# A Taylor's power law in the Wenchuan earthquake sequence with fluctuation scaling

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4 Peijian Shi<sup>1</sup>, Mei Li<sup>2\*</sup>, Yang Li<sup>3\*</sup>, Jie Liu<sup>2</sup>, Haixia Shi<sup>2</sup>, Tao Xie<sup>2</sup>, Chong Yue<sup>2</sup>

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6 <sup>1</sup>Co-Innovation Center for Sustainable Forestry in Southern China, College of Biology and the

7 Environment, Bamboo Research Institute, Nanjing Forestry University, Nanjing 210037, China

8 <sup>2</sup>China Earthquake Networks Center, China Earthquake Administration, Beijing 100045, China

9 <sup>3</sup>Department of Mathematics and Statistics, University of Minnesota Duluth, Duluth, MN 55812, USA

10 Correspondence to: mei\_seis@163.com (M. Li); yangli@d.umn.edu (Y. Li)

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Abstract Taylor's power law (TPL) describes the scaling relationship between the 12 temporal or spatial variance and mean of population densities by a simple power law. 13 TPL has been widely testified across space and time in biomedical sciences, botany, 14 15 ecology, economics, epidemiology, and other fields. In this paper, TPL is analytically reconfirmed by testifying the variance as a function of the mean of the released 16 energy of earthquakes with different magnitudes on varying timescales during the 17 Wenchuan earthquake sequence. Estimates of the exponent of TPL are approximately 18 2, showing that there is mutual attraction among the events in the sequence. On the 19 other hand, the spatial-temporal distribution of the Wenchuan aftershocks tends to be 20 21 nonrandom but approximately definite and deterministic, which highly indicates a 22 stable spatial-temporal dependent energy release caused by regional stress adjustment and redistribution during the fault revolution after the main shock. Effect of different 23 24 divisions on estimation of the intercept of TPL straight line has been checked while the exponent is kept to be 2. The result shows that the intercept acts as a logarithm 25

function of the time division. It implies that the mean–variance relationship of the energy release from the earthquakes can be predicted although we cannot accurately know the occurrence time and locations of imminent events.

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## 30 1 Introduction

The Wenchuan  $M_{\rm S}$  8.0 earthquake on May 12, 2008 was the result of the 31 intensively compressive movement between the Qinghai-Tibet Plateau and the 32 Sichuan basin. It ruptured the middle segment of the Longmenshan (LMS) thrust belt 33 34 (Burchfiel et al., 2008), with a total length of fault trace of approximately 400 km along the edge of the Sichuan basin and the eastern margin of the Tibetan plateau, in 35 the middle of the north-south seismic belt of China. Millions of aftershocks have 36 37 occurred after the main event. Up to now, the focus zone tends to be quiet with only small ones occurring occasionally. A complete Wenchuan earthquake sequence has 38 been attained. 39

40 Statistical seismology applies statistical methods to the investigation of seismic activities, and stochastic point process theory promotes the development of statistical 41 seismology (Vere-Jones et al., 2005). After some improvement, most of the 42 point process theories and methods can be used to analyze spatio-temporal data of 43 earthquake occurrence and to describe active laws of aftershocks. The term 44 "aftershock" is widely used to refer to those earthquakes which follow the occurrence 45 of a large earthquake and aggregately take place in abundance within a limited 46 interval of space and time. This population of earthquakes is usually called an 47 earthquake sequence. In seismological investigations, one important subject has long 48

