DETERMINATION OF GEOMETRIC STANDARD DEVIATION FOR

DISSOLUTION

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ABSTRACT

The two important instances in which the scientist experimental data to a logarithmic converts his/her prior to computing the mean and deviation, are (i) when the distribution of the data is asymmetrical (e.g. percentage data) and (ii) when he/she intends to compare statistically the averages of two or more groups with unequal standard deviations. In either case, the mean is restored to its original scale by taking the anti-log of the log mean, which geometric However, this procedure cannot mean. applied for computing the geometric standard deviation. The author of reference(1) erroneously claims that the anti-log of log standard deviation is the geometric standard deviation. This paper demonstrates procedure in incorrectness of the reference exhibits the exact statistical formula and introduces a novel method called "jackknife statistic" to confirm the results, based on the dissolution data associated with Product-C.

1381



INTRODUCTION

The two important reasons for transforming the original scale of measurement of any pharmaceutical experimental data (e.g. dissolution, tablet strength) to a logarithmic scale prior to determining the mean and standard deviation are (i) when the frequency distribution of the response data is asymmetrical (skewed) (2) and (ii) when a statistical test for comparing the means of two or more groups (e.g. formulations) with distinctly unequal standard deviations, is undertaken (2,3). The logarithmic transformation has the unique property of achieving symmetricity of the data and of stabilizing the variances of the groups to be compared (2,3). log mean (LM) and log standard deviation (LSD) are not meaningful for the purpose of interpretation, one computes the anti-log of LM to obtain the geometric mean (GM) [GM = $10^{\overline{X}}$, where \overline{X} = LM, X = log Y and Y = % dissolution] and computes the following formula to obtain the geometric standard deviation (GSD): $GSD = [(2.303)(GM)(S_X)], where S_X = LSD (see derivation)$ next section). However, the author (TK) of reference (1) advocates that the geometric standard deviation should be computed by taking the anti-log of LSD (10Sx) just as one obtains the GM by taking the anti-log of LM (10^{X}) (1). (In the text this formula will be referred to as "ref(1) formula"). Because this formula is simpler and seems logical, several pharmaceutical scientists and statisticians associated with the pharmaceutical development, production and quality control departments brought the paper to the attention of the author (NRB) to examine the validity of the ref (1) formula. In the course of this paper, it will be shown conclusively that the ref(1) formula is absolutely To begin with, consider a sample of 5 incorrect.



dissolution values, 90, 80, 90, 90 and 90, and their respective log values, 1.9542, 1.9031, 1.9542, 1.9542 The GM and GSD are 87.90 and 4.6311, respectively. However, the ref(1) formula yields 1.054 The mean (M) and standard deviation since LSD = 0.0229. (SD) for the same data without the log transformation are 88.0 and 4.472, respectively. The GSD value is much closer to the SD than the value given by the ref(1) Indeed, if the dissolution values were all 90, then the ref(1) formula would yield a value of 1.0, since LSD = 0.0. Therefore this is a contradiction, and the ref(1) formula is incorrect and should not be used In the following sections, the derivation in practice. of the GSD formula and the results of a dissolution experiment will be presented. In addition, a new and more accurate method of determining the GSD, called the jackknife statistic (JK) will be introduced (2,4,5). Note that, GM(JK) and GSD(JK) for the above data are 87.9 and 4.6716, respectively. This confirms the values of the GM and GSD given above.

THEORY

Let Y_1 , Y_2 , ---, Y_n be a random sample of n dissolution values and let $X_i = log_{10}Y_i$ (i = 1,2---n). Then, by definition, $e^{2.303x}$ is the geometric mean of the Y_i 's, where, $\bar{X} = \Sigma X/n$. The interest here is to derive the variance (V) of $e^{2.303\overline{x}}$, a function of \overline{x} . (Note that the square root of V provides the GSD). general terms (3), let the function be $F(\theta_1, \theta_2, --- \theta_k)$ and $V[F(\Theta)] = (\delta F/\delta \Theta_1)^2 V(\Theta_1) + (\delta F/\delta \Theta_2)^2 V(\Theta_2) + ---- +$ $(\delta F/\delta \theta_k) 2V(\theta_k) + \text{covariance terms},$ $(\delta F/\delta \Theta_{\dot{1}})(\delta F/\delta \Theta_{\dot{1}})COV(\Theta_{\dot{1}},\Theta_{\dot{1}})$ i,j=1,2,---,k $(i \neq j)$. Note that, covariance terms do not appear in the case being considered.



For the present derivation, F $(\bar{X}) = e^{2.303\bar{X}}$ $V[F(\bar{x})] = [\delta F/\delta \bar{x}]^2 V(\bar{X}) = (2.303)^2 [e^{2.303\bar{x}}]^2 V(\bar{X})$ = $(2.303)^2[e^{2.303x}]^2S^2_x/n$ $= (2.303)^2 (GM)^2 S^2 v/n$

From the above expression, the formula for the GSD is obtained as.

Geometric Standard Deviation = $(2.303)(GM)(S_X)$ Geometric Standard Error = $(2.303)(GM)(S_{y})$ where $(S_{\mathbf{x}}^-) = S_{\mathbf{x}}/\sqrt{n}$.

