Earthquake slip distribution: A statistical model

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[1] The purpose of this paper is to interpret slip statistics in a framework of extended earthquake sources. We first discuss deformation pattern of the Earth’s surface from earthquakes and suggest that continuum versus block motion controversy can be reconciled by a model of the fractal distribution of seismic sources. We consider earthquake slip statistical distributions as they can be inferred from seismic moment-frequency relations and geometrical scaling for earthquakes. Using various assumptions on temporal earthquake occurrence, these distributions are synthesized to evaluate the accuracy of geologic fault slip determinations and to estimate uncertainties in long-term earthquake patterns based on paleoseismic data. Because the seismic moment distribution is a power law (Pareto), a major part of the total seismic moment is released by major earthquakes, $M \geq 10^{19.5}$ N m (moment magnitude $m \geq 7$); for these large earthquakes the rupture is confined to the upper brittle crust layer. We review the various moment-frequency and earthquake scaling relationships and apply them to infer the slip distribution at area- and site-specific regions. Simulating the seismic moment and strain accumulation process demonstrates that some synthetics can be interpreted as examples of a quasiperiodic sequence. We demonstrate the application of the derived slip statistical relations by analyzing the slip distribution and history of the San Andreas fault at Wrightwood, California.


1. Introduction

[2] This paper is meant to be the first in a series that will describe statistically deformation of the Earth’s surface from earthquakes. When completed, the statistical model can be compared with results of detailed GPS [Jackson et al., 1997; Sagiya, 2004] and interferometric synthetic aperture radar (InSAR) measurements. This task cannot be undertaken now because needed theoretical tools are still unavailable and observational results are not sufficient for extensive statistical analysis. However, we discuss below (see sections 2, 4.1, 6.1, and especially in section 7) what statistical methods should be developed and what measurements should be attempted to enable construction of a complex, comprehensive model.

[3] We have a much more modest aim here: to present a statistical model of slip distributions due to earthquakes. Slip measurements are being carried out at particular points along faults in California [Sieh et al., 1989; Yeats et al., 1997; Rockwell et al., 2000; Weldon et al., 2002; Rockwell et al., 2002; Scharer et al., 2003; Liu et al., 2004] and in other places. Histories of vertical displacements have been investigated in coral reefs [Natawidjaja et al., 2004] and in uplifts of marine/fluvial terraces along many Pacific subduction zones. These observations can be compared to the theoretical derivations of this paper to evaluate the statistical properties of these time series. Here we try to create only the methodological tools for such an exercise; the few applications of the statistical analysis discussed in section 6.3 are meant only as illustrations. A full statistical analysis of these data would need cooperation between several Earth science specialties.

[4] To create the statistical model for earthquake slip distribution, we consider three interrelated problems: earthquake size distribution, scaling of geometric parameters with earthquake size, and the temporal behavior of seismicity, especially in the largest earthquakes. These problems need to be solved to estimate the accumulation and release of strain by earthquakes in plate boundary zones and zones of distributed deformation. Until recently, studies of earthquake statistical distributions presented earthquakes as points in space and time. Even in those cases where the extent of rupture has been investigated, it was done by studying the data for individual earthquakes. Here we combine known earthquake distributions to present a geometrical and statistical picture of deformation at certain points along seismic fault.

[5] Most tectonic strain is released by the largest earthquakes [Brune, 1968; Kagan, 2002a]. There is little uncertainty about the size distribution for small and moderate earthquakes: they follow the Gutenberg-Richter (G-R) relation. A similar conclusion is valid for moderate and large earthquakes distributed over large seismogenic regions [Utsu, 1999]. Kagan [1999, 2002a, 2002b] and Bird and Kagan [2004] argue that earthquake size distribution has a universal $b$ value. An upper bound or maximum magnitude needs to be introduced for earthquake size distributions...
[Kagan, 2002a, 2002b]. We review several distributions in which this limit is applied and consider its influence on estimating rate of seismic deformation and slip.

[6] Despite extensive investigations and an extended debate [see, e.g., Scholz, 1997, 1998; Romanowicz, 1994; Abercrombie, 1995; Main, 2000; Romanowicz and Ruff, 2002], no consensus exists on scaling relationships between earthquake magnitude and geometric variables such as the rupture length, width, and average slip. Therefore we will explore different scaling relations. What are their implications for statistical distributions of slip?

[7] The G-R relation for magnitudes is equivalent to the power law (Pareto) distribution for the seismic moment [Utsu, 1999; Kagan, 2002a, 2002b]. This implies that the geometry of earthquake rupture is also controlled by power law distributions. Depending on the value of the exponent, sums of power law variables may not converge to any finite value [Zaliapin et al., 2005] or may be highly fluctuating.

[8] If a maximum size in a power law earthquake size distribution is present, the sum distribution and its random fluctuations strongly differ from that for regular statistical variables, following the Gaussian law which is mostly (explicitly or more often implicitly) used in geophysical practice. For example, Holt et al. [2000] found that a sum of earthquake seismic moments for some regions of southeastern Asia may exceed the tectonic rate by a factor of 3–5 and more. This difference cannot be explained by any errors in tectonic rate calculations or seismic moment evaluation. The mismatch is clearly caused by random fluctuations of the power law distributed variable. This highly random behavior of cumulative seismic moment necessitates a new way to compare seismic and tectonic rates [Kagan, 1999, 2002a, 2002b; Kreemer et al., 2003; Bird and Kagan, 2004]. Thus we will discuss statistical properties of accumulated earthquake slip and consider appropriate estimates of statistical error for measuring it.

[9] In studying slip distribution we should proceed from representing a point source to the approach based on extended earthquake rupture. This change requires applying statistical distributions of the rupture length, width, and slip to infer the pattern of average slip not only for a region or an area but also for a specific place on a fault.

[10] The temporal pattern of large earthquakes is key to the distribution of cumulative earthquake slip. Unfortunately, it is still not known whether large events cluster in time or follow a quasiperiodic pattern. Long-term earthquake hazard estimates are usually based on the Poisson assumption: Occurrence of one event is statistically unrelated to occurrence of others. Thus we need to consider various hypotheses of moment and slip release in time and their impact on slip distribution.

[11] An additional issue needs to be discussed: How earthquake rupture is concentrated in space. Two models are commonly used: tectonic block motion and continuum deformation. The former model envisions a very narrow concentration of ruptures, whereas the latter expects the deformation to be broadly distributed. As we show in section 2, in reality, both points of view are somewhat valid; the spatial fractal framework combines and synthesizes both these hypotheses.

