

THE STANDARD ERROR OF THE MAGNITUDE-FREQUENCY b VALUE

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ABSTRACT

Estimated b values in $\log N = a - bM$ are widely used in seismicity comparisons and risk analysis, but uncertainties have been little explored. In this paper, the usual F probability density distribution for b is given and compared with an asymptotic form for temporally varying b . Convenient tables for the standard error of b are given that allow statistical tests to accompany investigations of both temporal and spatial variations of b . With large samples and slow temporal changes in b , the standard error of b is

$$\sigma(\hat{b}) = 2.30b^2\sigma(\bar{M}),$$

where

$$\sigma^2(\bar{M}) = \sum_{i=1}^n (M_i - \bar{M})^2 / n(n-1).$$

In an example from central California, stable estimates of b require a space-time window containing about 100 earthquakes. From 1952 to 1978, the average b and 90 per cent confidence limits are 0.95 (+0.94, -0.30). Some fluctuations of b are statistically significant but some are not. Within 90 per cent confidence limits, b changes from a low of 0.60 (+0.11, -0.09) in 1955 to a high of 1.39 (+0.25, -0.21) in 1967 and drops to 0.72 (+0.13, -0.10) in 1975. In this example, no correlation between large earthquakes ($M > 5$) and b variations occurred.

INTRODUCTION

The b value in Gutenberg and Richter's relation (1954)

$$\log N = a - bM \tag{1}$$

has been widely used in research on seismicity and seismic risk analysis. In laboratory experiments on rocks, the occurrence of microcracks follows a similar form, and the b value is inversely related to the stress level (Scholz, 1968). For some earthquake sequences, the b value is reported to change before the main shock (e.g., Suyehiro, 1966; Gibowicz, 1973; Robinson, 1979; Smith, 1981). This property is consistent with an initially high stress level followed by a stress drop due to the main shock and it has been considered useful in earthquake prediction (Scholz, 1968; Li *et al.*, 1978).

For use in prediction and comparative mechanism studies, standard errors of b must be computed so that meaningful statistical tests can be performed. Because earthquakes are stochastic processes and the b value is a random variable, the probability distribution and the variance of b are essential in studying its temporal and spatial variations.

The b value can be calculated by least-squares regression, but the presence of even a few large earthquakes influences the resulting b value significantly. In least-squares regression of equation (1), the variance of one observation, $\sigma^2 = \sum r_i^2 / (n-1)$, is only a measure of the deviation of the data points from the linear curve. The problem is to derive a formula for an estimation of the variance of b itself.

As an alternative, the maximum likelihood method has been suggested as pref-

erable for calculating the b value because it yields a more robust estimate when the number of the infrequent large earthquakes changes. There will be cases, however, such as estimating the probability of the largest magnitude of earthquakes from relation (1), where the least-squares method is more suitable.

Suppose the probability density of magnitude M is exponential with mean $1/\beta$, i.e.,

$$\beta \exp\{-\beta(M - M_0)\}, \quad M > M_0. \quad (2)$$

The maximum likelihood estimation of β ($b = \beta \log e$) is (Aki, 1965; Utsu, 1965)

$$\hat{\beta} = \frac{1}{\bar{M} - M_0}, \quad (3)$$

where \bar{M} is the mean magnitude and M_0 is the smallest (or "threshold") magnitude considered. In these seminal papers, only an approximate formula to estimate the variance of b was given.

First, confidence limits need to be defined. If we form a linear combination Y so that

$$Y = \sum_{i=1}^n y_i, \quad (4)$$

where, in terms of the density function, $y_i = \partial/\partial\beta \log f(M_i, \beta)$, it can then be shown (e.g., Aki, 1965) that for large n , Y has a normal distribution with a zero mean and variance n/β^2 .

The probability that

$$-d_\epsilon \leq \frac{\beta Y}{\sqrt{n}} \leq d_\epsilon \quad (5)$$

is given by

$$\epsilon = \frac{1}{\sqrt{(2\pi)}} \int_{-d_\epsilon}^{d_\epsilon} \exp(-x^2/2) dx. \quad (6)$$

Then $\epsilon = 95$ per cent if $d_\epsilon = 1.96$ and $\epsilon = 90$ per cent if $d_\epsilon = 1.65$. When defined in this way, the confidence level is directly related only to Y , and consequently, it is difficult to assess information about the distribution of b itself.