been the statistical properties of the aftershocks. Spatial and temporal distribution of 49 aftershocks after a destructive earthquake is usually performed in a general survey 50 51 (Utsu, 1969). In seismology, one of the most famous theories describing the activities of aftershocks is the Gutenberg-Richter law (Gutenberg and Richter, 1956), which 52 expresses the relationship between the magnitude and the total number of earthquakes 53 with at least that magnitude in any given region and time interval. Another one is the 54 Omori's law, which was first depicted by Fusakichi Omori in 1894 (Omori, 1894) and 55 shows that the frequency of aftershocks decreases roughly with the reciprocal of time 56 57 after the main shock. Utsu (1969) and Utsu et al. (1995) developed this law and proposed the modified Omori formula afterwards. Since the 1980s, with the 58 development of nonlinear theory, an epidemic-type aftershock sequence (ETAS) 59 60 model has been proposed by Ogata (1988, 1989, 1999), which is based on the empirical laws of aftershocks and quantifies the dynamic forecasting of the induced 61 effects. This model has been used broadly in earthquake sequence study (Kumazawa 62 63 and Ogata, 2013; Console, 2010).

An increasing number of investigations show that there is an interaction effect for the occurrence of aftershocks in a given area. Stress triggering model is usually used to depict interaction between larger earthquakes by the view of physics (Haris, 1998; Stein, 1999). More and more results show that obvious enhancement in Coulomb stress not only promotes the occurrence of upcoming mid or strong events of an earthquake sequence but also affects their spatial distribution to some degree (Robinson and Zhou, 2005).

The goal of this paper is to introduce a different statistical method called Taylor's power law into the statistical seismology field by analyzing the Wenchuan earthquake sequence from the point of view of energy distribution or energy release. We aim to find out whether or not the energy distribution or energy release of the Wenchuan earthquake sequence complies with a specific power–law function of TPL for different scaled samples, and what the spatial and temporal properties are.

In statistics, there are two important moments in a distribution, the mean  $(\mu)$ 77 and the variance ( $\sigma^2$ ). It is common to describe the types of the distributions using the 78 relationship between these two parameters. For instance, we have  $\sigma^2 = \mu$  for a Poisson 79 distribution. In nature, however, the variance is not always equal to or proportional to 80 the mean. Mutual attraction or mutual repulsion for individuals in natural populations, 81 82 e.g., the intra-specific completion of plants, makes variance different from the mean. After examining many sets of samples of animal and plant population spatial densities, 83 Taylor (1961) found that the variance appears to be related to the mean by a power-84 85 law function: the variance is proportional to the mean raised to a certain power

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$$\sigma^2 = a\mu^b \tag{1}$$

87 or equivalently as a linear function when the mean and variance are both

88 logarithmically transformed

89 
$$\log_{10}(\sigma^2) = \log_{10}(a) + b \times \log_{10}(\mu) = c + b \times \log_{10}(\mu)$$
 (2)

where *a* and *b* are constants and  $c = \log_{10}(a)$ . Eqs. 1 or 2 is called Taylor's power law (henceforth TPL) or Taylor's power law of fluctuation scaling (Eisler et al., 2008).

Eqs. 1 and 2 may be exact if the mean and variance are population moments

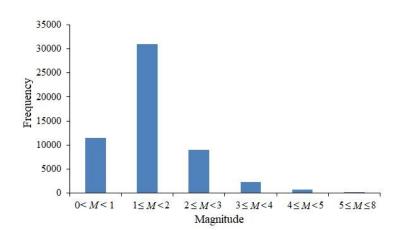
93	calculated from certain parametric families of skewed probability distributions
94	(Cohen and Xu, 2015). TPL describes the species-specific relationship between the
95	spatial or temporal variance of populations and their mean abundances (Kilpatrick and
96	Ives, 2003). It has been verified for hundreds of biological species and abiotic
97	quantities in biomedical sciences, botany, ecology, epidemiology, biomedical sciences,
98	botany, and other fields (Taylor, 1961, 1984; Kendal, 2002; Eisler et al., 2008; Cohen
99	and Xu, 2015; Shi et al., 2016, 2017; Lin et al., 2018). Most of the scientific
100	investigations of TPL mainly focus on the power-law exponent $b$ (or slope $b$ in the
101	linear form), which has been believed to contain information on aggregation in space
102	or time of populations for a certain species (Horne and Schneider, 1995).

In this study, we also concentrate on the parameter b of TPL. We expect that b is independent of the temporal block size A which is used to divide the Wenchuan sequence into different temporal blocks because the aftershock area is invariable during this period.