[12] In this paper, first we will discuss the earthquake size distribution (seismic moment-frequency relation) and review the major theoretical relations proposed to approximate observational data (section 3). Section 4 reviews models for earthquake scaling, or the relationship between the seismic moment and rupture geometry. We obtain statistical expressions for rupture lengths and slip distributions based on scaling and magnitude-frequency relations. Section 5 applies the relations derived in sections 3 and 4 to earthquake ruptures at specific sites of a fault. Temporal distributions of large earthquakes and slip accumulation at a specific point of a fault are considered in section 6. We illustrate the obtained results by evaluating statistical properties of slip distribution at Wrightwood, California, on the San Andreas fault. Sections 7 and 8 summarize our results. Appendices A–C (see auxiliary material1) contain some technical statistical formulas.

2. Deformation Pattern: Continuum Versus Block Motion

[13] King et al. [1994] ask whether block or continuum deformation should be used to describe patterns of seismicity and deformation in the western United States. A continuous representation is usually used for large seismic regions [e.g., Holt et al., 2000; Kreemer et al., 2003]. For smaller, extensively studied regions such as California, earthquake deformation is normally represented as occurring on specific faults [Wesnousky, 1986].

[14] Earthquake data are traditionally displayed as epicenter points or focal mechanisms on a map, although the rupture’s extent can be estimated from surface rupture traces and aftershock distributions. To analyze strain at the Earth’s surface, we need to display earthquake sources as extended ruptures, at least for large earthquakes. To exemplify earthquake data, Kagan et al. [2005, Figure 1] show focal mechanisms for moderate and large earthquakes in southern California from 1850 to 2003. Kagan et al.’s Figure 1 suggests that earthquakes are not concentrated on a few faults. The mechanisms of neighboring events may have a very different orientation. Even in a neighborhood of major faults, some focal mechanisms significantly disagree with fault surface traces. This mismatch confirms the idea that major faults do not fully represent the deformation pattern, even in a region with relatively simple and well-studied tectonics. In contrast, measuring deformation on major faults and in their neighborhood indicates that most strain is within confined zones (often narrow enough to trench in paleoseismic investigations). Both findings suggest that block and continuum hypotheses need to be combined in a joint model.

[15] Kagan and Knopoff [1980] and Kagan [1991] show that the spatial distribution of epicenters and hypocenters can be described by a fractal, scale-invariant relation with the value of the correlation dimension $d_\text{corr} = \approx 2.2–2.3$. These results imply that neither block models (with a finite number of faults) nor continuous models can accurately describe the tectonic deformation pattern.

[16] In this paper we concentrate on studying slip on individual fault segments. Observations [e.g., Sieh et al., 1989; Weldon et al., 2002] and modeling of fractal fault systems [Kagan, 1982; Libicky and Ben-Zion, 2005] suggest

that while faults generally show a complicated fractal branching structure, much tectonic deformation is concentrated in very narrow zones in certain places. These zones can be represented by a planar fault intersecting the Earth’s surface, displacing river valleys and other geomorphic features by measurable distances. Such displacements are often used to estimate the average fault slip rate. Given that no statistical description of a fault displacement exists, it is difficult to estimate uncertainties in evaluating the rate, at least those uncertainties arising from random fluctuations. Thus we will model fault deformation statistically and estimate random errors in tectonic rate calculation quantitatively. As explained in section 1, for power law distributed variables, statistical fluctuations can be very large. To evaluate them, we should know their statistical properties.

3. Seismic Moment Statistical Distribution

[17] We need to find the most appropriate statistical relationship that describes earthquake size distribution. The relation between magnitude and moment-frequency is one of the most studied statistical properties of earthquake occurrence. However, many unresolved controversies remain open in the properties of earthquake size distribution. We review these, emphasizing the moment distribution main open in the properties of earthquake size distribution. However, many unresolved controversies remain open in the properties of earthquake size distribution. The relation between magnitude and moment-frequency is one of the most studied statistical properties of earthquake occurrence.

[18] We use the notation $M$ for the scalar seismic moment and $m$ for the moment magnitude [Hanks, 1992]:

$$m = \frac{2}{3} \log_{10} M - 6.0, \quad (1)$$

where $M$ is measured in Newton meters (N m). The magnitude calculated by equation (1) is used here for illustration and as a proxy for the seismic moment. All pertinent computations are carried out with moment estimates.

3.1. Theoretical Distributions

[19] In this section we briefly review statistical distributions for approximating the moment-frequency relation. For a more complete discussion, see Kagan [2002a, 2002b] and Bird and Kagan [2004]. The earthquake size distribution is usually described by the G-R magnitude-frequency law

$$\log_{10} N(m) = a - bm, \quad (2)$$

where $N(m)$ is the number of earthquakes with magnitude $\geq m$, and $b \approx 1$. Utsu [1999, equation (5)] [see also Aki, 1965] explains that the G-R law (equation (2)) is equivalent to the exponential statistical distribution [see, e.g., Evans et al., 1993] for magnitudes. For earthquake energy or seismic moment, the exponential distribution can be converted to a power law relation [Utsu, 1999, equation (2)] or, as this distribution is known in statistics, the Pareto distribution [Evans et al., 1993], with the power exponent $\beta = b/1.5$. Because seismographic networks have limited sensitivity, small earthquakes are not completely sampled in earthquake catalogs. Thus we must introduce a catalog completeness threshold (cutoff) $M_t$ and truncate the distribution from the left:

$$\phi(M) = \beta M_t^{\beta} M^{-1-\beta}, \quad M_t \leq M < \infty. \quad (3)$$

The threshold magnitude varies from about 1.0 for modern local catalogs like that of southern California to about 5.5 for the Harvard centroid moment tensor catalog [Ekström et al., 2003].

[20] Two types of errors need to be considered in fitting the theoretical earthquake size distribution to data: uncertainty in estimating the $\beta$ value and random fluctuations of earthquake numbers. Figure 1 displays the moment-frequency relation in the southern California area for 1800–1999, using the Toppozada et al. [2000] catalog. We approximate the observational curve by the Pareto distribution (3). The displayed 95% confidence limits for $\beta$ [Aki, 1965; Kagan, 2002a] are conditioned by the total number of earthquakes observed. To calculate full uncertainty bounds, it is necessary to convolve $\beta$ errors with the event number distribution. We make the simplest assumption that this distribution is Poisson. For the number ($N$) of samples in the Poisson distribution greater than 30, one can use the Gaussian approximation with the variance equal to $N$. Thus the total number of events in the subcatalog has the standard error $197 \pm 14.0$. To make the calculations easy for this illustrative diagram, we extend this approximation down to $N = 1$. For the upper limit the resulting uncertainties shown by the upper dashed curve in the plot are smaller than the actual Poisson bounds, so our limit is underestimate of the actual uncertainty spread. For the lower limit the opposite is true, so that the real lower limit should be between the $\beta$ (solid) line and the dashed line shown in the plot.