More directly, the distribution of $\hat{\beta}$ is

$$\hat{\beta} \stackrel{d}{=} 2n\beta/\chi_{2n}^2 = \beta F_{\infty, 2n}. \quad (7)$$

This chi-squared form is given by Utsu (1966) and rederived by Zhang and Song (1981). In brief,

$$\begin{aligned} \bar{M} - M_0 &= \frac{1}{n} \sum_{i=1}^n (M_i - M_0) \\ &= \frac{1}{n} \sum_{i=1}^n e_i/\beta, \end{aligned}$$

where e_i are independent unit exponentials that are distributed like $\chi^2/2$. Then, from the additivity property of χ^2 ,

$$\bar{M} - M_0 \stackrel{d}{=} \chi_{2n}^2/2n\beta.$$

It follows from (7) (see Abramowitz and Stegun, 1965, 26.6.3) that

$$E(\hat{\beta}) = \beta n/(n-1), \quad n > 1 \quad (8)$$

$$\text{var } \hat{\beta} = \beta^2 n^2 / [(n-1)^2(n-2)], \quad n > 2. \quad (9)$$

Zhang and Song (1981) concluded from (8) that the maximum likelihood method yields a biased estimation of b , and they also investigated, by a Monte Carlo procedure, the least-squares estimate of the standard error of b . The formula β/\sqrt{n} (Aki, 1965) is an approximation to (9).

Once the distribution is derived, confidence intervals can be found by using tables of F or χ^2 distributions, e.g., for 0.90 per cent confidence level

$$F_{\infty, 2n}(0.05) < \frac{\hat{b}}{b} < F_{\infty, 2n}(0.95) \quad (10)$$

or

$$\chi_{2n}^2(0.05) < \frac{2nb}{\hat{b}} < \chi_{2n}^2(0.95). \quad (11)$$

Appropriate tables are not readily available in the seismological literature for $n > 50$, but graphs giving the cumulative distribution of the ratio (10) are given by Utsu (1966).

In the above work, b was considered a constant in time. We will discuss the distribution of b and estimate its standard error for both constant and slowly changing b . Our results are then checked by comparison with real earthquake sequences.

NUMERICAL FORMULA

We will now take an alternative approach to the distribution of b and consider the distribution of M_i without any assumption of the constancy of b . By the central limit theorem, the distribution function of \bar{M} , $\bar{M} = \frac{1}{n} \sum_{i=1}^n M_i$, approaches a normal distribution for large n if each M_i has finite mean and variance, and the M_i are independent. Therefore, in the limit, the distribution of \bar{M} can be expressed as

$$f(\bar{M}) = \frac{1}{\sqrt{2\pi \text{var } \bar{M}}} \exp(-(\bar{M} - E(\bar{M}))^2/2 \text{var } \bar{M}).$$

From (2),

$$E(M) = M_0 + \frac{1}{\beta} \quad (12)$$

$$\text{var } M = \frac{1}{\beta^2}$$

$$E(\bar{M}) = E(M) = M_0 + \frac{1}{\beta}$$

$$\text{var } \bar{M} = \frac{1}{n} \text{var } M = 1/n\beta^2.$$

Because $\beta(\bar{M})$ is a monotonically decreasing function of \bar{M} , there exists a single-valued inverse function

$$\bar{M}(\hat{\beta}) = \frac{1}{\hat{\beta}} + M_0. \quad (13)$$

Then the distribution of \hat{b} is

$$\begin{aligned} g(\hat{b}) &= \left| f(\bar{M}(b)) \frac{d\bar{M}(\hat{b})}{d\hat{b}} \right| \\ &= \frac{1}{\sqrt{2\pi \text{var } \bar{M}}} \frac{\log e}{\hat{b}^2} \exp(-(1/b - 1/\hat{b})^2 \log^2 e / 2 \text{var } \bar{M}) \end{aligned} \quad (14)$$

where $\hat{b} = \log e / [E(\bar{M}) - M_0]$. Typical results are shown in Figure 1.

