Title

ksmirnov — Kolmogorov-Smirnov equality-of-distributions test

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Syntax

One-sample Kolmogorov-Smirnov test

```
ksmirnov varname = exp | if | in |
```

Two-sample Kolmogorov-Smirnov test

ksmirnov varname [if] [in], by(groupvar) [exact]

Menu

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Description

ksmirnov performs one- and two-sample Kolmogorov–Smirnov tests of the equality of distributions. In the first syntax, *varname* is the variable whose distribution is being tested, and *exp* must evaluate to the corresponding (theoretical) cumulative. In the second syntax, *groupvar* must take on two distinct values. The distribution of *varname* for the first value of *groupvar* is compared with that of the second value.

When testing for normality, please see [R] sktest and [R] swilk.

Options for two-sample test

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by (groupvar) is required. It specifies a binary variable that identifies the two groups. exact specifies that the exact p-value be computed. This may take a long time if n > 50.

Remarks and examples

stata.com

Example 1: Two-sample test

Say that we have data on x that resulted from two different experiments, labeled as group==1 and group==2. Our data contain

- . use http://www.stata-press.com/data/r13/ksxmpl
- . list

group	x
2	2
1	0
2	3
1	4
1	5
2	8
2	10
	2 1 2 1 1 1 2

We wish to use the two-sample Kolmogorov–Smirnov test to determine if there are any differences in the distribution of x for these two groups:

. ksmirnov x, by(gr	oup)					
Two-sample Kolmogor	ov-Smirno	v test for	r equality	of	distribution	functions
Smaller group	D	P-value	Corrected			
1:	0.5000	0.424				
2:	-0.1667	0.909				
Combined K-S:	0.5000	0.785	0.735			

The first line tests the hypothesis that x for group 1 contains *smaller* values than for group 2. The largest difference between the distribution functions is 0.5. The approximate p-value for this is 0.424, which is not significant.

The second line tests the hypothesis that x for group 1 contains *larger* values than for group 2. The largest difference between the distribution functions in this direction is 0.1667. The approximate *p*-value for this small difference is 0.909.

Finally, the approximate *p*-value for the combined test is 0.785, corrected to 0.735. The *p*-values ksmirnov calculates are based on the asymptotic distributions derived by Smirnov (1933). These approximations are not good for small samples (n < 50). They are too conservative—real *p*-values tend to be substantially smaller. We have also included a less conservative approximation for the nondirectional hypothesis based on an empirical continuity correction—the 0.735 reported in the third column.

That number, too, is only an approximation. An exact value can be calculated using the exact option:

. ksmirnov x, b	oy(group) exac	t				
Two-sample Kolm	nogorov-Smirno	v test for	equality	of	distribution functions	3
Smaller group	D	P-value	Exact			
1:	0.5000	0.424				
2:	-0.1667	0.909				
Combined K-S:	0.5000	0.785	0.657			

Example 2: One-sample test

Let's now test whether x in the example above is distributed normally. Kolmogorov–Smirnov is not a particularly powerful test in testing for normality, and we do not endorse such use of it; see [R] sktest and [R] swilk for better tests.

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In any case, we will test against a normal distribution with the same mean and standard deviation:

. summarize x					
Variable	Obs	Mean	Std. Dev.	Min	Max
x	7 4.	.571429	3.457222	0	10
. ksmirnov x = normal((x-4.571429)/3.457222)					
One-sample Kolmogorov-Smirnov test against theoretical distribution normal((x-4.571429)/3.457222)					
Smaller group	p D	P-value	Corrected		
x:	0.1650	0.683			
Cumulative:	-0.1250	0.803			
Combined K-S	. 0.1650	0.991	0.978		

Because Stata has no way of knowing that we based this calculation on the calculated mean and standard deviation of x, the test statistics will be slightly conservative in addition to being approximations. Nevertheless, they clearly indicate that the data cannot be distinguished from normally distributed data.

Stored results

ksmirnov stores the following in r():

Scalars			
r(D_1)	D from line 1	r(D)	combined D
r(p_1)	p-value from line 1	r(p)	combined <i>p</i> -value
r(D_2)	D from line 2	r(p_cor)	corrected combined p-value
r(p_2)	p-value from line 2	r(p_exact)	exact combined p-value
Macros			
r(group1)	name of group from line 1	r(group2)	name of group from line 2

Methods and formulas

In general, the Kolmogorov–Smirnov test (Kolmogorov 1933; Smirnov 1933; also see Conover [1999], 428–465) is not very powerful against differences in the tails of distributions. In return for this, it is fairly powerful for alternative hypotheses that involve lumpiness or clustering in the data.

The directional hypotheses are evaluated with the statistics

$$D^{+} = \max_{x} \left\{ F(x) - G(x) \right\}$$
$$D^{-} = \min_{x} \left\{ F(x) - G(x) \right\}$$

where F(x) and G(x) are the empirical distribution functions for the sample being compared. The combined statistic is

$$D = \max\left(\left|D^{+}\right|, \left|D^{-}\right|\right)$$

The *p*-value for this statistic may be obtained by evaluating the asymptotic limiting distribution. Let m be the sample size for the first sample, and let n be the sample size for the second sample. Smirnov (1933) shows that

$$\lim_{m,n\to\infty} \Pr\left\{\sqrt{mn/(m+n)}D_{m,n} \le z\right\} = 1 - 2\sum_{i=1}^{\infty} (-1)^{i-1} \exp\left(-2i^2 z^2\right)$$

The first five terms form the approximation P_a used by Stata. The exact *p*-value is calculated by a counting algorithm; see Gibbons and Chakraborti (2011, 236–238). A corrected *p*-value was obtained by modifying the asymptotic *p*-value by using a numerical approximation technique:

$$Z = \Phi^{-1}(P_a) + 1.04/\min(m, n) + 2.09/\max(m, n) - 1.35/\sqrt{mn/(m+n)}$$
p-value = $\Phi(Z)$

where $\Phi(\cdot)$ is the cumulative normal distribution.

Andrei Nikolayevich Kolmogorov (1903–1987), of Russia, was one of the great mathematicians of the twentieth century, making outstanding contributions in many different branches, including set theory, measure theory, probability and statistics, approximation theory, functional analysis, classical dynamics, and theory of turbulence. He was a faculty member at Moscow State University for more than 60 years.

Nikolai Vasilyevich Smirnov (1900–1966) was a Russian statistician whose work included contributions in nonparametric statistics, order statistics, and goodness of fit. After army service and the study of philosophy and philology, he turned to mathematics and eventually rose to be head of mathematical statistics at the Steklov Mathematical Institute in Moscow.

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Also see

- [R] runtest Test for random order
- [R] sktest Skewness and kurtosis test for normality
- [R] swilk Shapiro-Wilk and Shapiro-Francia tests for normality