

# ANALYSIS OF SEISMIC FIRE

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## ABSTRACT

This paper deals with the number of the seismic fire. The usual least squares method of curve fitting is not suitable to the analysis for counting numbers. Then more suitable method, which is based on the maximum likelihood estimation, is introduced for the curve fitting problem concerning to the non-negative discrete data such as the number of the seismic fire. It is concluded that this method is better than the usual method and that the results support the assumption of the new method. And the methodology which can give the most appropriate formula in view of statistical model is referred.

**Keywords ;** Seismic fire, earthquake, curve fitting, least squares method, maximum likelihood, estimation, statistical model, data analysis

## 1. INTRODUCTION

Several formulas on the number of the seismic fire have been reported till now<sup>1-7</sup>. However, it can be pointed out that former studies have a few of imperfections theoretically and it is insufficient to discuss for propriety of the formulas statistically. Then we consider more theoretically and deal with the curve fitting problem concerning to the non-negative discrete data as synthetically as possible. And the purpose of this paper is to investigate the methodology which can give the most appropriate formula in view of statistical model. Although we deal with the seismic fire, we confine to the number of it. It is also a significant problem to discuss the mechanism of the seismic fire, but it is omitted owing to limited space.

## 2. REVIEW OF LITERATURE

H.Kawasumi<sup>1</sup> researched the relationship between the ratio of serious fires outbreak and the ratio of destroyed houses following Kanto earthquake(1923) in Japan. So-called Kawasumi formula is expressed as below after M.Kobayashi<sup>2</sup>.

$$\log Y = 0.773 \log X - 1.854 \quad (1)$$

where X is the ratio of destroyed houses and Y is the ratio of serious fires outbreak. Here serious fires means fires which are not extinguished immediately, and which spread to surrounding buildings. Next, H.Mizuno<sup>3,4</sup> has researched the following relationship based on the data from 13 different earthquakes dating from 1923 to 1974 in Japan.

$$\log \{y / (\alpha \cdot N)\} = a \cdot \log(X) + r \cdot s + b \quad (2)$$

where y is the number of seismic fire, N is the number of all houses, X is the ratio of destroyed houses,  $\alpha$  is a factor of using fire instruments and s is a factor of season. a, b, and r are coefficients. The values of coefficients and factors are omitted.

Eq.(2) can be changed as follows:

$$y = \alpha \cdot \beta \cdot B \cdot N \cdot X^a \quad (3)$$

where  $\beta = e^b$  and  $B = e^{rs}$ . Furthermore

$$y/N=c \cdot X^a \quad (4)$$

or

$$\log Y = a \cdot \log X + \log c \quad (5)$$

where  $c = \alpha \cdot \beta \cdot B$  and  $Y=y/N$ . One of eq.(5) is as follows:

$$\log Y = 0.429 \log X - 6.231 \quad (6)$$

H.Kawasumi and H.Mizuno used the method of least squares but Building Research Institute<sup>6</sup>, Ministry of Construction, Japan, proposed the method of safety side estimation, is based on the method of least squares, and analyzed the following relationship based on the same data of H.Mizuno.

$$\log Y = 0.299 \log X - 4.656 \quad (7)$$

One the other hand, M.Kobayashi<sup>2</sup> has pointed out that other relationships have higher correlation coefficients than those of the above type functions.

$$Y = 0.00348X - 0.0163 \quad (8)$$

$$Y = 0.0114 \log X - 0.0210 \quad (9)$$

where  $X$  is the percentage of destroyed houses and  $Y$  is the percentage of serious fires outbreak.

Lastly, O.Koide<sup>8</sup> used the maximum likelihood estimation with the data supposed to be the Poisson distribution and found following formula.

$$Y = 0.020X - 0.872 \quad (10)$$

where  $X$  is the number of destroyed houses per 10000 houses and  $Y$  is the number of serious fires outbreak per 10000 houses. And it is reported by O.Koide that the data of seismic fire in the vicinity to zero follow the Poisson distribution. Here, we can summarize following two points of the problem in former studies on the number of seismic fire:

- a. Is the analysis method suitable theoretically?
- b. What type of function of formula is appropriate?