[21] Simple considerations of the finiteness of seismic moment flux or of deformational energy available for an
earthquake generation [Kagan, 2002a, 2002b] require that the Pareto relation (3) be modified at the large size end of the moment scale. We consider four theoretical distributions to describe seismic moment-frequency relations with an upper bound [Kagan, 2002a, 2002b]: (1) the characteristic distribution (model a); (2) the truncated Pareto distribution (model b); (3) the modified G-R distribution (model c), and (4) the gamma distribution (model d). These distributions have a scale-invariant, power law segment for small and moderate earthquakes; the right-hand tail of the distributions have a scale-invariant, power law segment for small and moderate earthquakes. The modified G-R distribution (model c), and (4) the gamma distribution (model d). These distributions apply a “soft” exponential taper to the distribution tail: in case c to the CDF and in case d to the PDF.

Figure 2. Seismic moment statistical distribution, Harvard catalog 1 January 1977 to 30 September 2002 [Ekström et al., 2003], subduction zones for (a) strike-slip earthquakes and (b) thrust earthquakes. Moment threshold is 10^{17.7} N m (m ≥ 5.8). Top four curves are earthquakes on ocean side; bottom curves are earthquakes on continental side. Solid line is approximation by a Pareto distribution (3), dashed lines are 95% confidence limits [Aki, 1965], conditioned by the total number of earthquakes observed. The limits show approximately the uncertainty due to the β value estimation error. The number of events should follow the Poisson distribution with these parameter values; hence the full uncertainties would be higher. For smaller time intervals the Poisson distribution is far from the Gaussian; the accurate uncertainties would be higher. For smaller time intervals the Poisson distribution is far from the Gaussian; the accurate uncertainties would be higher. For smaller time intervals the Poisson distribution is far from the Gaussian; the accurate uncertainties would be higher.

We classify earthquakes by their focal mechanisms: thrust earthquakes. The characteristic moment (magnitude) is used in these expressions: in case a 

\[ F(M) = \frac{(M_t/M_{\text{cm}})^\beta - (M_t/M_{\text{cm}})^\alpha}{1 - (M_t/M_{\text{cm}})^\beta} \]

where \( F(M) \) is a cumulative function and \( M_{\text{mc}} \) is the maximum moment. Equation (4) does not exactly correspond to the characteristic earthquake model as formulated by Schwartz and Coppersmith [1984] or by Wesnousky [1994]. In their model the characteristic earthquakes release about 87% of the total moment [Kagan, 1996]. In case a, only 33% (for \( \beta = 2/3 \)) of the total is released by the characteristic events [Kagan, 2002b]. Kagan [1996] criticizes using the characteristic model for seismicity analysis; see also Wesnousky [1996], Rong et al. [2003], and Stein and Newman [2004] for more discussion. We use model a here for illustration, as it is more compatible with other distributions (models b–d).

Similarly, for case b,

\[ \Phi(M) = \frac{(M_t/M_{\text{cm}})^\beta - (M_t/M_{\text{cm}})^\alpha}{1 - (M_t/M_{\text{cm}})^\beta} \]

For the tapered G-R and gamma distributions the expressions are

\[ \Phi(M) = (M_t/M_{\text{cm}})^\beta \exp \left( \frac{M_t - M}{M_{\text{cm}}} \right) \]

where the normalization coefficient C is defined by equation (17) of Kagan [2002a] (for \( M_{\text{eg}} \gg M_t \) the coefficient C ≈ 1) and \( \Gamma(a, b) \) is the incomplete gamma function [Abramowitz and Stegun, 1972, p. 260].

The first two distributions are utilized extensively in practical applications. A “hard” cutoff for the maximum moment (magnitude) is used in these expressions: in case a for the cumulative distribution function (CDF) and in case b for the probability density function (PDF). The two latter distributions apply a “soft” exponential taper to the distribution tail: in case c to the CDF and in case d to the PDF.

3.2. Empirical Evidence: Moment-Frequency Relation

Pacheco et al. [1992], Okal and Romanowicz [1994], Triep and Sykes [1997], and others (for more references, see Main [2000] and Leonard et al. [2001]) argue that the β value (3) increases from 2/3 to 1.0 over the moment range \( M = 10^{17.25} \rightarrow 10^{21} \) N m (\( m = 5.5 \rightarrow 7.5 \)). They believe that the finite thickness of a seismogenic crust explains this change.
earthquakes have a B axis more vertical than other axes, respectively. In addition, we subdivide the earthquake distribution into two groups: one on the ocean and one on the continental side of the subduction boundary. The β values for all the samples are statistically indistinguishable and correspond to the universal value for the exponent [Kagan, 2002a, 2002b; Bird and Kagan, 2004].

[27] The strike-slip earthquake distribution (Figure 2a) is of special interest; according to the standard interpretation their rupture width should approximately equal the thickness of the seismogenic upper crust. Thus we expect the distribution to exhibit a break at about $10^{18} - 10^{19}$ N m ($m = 6.0 - 6.7$). Pacheco et al. [1992] suggest that the $b$ value should change from 1.0 to 1.5 (corresponding limits are 2/3 to 1.0 for the $β$ value) at this magnitude range. No break of such amplitude is observed in the plots, at least by inspection. Sornette et al. [1996] discuss more rigorous testing techniques for the break in the magnitude-frequency relation. Closely inspecting the spatial distribution of the strike-slip earthquakes indicates that they are not concentrated in any of the subduction zones. Hence the absent break is not due to the seismicity alone in one or two subduction zones.

[28] The tectonic structure in subduction zones may be very complex (i.e., the seismogenic crustal thickness highly variable) so that any systematic variation of rupture width may not be discernable. However, only subduction zones yield a sufficient number of earthquakes for statistical comparison. The absence of a break in scaling relation for strike-slip earthquakes can also be demonstrated by Kagan [2002c] results on the relation between the aftershock zone length and the main shock magnitude. Strike-slip events exhibit no change in correlation behavior up to magnitude $m ≈ 8$ main shocks (see more in section 4.2).

[29] Thrust earthquakes (Figure 2b) demonstrate a similar pattern: There is no observable statistical difference in these diagrams. The upper observational curve decay (Figure 2b) for the largest earthquakes ($M > 10^{22}$ N m, $m > 7.7$) is caused by the upper bound for moment distribution (the maximum or corner magnitude; see above equation (4)). This decay is not seen in the lower curve (Figure 2b), perhaps because its number of earthquakes is small. Bird and Kagan [2004] discuss the corner magnitude distribution for global earthquakes in various tectonic zones.

4. Earthquake Scaling

[30] To relate the moment-frequency law to observed geometrical properties of earthquake rupture and its statistics, we should know the relationship between these variables. This relation depends on earthquake scaling: the connection between the moment and rupture length, width, and slip. If we know or assume the scaling relation, we can then calculate the distribution of earthquake geometrical parameters by extending the moment-frequency relation.

