

mvtest normality — Multivariate normality tests[Description](#)[Syntax](#)[Stored results](#)[References](#)[Quick start](#)[Options](#)[Methods and formulas](#)[Also see](#)[Menu](#)[Remarks and examples](#)[Acknowledgment](#)

Description

`mvtest normality` performs tests for univariate, bivariate, and multivariate normality.

See [\[MV\] mvtest](#) for more multivariate tests.

Quick start

Doornik–Hansen omnibus test for multivariate normality for `v1`, `v2`, `v3`, and `v4`

```
mvtest normality v1 v2 v3 v4
```

Also show Henze–Zirkler’s consistent test, Mardia’s multivariate kurtosis test, and Mardia’s multivariate skewness test

```
mvtest normality v1 v2 v3 v4,                               ///  
stats(dhansen hzirkler kurtosis skewness)
```

Same as above

```
mvtest normality v1 v2 v3 v4, stats(all)
```

Also show Doornik–Hansen test for bivariate normality for each pair of variables

```
mvtest normality v1 v2 v3 v4, stats(all) bivariate
```

As above, but show univariate normality tests from `sktest` instead of the bivariate tests

```
mvtest normality v1 v2 v3 v4, stats(all) univariate
```

Show all multivariate, bivariate, and univariate tests of normality for `v1`, `v2`, `v3`, and `v4`

```
mvtest normality v1 v2 v3 v4, stats(all) bivariate univariate
```

Menu

Statistics > Multivariate analysis > MANOVA, multivariate regression, and related > Multivariate test of means, covariances, and normality

Syntax

```
mvtest normality varlist [if] [in] [weight] [, options]
```

<i>options</i>	Description
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Options

univariate	display tests for univariate normality (sktest)
bivariate	display tests for bivariate normality (Doornik–Hansen)
stats(<i>stats</i>)	statistics to be computed

<i>stats</i>	Description
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dhansen	Doornik–Hansen omnibus test; the default
hzirkler	Henze–Zirkler’s consistent test
kurtosis	Mardia’s multivariate kurtosis test
skewness	Mardia’s multivariate skewness test
all	all tests listed here

[bootstrap](#), [by](#), [collect](#), [jackknife](#), [rolling](#), and [statsby](#) are allowed; see [U] [11.1.10 Prefix commands](#).

Weights are not allowed with the [bootstrap](#) prefix; see [R] [bootstrap](#).

[aweight](#)s are not allowed with the [jackknife](#) prefix; see [R] [jackknife](#).

[fweight](#)s are allowed; see [U] [11.1.6 weight](#).

Options

Options

[univariate](#) specifies that tests for univariate normality be displayed, as obtained from [sktest](#); see [R] [sktest](#).

[bivariate](#) specifies that the Doornik–Hansen (2008) test for bivariate normality be displayed for each pair of variables.

[stats\(*stats*\)](#) specifies one or more test statistics for multivariate normality. Multiple *stats* are separated by white space. The following *stats* are available:

[dhansen](#) produces the Doornik–Hansen (2008) omnibus test.

[hzirkler](#) produces Henze–Zirkler’s (1990) consistent test.

[kurtosis](#) produces the test based on Mardia’s (1970) measure of multivariate kurtosis.

[skewness](#) produces the test based on Mardia’s (1970) measure of multivariate skewness.

[all](#) is a convenient shorthand for `stats(dhansen hzirkler kurtosis skewness)`.

Remarks and examples

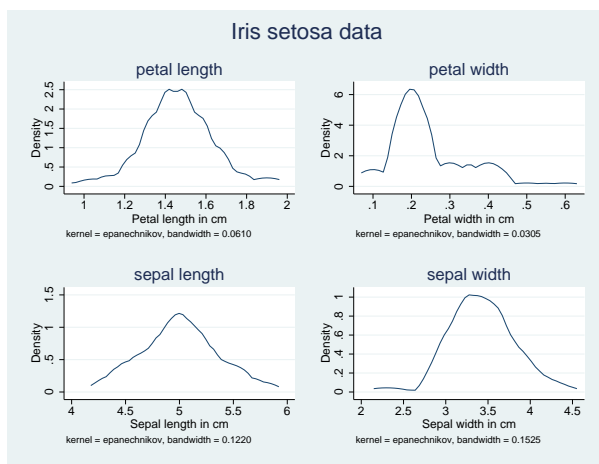
stata.com

Univariate and multivariate tests of normality are provided by the `mvtest normality` command.

