

Real Statistics Using Excel

*Everything you need to do real
statistical analysis using Excel*

Lilliefors Test for Normality

When the population mean and standard deviation are known we can use the one sample Kolmogorov-Smirnov test to test for normality, as described in [Kolmogorov-Smirnov Test for Normality](#).

However, when the population mean and standard deviation are not known, but instead are estimated from the sample data, then the usual Kolmogorov-Smirnov test based on the critical values in the [Kolmogorov-Smirnov Table](#) yields results that are too conservative. Lilliefors created a related test that gives more accurate results in this case (see [Lilliefors Test Table](#)).

The Lilliefors Test uses the same calculations as the Kolmogorov-Smirnov Test, but the table of critical values in the [Lilliefors Test Table](#) is used instead of the [Kolmogorov-Smirnov Table](#). Since the critical values in this table are smaller, the Lilliefors Test is less likely to show that data is normally distributed.

Example 1: Repeat Examples 1 and 2 of the [Kolmogorov-Smirnov Test for Normality](#) using the Lilliefors Test.

For Example 1 of [Kolmogorov-Smirnov Test for Normality](#), using the [Lilliefors Test Table](#), we have

$$D_{n,\alpha} = \frac{.895}{f(n)} \qquad f(n) = \frac{.83 + n}{\sqrt{n}} - .01$$

$$D_{1000,.05} = \frac{.895}{f(1000)} = \frac{.895}{31.64} = .0283 \qquad f(1000) = \frac{.83 + 1000}{\sqrt{1000}} - .01 = 31.64$$

Since $D_n = 0.0117 < 0.0283 = D_{n,\alpha}$, once again we conclude that the data is a good fit with the normal distribution. (Note that the critical value of .0283 is smaller than the critical value of .043 from the KS Test.)

For Example 2 of [Kolmogorov-Smirnov Test for Normality](#), using the [Lilliefors Test Table](#) with $n = 15$ and $\alpha = .05$, we find that $D_n = 0.1875 < 0.2196 = D_{n,\alpha}$, which confirms that the data is normally distributed.

Real Statistics Functions: The following functions are provided in the Real Statistics Resource Pack to automate the table lookup:

LCRIT($n, \alpha, tails, interp$) = the critical value of the Lilliefors test for a sample of size n , for the given value of alpha (default .05) and $tails = 1$ (one tail) or 2 (two tails, default) based on the [Lilliefors Test Table](#). If $interp = TRUE$ (default) harmonic interpolation is used; otherwise linear interpolation is used.

LPROB($x, n, tails, iter, interp, txt$) = an approximate p-value for the Lilliefors test for the D_n value equal to x for a sample of size n and $tails = 1$ (one tail) or 2 (two tails, default) based on a linear interpolation (if $interp = FALSE$) or harmonic interpolation (if $interp = TRUE$, default) of the critical values in the [Lilliefors Test Table](#), using $iter$ number of iterations (default = 40).

Note that the values for α in the table in the [Lilliefors Test Table](#) range from .01 to .2 (for $tails = 2$) and .005 to .1 for $tails = 1$. When $txt = FALSE$ (default), if the p-value is less than .01 ($tails = 2$) or .005 ($tails = 1$) then the p-value is given as 0 and if the p-value is greater than .2 ($tails = 2$) or .1 ($tails = 1$) then the p-value is given as 1. When $txt = TRUE$, then the output takes the form “< .01”, “< .005”, “> .2” or “> .1”.

For Example 2 of [Kolmogorov-Smirnov Test for Normality](#), $D_{n,\alpha} = \text{LCRIT}(15, .05, 2) = .2196 > .184 = D_n$ and p-value = $\text{LPROB}(0.184, 15) = .182858 > .05 = \alpha$, and so once again we can't reject the null hypothesis that the data is normally distributed.

Real Statistics Support for KS Test

[Click here](#) for information about the Real Statistics functions that perform the Kolmogorov-Smirnov test both when the mean and standard deviation are specified and when they are estimated from the data. Both raw data and a data in the form of a frequency table are supported.