### 3. THEORETICAL REMARKS

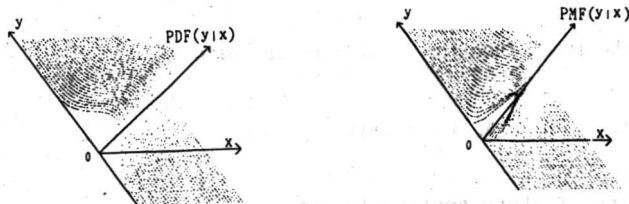
#### (1) Maximum Likelihood Estimation<sup>9</sup>

We suppose now that  $X$  is a random variables, discrete or continuous, whose distribution depends upon the parameters  $\theta_1, \theta_2, \dots, \theta_p$ . Let  $x_1, x_2, \dots, x_n$  be an observed sample. The idea of maximum likelihood estimation of the parameters  $\theta_1, \theta_2, \dots, \theta_p$ , which characterize a random variable  $X$ , is to choose the values which make the observed sample  $x_1, x_2, \dots, x_n$  most probable. If  $X$  is discrete, the probability that a random sample consists of these values exactly is given by

$$L = \prod_{i=1}^n f(x_i | \theta_1, \theta_2, \dots, \theta_p) \quad (11)$$

where  $f(\cdot)$  is the probability mass function of  $X$ .  $L$  is called the likelihood function and is a function of the parameters  $\theta_1, \theta_2, \dots, \theta_p$ . If  $X$  is continuous,  $f(\cdot)$  is the probability density function of  $X$ . In many cases it is more convenient to work with  $l = \ln L$ . If we suppose the random variable  $X$  has a normal distribution, the maximum likelihood estimation is the same as the method of least squares. The normal distribution consists of the continuous data from  $-\infty$  to  $+\infty$  theoretically. By the way, we have often applied the method of least squares to the curve fitting problem till now even if the data are discrete. It is because that the distribution can be approximated by normal one under certain conditions. But if the data are confined to the non-negative integer and in the vicinity to zero, it is known by Fig.1 that the distribution is not suitable by reason of being out of range. T.Awaya<sup>9-11</sup> have developed the new method which is suitable to analysis for the intensity estimation of such data with low statistics. This method can be

precisely applied to the data which follow the Poisson distribution and is based on the maximum likelihood estimation. He has analyzed the number of radioactive nuclear and of their half-lives. It has then been concluded that the method gives good results. The method of least squares and Aways method are summarized as follows.



(a) Normal distribution (b) Poisson distribution  
Fig.1 The typical example of distribution function

(2) The method of least squares<sup>12, 13</sup>

Let  $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$  be the data series. The fitting function is written by

$$y = g(x | a_1, a_2, \dots, a_m) \quad (12)$$

where  $a_1, a_2, \dots, a_m$  are the parameters to be searched. Then the likelihood function is given by

$$L = \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp \left\{ -\frac{1}{2\sigma^2} [y_i - g(x_i | a_1, \dots, a_m)]^2 \right\} \quad (13)$$

$$= \left\{ \frac{1}{\sqrt{2\pi}\sigma} \right\}^n \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^n [y_i - g(x_i | a_1, \dots, a_m)]^2 \right\} \quad (14)$$

where  $\sigma^2$  is the variance. The logarithmic likelihood function can be written as follows:

$$l = -\frac{n}{2} \ln(2\pi) - \frac{n}{2} \ln(\sigma^2) - \frac{1}{2\sigma^2} \sum_{i=1}^n [y_i - g(x_i | a_1, \dots, a_m)]^2 \quad (15)$$

The maximum likelihood estimators of  $a_1, \dots, a_m$  are the values that maximize  $l$ . The optimal values of the parameters can be obtained by solving normal equations, i.e. a simultaneous equation in  $m$  unknowns which is obtained by setting the partial derivatives of  $S$  with respect to the parameters  $a_1, \dots, a_m$  equal to zero. If  $g$  is a linear function with respect to the parameters, it is easy to solve it. If  $g$  is a nonlinear function with respect to the parameters, it is difficult to solve it. Then we must substitute the initial approximate values of the parameters and repeat the try-and-error calculation for that case. It is important to take care in choosing these initial values in order to guarantee fast convergence and to get reliable results. If we could solve it anyway, we describe the estimators as  $\hat{a}_1, \dots, \hat{a}_m$ . On the other hand, in order to maximize  $l$  the following equation is essential.