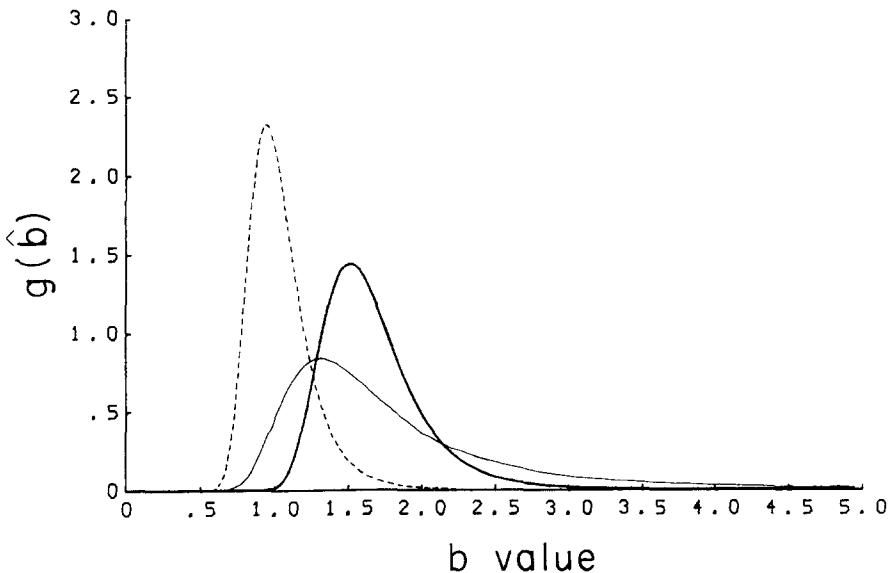


FIG. 1. Density functions $g(\hat{b})$ calculated from (14) and (15a). Light line for $b = 1.6$, $\sigma\bar{M} = 0.1$; heavy line for $b = 1.6$, $n = 20$ (or $\sigma\bar{M} = 0.05$); dashed line for $b = 1.0$, $n = 30$.

Here the distribution function of \hat{b} , is given for large n ; the mean and variance of \bar{M} , and then of \hat{b} , can be calculated from earthquake catalogs. Statistical tests can, as a consequence, be carried out conveniently by table reference. Because the value of b in a region is observed to change with time and location, its mean and variance might also be expected to change. In other words, b should be regarded, in general, as a nonstationary stochastic process $b(t)$. When sampling with a small time and space window, however, b can be usually taken as stationary and with constant expectation. Then, the above formulas hold and the distribution function $g(\hat{b})$ can be expressed simply as

$$g(\hat{b}) = \sqrt{\frac{n}{2\pi}} \frac{b}{\hat{b}^2} \exp(-n(b/\hat{b} - 1)^2/2). \quad (15a)$$

For comparison, the exact formula for $g(\hat{b})$ for constant b is (Utsu, 1966)

express the sampling weight at different times. Then

$$g(\hat{b}) = \frac{n^n}{(n-1)!} \frac{b^n}{\hat{b}^{n+1}} \exp(-nb/\hat{b}). \quad (15b)$$