4.1. Theoretical Principles

[31] Earthquake rupture is characterized by three geometric quantities: length of the rupture, $L$, width $W$, and average slip, $u$. The seismic moment is expressed through these quantities as [Scholz, 2002]

$$M = \mu u W L,$$  \hspace{1cm} (8)

where $\mu$ is an elastic shear modulus. The empirical connection between the length and the moment $M$ is expressed as

$$M \propto L^d,$$  \hspace{1cm} (9)

i.e., $M$ is proportional to $L^d$. From the scaling arguments and the observational evidence, it is generally agreed that $d = 3$ for small and moderate earthquakes whose size does not exceed the thickness of seismogenic layer [Scholz, 1997, 1998; Romanowicz, 1994; Abercrombie, 1995; Romanowicz and Ruff, 2002]. Since the stress or strain drop is generally assumed to be independent of $M$ and the rupture unbounded, these variables should depend on the moment only

$$L, W, u \propto M^{1/3}.$$  \hspace{1cm} (10)

[32] Let us consider the slip distribution at a particular site on the Earth’s surface along the San Andreas fault. Since total slip can be partitioned between the major fault (the San Andreas in our case) and subsidiary faults, slip distribution is for a thin slice orthogonal to the San Andreas fault. Most of the total moment is released by the largest earthquakes, $M > 10^{19.5}$ N m ($m > 7.0$) [Brune, 1968; Kagan, 2002a]; for these large earthquakes the rupture is usually considered confined to the upper crust layer of thickness $W$ [Scholz, 1997, 1998; Romanowicz, 1994; Wells and Coppersmith, 1994; Bodin and Brune, 1996; Stock and Smith, 2000]. Thus, if

$$W = W_0 = \text{const},$$  \hspace{1cm} (11)

the rupture can be characterized by two geometric quantities: length $L$, and average slip, $u$.

[33] Main [2000] and Leonard et al. [2001] discuss various methods of investigating geometrical scaling of the earthquake rupture. They propose evidence for a break in the scaling corresponding to the brittle-plastic transition at the base of the crust. Below we briefly review both of these relations: geometrical features of earthquakes and a supposed break in the magnitude(moment)-frequency relation.

[34] Expressions (8)–(11) implicitly assume that the hypocenter or centroid depth distribution is uniform over the thickness of the seismogenic layer; otherwise, the width $W$ is not properly defined. In reality, earthquake depth distribution is highly nonuniform. In Figure 3 we show the depth histogram for one of the southern California catalogs [Shearer et al., 2003], arguably the catalog with the most accurate earthquake locations. Most of its earthquakes are concentrated within the upper 2–12 km zone. However, slip inversions for several large earthquakes in California and elsewhere [Mai and Beroza, 2002] show that their slip distributions extend to greater depth. We know that slip inversions are not well constrained, especially for the lower part of the rupture width. Thus the exact slip distribution over depth still needs to be established.

[35] Several studies [see, e.g., Mori and Abercrombie, 1997; Gerstenberger et al., 2001; Wyss et al., 2004] suggest that the magnitude-frequency relation varies with depth. In
assuming the simple G-R law (2) these studies show that the $b$ value significantly decreases with depth. However, Kagan, [2002a, p. 539] argues that this decrease may be caused by an inappropriate use of relation (2) for approximating earthquake size distribution. If the maximum earthquake size is taken into consideration (equations (4)–(7)), a more reasonable explanation for the apparent $b$ value change would be an increase in the maximum size with depth [Kagan, 2002a, p. 538]. Centroid depth for a large earthquake on a vertical fault cannot be smaller than a half of the rupture width; inspecting seismic maps (such as Figure 2 of Wyss et al. [2004]) suggests that hypocenters of larger events are on average deeper than those for small earthquakes.

### 4.2. Empirical Evidence: Length Scaling

[36] Although other parameters of rupture have been correlated with the seismic moment, rupture length is determined with better accuracy. This accuracy is reflected, for example, in a higher correlation coefficient of $M$ versus $L$, as compared to the correlation of the moment with $W$, or $u$ [Wells and Coppersmith, 1994]: $\rho = 0.95$ for the first case, and $\rho = 0.84$ and $\rho = 0.75$ for the second and third case, respectively. Thus here we use only the $M$ versus $L$ correlation.

[37] Two models are usually proposed for length scaling of large earthquakes. The $W$ model [Romanowicz, 1994, and references therein] assumes that slip is proportional to the rupture width $W$ and thus is constant as long as equation (11) holds. The second or $L$ model [Scholz, 1997, 1998, and references therein] assumes that $u \propto L$. According to the $W$ model, $d = 1$, whereas the $L$ model requires that $d = 2$ (see equation (8)).

[38] Kagan [2002c] investigated the distribution of aftershock zones for large earthquakes in global catalogs (scalar seismic moment $M \geq 10^{19.5} \text{Nm}$, moment magnitude, $m \geq 7$). How the aftershock zone length, $l$, depends on earthquake size was studied for three representative focal mechanisms: thrust, normal, and strike slip. It was found that all earthquakes show the same scaling ($M \propto l^4$). No observable scaling break or saturation occurs for the largest earthquakes ($M \geq 10^{21} \text{Nm}$, $m \geq 8$). Henry and Das [2001] obtained an analogous scaling result. It is natural to assume that the aftershock zone length $l$ is equal or proportional to the rupture length $L$. It seems that earthquake geometrical focal zone parameters are self-similar.

[39] Preliminary data on the recent (26 December 2004) Sumatra great earthquake as well as addition of 2001–2004 earthquake catalogs allow us to extend Kagan [2002c] results. New regression curves (see http://scec.ess.ucla.edu/~ykagan/scal_update_index.html) show that $M \propto l^4$ dependence continues up to $m = 9$ earthquakes. Although the Sumatra earthquake is of thrust type, a few large $m \approx 8$ strike-slip events occurred in the 2001–2004 period. Estimated regression parameters for strike-slip and normal earthquakes are similar to those of thrust events, supporting the conjecture that the scaling relation is identical for earthquakes of various focal mechanisms.

### 4.3. Earthquake Scaling Relations

[40] As the discussion in sections 4.1 and 4.2 shows, geometrical scaling relations still challenge our understanding. The major difficulty in studying $L$, $W$, and $u$ is that these quantities are often not subject to direct measurement, especially for earthquakes occurring under the ocean. Seismic moment is the best statistically studied variable among the earthquake scaling parameters. Therefore we may try to infer statistical properties of other geometrical quantities by using their relations to the moment. If the assumed functional dependence between the moment and a variable $x$ has a form

$$x = f(M),$$

then the PDF for $x$ can be expressed as

$$\varphi_x(x) = \varphi_M[f^{-1}(x)] \left| \frac{\partial f^{-1}(x)}{\partial x} \right|,$$

where $\varphi_M$ is the PDF of the seismic moment and $f^{-1}(x)$ is the inverse function of (12).