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Wenchuan earthquake sequence

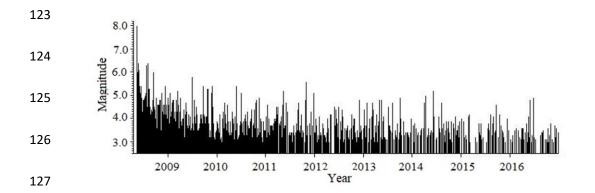


**Figure 1.** Histogram of earthquakes with different magnitudes of the Wenchuan sequence.

111 A large earthquake of magnitude  $M_{\rm S}$  8.0 hit Wenchuan, Sichuan province of 112 China at 14:28:01 CST (China Standard Time) on May 12, 2008 with an epicenter 113 located at 103.4 N and 31.0  $\oplus$  and a depth of 19 km.

According to the earthquake catalogue of the China Earthquake Networks Center 114 (CENC) (http://www.csi.ac.cn/), there have been 54,554 earthquakes of magnitudes 115 M > 0 recorded for the Wenchuan sequence by December 31, 2016. Figure 1 shows 116 the frequency of aftershocks with different magnitudes. Here, aftershocks with M < M117 2.0 account for 77.9% of the total sequence due to the fact that only weak ones occur 118 119 after a long period of time after the main shock. In addition, except for the main shock, the number of aftershocks is 733 for magnitudes 4.0  $\leq M < 5.0$ , and 86 for 5.0  $\leq$ 120 M < 8.0, respectively. They account for a very small percentage of the total. 121





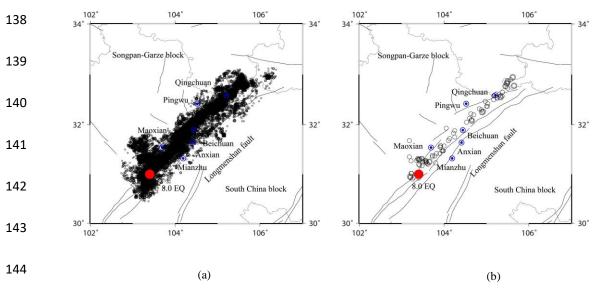
**Figure 2.** Series plot of the Wenchuan earthquake sequence with  $M \ge 3.0$  from May 12, 2008 to December 31, 2016.

Figure 2 displays the fluctuation variability of the Wenchuan earthquake sequence with  $M \ge 3.0$  from May 12, 2008 to December 31, 2016. The temporal distribution of the magnitudes of aftershocks attenuates quickly after the main shock.

The three larger aftershocks all occurred in 2008 with M 6.4 on May 25, M 6.1 on August 1, and M 6.1 on August 5, respectively. Eighty-five percent of aftershocks with  $M \ge 3.0$  occurred by the end of 2011, about 2.5 years after the main shock.

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145Figure 3. Spatial distribution of epicenters of the Wenchuan earthquake sequences with (a) M > 0 and146(b)  $M \ge 5.0$  from May 12, 2008 to December 31, 2016. The main shock on May 12, 2008 is labeled147by a red solid circle.

Figure 3a shows the spatial distribution of epicenters of the Wenchuan 149 earthquake sequence with M > 0 from May 12, 2008 to December 31, 2016. The 150 aftershocks are distributed in the region with latitude 102 E-107 E and longitude 151 30 N-34 N, mainly along the Longmenshan thrust fault, which is a junction region of 152 Songpan–Garze block and South China block and extends along north–east–east 153 (NEE) direction for more than 400 km. The size of the aftershocks on different scales 154 is characterized by a population density of the events distributed in space and time 155 after the Wenchuan  $M_{\rm S}$  8.0 earthquake but we neglect the variations of the aftershock 156 area in the next step. The distribution of strong aftershocks is of different segment 157