Sample confidence levels, calculated from (15) are given in Table 1 and may be

TABLE 1
PERCENTAGE POINTS FOR THE DISTRIBUTION OF $g(\hat{b})$ AND n

b/n	Confidence Level 0.90			
	30	50	100	200
0.4	-0.10	-0.08	-0.06	-0.04
	0.14	0.11	0.07	0.05
0.5	-0.13	-0.10	-0.07	-0.05
	0.18	0.13	0.09	0.06
0.6	-0.15	-0.12	-0.09	-0.06
	0.21	0.16	0.11	0.07
0.7	-0.18	-0.14	-0.10	-0.08
	0.25	0.19	0.13	0.09
0.8	-0.20	-0.16	-0.12	-0.09
	0.29	0.21	0.14	0.10
0.9	-0.23	-0.18	-0.13	-0.10
	0.32	0.24	0.16	0.11
1.0	-0.25	-0.20	-0.15	-0.11
	0.36	0.27	0.18	0.12
1.1	-0.28	-0.22	-0.16	-0.12
	0.39	0.29	0.20	0.14
1.2	-0.30	-0.24	-0.18	-0.13
	0.43	0.32	0.22	0.15
1.3	-0.33	-0.26	-0.19	-0.14
	0.46	0.35	0.24	0.16
1.4	-0.35	-0.28	-0.21	-0.15
	0.50	0.37	0.25	0.17
1.5	-0.38	-0.30	-0.22	-0.16
	0.54	0.40	0.27	0.19
1.6	-0.40	-0.32	-0.24	-0.17
	0.57	0.43	0.29	0.20
1.7	-0.43	-0.35	-0.25	-0.18
	0.61	0.45	0.31	0.21
1.8	-0.45	-0.37	-0.27	-0.19
	0.64	0.47	0.33	0.22
1.9	-0.48	-0.39	-0.28	-0.21
	0.68	0.51	0.34	0.24
2.0	-0.50	-0.41	-0.30	-0.22
	0.71	0.53	0.36	0.25

checked against standard χ^2 tables. Differences between (15a) and (15b) are not large for large n .

Formula (15b) is important not only because of the assumption that a constant b within a small sampling window may be acceptable in many cases, but also because most published b values are given without standard errors and var \bar{M} . This omission means that they cannot be used with formula (14) nor for quantitative statistical tests. Formula (15b) allows, however, the upper limits of the standard error of b

values to be estimated, as illustrated later. In this way, b values can be used for statistical tests.

When b changes with time, a distribution function of b , $f(b, t)$, can be assigned and

$$\text{var } b = \int (b - E(b))^2 f(b, t) db dt. \quad (16)$$

If b changes slowly with time, $f(b, t)$ can be decomposed to

$$f(b, t) = g(b)w(t), \quad (17)$$

where $g(b)$ is defined earlier. At any time, b has the same form of distribution function as (14), although its mean and variance may change. Let function $w(t)$ express the sampling weight at different times. Then

$$\begin{aligned} \text{var } b &= \int \{b - E(b)\}^2 g(b)w(t) db dt \\ &= \int \{(b - \bar{b}_t) + (\bar{b}_t - E(b))\}^2 g(b)w(t) db dt \\ &= \langle \text{var } b_t \rangle + \text{var } \bar{b}, \end{aligned} \quad (18)$$

where b_t is the expectation of b at time t , i.e.,

$$\bar{b}_t = \int b g(b) db \quad (19)$$

and $\langle \rangle$ indicates the time average. Thus, because the cross-product terms vanish, the variance of b is composed of two parts. The first part is the time average of the variance of b at different times as calculated previously. The second part is the variance of \bar{b} , the mean of b averaged over the whole time range.

Because $\text{var } b \geq \langle \text{var } b_t \rangle$, formula (15b) may give an underestimated standard error when the temporal and spatial window is too large for b to be constant. In this

case, we have to estimate the average \bar{M} and $\text{var } \bar{M} = \sum_{i=1}^n (M_i - \bar{M})^2 / n(n-1)$, and then determine the probability interval of b from Table 2, computed using (14).

In the general case, a large sample distribution of $\hat{\beta}$ can be derived in the usual way. Let $X_i = M_i - M_0$. Then $\hat{\beta} = 1/\sqrt{X}$. If X_i has mean μ and variance σ^2 , it can be shown as a special case of (14) that $\hat{\beta}$ is asymptotically normal with mean $1/\mu$ and variance

$$(1/\mu^2)^2 \sigma^2 / n = \beta^4 \sigma^2 / n. \quad (20)$$

Here σ^2 may be estimated, as above, from $\sum_{i=1}^n (M_i - \bar{M})^2 / (n-1)$. Note that $\log \hat{\beta}$ is also asymptotically normal with mean $\log \beta$ and variance $\beta^2 \sigma^2 / n$. These functions may also fluctuate with time. It also follows that for large n ,

$$\sigma(\hat{b}) = \frac{\hat{b}^2}{\log e} \sigma(\bar{M}) = 2.30 \hat{b}^2 \sigma(\bar{M}). \quad (21)$$

In practice, starting from an earthquake catalog, we can estimate b by maximum likelihood from \bar{M} using (3). We can also obtain a probability interval for b from

formula (15), (Table 1) or formula (14), (Table 2) depending on whether the b value can be regarded as an approximate constant or not. Brillinger (personal communication) describes the model contemplated as "doubly stochastic." The process $b(M, t)$ may, of course, depend not only on t , but also on M , and it is an assumption that the distribution of \hat{b} remains exponential in these cases.