Lilliefors Distribution

Especially for values of α not found in the [Lilliefors Test Table](#), we can use an approximation to the Lilliefors distribution. [Click here](#) for more information about this distribution, including some useful functions provided by the Real Statistics Resource Pack.

11 Responses to *Lilliefors Test for Normality*



Chris says:

August 10, 2019 at 6:43 am

Dear Charlie,

Thank you for your website, which is well written and particularly pedagogical.

I see a problem of principles in these tests of normality. In fact we don't test the hypothesis H_0 with an accuracy of α but we test the hypothesis H_1 (rejection) with this percentage.

For the KS test for example the higher the % (0.95 ; 0.99 ; 0.995 ; ...) and the lower the chance not to conclude H_1 and reject H_0 , so the "easier" to conclude it would be a Gaussian! That makes no sense.

When the test passes with success, that does not mean we have 95 % (or more) it is a Gaussian. It means that we can't say with 95 % chance it is something different. But the probability it is really a normal distribution is not known.

So shouldn't we always take at least 50 % (meaning 50 % or *less*) if we want to conclude distribution is a Gaussian ?
Indeed, to fairly conclude we have a "good" chance that it is a Gaussian, we should at least be allowed to say there is no 50 % chance it is something else...

[Reply](#).



Charles says:

August 10, 2019 at 9:47 am

Hello Chris,

This is the sort of issue we have with all statistical tests (at least the non-Bayesian tests). We don't know whether the data is really coming from a normal distribution whether the p-value is 50% or 2%. The value of 5% is arbitrary, but commonly used, compromise. Since rejection occurs for values less than α , the lower the α value the more

likely you are to declare the data as normally distributed. An alpha of 50% would increase the likelihood that you would declare the data as not normally distributed.

Charles

[Reply](#)



Mark G Filler says:

November 5, 2017 at 12:41 am

For LCRIT, I can't seem to get a value if $n > 50$. What am I doing wrong?

[Reply](#)



Charles says:

November 5, 2017 at 8:18 am

Mark,

I am not sure what you are doing wrong, but I just tried to use =LCRIT(60), and I got the value .114113. What version of Real Statistics are you using? You can find this out by entering the formula =VER()

Charles

[Reply](#)



Mark G Filler says:

November 5, 2017 at 8:46 pm

Charles

I am using 4.14 2010.

When I use Excel 2013 with the corresponding Real Statistics version, it works OK.

I don't like Excel 2013, so I guess this a cost of that attitude.

[Reply](#)



Mark G Filler says:

November 5, 2017 at 9:50 pm

Charles

Problem solved – I installed version 5.2 for Excel 2010 and LCRIT works for a sample size of 300.

Mark

[Reply](#)



[Charles](#) says:

November 6, 2017 at 8:28 am

Mark,
Good to hear.
Charles



David says:

August 7, 2017 at 9:28 pm

Hey Charles,

If I'm not mistaken, D_n from the Kolmogorov-Smirnov Test for Normality page should be $D_n = 0.1875$, not $D_n = 0.184$.

Thanks.

[Reply](#)



[Charles](#) says:

August 7, 2017 at 11:23 pm

David,
Yes you are correct. Thanks for catching this mistake. I really appreciate your helping in improving the Real statistics website.

Charles

[Reply](#)



Keith Wild says:

June 28, 2017 at 4:44 pm

Of the many tests regimes there are for tests for normality. Is there a list illustrating the order of preference for the test method according to the type of data you have?

I mean which test should I use for what type of data? It seems to be so easy to fudge a result as necessary according to the test method.

[Reply](#).



Charles says:

June 28, 2017 at 9:36 pm

Keith,

In general, I believe that the Shapiro-Wilk test is the best one to use. If you have a number of ties, then d'Agostino-Pearson is probably better.

Charles

[Reply](#).

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