[41] If $x$ itself has a power law dependence on the moment, say

$$x \propto M^{1/\nu} \quad \text{or} \quad M \propto x^\nu,$$

then for the scale-invariant part of the moment PDF (equations (4), (5), (6), and (7)) we obtain the following relations for variable $x$:

$$\varphi_x(x) \propto (x^\nu)^{-1-\beta} x^{\nu-1} = x^{-1-\nu\beta},$$

i.e., the $x$ variable is power law distributed with the exponent value $\nu\beta$. Using (9) for the rupture length, we obtain

$$\varphi(L) \propto (L^d)^{-1-\beta} L^{d-1} = L^{-1-d\beta}.$$
Hence, for earthquakes smaller than the maximum or corner moment $M_0$, the length $L$ is distributed according to a power law (see Table 1 for exponent values). Similarly, the PDF for the other variables, like width $W$ and average slip $u$, can be calculated as long as they follow the same proportionality relations as the moment (see equation (10)).

[43] If we assume that (11) is true,

$$L = L_0 \left( \frac{M}{M_0} \right)^{1/d} \quad (17)$$

and

$$u = u_0 \left( \frac{M}{M_0} \right)^{(d-1)/d} \quad (18)$$

where $M_0$, $L_0$, and $u_0$ are the seismic moment, length, and average slip for a reference earthquake, taken here for illustration to be $M_0 = 10^{19.5}$ N m ($m = 7$), i.e., the smallest earthquake for which its rupture width reaches the maximum value $W_0$. For an $m7$ earthquake, assuming a width $W_0$ of 15 km and elastic modulus $\mu$ equal to 30 GPa [Scholz, 2002, p. 207], we take $u_0 = 1.87$ m and $L_0 = 37.5$ km. We use these reference earthquake parameters in discussion below as an illustration. The derivations can be easily repeated using other values.

[45] The ratio $\theta = u/L$ is an important parameter which determines the stress drop $\Delta \sigma$

$$\Delta \sigma = C_\mu \theta \quad (19)$$

where $C$ is a coefficient of the order of unity [Scholz, 2002, equation 4.29].

[44] Using (17) and (18), we obtain the following expression for $\theta$:

$$\theta = u_0 \left[ \frac{M}{L_0} \right]^{(d-2)/d} \quad (20)$$

From equation (20) the stress drop depends on the moment of an earthquake with the exponent equal to $-1/3$, 0, and $1/3$ for $d = 1.5$, 2.0, and 3.0, respectively. The independence of the stress drop from earthquake size seems well established [Scholz, 2002; Abercrombie, 1995]. Therefore assumption (11) leads us to accept the $L$ model of scaling ($d = 2$).

[45] However, assumption (11) may not be true, as suggested by the evidence discussed in sections 4.1 and 4.2. Then using (10), for $d = 3$ we obtain

$$\theta = \left\{ \frac{u_0}{L_0} \right\}, \quad (21)$$

i.e., the stress drop is again constant.

[46] From (18) similar to (13) and (16), we obtain the scale-invariant part of the PDF for the slip

$$\varphi(u) \propto \left[ u^{1/(d-1)} \right]^{-(3-\beta)} u^{1/(d-1)} = u^{1-(3\beta)/(d-1)}. \quad (22)$$

Hence for earthquakes smaller than the maximum or corner moment $M_0$, slip $u$ is distributed according to a power law (see Table 1).

[47] Table 1 shows the exponent values for three choices of the moment-frequency exponent. Value $\beta = 2/3$, corresponding to the G-R $b = 1$, is the most often quoted quantity. Kagan’s [2002a, 2002b] results and those of Bird and Kagan [2004] suggest that a slightly lower $\beta$ value of 0.63 characterizes shallow earthquake seismicity. The index value $\beta = 0.6$ may be appropriate for shallow earthquake sequences or deep earthquakes [Kagan, 1999]. Finally, $\beta = 1/2$ is the theoretically derived value [Vere-Jones, 1976].

5. Earthquakes at a Specific Site

[48] Most known statistical distributions for earthquake size (magnitudes or seismic moment) are constructed for areas or regions. However, earthquake rupture extends over the length of a fault. Thus, if we are interested in earthquake phenomena observed at a particular point on a fault, we need to “translate” these distributions into a site-specific form. Large earthquakes have a higher probability of intersecting the site than smaller ones, and this probability should be accounted for. Below we calculate site distributions for seismic moment and earthquake slip.

5.1. Earthquake Moment Distribution at a Specific Site

[49] To calculate the distribution of the seismic moment at a single site $\nu(M)$, we consider two cases: a potential width of the rupture $W$ is greater than $W_0$, the width of the seismogenic zone, and $W \leq W_0$. In the first case we multiply the distribution density function of the seismic moment $\phi(M)$ by the length of the rupture $L(M)$ as described by

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*For $u \propto M^{d-3}$ ($d = 3$), width $W$ is not limited, whereas all other columns assume rupture width $W \leq W_0$ (15 km); see equation (11); $x_s$ is area-specific distribution; $x_t$ is area-specific distribution; $x_p$ is area-specific distribution; and $x_t$ is reduced site-specific slip for small earthquakes, $m \leq 7$. 

### Table 1. Values of Exponents for the Scale-Invariant Part of Various Fault Geometrical Distributions

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crust’s upper 1.5 km layer, then the smallest earthquake that could rupture the surface would be about m5. Since the total displacement of seismogenic crust at any depth interval over long time should be uniform, this would mean that the slip of large earthquakes should “catch up” on the slip deficit at the Earth’s surface left by smaller events; that is, large earthquake slip at the surface should be on average greater than the slip in the crust middle layers. If this conjecture is true, then the slip and, by inference, moment distributions in paleoseismic studies would significantly differ from the power law (like equation (3)), which explains instrumental seismology data. This deviation may provide a misleading evidence in support of claims in the characteristic earthquake model.

[55] In general, the value of the exponent equal to 1.0 means that in the logarithmic scale the total sum for a variable is uniformly distributed. For example, the total length (Ls) for d = 3 and β = 2/3 is equal for all magnitudes. The value of the exponent equal to 0.0 means that the same condition is satisfied for a linear scale.

[53] In the second case (W < W0), several earthquakes are needed to fill out the vertical zone of length W(M). Thus, in addition to (24) we multiply the density by a term W(M) × Md/2:

\[
v(M) \propto L(M) W(M) \phi(M) = M^{d/2} \phi(M).
\]

where we use d = 3 as in equation (10), appropriate for this case. We assume that small earthquakes are distributed uniformly in width W0.

[54] The latter expression may be considered as a PDF of a fault surface covered by earthquake ruptures

\[
v(S(M)) \propto M^{-1-β+2/3},
\]

i.e., for β = 2/3, earthquakes of any magnitude or log moment range cover the same total surface [Rundle, 1989].