characteristics. Earthquakes with magnitude  $M \ge 5.0$  mainly spread in south 158 Miaoxian and Mianzhu area and north Pingwu area. There are no strong aftershocks 159 occurring in the middle areas such as Beichuan and Anxian (see Figure 3b). 160 According to the primary investigation results of the Wenchuan rupture process 161 conducted by Chen et al. (2008), the rupture of the Wenchuan 8.0 earthquake 162 originated from Wenchuan thrust fault with a little right lateral slip component and 163 extended mainly in north-east (NE) orientation. The whole process formed two areas 164 with larger dislocations. One is the south area of Miaoxian located in the bottom 165 166 section in Figure 3b. The other one lies near Beichuan area (the middle segment in Figure 3b) but no strong shocks happened there. 167

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169 **3 Data processing method** 

For the complete Wenchuan earthquake sequence, we denote the number of all earthquakes by *N*, i.e., N = 54,554, and use q = 1, ..., N to index each earthquake. For each earthquake with magnitude  $M_q$ , its corresponding energy release is labeled by  $E_q$ and it can be attained in the light of the following relationship (Xu and Zhou, 1982)

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$$\log_{10}(E_q) = 11.8 + 1.5M_q$$
 (3)

We use  $t_q$  to index the time lag of the *q*-th aftershock from the main shock (in days), i.e.,  $t_1 = 0$  for the main event. The last aftershock occurred at 18:05:57 CST (China Standard Time) on December 31, 2016, and its  $t_q$  value is 3155.

178 In order to study the relationship between the variance and mean of the energy 179 sequence  $E_q$ , we first divide it into equally–spaced short temporal blocks with size A (in days). For example, if A = 10, then the number of blocks is N/A = 3155/10 = 315.5which is rounded to the nearest integer. Now the complete energy sequence  $E_q$  is partitioned into n = 316 blocks of short energy subsequences. We use *i* to index each block, i.e., i = 1, ..., n and  $h_i$  to denote the number of data points in each block which is variable because earthquakes occurred stochastically in the sequence. Now we can calculate the mean ( $\mu$ ) and variance ( $\sigma^2$ ) for each block using

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$$\mu_i = \frac{\sum_{j=1}^{h_i} E_{i,j}}{h_i}$$
(4)

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$$\sigma_i^2 = \frac{\sum_{j=1}^{h_i} (E_{i,j} - \mu_i)^2}{h_i - 1}$$
(5)

where  $E_{i,j}$  denotes the energy of the *j*-th earthquake in the *i*-th block.

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# 190 **4 Results**

191 The data processing procedure has been performed with different block size A =4, 5, 6, ..., 100. The number of sample points in each block decreases as the block 192 size increases. The relationships between the mean and variance of the released 193 194 energies from earthquakes in 6 representative temporal blocks are shown in Figure 4 on a log-log scale. The red line stands for the fitted linear function of TPL's power 195 law  $\log_{10}(\sigma^2) = c + b \log_{10}(\mu)$  using least squares. The 95% confidence intervals (CI) 196 of the slope and the coefficients of determination  $R^2$  are shown in Table S1. For 197 instance, Figure 4a shows the variance as a function of the mean for 316 time 198 intervals when A = 10. The estimated intercept is 0.702 and the estimated slope is 199 2.060 with 95% CI (1.989, 2.076), and  $R^2 = 0.963$ . The root-mean-square error 200 (RMSE) was also calculated to exhibit the feasibility of using a TPL with the 201

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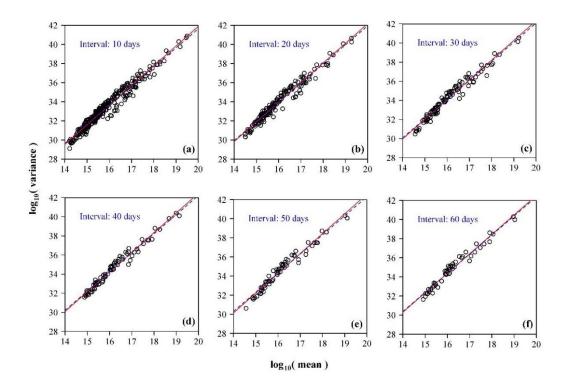


