



`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:

$$C_{11} \ C_{21} \ C_{22} \ C_{31} \ C_{32} \ C_{33} \ \dots \ C_{k1} \ C_{k2} \ \dots \ C_{kk}$$

`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:

$$C_{11} \ C_{12} \ C_{13} \ \dots \ C_{1k} \ C_{22} \ C_{23} \ \dots \ C_{2k} \ \dots \ C_{(k-1)(k-1)} \ C_{(k-1)k} \ C_{kk}$$

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)`.

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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

## ▷ 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 1000)
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
x	1000	5.001715	3.00608	-4.572042	13.66046
y	1000	-5.980279	2.004755	-12.08166	-.0963039
z	1000	.5271135	1.011095	-2.636946	4.102734

```
. correlate, cov
(obs=1000)
```

	x	y	z
x	9.03652		
y	5.04462	4.01904	
z	2.10142	1.08773	1.02231

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## □ 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,  $\mathbf{C}$ , of 4 variables. In full storage mode, this structure can be entered as a  $4 \times 4$  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  $1 \times 10$  or  $10 \times 1$  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  $1 \times 10$  or  $10 \times 1$  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 )
```

◀

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

## References

- 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/>.

## Also see

- [D] **corr2data** — Create dataset with specified correlation structure
- [R] **set seed** — Specify initial value of random-number seed