▷ Example 1

The classic Fisher iris data from [Anderson \(1935\)](#) consists of four features measured on 50 samples from each of three iris species. The four features are the length and width of the sepal and petal. The three species are *Iris setosa*, *Iris versicolor*, and *Iris virginica*. We hypothesize that these features might be normally distributed within species, though they are likely not normally distributed across species. We will examine the *Iris setosa* data.

```
. use https://www.stata-press.com/data/r17/iris
(Iris data)
. kdensity petlen if iris==1, name(petlen, replace) title(petal length)
. kdensity petwid if iris==1, name(petwid, replace) title(petal width)
. kdensity sepwid if iris==1, name(sepwid, replace) title(sepal width)
. kdensity seplen if iris==1, name(seplen, replace) title(sepal length)
. graph combine petlen petwid seplen sepwid, title("Iris setosa data")
```



We perform all multivariate, univariate, and bivariate tests of normality.

```
. mvtest norm pet* sep* if iris==1, bivariate univariate stats(all)
Test for univariate normality
```

Variable	Pr(Skewness)	Pr(Kurtosis)	adj chi2(2)	joint Prob>chi2
petlen	0.7403	0.1447	2.36	0.3074
petwid	0.0010	0.0442	12.03	0.0024
seplen	0.7084	0.8157	0.19	0.9075
sepwid	0.8978	0.1627	2.07	0.3553

Doornik-Hansen test for bivariate normality

Pair of variables	chi2	df	Prob>chi2
petlen petwid	17.47	4	0.0016
petlen seplen	5.76	4	0.2177
petlen sepwid	8.50	4	0.0748
petwid seplen	14.97	4	0.0048
petwid sepwid	19.15	4	0.0007
seplen sepwid	5.92	4	0.2049

Test for multivariate normality

```
Mardia mSkewness = 3.079721 chi2(20) = 27.860 Prob>chi2 = 0.1128
Mardia mKurtosis = 26.53766 chi2(1) = 1.677 Prob>chi2 = 0.1953
Henze-Zirkler = .9488453 chi2(1) = 2.707 Prob>chi2 = 0.0999
Doornik-Hansen chi2(8) = 24.414 Prob>chi2 = 0.0020
```

From the univariate tests of normality, `petwid` does not appear to be normally distributed: p -values of 0.0010 for skewness, 0.0442 for kurtosis, and 0.0024 for the joint univariate test. The univariate tests of the other three variables do not lead to a rejection of the null hypothesis of normality.

The bivariate tests of normality show a rejection (at the 5% level) of the null hypothesis of bivariate normality for all pairs of variables that include `petwid`. Other pairings fail to reject the null hypothesis of bivariate normality.

Of the four multivariate normality tests, only the Doornik–Hansen test rejects the null hypothesis of multivariate normality, p -value of 0.0020.

◀

The Doornik-Hansen (2008) test and Mardia's (1970) test for multivariate kurtosis take computing time roughly proportional to the number of observations. In contrast, the computing time of the test by Henze-Zirkler (1990) and Mardia's (1970) test for multivariate skewness are roughly proportional to the square of the number of observations.

Stored results

mvtest normality stores the following in `r()`:

Scalars

<code>r(p_dh)</code>	<i>p</i> -value for Doornik–Hansen test (<code>stats(dhansen)</code>)
<code>r(df_dh)</code>	degrees of freedom of <code>chi2_dh</code> (<code>stats(dhansen)</code>)
<code>r(chi2_dh)</code>	Doornik–Hansen statistic (<code>stats(dhansen)</code>)
<code>r(rank_hz)</code>	rank of covariance matrix (<code>stats(hzirkler)</code>)
<code>r(p_hz)</code>	<i>p</i> -value for two-sided Henze–Zirkler’s test (<code>stats(hzirkler)</code>)
<code>r(z_hz)</code>	normal variate associated with <code>hz</code> (<code>stats(hzirkler)</code>)
<code>r(V_hz)</code>	expected variance of <code>log(hz)</code> (<code>stats(hzirkler)</code>)
<code>r(E_hz)</code>	expected value of <code>log(hz)</code> (<code>stats(hzirkler)</code>)
<code>r(hz)</code>	Henze–Zirkler discrepancy statistic (<code>stats(hzirkler)</code>)
<code>r(rank_mkurt)</code>	rank of covariance matrix (<code>stats(kurtosis)</code>)
<code>r(p_mkurt)</code>	<i>p</i> -value for Mardia’s multivariate kurtosis test (<code>stats(kurtosis)</code>)
<code>r(z_mkurt)</code>	normal variate associated with Mardia <i>m</i> Kurtosis (<code>stats(kurtosis)</code>)
<code>r(chi2_mkurt)</code>	χ^2 of Mardia <i>m</i> Kurtosis (<code>stats(kurtosis)</code>)
<code>r(mkurt)</code>	Mardia <i>m</i> Kurtosis test statistic (<code>stats(kurtosis)</code>)
<code>r(rank_mskew)</code>	rank for Mardia <i>m</i> Skewness test (<code>stats(skewness)</code>)
<code>r(p_mskew)</code>	<i>p</i> -value for Mardia’s multivariate skewness test (<code>stats(skewness)</code>)
<code>r(df_mskew)</code>	degrees of freedom of Mardia <i>m</i> Skewness test (<code>stats(skewness)</code>)
<code>r(chi2_mskew)</code>	χ^2 of Mardia <i>m</i> Skewness test (<code>stats(skewness)</code>)
<code>r(mskew)</code>	Mardia <i>m</i> Skewness test statistic (<code>stats(skewness)</code>)