[55] In Appendix A (available as auxiliary material) we derive several formulas of the complementary moment functions for cases a and b. The expressions for cases c and d can be obtained by the same method, not shown because of their length. In Figure 4 we show several examples of these functions. The functions are not normalized in this case. It is to be expected that for a smaller value of the d exponent the rupture length of large earthquakes will increase. Thus more ruptures will intersect the site.

[56] In Figure 4 we adjust the maximum moment M0 or the corner moment Mc of the distribution according to equation (7) of Kagan [2002b], so that the total moment rate M is identical for all four distributions: Mc = 1021 and 3.38 × 1021 and M0 = 1.40 × 1021 and 4.74 × 1021 N m, for models a–d, respectively. These moment values correspond approximately to magnitude values m0 = 8.0, 8.35, 8.10, and 8.45, appropriate, for instance, to southern California [Jackson et al., 1995; Field et al., 1999]. Kagan [2002b, Figure 2 and equation 21] estimates that for such conditions, m < 7 earthquakes contribute about 21% of the total seismic moment. Bird and Kagan [2004] also obtained the corner magnitude (case c, see equation (6)) values of mcm = 8.04–0.21 for

**Figure 4.** Probability functions for the site-specific moment distribution for M ∝ L1.9. Solid line is truncated G-R cumulative distribution, model a in section 3.1; dashed line is truncated G-R distribution density (model b); dash-dotted line, modified gamma distribution (model d). Dotted line corresponds to the unrestricted G-R distribution.
obtained numbers of earthquake ruptures are comparable with the number of $m \geq 7$ earthquakes found during paleoseismic explorations of the San Andreas fault [e.g., Sieh et al., 1989; Weldon et al., 2002] (see also Figure 9 in section 6.3).

5.2. Earthquake Slip Distribution at a Specific Site

Again, assuming $W = W_0$, as in (16) using (24) as the PDF for distribution ($\varphi_M$), we obtain the following site-specific relations for the rupture length ($L_s$):

$$\varphi(L_s) \propto (L_s^d - 1)^{-1} \frac{M_s^3}{L_s^{d-1}} = L_s^{-d} = L_s^{1-\kappa}. \quad (28)$$

The length $L_s$ is distributed according to a power law (see Table 1).

In the second case ($W < W_0$), several earthquakes $n \propto M_s^{1.5}$ are needed to fill out the rupture width. We assume that their slip distribution is the same as that for larger ($m \geq 7$) events. Each of these smaller events on average contributes $u/n$ slip to total displacement of the surface.

For illustration we calculate the number of earthquakes at a site for conditions similar to southern California, setting in equation (A4) $M = M_0 = 1$. We take $\alpha_0 = 0.04$ (rate of $M \geq 10^{19.5}$ N m, $m \geq 7$ earthquakes, see Figure 1), the total length of the fault system $L = 700$ km, and $\beta = 2/3$ [Kagan, 1996]. To illustrate, for the truncated Pareto distribution,

$$n = \frac{\alpha_0 L_0}{L} \frac{\beta^d}{M_s^{d-1}} \frac{M_s^3}{M_s^{d-1} - 1} \left(1 - M_s^{-\kappa}\right). \quad (27)$$

$M_s = 106.7$, if we normalize $M_0 = 1$. Then for $d = 3$ we obtain 3.61, 3.54, 3.52, and 3.5 (cases a, b, c, and d, respectively) events that rupture an arbitrary site per millennium. The difference between various distributions is insignificant. For $d = 2$ the numbers are approximately 4.9, and for $d = 1.5$ the numbers are around 7.0. The analogous tectonic regions (continental transform boundaries) in global earthquake distribution.

Comparing the moment-frequency log-log plots [see, e.g., Kagan, 2002a, Figure 1] with Figure 4, we see another significant difference between the moment distributions for an area and a site-specific distribution: Even for the characteristic model, the cumulative curve is no longer linear in a log-log plot. The reason is that site-specific distributions are defined for earthquakes close to the upper limit of the distribution. However, the right-hand tail of the cumulative distributions is not linear for area-specific curves. For the characteristic law we adjusted the number of characteristic events in such a way as to produce a linear area-specific curve. This adjustment is no longer valid for a site-specific distribution.

For illustration we calculate the number of earthquakes at a site for conditions similar to southern California, setting in equation (A4) $M = M_0 = 1$. We take $\alpha_0 = 0.04$ (rate of $M \geq 10^{19.5}$ N m, $m \geq 7$ earthquakes, see Figure 1), the total length of the fault system $L = 700$ km, and $\beta = 2/3$ [Kagan, 1996]. To illustrate, for the truncated Pareto distribution,

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In Figures 6a and 6b we display the distributions of the total slip (see Appendix C for the expressions). In Figure 6a it is assumed that all surface slip is due to large \((m \geq 7)\) earthquakes; in Figure 6b we estimate the contribution of smaller events. For the reasons explained in section 4.1, the slip accumulated close to the Earth’s surface due to small and moderate \((m < 7)\) earthquakes is probably released during the rupture of large events.

In both Figures 6a and 6b, the distribution of the total slip for \(d = 1.5\) is linear, demonstrating that in each linear interval, earthquakes contribute an equal amount to the total slip budget. However, the length of the rupture for \(d = 1.5\) is too large for an \(m = 8.35\) earthquake, whose \(L = 840\) km. For the greater \(d\) values the larger part of the slip total is released by more extensive slip events. For example, for \(d = 3\) about 50% of the total slip is caused by slips in individual earthquakes exceeding 15 m. The largest slip in the \(d = 3\) case exceeds 50 m for the largest earthquakes: a value that seems unrealistic, especially for the Gamma distribution.

However, if we assume that \(d = 3\) and allow for an unlimited width \(W\), the total slip for \(m = 8.35\) earthquake is only 8.87 m, although its width should reach 71.1 km, and the length \(L = 177.9\) km. These parameter values may be appropriate for subduction zones. For such \(m > 8\) earthquakes the scaling is most likely a mixture of condition (11) and unlimited \(W\).

6. Seismic Moment and Slip Release in Time

The seismic moment and slip distributions obtained in section 5 cannot be used for theoretical modeling and practical purposes of seismic hazard evaluation without a proper understanding of earthquakes occurrence in time. Unfortunately, there is no present consensus on even basic facts about the temporal distribution of the large earthquakes or principles for their mathematical representation. Below we briefly discuss available data and their interpretation and show one example of applying results of this work.

6.1. Temporal Evolution of Seismic Moment Release

There are several models for stress accumulation and release by earthquakes. Most of these models assume that stress is scalar, that is, the tensor properties of stress are disregarded. These models are descendants of Reid’s [1910] “elastic rebound theory.” Two recent modifications of this model are presented by Shimazaki and Nakata [1980]: the “time-predictable” and “slip-predictable” earthquake occurrence hypotheses. In the former, stress accumulates until it reaches an upper critical value and is then released by earthquakes of various magnitudes. In the latter model the release of stress during an earthquake is limited by a lower critical quantity; following stress release, it accumulates until a new earthquake is triggered. In both models the total variation of stress cannot exceed the moment released by the largest earthquakes. It is also assumed that after such an earthquake the tectonic stress level is close to zero [Jaume and Sykes, 1996, Figure 4].

