

## Statistical notes for clinical researchers: assessing normal distribution (2) using skewness and kurtosis

**Hae-Young Kim**

Department of Dental Laboratory  
Science & Engineering, Korea  
University College of Health  
Science, Seoul, Korea

As discussed in the previous statistical notes, although many statistical methods have been proposed to test normality of data in various ways, there is no current gold standard method. **The eyeball test** may be useful for medium to large sized (e.g.,  $n > 50$ ) samples, however may not be useful for small samples. **The formal normality tests** including Shapiro-Wilk test and Kolmogorov-Smirnov test may be used from small to medium sized samples (e.g.,  $n < 300$ ), but may be unreliable for large samples. Moreover we may be confused because 'eyeball test' and 'formal normality test' may show incompatible results for the same data. To resolve the problem, another **method of assessing normality using skewness and kurtosis of the distribution may be used, which may be relatively correct in both small samples and large samples.**

### 1) Skewness and kurtosis

**Skewness is a measure of the asymmetry and kurtosis is a measure of 'peakedness' of a distribution.** Most statistical packages give you values of skewness and kurtosis as well as their standard errors.

In SPSS you can find information needed under the following menu:  
Analysis – Descriptive Statistics – Explore

Descriptives			Statistic	Std. Error
Y	Mean		100,8097	2,17292
	95% Confidence Interval for Mean	Lower Bound	96,4982	
		Upper Bound	105,1213	
	5% Trimmed Mean		100,7981	
	Median		97,7470	
	Variance		472,158	
	Std. Deviation		21,72920	
	Minimum		53,21	
	Maximum		151,76	
	Range		98,55	
	Interquartile Range		34,12	
	Skewness		,105	,241
	Kurtosis		-,621	,478

### \*Correspondence to

Hae-Young Kim, DDS, PhD.  
Associate Professor,  
Department of Dental Laboratory  
Science & Engineering, Korea  
University College of Health  
Science, San 1 Jeongneung 3-dong,  
Seongbuk-gu, Seoul, Korea 136-703  
TEL, +82-2-940-2845; FAX, +82-2-  
909-3502, E-mail, kimhaey@korea.  
ac.kr

Skewness is a measure of the asymmetry of the distribution of a variable. The skew value of a normal distribution is zero, usually implying symmetric distribution. A positive skew value indicates that the tail on the right side of the distribution is longer than the left side and the bulk of the values lie to the left of the mean. In contrast, a negative skew value indicates that the tail on the left side of the distribution is longer than the right side and the bulk of the values lie to the right

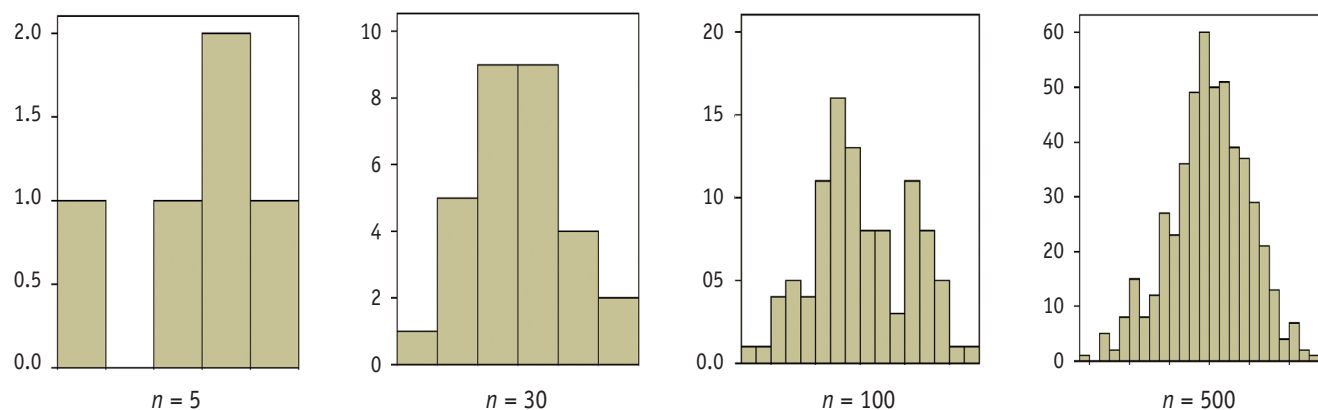
of the mean. West *et al.* (1996) proposed a reference of substantial departure from normality as an absolute skew value  $> 2$ .<sup>1</sup>

