this multinormal distribution. This is much faster than the static method as it computes the singular value decomposition matrix only once in the constructor.

MultinormalPCAGen

Extends `MultinormalGen` for a *multivariate normal* distribution [1], generated via the method of principal components analysis (PCA) of the covariance matrix. The covariance matrix $\mathbf{\Sigma}$ is decomposed (by the constructor) as $\mathbf{\Sigma} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^t$ where $\mathbf{V}$ is an orthogonal matrix and $\mathbf{\Lambda}$ is the diagonal matrix made up of the eigenvalues of $\mathbf{\Sigma}$. $\mathbf{V}^t$ is the transpose matrix of $\mathbf{V}$. The eigenvalues are ordered from the largest (λ_1) to the smallest (λ_d). The random multinormal vector $\mathbf{X}$ is generated via

$$\mathbf{X} = \boldsymbol{\mu} + \mathbf{A}\mathbf{Z},$$

where $\mathbf{A} = \mathbf{V}\sqrt{\mathbf{\Lambda}}$, and $\mathbf{Z}$ is a d-dimensional vector of independent standard normal random variates. The decomposition method uses the `SingularValueDecomposition` class in `colt`.

```
package umontreal.iro.lecuyer.randvarmulti;

   import cern.colt.matrix.DoubleMatrix2D;
   import cern.colt.matrix.linalg.SingularValueDecomposition;


public class MultinormalPCAGen extends MultinormalGen
```

Constructors

`public MultinormalPCAGen (NormalGen gen1, double[] mu, double[][] sigma)`

Equivalent to `MultinormalPCAGen(gen1, mu, new DenseDoubleMatrix2D(sigma))`.

`public MultinormalPCAGen (NormalGen gen1, double[] mu, DoubleMatrix2D sigma)`

Constructs a multinormal generator with mean vector `mu` and covariance matrix `sigma`. The mean vector must have the same length as the dimensions of the covariance matrix, which must be symmetric and positive semi-definite. If any of the above conditions is violated, an exception is thrown. The vector $\mathbf{Z}$ is generated by calling d times the generator `gen1`, which must be a *standard normal* 1-dimensional generator.

Methods

`public static DoubleMatrix2D decompPCA (double[][] sigma)`

Computes the decomposition `sigma` $= \mathbf{\Sigma} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^t$. Returns $\mathbf{A} = \mathbf{V}\sqrt{\mathbf{\Lambda}}$.

`public static DoubleMatrix2D decompPCA (DoubleMatrix2D sigma)`

Computes the decomposition `sigma` $= \mathbf{\Sigma} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^t$. Returns $\mathbf{A} = \mathbf{V}\sqrt{\mathbf{\Lambda}}$.

`public DoubleMatrix2D getPCADecompSigma()`

Returns the matrix $\mathbf{A} = \mathbf{V}\sqrt{\mathbf{\Lambda}}$ of this object.

`public static double[] getLambda (DoubleMatrix2D sigma)`

Computes and returns the eigenvalues of `sigma` in decreasing order.

`public double[] getLambda()`

Returns the eigenvalues of $\mathbf{\Sigma}$ in decreasing order.

`public void setSigma (DoubleMatrix2D sigma)`

Sets the covariance matrix $\mathbf{\Sigma}$ of this multinormal generator to `sigma` (and recomputes $\mathbf{A}$).

```
public static void nextPoint (NormalGen gen1, double[] mu,
                              DoubleMatrix2D sigma, double[] p)
```

Generates a d-dimensional vector from the multinormal distribution with mean vector `mu` and covariance matrix `sigma`, using the one-dimensional normal generator `gen1` to generate the coordinates of $\mathbf{Z}$, and using the PCA decomposition of $\mathbf{\Sigma}$. The resulting vector is put into `p`. Note that this static method will be very slow for large dimensions, because it recomputes the singular value decomposition at every call. It is therefore recommended to use a `MultinormalPCAGen` object instead, if the method is to be called more than once.

```
public static void nextPoint (NormalGen gen1, double[] mu,
                              double[][] sigma, double[] p)
```

Equivalent to `nextPoint(gen1, mu, new DenseDoubleMatrix2D(sigma), p)`.

`public void nextPoint (double[] p)`

Generates a point from this multinormal distribution. This is much faster than the static method as it computes the singular value decomposition matrix only once in the constructor.

DirichletGen

Extends `RandomMultivariateGen` for a *Dirichlet* [1] distribution. This distribution uses the parameters $\alpha_1, \ldots, \alpha_k$, and has density

$$f(x_1, \ldots, x_k) = \frac{\Gamma(\alpha_0) \prod_{i=1}^{k} x_i^{\alpha_i - 1}}{\prod_{i=1}^{k} \Gamma(\alpha_i)}$$

where $\alpha_0 = \sum_{i=1}^{k} \alpha_i$.

Here, the successive coordinates of the Dirichlet vector are generated [1] via the class `GammaAcceptanceRejectionGen` in package `randvar`, using the same stream for all the uniforms.

```
package umontreal.iro.lecuyer.randvarmulti;

public class DirichletGen extends RandomMultivariateGen
```

Constructor

`public DirichletGen (RandomStream stream, double[] alphas)`

Constructs a new Dirichlet generator with parameters $\alpha_{i+1} =$ `alphas[i]`, for $i = 0, \ldots, k-1$, and the stream `stream`.

Methods

`public double getAlpha (int i)`

Returns the α_{i+1} parameter for this Dirichlet generator.

```
public static void nextPoint (RandomStream stream, double[] alphas,
                              double[] p)
```

Generates a new point from the Dirichlet distribution with parameters `alphas`, using the stream `stream`. The generated values are placed into `p`.

`public void nextPoint (double[] p)`

Generates a point from the Dirichlet distribution.

---

[1] From Pierre: How?

References

[1] N. L. Johnson and S. Kotz. *Distributions in Statistics: Continuous Multivariate Distributions*. John Wiley, New York, NY, 1972.