The often preferred time-predictable model implies that the probability of a new earthquake is lower, especially after a large event, since the stress field is significantly depleted: A stress shadow is created. This conclusion contradicts the universally observed clustering of shallow earthquakes. The most obvious manifestation of such clustering is aftershock sequences subsequent to strong events [Kagan and Jackson, 1999; Rong et al., 2003].

A commonly used explanation of the contradiction is that the shadow effects start following an aftershock sequence and outside the aftershock zone. However, it is difficult to draw temporal and spatial boundaries for aftershock sequences. Some of these sequences (such as the ones following the 1952 Kern County earthquake, or the Nobi earthquake of 1891 [see Utsu et al., 1995]) are still active after several decades. There are many indications that intraplate earthquakes, where the tectonic deformation rate is low, have aftershock sequences decades and centuries long [Ebel et al., 2000]. There are also distant aftershocks, which further complicate determination of the aftershock zone.

Another attempt at explaining the above contradiction is that clustering is a property of small earthquakes only; large events rupturing the whole crust follow a time- or slip-predictable model. Yet Kagan and Jackson [1999] show that even \(M \geq 10^{20.25}\) N m \((m \geq 7.5)\) earthquakes continue to exhibit time-distance clustering. The time interval between closely spaced pairs of earthquakes is significantly less than the time span needed for the plate motion to accumulate the strain released by the first event. Some \(M \geq 10^{20.25}\) N m \((m \geq 7.5)\) earthquake pairs listed by Kagan and Jackson [1999] clearly qualify as members of a foreshock-main shock-aftershock sequence, which suggests that strong earthquakes can recur at a small time interval in essentially the same area. Pérez and Scholz [1997] demonstrate that very large earthquakes \(M \geq 10^{21.75}\) N m \((m \geq 8.5)\) are still accompanied by foreshock/aftershock sequences that include large earthquakes. Hence these earthquakes do not fully release accumulated deformation. It is still possible that earthquakes can become large enough to deter others at some greater threshold. However, if an earthquake releases only a small part of tectonic stress, then many assumptions of earthquake temporal behavior need revision.

To illustrate possible stochastic variations in the seismic moment release, in Figure 7 we display four simulations of the process. Their earthquake size distribution is assumed to follow the truncated density G-R law (5), with the maximum moment equal to \(10^{31}\) N m \((m = 8)\) and the minimum size earthquake equal to \(10^{19}\) N m \((m = 6)\). The earthquake temporal distribution is assumed to be governed by the Poisson law with the rate of occurrence of one event per unit of time. We use the technique described by Kagan [2002a, equation 40] to simulate truncated Pareto distribution for moment and the standard methods to obtain the Poisson variable.

The curves’ behaviors significantly differ from the idealized pictures of stress-time history in the time- or slip-predictable models [Jaume and Sykes, 1996, Figure 4]. However, some of the trajectories (like the solid curve in Figure 7) can be interpreted using these hypotheses. Three of the four series in Figure 7 imply stress buildups that look implausible.

The possible reason for such moment behavior is use of the Poisson process to model earthquake occurrence. The
problem with the Poisson distribution utilized in Figure 7 is that regional stress accumulation would change drastically, since the Poisson process allows a potentially infinite accumulation of strain in a region. Clearly, some modification of Poisson model strain accumulation is necessary. Zheng and Vere-Jones [1994] propose a stochastic version of the elastic rebound model for describing large earthquake occurrence within a seismic region. This model replaces a deterministic “hard” threshold with a soft statistical critical cutoff to allow within certain stochastic limits clustering of large earthquakes. Further development of this model and review of results is presented by Bebbington and Harte [2003]. However, the results of this hypothetical application are still ambiguous, perhaps because the model does not fully incorporate earthquake spatial parameters. Thus the problem persists. Are the largest earthquakes quasiperiodic or clustered? Is one pattern changed by another for large enough time intervals?

6.2. Slip Accumulation in Time

[73] Difficulties in describing earthquake temporal distribution remain for slip accumulation as well. Chery et al. [2001] discuss many qualitative examples of long-term clustering for large earthquakes. Rockwell et al. [2000] and Dawson et al. [2003] discover a similar temporal earthquake clustering in eastern California. These researchers define clustering as a deviation of earthquake occurrence from a periodic or quasiperiodic pattern. In statistics, temporal clustering is normally assumed if a point process has a larger coefficient of variation than the Poisson process [Kagan and Jackson, 1991, Figure 1]. For example, Sieh et al.’s [1989] results are usually interpreted as evidence for earthquake clustering, although they find that the Poisson hypothesis cannot be rejected as a model for the temporal event series they describe. Generally, paleoseismic investigations cannot distinguish earthquakes occurring closely in time. Hence their coefficient of variation estimates should be biased toward smaller quantities: a more periodic pattern.

[74] Quantitative studies of earthquake temporal occurrence [Kagan and Jackson, 1991, 1999] suggest that strong earthquakes are more clustered than a Poisson process. Such clustering can presently be established for instrumental catalogs: for time intervals of a few decades at maximum. Paleoseismic investigations [Rockwell et al., 2000; Dawson et al., 2003] also suggest that earthquakes follow a long-term clustering pattern at least in regions of slow tectonic deformation. Average slip rates are known to be stable for most significant faults over 10,000 to 10^6 years. However, earthquake long-term clustering implies that the rates should fluctuate at least on the order of hundreds or thousands of years.

[75] Recent paleoseismic investigations of slip distribution on California faults [Weldon et al., 2002; Scharer et al., 2003] cast doubt on the quasiperiodic model of slip accumulation. Although the Poisson distribution for slip events on earthquake faults may be incorrect over long time intervals, this distribution approximates the observed temporal sequences. Thus, as a provisional model of seismic temporal behavior, we suggest that at relatively short time intervals (less than a millennium), strong earthquakes are clustered. For longer time intervals (several thousand years) their pattern can be approximated by the Poisson process, and, finally, for even longer times, 2-D continuum variants of the stochastic stress release model Zheng and Vere-Jones [1994] may fit better.

[76] In Figure 8 we show four examples of synthetic slip history obtained by Monte Carlo simulations. These realizations demonstrate a significant random variability of slip release when one assumes that their time distribution is

Figure 7. Simulations of the seismic moment release history. Four examples are shown. Each represents an independent simulation of moment release history. Earthquakes are assumed to occur according to the Poisson law; the moment distribution is assumed to be the truncated power law distribution density (5) with $\beta = 2/3$. For illustration we use the maximum moment equal to 1000, and the minimum size earthquake is equal to 1.0. If the moment unit is assumed to equal 10^{18} N m, the maximum magnitude (or the maximum jump in the diagram) corresponds to an m8 earthquake.

