Title

drawnorm — Draw sample from multivariate normal distribution

Description

`drawnorm` draws a sample from a multivariate normal distribution with desired means and covariance matrix. The default is orthogonal data with mean 0 and variance 1. The covariance matrix may be singular. The values generated are a function of the current random-number seed or the number specified with `set seed()`; see [R] **set seed**.

Quick start

Generate independent variables `x` and `y`, where `x` has mean 2 and standard deviation 0.5 and `y` has mean 3 and standard deviation 1

```
drawnorm x y, means(2,3) sds(.5,1)
```

Same as above, but create dataset of 1,000 observations on `x` and `y` with means stored in vector `m` and standard deviations stored in vector `sd`

```
drawnorm x y, means(m) sds(sd) n(1000)
```

Same as above, and set the seed for the random-number generator to reproduce results

```
drawnorm x y, means(m) sds(sd) n(1000) seed(81625)
```

Sample from bivariate standard normal distribution with covariance between `x` and `y` of 0.5 stored in variance–covariance matrix `C`

```
matrix C = (1, .5 \ .5, 1)
drawnorm x y, cov(C)
```

Sample from a trivariate standard normal distribution with correlation between `x` and `y` of 0.4, `x` and `z` of 0.3, and `y` and `z` of 0.6 stored in correlation matrix `C`

```
matrix C = (1, .4, .3 \ .4, 1, .6 \ .3, .6, 1)
drawnorm x y z, corr(C)
```

Same as above, but avoid typing full matrix by specifying correlations in vector `v` treated as a lower triangular matrix

```
matrix v = (1, .4, 1, .3, .6, 1)
drawnorm x y z, corr(v) cstorage(lower)
```

Menu

Data > Create or change data > Other variable-creation commands > Draw sample from normal distribution

Syntax

```
drawnorm newvarlist [ , options ]
```

options	Description
Main	
`clear`	replace the current dataset
`double`	generate variable type as `double`; default is `float`
`n(#)`	generate # observations; default is current number
`sds(`*vector*`)`	standard deviations of generated variables
`corr(`*matrix*`|`*vector*`)`	correlation matrix
`cov(`*matrix*`|`*vector*`)`	covariance matrix
`cstorage(full)`	store correlation/covariance structure as a symmetric $k \times k$ matrix
`cstorage(lower)`	store correlation/covariance structure as a lower triangular matrix
`cstorage(upper)`	store correlation/covariance structure as an upper triangular matrix
`forcepsd`	force the covariance/correlation matrix to be positive semidefinite
`means(`*vector*`)`	means of generated variables; default is `means(0)`
Options	
`seed(#)`	seed for random-number generator

Options

Main

`clear` specifies that the dataset in memory be replaced, even though the current dataset has not been saved on disk.

`double` specifies that the new variables be stored as Stata `doubles`, meaning 8-byte reals. If `double` is not specified, variables are stored as `floats`, meaning 4-byte reals. See [D] **Data types**.

`n(#)` specifies the number of observations to be generated. The default is the current number of observations. If `n(#)` is not specified or is the same as the current number of observations, `drawnorm` adds the new variables to the existing dataset; otherwise, `drawnorm` replaces the data in memory.

`sds(`*vector*`)` specifies the standard deviations of the generated variables. `sds()` may not be specified with `cov()`.

`corr(`*matrix*`|`*vector*`)` specifies the correlation matrix. If neither `corr()` nor `cov()` is specified, the default is orthogonal data.

`cov(`*matrix*`|`*vector*`)` specifies the covariance matrix. If neither `cov()` nor `corr()` is specified, the default is orthogonal data.

`cstorage(full | lower | upper)` specifies the storage mode for the correlation or covariance structure in `corr()` or `cov()`. The following storage modes are supported:

`full` specifies that the correlation or covariance structure is stored (recorded) as a symmetric $k \times k$ matrix.