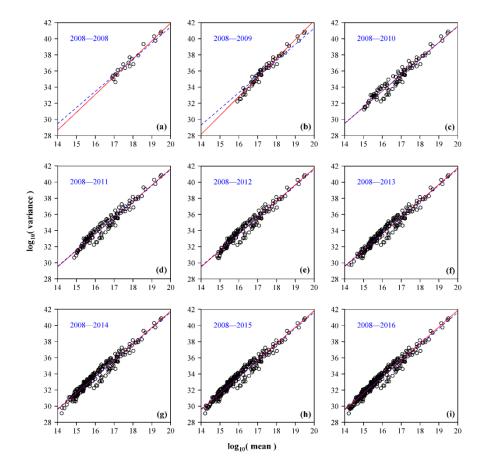
Figure 4. Calculated variance as a function of the observed mean of the energies from earthquakes in each time interval on a log-log coordinate (open circles), for different values of *A*. The red straight line corresponds to the fitted Taylor's power law with an unknown exponent, i.e.  $\log_{10}(\sigma^2) =$  $c + b \log_{10}(\mu)$ , using least squares. The blue dashed line corresponds to the fitted Taylor's power law with the exponent 2, i.e.  $\log_{10}(\sigma^2) = d + 2 \log_{10}(\mu)$ . There are 97 different values of *A* in total, and only 6 are shown here. (a) A = 10; (b) A = 20; (c) A = 30; (d) A = 40; (e) A = 50; and (f) A = 60.

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Figure 4 and Table S1 show that there is an apparent linear relationship between the common logarithm of the variance and that of the mean for all earthquakes occurring within different temporal blocks, characterized by a property of aggregation on different timescales. The estimated value of the intercept, c (or  $\log_{10}(a)$ ), which is mainly influenced by the number of samples, overall increases with A from 0.016 to 3.249 (Table S1). The estimates of slope b, on the other hand, are roughly 2 for all

block sizes used in the study. All  $R^2$  values are greater than 0.96, showing a very strong linear relationship. These results indicate that the energy release of aftershocks of the Wenchuan sequence complies well with a temporal TPL.



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Figure 5. The calculated variance as a function of the observed mean of the energies from earthquakes in each block on a log-log scale (open circles) when A is fixed to be 10. The red straight line corresponds to the fitted Taylor's power law with an unknown exponent, i.e.  $\log_{10}(\sigma^2) = c + b \log_{10}(\mu)$ , using least squares. The blue dashed line corresponds to the fitted Taylor's power law with the exponent 2, i.e.  $\log_{10}(\sigma^2) = d + 2 \log_{10}(\mu)$ . (a) 2008–2008; (b) 2008–2009; (c) 2008–2010; (d) 2008– 2011; (e) 2008–2012; (f) 2008–2013; (g) 2008–2014; (h) 2008–2015; and (i) 2008–2016.

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Next, we divide the Wenchuan earthquake sequence into 9 time stages in years:
2008–2008, 2008–2009, 2008–2010, 2008–2011, 2008–2012, 2008–2013, 2008–2014,
2008–2015, and 2008–2016. For each stage, we follow a similar procedure leading to
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Figure 4. That is, we first transform all earthquakes into their energy forms using the relationship between earthquake magnitude M and energy E. Then the energy sequence are partitioned into temporal blocks with a fixed block size A = 10 days. The calculated variances and means are plotted on a log-log scale as shown in Figure 5. Again, TPL comes into play for all time stages. The estimates of the parameters in Eq. (2) for the data in different stages were listed in Table S2.

Figure 5 shows a strong linear relationship between the variance and mean of the earthquake energy populations on a log-log scale, especially for those large samples. The estimates summarized in Table S2 (red fitted lines in Figure 5) show similar results as in Table S1. The intercept gradually increases as the total number of samples increases but with a little more fluctuation. Meanwhile, the estimate of slope *b* is still roughly a constant around 2.