TABLE 2
PERCENTAGE POINTS FOR THE DISTRIBUTION OF $g(\hat{b})$ AND $\sigma\bar{M}$

$b/\sigma\bar{M}$	Confidence Level 0.90			
	0.15	0.10	0.05	0.03
0.4	-0.07	-0.05	-0.03	-0.01
	0.12	0.07	0.04	0.02
0.5	-0.11	-0.08	-0.04	-0.02
	0.20	0.12	0.05	0.03
0.6	-0.15	-0.11	-0.06	-0.04
	0.31	0.18	0.08	0.05
0.7	-0.20	-0.14	-0.08	-0.05
	0.46	0.26	0.11	0.06
0.8	-0.25	-0.18	-0.10	-0.06
	0.67	0.35	0.15	0.08
0.9	-0.30	-0.23	-0.13	-0.08
	0.94	0.47	0.19	0.11
1.0	-0.36	-0.27	-0.16	-0.10
	1.32	0.61	0.24	0.13
1.1	-0.42	-0.32	-0.19	-0.12
	1.84	0.79	0.29	0.16
1.2	-0.50	-0.38	-0.22	-0.14
	3.20	0.99	0.36	0.19
1.3	-0.55	-0.43	-0.25	-0.16
	3.67	1.26	0.43	0.23
1.4	-0.62	-0.48	-0.29	-0.19
	5.44	1.58	0.51	0.27
1.5	-0.69	-0.54	-0.33	-0.22
	8.64	1.98	0.60	0.31
1.6	-0.76	-0.60	-0.37	-0.24
	15.97	2.46	0.70	0.36
1.7	-0.84	-0.66	-0.41	-0.27
	47.97	3.08	0.81	0.41
1.8	-0.91	-0.73	-0.46	-0.30
	—*	3.86	0.93	0.47
1.9	-0.99	-0.79	-0.50	-0.32
	—	4.88	1.07	0.53
2.0	-1.07	-0.86	-0.55	-0.37
	—	6.26	1.22	0.59

* — > 100.

CENTRAL CALIFORNIA SEISMICITY

The seismicity in central California is taken as an example. The catalog is considered complete for earthquakes with $M \geq 2.5$ in this region since 1952 (Bolt and Miller, 1975). The sample has 2572 earthquakes in 22 yr.

Through a time window containing $n = 30, 50, 100$, and 200 earthquakes, respectively, b values were scanned in the time domain. Two distinct $b(t)$ curves for $n = 30$ and $n = 200$ are plotted in Figures 2 and 3. Error bars for the 90 per cent