Matrices

<code>r(U_dh)</code>	matrix with the skewness and kurtosis of orthonormalized variables (used in the Doornik–Hansen test): <code>b1</code> , <code>b2</code> , <code>z(b1)</code> , and <code>z(b2)</code> (<code>stats(dhansen)</code>)
<code>r(Btest)</code>	bivariate test statistics (<code>bivariate</code>)
<code>r(Utest)</code>	univariate test statistics (<code>univariate</code>)

Methods and formulas

There are N independent k -variate observations, \mathbf{x}_i , $i = 1, \dots, N$. Let \mathbf{X} denote the $N \times k$ matrix of observations. We wish to test whether these observations are multivariate normal distributed, $MVN_k(\boldsymbol{\mu}, \boldsymbol{\Sigma})$. The sample mean is $\bar{\mathbf{x}} = 1/N \sum_i \mathbf{x}_i$, and the sample covariance matrix is $\mathbf{S} = 1/N \sum (\mathbf{x}_i - \bar{\mathbf{x}})(\mathbf{x}_i - \bar{\mathbf{x}})'$.

Methods and formulas are presented under the following headings:

Mardia mSkewness and mKurtosis

Henze–Zirkler

Doornik–Hansen

Mardia mSkewness and mKurtosis

Mardia (1970) defined multivariate skewness, $b_{1,k}$, and kurtosis, $b_{2,k}$, as

$$b_{1,k} = \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N g_{ij}^3 \quad \text{and} \quad b_{2,k} = \frac{1}{N} \sum_{i=1}^N g_{ii}^2$$

where $g_{ij} = (\mathbf{x}_i - \bar{\mathbf{x}})' \mathbf{S}^{-1} (\mathbf{x}_j - \bar{\mathbf{x}})$. The test statistic

$$z_1 = \frac{(k+1)(N+1)(N+3)}{6\{(N+1)(k+1)-6\}} b_{1,k}$$

is approximately χ^2 distributed with $k(k+1)(k+2)/6$ degrees of freedom. The test statistic

$$z_2 = \frac{b_{2,k} - k(k+2)}{\sqrt{8k(k+2)/N}}$$

is approximately $N(0, 1)$ distributed. Also see [Rencher and Christensen \(2012, 108\)](#); [Mardia, Kent, and Bibby \(1979, 20–22\)](#); and [Seber \(1984, 148–149\)](#).

Henze–Zirkler

The Henze–Zirkler (1990) test, under the assumption that \mathbf{S} is nonsingular, is

$$\begin{aligned} T &= \frac{1}{N} \sum_{i=1}^N \sum_{j=1}^N \exp \left\{ -\frac{\beta^2}{2} (\mathbf{x}_i - \mathbf{x}_j)' \mathbf{S}^{-1} (\mathbf{x}_i - \mathbf{x}_j) \right\} \\ &\quad - 2(1 + \beta^2)^{-k/2} \sum_{i=1}^N \exp \left\{ -\frac{\beta^2}{2(1 + \beta^2)} (\mathbf{x}_i - \bar{\mathbf{x}})' \mathbf{S}^{-1} (\mathbf{x}_i - \bar{\mathbf{x}}) \right\} \\ &\quad + N(1 + 2\beta^2)^{-k/2} \end{aligned}$$

where

$$\beta = \frac{1}{\sqrt{2}} \left\{ \frac{N(2k+1)}{4} \right\}^{1/(k+4)}$$

As $N \rightarrow \infty$, the first two moments of T are given by

$$E(T) = 1 - (1 + 2\beta^2)^{-k/2} \left\{ 1 + \frac{k\beta^2}{1 + 2\beta^2} + \frac{k(k+2)\beta^4}{2(1 + 2\beta^2)^2} \right\}$$

$$\begin{aligned} \text{Var}(T) &= 2(1 + 4\beta^2)^{-k/2} + 2(1 + 2\beta^2)^{-k} \left\{ 1 + \frac{2k\beta^4}{(1 + 2\beta^2)^2} + \frac{3k(k+2)\beta^8}{4(1 + 2\beta^2)^4} \right\} \\ &\quad - 4w^{-k/2} \left\{ 1 + \frac{3k\beta^4}{2w} + \frac{k(k+2)\beta^8}{2w^2} \right\} \end{aligned}$$

where $w = (1 + \beta^2)(1 + 3\beta^2)$.