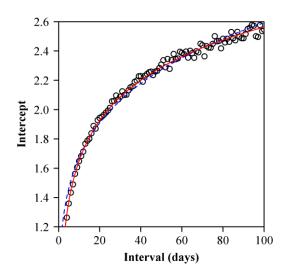
Here with the exponent b = 2 considered, the possible relationship between the 244 estimate of the intercept (namely d) in equation  $\log_{10}(\sigma^2) = d + 2 \log_{10}(\mu)$  and the 245 246 temporal block size A is also examined. The estimated intercepts of the Wenchuan sequence as A increases from 4 days to 100 days in 1-day increments are shown in 247 Figure 6. At the same time, a logarithm function and an exponential function are 248 employed respectively to fit the data (i.e.,  $d = \alpha + \beta \times \log_{10}(A)$  and  $d = m \times A^n$ , where 249  $\alpha$ ,  $\beta$ , m and n are constants), and the results show that the logarithm function has a 250 higher goodness of fit (namely a lower residual sum of squares). The estimate of 251 parameter  $\alpha$  is equal to 0.7398 with 95% CI (0.7246, 0.7581), and the estimate of 252 parameter  $\beta$  is equal to 0.9121 with 95% CI (0.9004, 0.9229). Because  $\log_{10}(a) = d =$ 253

254  $\alpha + \beta \times \log_{10}(A)$ , we will have:

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$$\sigma^2 = a\mu^2 = 10^{\alpha}A^{\beta}\mu^2$$
 (6)

It illustrates that the variance of energy releases from aftershocks depends on two factors: (i) the mean squared, and (ii) the size of temporal block defined. Up to now, we confirm that the mean-variance relationship of energy releases from an earthquake sequence can be quantified although the accurate prediction of the time and location of an imminent event is still not attainable.

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Figure 6 The effect of time division (time span) on the estimate of the intercept in the TPL with a fixed exponent of 2, i.e.  $\log_{10}(\sigma^2) = d + 2 \log_{10}(\mu)$ , where *d* denotes the intercept. Two equations were used to fit the data ( $d = \alpha + \beta \times \log_{10}(A)$  and  $d = m \times A^n$ , where  $\alpha$ ,  $\beta$ , *m* and *n* are constants). The residual sum of squares (= 0.0535) using the logarithm function (represented by the red curve) is lower than that (= 0.1460) using the exponential function (represented by the blue curve).

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### 269 **5 Discussion**

The evolutionary process of a large earthquake is characterized by some complex
features from stochastic to chaotic or pseudo-periodic dynamics (McCaffrey, 2011).

On the one hand, there is a long-term slow strain of accumulation and culminating of rocks in the rigid lithosphere prior to the event with a sudden rupture and displacement of blocks. On the other hand, there is another long-term slow strain of redistribution and energy release with a large number of aftershock occurrences in an extensive area, which generally lasts for several months, sometimes even years, after the main shock.

It has been statistically established that in populations, if individuals distribute 278 randomly and are independent of each other, then the variance is equal to the mean, 279 i.e.,  $\sigma^2 = \mu$ ; individuals show mutual attraction if the variance is proportional to the 280 mean to a power > 1, i.e.,  $\sigma^2 > \mu$ ; individuals mutually repel each other if the 281 variance is proportional to the mean to a power < 1, i.e.,  $\sigma^2 < \mu$  (Taylor, 1961; Horne 282 283 and Schneider, 1995). The results obtained in this study show that the exponent of the TPL is around 2 in the Wenchuan energy sequence either with different time span A =284 4, 5, 6, ..., 100 days or with a fixed time span A = 10 days but for 9 time stages 285 286 between 2008 and 2016. This means earthquakes in the Wenchuan sequence are not distributed at random and independent of each other but with a mutual attraction. It 287 also indicates that there are possible interactions among different magnitudes in the 288 earthquake sequence. Cohen and Xu (2015) proposed analytically that observations 289 randomly sampled in blocks from any skewed frequency distribution with four finite 290 moments give rise to TPL because the variation in the sample mean and sample 291 292 variance between blocks are theoretically small if every block is randomly sampled from the same distribution. 293