This method of derivation is generally known as the Taylor series expansion method, error propagation method or statistical differential method (3,5).

Jackknife Statistic (2,4,5):

in the sample.

The derivation of the variance of a complicated function of the estimate, such as the one for the GM, results in a complex expression. The jackknife statistic not only provides a simpler means to determine the variance but also it has many desirable statistical properties. Because it requires a large number of repetitive computations, one would need to use either a programmable calculator or a desk-top PC. X_1 , X_2 , ---- X_n be a sample with n observations. Q(all) denote the estimator of the parameter GM based on all the available data. Let Q(i) denote the estimator based on the sample size of (n-1), where the ith observation has been deleted (skipped). The pseudo values arè computed as follows: $Q^*(i) = n[Q(all)] - (n-1)[Q(i)],$ i = 1, 2, ---, n. The jackknife estimate of $GM(JK) = \Sigma Q^*(i)/n$ and the jackknife estimate of $GSD(JK) = [\Sigma[Q^*(i) - GM(JK)]^2/n-1]^{\frac{1}{2}}$. The added accuracy of the estimate emanates from the fact that one has to compute as many pseudo-values as there are observations



TABLE I GEOMETRIC STANDARD DEVIATION AND GEOMETRIC CONFIDENCE LIMITS FOR PRODUCT-C DISSOLUTION

NO.	STATISTIC	BATCH-P	BATCH-Q	BATCH-R
1	n	10	10	10
2	range	95-99	86-100	83-102
3	GŃ	97.3868	96.4051	96.6496
4	GSD	1.4354	4.3738	5.9651
5	95% GCL	96.36-98.41	93.28-99.53	92.38-100.92
6	GM(JK)	97.3859	96.3353	96.7114
7	GSD(JK)	1.4430	4.4391	5.9842
8	95% GCL(JK)	96.35-98.42	93.22-99.45	92.43-100.99
9	LSD	0.0064	0.0197	0.0268
10	Ref(1) formula for GSD	1.0150	1.0464	1.0637

GM= Geometric Mean, GSD = Geometric Standard ion, GCL = Geometric Confidence , JK - Jackknife, LSD = Log Standard Deviation

DISSOLUTION STUDY: RESULTS AND DISCUSSION

The dissolution testing was performed on the three selected batches, P, Q and R, of Product-C, using 100 RPM basket method with 900 ml. HCl (1:100) medium at 30 The statistical analysis was conducted on the minutes. spectrophotometric assay values of the 30-minute samples expressed as a percent of label claim. The statistical results are presented in Table I.

A cursory examination of the results indicates that the two methods of determination, statistical differential and jackknife statistic, yielded essentially identical The unanimity of the results of the two methods unequivocally demonstrates that the ref(1) formula is erroneous and the values given in the last row of Table I are absolutely meaningless. depicts the detailed computation of the GSD(JK) for The contents are self-explanatory. Batch-R.



1386 BOHIDAR

ABLE-II

COMPUTATION OF JACKKNIFE STATISTICS FOR BATCH-R

Skip 10	2.0000 1.9956 1.9868 1.9638 2.0043 1.9823 2.0086 1.9191	1.9830 96.161 0.0275 101.05
Skip 9	2.0000 1.9956 1.9868 1.9638 2.0043 1.9823 2.0086 1.9191	1.9850 96.605 0.0284 97.050
Skip 8	2.0000 1.9956 1.9868 1.9638 2.0043 1.9823 2.0086 1.9868	1.9925 98.288 0.0141 81.905
Skip 7	2.0000 1.9956 1.9868 1.9638 2.0043 1.9823 1.9191 1.9868	1.9826 96.073 0.0270 101.84
Skip 6	2.0000 1.9956 1.9868 1.9638 2.0043 2.0086 1.9191 1.9868 2.0043	1.9855 96.716 0.0284 96.048
Skip 5	2.0000 1.9956 1.9868 1.9638 1.9823 2.0086 1.9191 1.9868 2.0043	1.9830 96.161 0.0275 101.045
Skip 4	2.0000 1.9956 1.9868 2.0043 1.9823 2.0086 1.9191 1.9868 2.0043	1.9875 97.163 0.0273 92.031
Skip 3	2.0000 1.9956 1.9638 2.0043 1.9823 2.0086 1.9191 1.9868 2.0043	1,9850 96,605 0,0284 97,050
Skip 2	2.0000 1.9868 1.9638 2.0043 1.9823 2.0086 1.9191 1.9868	1.9840 96.383 0.0281 99.050
Skip 1	1.9956 1.9868 1.9638 2.0043 1.9823 2.0086 1.9191 1.9868 2.0043	1.9835 96.272 0.0279 100.05 96.711 5.9842
LOG Y	2.000 1.9956 1.9868 1.9638 2.0043 1.9823 2.0086 1.9191 1.9868	1.9852 96.650 0.0268 7alues:*
¥	100 99 97 92 101 102 83 83	LM: GN: LSD: Pseudo Val GN(JK): GSD(JK)

* Pseudo Value = $10 \times G.M.$ (all) - $9 \times (G.M.$ (i), where i = 1,2,....10



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