Kurtosis is a measure of the peakedness of a distribution. The original kurtosis value is sometimes called **kurtosis (proper)** and West *et al.* (1996) proposed a reference of substantial departure from normality as an absolute kurtosis (proper) value  $> 7$ .<sup>1</sup> For some practical reasons, most statistical packages such as SPSS provide '**excess kurtosis**' obtained by subtracting 3 from the kurtosis (proper). The excess kurtosis should be zero for a perfectly normal distribution. Distributions with positive excess kurtosis are called leptokurtic distribution meaning high peak, and distributions with negative excess kurtosis are called platykurtic distribution meaning flat-topped curve.

## 2) Normality test using skewness and kurtosis

A z-test is applied for normality test using skewness and kurtosis. A z-score could be obtained by dividing the skew values or excess kurtosis by their standard errors.

$$Z = \frac{\text{Skew value}}{SE_{\text{skewness}}}, Z = \frac{\text{Excess kurtosis}}{SE_{\text{excess kurtosis}}}$$



**Figure 1.** Histograms of a characteristic of interests in various sizes of samples.

**Table 1.** Skewness, kurtosis and normality tests for a characteristic of interests in various sizes of samples

Sample size (n)	Skewness	SE <sub>skewness</sub>	Z <sub>skewness</sub>	Kurtosis	SE <sub>kurtosis</sub>	Z <sub>kurtosis</sub>	Kolmogorov-Smirnov*		Shapiro-Wilk	
							Statistics	p-value	Statistics	p-value
5	-0.971	0.913	-1.064	0.783	2.000	0.392	0.191	0.200	0.948	0.721
30	0.285	0.427	0.667	0.463	0.833	0.556	0.068	0.200	0.988	0.976
100	0.105	0.241	0.436	-0.621	0.478	-1.299	0.076	0.167	0.983	0.216
500	-0.251	0.109	-2.303	0.094	0.218	0.431	0.044	0.020	0.993	0.029

\*Lilliefors significance correct

As the standard errors get smaller when the sample size increases, z-tests under null hypothesis of normal distribution tend to be easily rejected in large samples with distribution which may not substantially differ from normality, while in small samples null hypothesis of normality tends to be more easily accepted than necessary. Therefore, critical values for rejecting the null hypothesis need to be different according to the sample size as follows:

1. For small samples ( $n < 50$ ), if absolute z-scores for either skewness or kurtosis are larger than 1.96, which corresponds with a alpha level 0.05, then reject the null hypothesis and conclude the distribution of the sample is non-normal.
2. For medium-sized samples ( $50 < n < 300$ ), reject the null hypothesis at absolute z-value over 3.29, which corresponds with a alpha level 0.05, and conclude the distribution of the sample is non-normal.
3. For sample sizes greater than 300, depend on the histograms and the absolute values of skewness and kurtosis without considering z-values. Either an absolute skew value larger than 2 or an absolute kurtosis (proper) larger than 7 may be used as reference values for determining substantial non-normality.

Referring to Table 1 and Figure 1, we could conclude all the data seem to satisfy the assumption of normality

despite that the histogram of the smallest-sized sample doesn't appear as a symmetrical bell shape and the formal normality tests for the largest-sized sample were rejected against the normality null hypothesis.

### 3) How strict is the assumption of normality?

Though the humble  $t$  test (assuming equal variances) and analysis of variance (ANOVA) with balanced sample sizes are said to be 'robust' to moderate departure from normality, generally it is not preferable to rely on the feature and to omit data evaluation procedure. A combination of visual inspection, assessment using skewness and kurtosis, and formal normality tests can

be used to assess whether assumption of normality is acceptable or not. When we consider the data show substantial departure from normality, we may either transform the data, e.g., transformation by taking logarithms, or select a nonparametric method such that normality assumption is not required.

### Reference

1. West SG, Finch JF, Curran PJ. Structural equation models with nonnormal variables: problems and remedies. In RH Hoyle (Ed.). Structural equation modeling: Concepts, issues and applications. Newbery Park, CA: Sage; 1995. p56-75.