There are various types of interpretations for the value of parameter b. Ford and 294 Andrew (2007) suggested that individuals' reproductive correlation determines the 295 size of b. While Kilpatrick and Ives (2003) proposed that interspecific competition 296 could reduce the value of b. Above all, empirically, b usually lies between 1 and 2 297 (Maurer and Taper, 2002). However, it is expected that TPL holds with b = 2 exactly 298 in a population with a constant coefficient of variation (CV) of population density. 299 This expectation derives from the well-known relationship: SD (standard deviation) 300 equals to square root of variance ( $\sigma^2$ ), i.e., SD =  $\sigma$  and the coefficient of variation 301  $CV = SD/\mu = k$ , here k is a constant. Then we can obtain  $\sigma^2 = (k\mu)^2$ . The relationship 302 between  $\log_{10}(\sigma^2)$  and  $\log_{10}(k\mu)$  is a straight line with slope 2 on a log-log scale. 303

It is well established that there is a specific property on the population either in 304 305 space or in time when b equals 2. Ballantyne (2005) proposed that b = 2 is a consequence of deterministic population growth. While Cohen (2013) showed that b =306 2 arose from exponentially growing, noninteracting clones. Furthermore, using the 307 308 Lewontin–Cohen (LC) model of stochastic population dynamics, Cohen et al. (2013) provided an explicit, exact interpretation of its parameters of TPL. They proposed that 309 the exponent of TPL will be equal to 2 if and only if the LC model is deterministic; it 310 will be greater than 2 if the model is supercritical (growing on average) and be less 311 than 2 if the model is subcritical (declining on average). This property indicates that 312 parameter b = 2 in our investigation on the Wenchuan earthquake sequence depends 313 exactly on its specific distribution of aftershocks. In other words, the law of 314 occurrence of all events or energy release in space and time is deterministic following 315

the main shock on May 12, 2008.

Although various empirical confirmations suggest that no specific biological, 317 physical, technological, or behavioral mechanism can explain all instances of TPL, 318 there has been some improvement in understanding the distribution and duration time 319 of aftershocks after the main event. Jiang et al. (2008) studied the Wenchuan 320 earthquake sequence using Gutenberg–Richter law (Gutenberg and Richter, 1956) and 321 the Omori's law (Omori, 1894). Their investigation attained a specific relationship 322 between the magnitude and the total number of earthquakes for a stable b value, 323 324 which indicates that the frequency of aftershocks decreases roughly with the reciprocal of time after the main shock. One of the models with physical parameters is 325 the stress triggering mechanism put forward by Dieterich (1994) with Dieterich and 326 327 Kilgore (1996). Shen et al. (2013) achieved a good fit between the observed Wenchuan aftershocks and the analytic solution of the modified Dieterich model. 328 Their results suggested that the generation of earthquakes is actually related to the 329 330 state of fault and can quantitatively describe the temporal evolution of the aftershock decay. In this sense, the Wenchuan energy sequence satisfies TPL with slope b = 2, 331 indicating a stable spatial-temporal dependent energy release caused by regional 332 stress adjustment and redistribution during the fault revolution after the main shock. 333 These results are of high coherence with what has been attained by Christensen et al. 334 (2002), who proposed a unified scaling law linking together the Gutenberg-Richter 335 Law, the Omori Law of aftershocks, and the fractal dimensions of the faults. Their 336 results show that nonzero driving force in the crust of the Earth leads to an earthquake 337

as a sequence of hierarchical correlated processes and this mechanism responsible
for small events also is responsible for large events. In other words, a main shock and
an aftershock are consequences of the same process.