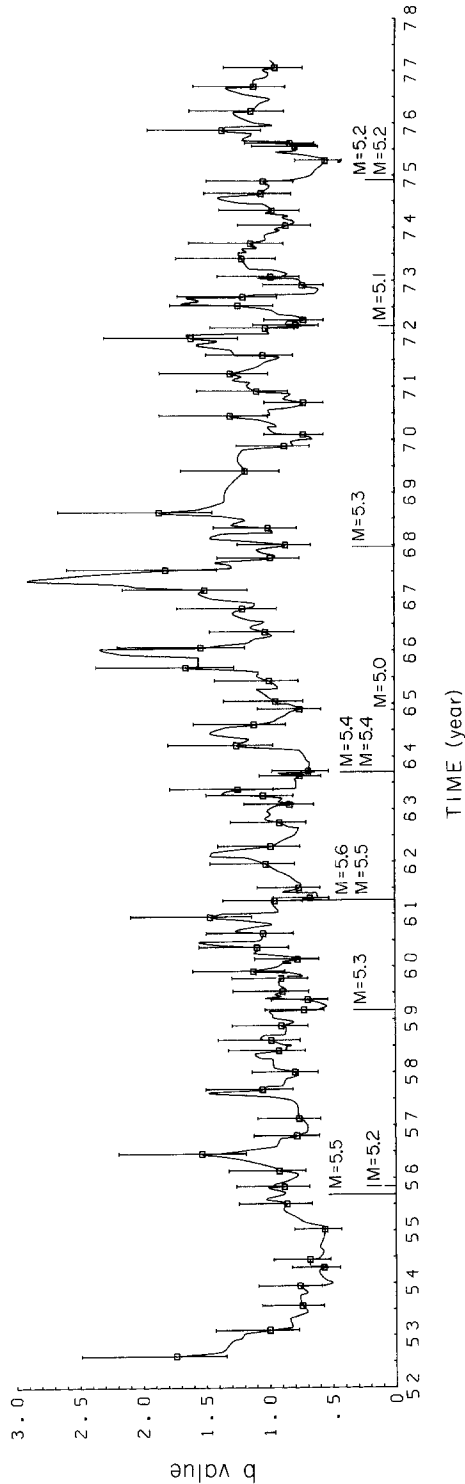


FIG. 2. Variation in b values and 90 per cent probability intervals for central California earthquakes ($M_L > 2.5$) in the period 1952 through 1978. Scanned with a time window containing $n = 30$ earthquakes. Error bars at the 90 per cent level are plotted for every five data points.

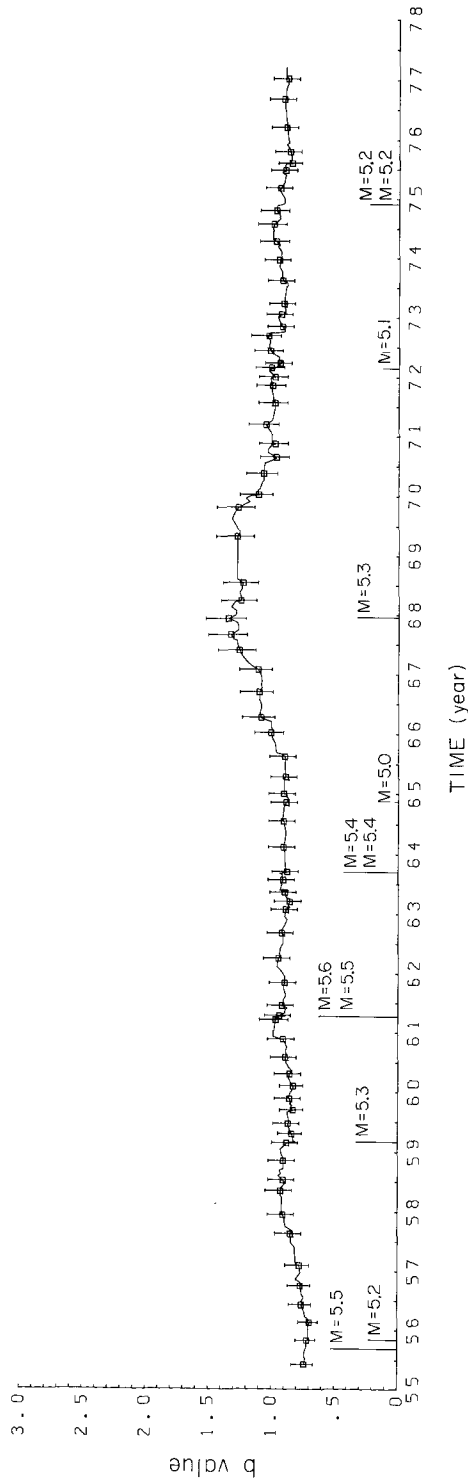


FIG. 3. As in Figure 2, but scanned with a time window containing $n = 200$ earthquakes.

confidence level are evaluated from Table 1 (formula 15). For comparison, error intervals were also evaluated from Table 2 (formula 14). These comparisons show that although for small time windows (e.g., $n = 30$), the estimates sometimes differ by over 100 per cent, for larger time windows (e.g., $n = 100$), the error intervals calculated from the two formulas are almost the same. This implies that the variance of b can be estimated from previous published data of b , even if $\text{var } \bar{M}$ is not given, provided n is not too small.