Henze–Zirkler suggest obtaining a p -value from the assumption, supported by a series of simulations, that T is approximately lognormal distributed. Thus let $VZ = \ln \{1 + \text{Var}(T)/E(T)^2\}$ and $EZ = \ln \{E(T)\} - VZ/2$. The transformation $Z = \{ \ln(T) - EZ \} / \sqrt{VZ}$. The p -value of Z is computed as $p = 2\Phi(-|Z|)$, where $\Phi(\cdot)$ is the cumulative normal distribution.

Doornik–Hansen

For the Doornik–Hansen (2008) test, the multivariate observations are transformed, then the univariate skewness and kurtosis for each transformed variable is computed, and then these are combined into an approximate χ^2 statistic.

Let \mathbf{V} be a matrix with i th diagonal element equal to $S_{ii}^{-1/2}$, where S_{ii} is the i th diagonal element of \mathbf{S} . $\mathbf{C} = \mathbf{V}\mathbf{S}\mathbf{V}$ is then the correlation matrix. Let \mathbf{H} be a matrix with columns equal to the eigenvectors of \mathbf{C} , and let $\mathbf{\Lambda}$ be a diagonal matrix with the corresponding eigenvalues. Let $\check{\mathbf{X}}$ be the centered version of \mathbf{X} , that is, \bar{x} subtracted from each row. The data are then transformed using $\dot{\mathbf{X}} = \check{\mathbf{X}}\mathbf{V}\mathbf{H}\mathbf{\Lambda}^{-1/2}\mathbf{H}'$.

The univariate skewness and kurtosis for each column of $\dot{\mathbf{X}}$ is then computed. The general formula for univariate skewness is $\sqrt{b_1} = m_3/m_2^{3/2}$ and kurtosis is $b_2 = m_4/m_2^2$, where $m_p = 1/N \sum_{i=1}^N (x_i - \bar{x})^p$. Let \dot{x}_i denote the i th observation from the selected column of $\dot{\mathbf{X}}$. Because by construction the mean of \dot{x} is zero and the variance m_2 is one, the formulas simplify to $\sqrt{b_1} = m_3$ and $b_2 = m_4$, where $m_p = 1/N \sum_{i=1}^N \dot{x}_i^p$.

The univariate skewness, $\sqrt{b_1}$, is transformed into an approximately normal variate, z_1 , as in D'Agostino (1970):

$$z_1 = \delta \log \left(y + \sqrt{1 + y^2} \right)$$

where

$$y = \left\{ \frac{b_1(\omega^2 - 1)(N + 1)(N + 3)}{12(N - 2)} \right\}^{1/2}$$

$$\delta = \left(\log \sqrt{\omega^2} \right)^{-1/2}$$

$$\omega^2 = -1 + \sqrt{2(\beta - 1)}$$

$$\beta = \frac{3(N^2 + 27N - 70)(N + 1)(N + 3)}{(N - 2)(N + 5)(N + 7)(N + 9)}$$

The univariate kurtosis, b_2 , is transformed from a gamma variate into a χ^2 -variate and then into a standard normal variable, z_2 , using the Wilson–Hilferty (1931) transform:

$$z_2 = \sqrt{9\alpha} \left\{ \left(\frac{\chi}{2\alpha} \right)^{1/3} - 1 + \frac{1}{9\alpha} \right\}$$

where

$$\chi = 2f(b_2 - 1 - b_1)$$

$$\alpha = a + b_1c$$

$$f = \frac{(N + 5)(N + 7)(N^3 + 37N^2 + 11N - 313)}{12\delta}$$

$$c = \frac{(N - 7)(N + 5)(N + 7)(N^2 + 2N - 5)}{6\delta}$$

$$a = \frac{(N - 2)(N + 5)(N + 7)(N^2 + 27N - 70)}{6\delta}$$

$$\delta = (N - 3)(N + 1)(N^2 + 15N - 4)$$

The z_1 and z_2 associated with the columns of $\dot{\mathbf{X}}$ are collected into vectors \mathbf{Z}_1 and \mathbf{Z}_2 . The statistic $\mathbf{Z}'_1\mathbf{Z}_1 + \mathbf{Z}'_2\mathbf{Z}_2$ is approximately χ^2 distributed with $2k$ degrees of freedom.

Acknowledgment

An earlier implementation of the [Doornik and Hansen \(2008\)](#) test is the omninorm package of [Baum and Cox \(2007\)](#).

References

- Anderson, E. 1935. The irises of the Gaspé Peninsula. *Bulletin of the American Iris Society* 59: 2–5.
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Also see

[R] [sktest](#) — Skewness and kurtosis tests for normality

[R] [swilk](#) — Shapiro–Wilk and Shapiro–Francia tests for normality