It is possible that there are some interactions among earthquakes with different 341 magnitudes in an earthquake sequence. This kind of interaction probably derives from 342 medium stress state of the focus zone where earthquakes happen. The stress field in 343 the aftershock area is in a rapidly adjusting state when a lager earthquake occurred. It 344 is probable that a light stress adjustment caused by a small earthquake most likely 345 346 induces an obvious event in its surroundings in the near future. This process can lead to aggregation of aftershocks in space and time in extensive areas, causing TPL to 347 hold for the Wenchuan earthquake energy sequence. However, whether TPL accords 348 with all earthquake sequences and complies with specific parameters, e.g., b = 2, 349 needs further investigation. Up to now, one thing we can confirm is that the missing 350 events can lead to the exponent in TPL increase. For example, the estimated b is 351 approximately 2.1 to 2.2 if the events with magnitude M > 1.0 are used. It indicates 352 missing events can change the state of energy release from a stable (deterministic) 353 state to an unstable (supercritical) state as Cohen et al. (2013) have proposed. 354

The current study shows that the exponents of TPL for different temporal blocks for the Wenchuan earthquake sequence are approximately equal to 2 universally. The estimated intercept could be expressed as a linear equation of the log-transformation of temporal block *A* (Figure 6). The goodness of fit of the nonlinear regression is fairly high ( $R^2 = 0.9940$  in Figure 6), indicating some interesting underlying

mechanism leading to the occurrence of the aftershocks. The distribution of the 360 energy releases from aftershocks should be a right-skewed unimodal curve that can be 361 362 reflected by magnitude frequency distribution as shown in Figure 1. In fact, Cohen and Xu (2015) have demonstrated that the correlated sampling variation of the mean 363 and variance of skewed distributions could account for TPL under random sampling 364 and the estimated exponent of TPL was proportional to the skewness of the 365 distribution curve. For an exponential distribution, the variance equals its mean 366 squared. However, in our study, although the variance of energy releases from 367 368 aftershocks is similarly proportional to its mean squared, the coefficient of proportionality (i.e., a in Eq. [1]) does rely on the size of the temporal block. This 369 means that the energy releases from aftershocks might follow a temporal 370 371 block-dependent generalized exponential distribution, which should be more complex than the generalized exponential distribution (Gupta and Kundu, 2007). However, the 372 distribution function for the energy releases from aftershocks has not been well 373 defined so far. The existing functions for describing a skewed distribution of energy 374 releases or magnitudes usually belong to pure statistical models that lack clear 375 physical dynamic mechanism. Our study suggests that further studies should focus on 376 a temporal block-dependent or a sub-region-dependent distribution. However, to 377 provide a clear mathematical expression for this distribution function is beyond the 378 topic of this paper. It deserves further investigation. 379

380 6 Conclusions



In summary, we attempt to use a new way to investigate a spatio-temporal

distribution property of aftershocks of the Wenchuan earthquake sequence during 382 2008–2016. In terms of the energy release, the variance of samples in the earthquake 383 population is shown to have a simple power law relationship as a function of the mean 384 on different timescales, which gives rise to a TPL, i.e.,  $\sigma^2 = a\mu^b$ , with b = 2. On the 385 one hand, the results show that the intercept of the fitted line in linear form  $\log_{10}(\sigma^2)$ 386  $= c + b \times \log_{10}(\mu)$  on a log-log scale, increases as the number of samples and it is 387 reconfirmed that parameter c (namely  $\log_{10}(a)$ ) predominantly depends upon the size 388 of the sampling units (Taylor, 1961). On the other hand, if TPL holds, the estimated 389 390 values of parameters a and b support the conclusion that the Wenchuan aftershocks mutually trigger each other and distribute in space and time not randomly but 391 determinantly and definitely. We fix the exponent of TPL to be 2, and check the 392 393 effects of different time divisions on the estimate of the intercept. The result shows that the intercept acts as a logarithm function of the timescale. It implies that the 394 mean-variance relationship of energy releases from the earthquakes can be predicted 395 396 even though we cannot accurately predict the time and location of imminent events.

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