For a time window containing $n = 30$ earthquakes, the $b(t)$ curve has more fluctuations than for the case $n = 200$. However, many such fluctuations are not statistically significant when the chance variation is considered, and this result must be taken into account if changes of b value are used as a precursor in earthquake prediction [c.f. Wyss and Lee (1973)].

Over a longer time interval, significant temporal variations of b are obvious in Figure 3. (Alternatively, $\log \hat{b}$ could be graphed in Figures 2 and 3 so that the

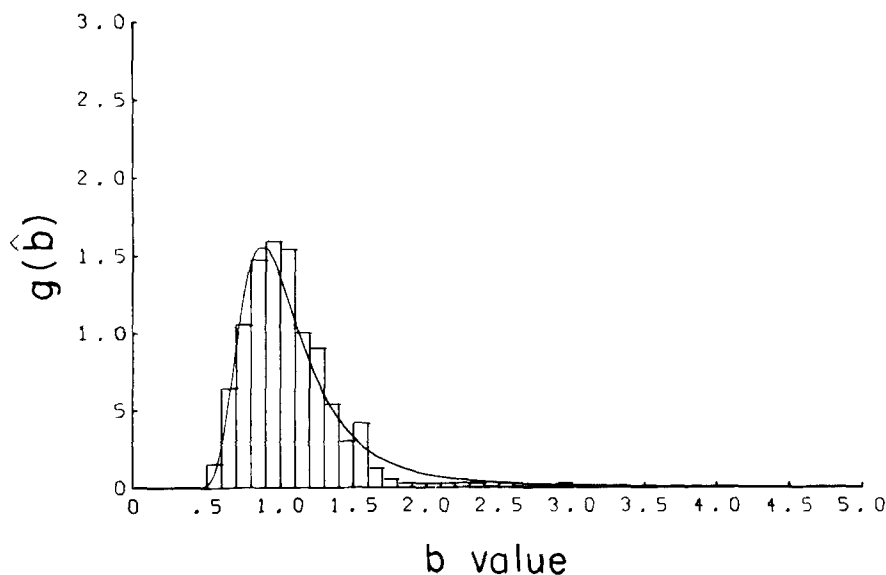


FIG. 4. Histogram of b distribution for central California seismicity, from 397 windows each containing 100 earthquakes, compared with theoretical curve $\hat{b} = 0.95$, $\sigma \bar{M} = 0.13$.

confidence intervals have constant width. The hypotheses of time invariant b may then be tested by calculating the proportion of \hat{b} values that exceed the intervals.) The times of occurrence of earthquakes larger than magnitude 5 are also plotted in the figure. The value of b changes from a low level of 0.60 (+0.11, -0.09) in 1955 to a high value of 1.39 (+0.25, -0.21) in 1967 and drops again to 0.72 (+0.13, -0.10) in 1975. However, we can find no correlation between changes of b and the time of occurrence of the larger earthquakes. The meaning of this longer term variation of b is not yet clear.

Finally, the histogram of observed b since 1952 is plotted in Figure 4. The theoretical distribution curve of $\hat{b} = 0.95$ and $\sigma \bar{M} = 0.13$ is also drawn. Because within the 22-yr period b changed with time, formula (15) cannot be used. The theoretical curve is calculated from (14) and a χ^2 test made with the histogram grouped into 10 intervals. The computed $\chi^2_{10} = 17.5$ is close to the expected value of 18.3 for 10 degrees of freedom at the 5 per cent significance level. Therefore, the

evidence does not suggest that the actual distribution differs from the derived theoretical form at this level.

It should be emphasized that the uncritical use of formula (15) leads to a 90 per cent confidence level estimate for b of 0.95 (+0.17, -0.14) and a standard error of 0.10. In a similar way, the simple application of the formula of Aki (1965) or Zhang and Song (1981) yields 0.95 (± 0.10) for the standard error. Slow changes of b with time entail a broader b distribution. From equation (21), the large sample approximation gives 0.95 (± 0.27) for the standard error for the observed distribution of slowly varying b in central California. We calculate also that the 90 per cent confidence level of b is more realistically 0.95 (+0.94, -0.30). The result implies that even large differences between long-term average b values for different regions or epochs or focal depths (Gutenberg and Richter, 1954) may not be significant.

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