$$\frac{\partial l}{\partial \sigma^2} \Big|_{\sigma^2 = \hat{\sigma}^2} = -\frac{n}{2\hat{\sigma}^2} + \frac{1}{2(\hat{\sigma}^2)^2} \sum_{i=1}^n [y_i - g(x_i | a_1, \dots, a_m)]^2 = 0 \quad (16)$$

Then

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n [y_i - g(x_i | \hat{a}_1, \dots, \hat{a}_m)]^2 \quad (17)$$

(3) Aways method

(A) the case of the data with the Poisson errors in ordinate

We use the same data and function as mentioned previous section. The datum  $y_i$  ( $i=1, \dots$

n) is supposed to be a sample from the Poisson distribution. Then the likelihood function is

$$L = \prod_{i=1}^n \frac{\{g(x_i | a_1 \dots a_m)\}^{y_i}}{y_i!} \exp\{-g(x_i | a_1 \dots a_m)\} \quad (18)$$

and the logarithmic likelihood function is written as follows:

$$l = \sum_{i=1}^n [y_i \ln \{g(x_i | a_1 \dots a_m)\} - \ln(y_i!) - g(x_i | a_1 \dots a_m)] \quad (19)$$

The similar calculation of above section is put into operation and the estimators of parameters  $\hat{a}_1 \dots \hat{a}_m$  are solved.

(B) the case of the data with the Poisson errors in ordinate and abscissa;  
called two-dimensional fitting

This part of section is improved the Awaya method by authors. Let the pairs of observed data be  $(x_1, y_1), \dots, (x_n, y_n)$  and the datum  $y_i$  ( $i=1 \dots n$ ) and the datum  $x_i$  ( $i=1 \dots n$ ) are supposed to be observed variables from the Poisson distributions whose parent means are  $\Phi_i$  and  $\Psi_i$ , respectively. Then the likelihood function is

$$L = \prod_{i=1}^n \frac{\Psi_i^{x_i}}{x_i!} \exp(-\Psi_i) \frac{\Phi_i^{y_i}}{y_i!} \exp(-\Phi_i) \quad (20)$$

and the logarithmic likelihood function is written as follows:

$$l = \sum_{i=1}^n [x_i \ln(\Psi_i) - \ln(x_i!) - \Psi_i + y_i \ln(\Phi_i) - \ln(y_i!) - \Phi_i] \quad (21)$$

Since both  $x_i$  and  $y_i$  include errors, the function to be fitted to the data should be written as an implicit function of two variables,  $x$  and  $y$ .

$$h(x, y | a_1, \dots, a_m) = 0 \quad \text{or} \quad y - g(x | a_1, \dots, a_m) = 0 \quad (22)$$

In order to get the maximum values of  $l$ , it is convenient to use the method of Lagrange multipliers. Then the logarithmic likelihood function is changed as follows:

$$l = \sum_{i=1}^n [x_i \ln(\Psi_i) - \Psi_i + y_i \ln(\Phi_i) - \Phi_i + \lambda_i h(x_i, y_i | a_1, \dots, a_m)] \quad (23)$$

where  $\lambda_i$  is Lagrange's multipliers and the term which play no part in calculating are omitted. The procedure to get the best values of parameters is the same as the previous section. If the estimators of parameters  $\hat{a}_1, \dots, \hat{a}_m$  are obtained, the maximum value of the logarithmic likelihood function can be given.

(4) Indexes of appropriateness and propriety of fitting function

We consider statistically the following indexes of appropriateness and propriety of fitting function.