`lower` specifies that the correlation or covariance structure is recorded as a lower triangular matrix. With k variables, the matrix should have $k(k+1)/2$ elements in the following order:

$$\mathrm{C}_{11}\ \mathrm{C}_{21}\ \mathrm{C}_{22}\ \mathrm{C}_{31}\ \mathrm{C}_{32}\ \mathrm{C}_{33}\ \ldots\ \mathrm{C}_{k1}\ \mathrm{C}_{k2}\ \ldots\ \mathrm{C}_{kk}$$

`upper` specifies that the correlation or covariance structure is recorded as an upper triangular matrix. With k variables, the matrix should have $k(k+1)/2$ elements in the following order:

$$\mathrm{C}_{11}\ \mathrm{C}_{12}\ \mathrm{C}_{13}\ \ldots\ \mathrm{C}_{1k}\ \mathrm{C}_{22}\ \mathrm{C}_{23}\ \ldots\mathrm{C}_{2k}\ \ldots\ \mathrm{C}_{(k-1k-1)}\ \mathrm{C}_{(k-1k)}\ \mathrm{C}_{kk}$$

Specifying `cstorage(full)` is optional if the matrix is square. `cstorage(lower)` or `cstorage(upper)` is required for the vectorized storage methods. See *Example 2: Storage modes for correlation and covariance matrices*.

`forcepsd` modifies the matrix C to be positive semidefinite (psd), and so be a proper covariance matrix. If C is not positive semidefinite, it will have negative eigenvalues. By setting negative eigenvalues to 0 and reconstructing, we obtain the least-squares positive-semidefinite approximation to C. This approximation is a singular covariance matrix.

`means(`*vector*`)` specifies the means of the generated variables. The default is `means(0)`.

Options

`seed(#)` specifies the initial value of the random-number seed used by the `runiform()` function. The default is the current random-number seed. Specifying `seed(#)` is the same as typing `set seed #` before issuing the `drawnorm` command.

Remarks and examples

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▷ Example 1

Suppose that we want to draw a sample of 1,000 observations from a normal distribution $N(\mathbf{M}, \mathbf{V})$, where $\mathbf{M}$ is the mean matrix and $\mathbf{V}$ is the covariance matrix:

```
. matrix M = 5, -6, 0.5
. matrix V = (9, 5, 2 \ 5 , 4 , 1 \ 2, 1, 1)
. matrix list M
M[1,3]
    c1  c2  c3
r1   5  -6  .5
. matrix list V
symmetric V[3,3]
    c1  c2  c3
r1   9
r2   5   4
r3   2   1   1
. drawnorm x y z, n(1000) cov(V) means(M)
(obs 1,000)
```

```
. summarize
    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
           x |      1,000       5.0424    3.061953  -5.065592   15.96129
           y |      1,000   -5.914462    2.012488  -12.25234   .3326397
           z |      1,000    .5181909    1.017397   -2.59316   3.884182

. correlate, cov
(obs=1,000)
             |        x        y        z
-------------+---------------------------
           x |  9.37556
           y |  5.14201  4.05011
           z |  2.17972  1.07222   1.0351
```

◁

❑ Technical note

The values generated by `drawnorm` are a function of the current random-number seed. To reproduce the same dataset each time `drawnorm` is run with the same setup, specify the same seed number in the `seed()` option.

❑

▷ Example 2: Storage modes for correlation and covariance matrices

The three storage modes for specifying the correlation or covariance matrix in `corr2data` and `drawnorm` can be illustrated with a correlation structure, `C`, of 4 variables. In full storage mode, this structure can be entered as a 4×4 Stata matrix:

```
. matrix C = ( 1.0000,  0.3232,  0.1112,  0.0066 \ ///
               0.3232,  1.0000,  0.6608, -0.1572 \ ///
               0.1112,  0.6608,  1.0000, -0.1480 \ ///
               0.0066, -0.1572, -0.1480,  1.0000 )
```