- Sum of squares residual;  $S$
- Coefficient of correlation between datum  $y_i$  and estimator  $\hat{y}_i$ ;  $R$
- The maximum value of the logarithmic likelihood function;  $l$
- Akaike information criterion<sup>12, 14, 15</sup>; AIC, defined by  
AIC =  $-2 \times l + 2 \times (\text{number of free parameters})$
- Sum of estimator  $y_i$ ;  $S_y$
- Upper and lower side of deviation between  $y_i$  and estimator  $\hat{y}_i$ ;  $\delta_+, \delta_-$   
 $\delta_+ = \max (y_i - \hat{y}_i) \quad [i=1 \dots n]$   
 $\delta_- = \max (\hat{y}_i - y_i) \quad [i=1 \dots n]$

We should choose the better analysis method by using these indexes and judge the best fitting function with due regard to the conformity between the function and the

phenomenon, for example,

- a. contradictions in the range of variables
- b. geometrical characteristics of the fitting function

#### 4. CALCULATION AND RESULTS

##### (1) Method of calculation

As it is difficult to compare the results from different data, we use the data after M. Kobayashi which are shown in Tab.1. The fitting function, the form of  $Y=g(X)$ , to be investigated are as follows:

- I.  $g_1(X) = aX + b$
- II.  $g_2(X) = a \ln X + b$
- III.  $g_3(X) = aX^b$
- IV.  $g_4(X) = aX^b + c$

where  $X=x/N$  is the ratio of destroyed houses,  $Y=y/N$  is the ratio of serious fires outbreak and  $a, b$  and  $c$  are the parameters to be searched. And  $x$  is the number of destroyed houses,  $N$  is the number of all houses,  $y$  is the number of serious fires outbreak. The sample data points are scattered in Fig.2. We use the function  $Y=g(X)$  for the method of least squares and use the function  $y=N \cdot g(X)$  for the Aways method in view of property of data. The fitting functions given by the form of  $g(X)$  are compared. Furthermore we consider the case of the data with the Poisson errors in ordinate and abscissa, i.e. two-dimensional curve fitting.

##### (2) Results and discussion

The eight functions are obtained from the calculations by the two methods about four types of  $g(X)$  if we supposed the data with the errors in ordinate. The results are shown in Tab.2. The indexes which play important roles are the maximum value of the logarithmic likelihood function and the value of AIC from standpoint of statistical model and other indexes are used only for the information of tendency. The model whose value of the logarithmic likelihood function is greater is the better one. The model whose value of AIC is less is the better one. It is reported that the significant difference is the value over one or two in view of entropy. About of all four functions, the maximum value of the logarithmic likelihood function given by Aways method is greater than that of the method of least squares and the value of AIC given by Aways method is less than that of the method of least squares. It became clear that the Aways method is better than the method of least squares in this case. A result which supports the assumption of the new method was obtained. Next, it has been shown that the function of ⑧ is the best fitted in view of the maximum value of the logarithmic likelihood function and that the function of

⑥ is the best fitted from standpoint of the value of AIC. The functions of ②, ④, ⑥ and ⑧ are traced in Fig.2. Lastly, we applied the method of two-dimensional fitting. The results are shown in Tab.3. The similar values of parameters to Tab.2 have been obtained and these table could hardly be compared in this case. In the former studies, only an arbitrary function was often fitted by the unsuitable method. However, it is important to compare several functions to be fitted and to investigate not only correlation coefficient but also the other indexes, the maximum value of logarithmic likelihood function and the value of AIC.

Tab.1 Data of Kanto earthquake(1923)

District	Item	All houses	Destroyed houses	Serious fires*
1.	Kojimachi	15430	409	6
2.	Nihonbashi	19425	82	0
3.	Kyobashi	21551	86	8
4.	Shiba	27755	415	6
5.	Azabu	14382	352	1
6.	Akasaka	11691	360	4
7.	Yotsuya	11417	48	1
8.	Hongo	21321	149	4
9.	Shimoya	30239	733	7
10.	Kanda	18430	1262	10
11.	Asakusa	32051	2218	19
12.	Kojishikawa	23686	144	3
13.	Ushigome	19073	247	4
14.	Honjo	32948	4426	17
15.	Hukagawa	25570	2064	9
Total		324969	12995	99

\* excluding chemical fires

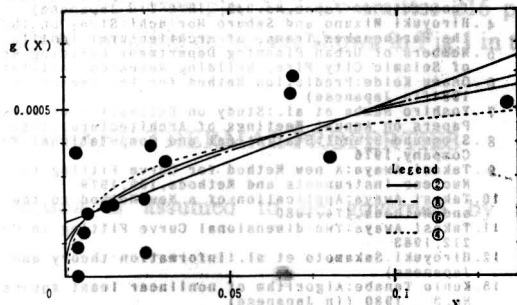


Fig.2 Scattered data and fitted curves

Tab.2 Results of the fitted function and these indexes

Fitted functions	Item	R	S	AIC	$\Sigma \delta$	$\delta - \delta$
① $g_1(X) = 0.348 \times 10^{-2} X + 0.163 \times 10^{-3}$		0.865	107.39	-36.05	78.09	98.2 3.8 6.1
② $g_2(X) = 0.364 \times 10^{-2} X + 0.159 \times 10^{-3}$		0.864	109.72	-34.66	73.32	99.0 4.4 5.8
③ $g_3(X) = 0.114 \times 10^{-3} \ln X + 0.735 \times 10^{-3}$		0.880	96.30	-35.23	76.46	97.3 3.5 5.7
④ $g_4(X) = 0.102 \times 10^{-3} \ln X + 0.694 \times 10^{-3}$		0.882	98.47	-35.09	74.18	98.8 3.5 5.5
⑤ $g_5(X) = 0.132 \times 10^{-2} X^{0.417}$		0.887	90.04	-34.73	75.45	97.8 3.0 5.2
⑥ $g_6(X) = 0.125 \times 10^{-2} X^{0.398}$		0.888	89.37	-34.27	72.55	99.3 3.1 5.1
⑦ $g_7(X) = 0.150 \times 10^{-2} X^{0.514} + 0.524 \times 10^{-4}$		0.885	91.29	-34.83	77.66	98.3 3.0 5.1
⑧ $g_8(X) = 0.158 \times 10^{-2} X^{0.588} + 0.816 \times 10^{-4}$		0.885	91.71	-34.19	74.37	99.3 3.0 5.2

(Note) Least squares method: ①, ③, ⑤, ⑦. Aways method: ②, ④, ⑥, ⑧

Tab.3 Results of two-dimensional fitting by Aways method

Fitted functions	Item	R	S	AIC	$\Sigma \delta$	$\delta - \delta$
① $g_1(X) = 0.365 \times 10^{-2} X + 0.159 \times 10^{-3}$		0.864	109.90	-34.66	73.32	99.1 4.4 5.8
② $g_2(X) = 0.104 \times 10^{-3} \ln X + 0.704 \times 10^{-3}$		0.881	97.59	-35.10	74.19	99.6 3.6 5.3
③ $g_3(X) = 0.125 \times 10^{-2} X^{0.398}$		0.888	89.47	-34.27	72.55	98.6 3.1 5.2
④ $g_4(X) = 0.154 \times 10^{-2} X^{0.522} + 0.763 \times 10^{-4}$		0.885	91.40	-34.19	74.37	98.9 3.0 5.3

### 5. CONCLUDING REMARKS

We have reviewed former studies on the number of seismic fire and pointed out the problems theoretically. Then we have applied the new method and obtained some findings. A summary of our conclusion is written as follows:

- 1) It has been shown by the theoretical consideration that the usual least squares curve fitting method is not suitable to the non-negative discrete data.
- 2) The new method proposed by Aways has been tried to analyze the data of the number on seismic fire, i.e. an example of Kanto earthquake(1923). Then it has been proved that the Aways method is more suitable than the method of least squares .
- 3) It has been shown by AIC that the function of ⑥ is the best fitted .
- 4) More theoretically we should apply the method of two-dimensional fitting. However similar values of parameters have been obtained in this case.
- 5) The former procedure to get the function is also not suitable. It is important to compare several functions to be fitted by using not only correlation coefficient but also the other indexes, the maximum value of logarithmic likelihood function and the value of AIC. This is the methodology which can give the most appropriate formula in view of statistical model.

Since this study is the basic one laying stress on the methodology, we should make these fruitful conclusions apply the more practical one.

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