Elements within a row are separated by commas, and rows are separated by a backslash, `\`. We use the input continuation operator `///` for convenient multiline input; see [P] **comments**. In this storage mode, we probably want to set the row and column names to the variable names:

```
. matrix rownames C = price trunk headroom rep78
. matrix colnames C = price trunk headroom rep78
```

This correlation structure can be entered more conveniently in one of the two vectorized storage modes. In these modes, we enter the lower triangle or the upper triangle of `C` in rowwise order; these two storage modes differ only in the order in which the $k(k+1)/2$ matrix elements are recorded. The lower storage mode for C comprises a vector with $4(4+1)/2 = 10$ elements, that is, a 1×10 or 10×1 Stata matrix, with one row or column,

```
. matrix C = ( 1.0000,  ///
               0.3232,  1.0000,  ///
               0.1112,  0.6608,  1.0000,  ///
               0.0066, -0.1572, -0.1480,  1.0000)
```

or more compactly as

```
. matrix C = ( 1, 0.3232, 1, 0.1112, 0.6608, 1, 0.0066, -0.1572, -0.1480, 1 )
```

C may also be entered in upper storage mode as a vector with $4(4+1)/2 = 10$ elements, that is, a 1×10 or 10×1 Stata matrix,

```
. matrix C = ( 1.0000,  0.3232,  0.1112,  0.0066, ///
                        1.0000,  0.6608, -0.1572, ///
                                 1.0000, -0.1480, ///
                                          1.0000 )
```

or more compactly as

```
. matrix C = ( 1, 0.3232, 0.1112, 0.0066, 1, 0.6608, -0.1572, 1, -0.1480, 1 )
```

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Methods and formulas

Results are asymptotic. The more observations generated, the closer the correlation matrix of the dataset is to the desired correlation structure.

Let $\mathbf{V} = \mathbf{A}'\mathbf{A}$ be the desired covariance matrix and $\mathbf{M}$ be the desired mean matrix. We first generate $\mathbf{X}$, such that $\mathbf{X} \sim N(\mathbf{0}, \mathbf{I})$. Let $\mathbf{Y} = \mathbf{A}'\mathbf{X} + \mathbf{M}$, then $\mathbf{Y} \sim N(\mathbf{M}, \mathbf{V})$.

References

Canette, I. 2013. Fitting ordered probit models with endogenous covariates with Stata's gsem command. *The Stata Blog: Not Elsewhere Classified.* http://blog.stata.com/2013/11/07/fitting-ordered-probit-models-with-endogenous-covariates-with-statas-gsem-command/.

Chen, M. 2015. Generating nonnegatively correlated binary random variates. *Stata Journal* 15: 301–308.

Gould, W. W. 2012a. Using Stata's random-number generators, part 2: Drawing without replacement. *The Stata Blog: Not Elsewhere Classified.* http://blog.stata.com/2012/08/03/using-statas-random-number-generators-part-2-drawing-without-replacement/.

——. 2012b. Using Stata's random-number generators, part 3: Drawing with replacement. *The Stata Blog: Not Elsewhere Classified.* http://blog.stata.com/2012/08/29/using-statas-random-number-generators-part-3-drawing-with-replacement/.

Huber, C. 2014. How to simulate multilevel/longitudinal data. *The Stata Blog: Not Elsewhere Classified.* http://blog.stata.com/2014/07/18/how-to-simulate-multilevellongitudinal-data/.

Lee, S. 2015. Generating univariate and multivariate nonnormal data. *Stata Journal* 15: 95–109.

Lindsey, C. 2015a. Probit model with sample selection by mlexp. *The Stata Blog: Not Elsewhere Classified.* http://blog.stata.com/2015/10/22/probit-model-with-sample-selection-by-mlexp/.

——. 2015b. Using mlexp to estimate endogenous treatment effects in a probit model. *The Stata Blog: Not Elsewhere Classified.* http://blog.stata.com/2015/11/05/using-mlexp-to-estimate-endogenous-treatment-effects-in-a-probit-model/.

Also see

[D] **corr2data** — Create dataset with specified correlation structure

[R] **set seed** — Specify random-number seed and state

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