Figure 8. Simulation of slip trajectory versus time at an arbitrary site (thick lines, solid and dot-dashed), Poisson in time, truncated Pareto in slip ($\zeta = 1/3$, $u_{\alpha p} = 10.33 m$). Dashed thin lines are Gaussian approximation (see equations (30) and (31)) for the sum, $\pm \sigma$; dash-dotted thin lines are Gaussian approximation for the sum, $\pm 2\sigma$. 

11 of 15
Poisson and the slip distribution is a power law. Simulation methods are analogous to that used in Figure 7. We also show the Gaussian approximations of random uncertainties [Kagan, 2002a, equation 10; Zaliapin et al., 2005, equations 60 and 61]

\[
\begin{align*}
\mu_y & = \begin{cases} 
\frac{\zeta}{1-\zeta} \left( y^{1-\zeta} - 1 \right) / (1 - y^{-\zeta}), & \zeta \neq 1, \\
\log(y) / (1 - y^{-1}), & \zeta = 1
\end{cases}, \\
\sigma_y^2 & = \begin{cases} 
\frac{\zeta}{2-\zeta} \left( y^{2-\zeta} - 1 \right) / (1 - y^{-\zeta}) - \mu_y^2, & \zeta \neq 2, \\
2 \log(y) / (1 - y^{-2}) - \mu_y^2, & \zeta = 2
\end{cases}
\end{align*}
\]

where \( \mu_y \) and \( \sigma_y^2 \) are the conditional mean and variance of each normalized summation, given the restriction on the maximum slip \( y \). These Gaussian estimates reasonably approximate the accumulated variable; for more, see Zaliapin et al. [2005].

6.3. Example: Slip Statistical Distribution, San Andreas Fault at Wrightwood, California

[Weldon et al. [2002] see also Scharer et al., 2003] present a detailed analysis of earthquake slip history on the San Andreas fault at Wrightwood, California. In Figure 9 we show the statistical distribution of slip events approximated by a power law (Pareto distribution). Since small slips may not be represented fully in the soil record, we employ a threshold slip 1.5 m, similar to an approximation of the earthquake seismic moment. Unfortunately, it is difficult to reliably estimate the slip threshold as can be done for the magnitudes. As explained in section 4.1, it seems likely that rupture at the Earth’s surface is caused mostly by large earthquakes. Therefore a deficit of small slips or an almost complete absence is to be expected. Liu et al. [2004] also find that the smallest offset at the Carrizo Plain (San Andreas fault) is 1.4 m. Hence we assume that the slip record is complete only for slip events larger than 1.5 m.

[79] The approximation of the slip distribution by the Pareto law yields \( \zeta = 1.52 \pm 0.51 \). When comparing these \( \zeta \) values to the theoretical estimates in Table 1, a few assumptions should be made. In Figure 9 the fault slip is measured at a point on the San Andreas fault. The distribution may differ from that of average slip \( \mu \) calculated in Table 1. Rockwell et al. [2002] indicate that surface slip may drastically change over relatively small distance over fault. Ward [2004] proposes approximating slip distribution along a fault by the Brownian bridge; applicability of this model needs to be explored. Moreover, as mentioned above, small slips have a larger chance of being overlooked, thus biasing the input data. Nevertheless, the obtained \( \zeta \) value is compatible with \( \zeta = 1 \) for the site-specific value shown for \( M^{1/3} \) dependence (Table 1).

[79] Figure 10 shows the slip history at Wrightwood with several uncertainties estimates. For the truncated distribution we assume that the maximum slip is 15 m. Bird and Kagan [2004] find that for continental transform faults the corner magnitude estimate is \( 8.04 \pm 0.47 \) (see section 5.1). An average slip of 15 m is consistent with this moment magnitude value. We also estimate the uncertainties of slip accumulation by a simulation performed similar to that of Figure 8. However, to estimate upper and lower bounds, we used 10,000 instead of 4 realizations as in Figure 8. For this choice of parameters the Gaussian estimates (equations (30) and (31)) reasonably approximate those obtained through simulation.

7. Discussion

[80] Three problems must be solved to explain earthquake recurrence: (1) the degree of allowed stress accumulation in seismic zones; (2) the distribution of strain release by earthquakes of various sizes; and (3) the distribution of

![Figure 9. Slip statistical distribution for paleoseismic measurements made in the San Andreas at Wrightwood, California. Threshold (minimum) slip is 1.5 m. Solid line is approximation by a power law Pareto distribution, \( \zeta = 1.52 \pm 0.51 \); dashed lines are 95% confidence limits [Aki, 1965], conditioned by the total number of earthquakes observed.](image1)

![Figure 10. Slip trajectory versus time for San Andreas at Wrightwood, California. Thick solid line is measurements; thin solid line is simulation; thin dotted line is Gaussian approximation (see equations (30) and (31)). Average estimate and 95% confidence limits are shown.](image2)
stress release along the fault during a large earthquake. The continuum/block character of Earth deformation (section 2) requires that all these problems be resolved within a framework of the 2-D or 3-D stochastic field process. As explained in section 1, both observational data and a theoretical foundation are not yet sufficient for solving the problems.

[81] We tried to solve a simpler problem: distribution of earthquake slip at tectonic fault sites where the slip is sufficiently concentrated to be measured by paleoseismic methods. Even this simpler task cannot be fully accomplished because several components of the earthquake process are not adequately known. Among these unknown or contested features are (1) temporal distribution for large and great earthquakes for time periods of decades, centuries, and millennia, (2) earthquake scaling for events $m > 8$, (3) dependence of earthquake surface slip on both the nonuniformity of earthquake depth distribution and a possible change of magnitude-frequency relation with depth, and (4) statistics of the slip distribution along fault rupture [cf. Mai and Beroza, 2002; Ward, 2004].

[82] The scheme of the stress evolution discussed above (particularly in section 6.2) raises two related questions. What is the maximum moment of earthquakes? What are the relations between the stress levels and earthquakes? Apparently these problems cannot be solved with available data, but accumulating paleoseismic results and space-geodetic measurements may provide the necessary input.

[83] Although solving the earthquake slip distribution problem in this paper cannot be considered complete and fully self-consistent, our theoretical results may be used to interpret various observational data. Even if consensus on earthquake size distribution, scaling of earthquake geometric parameters, and earthquake temporal behavior is not yet there, the derived formulas can be applied to different sets of assumed distributions.

[84] Finally, we briefly comment on issues needing resolution before we can create a viable model of the Earth’s surface deformation to compare with new and future GPS and InSAR measurements:

[85] 1. The fractal nature of fault systems should be explicitly used in constructing the model.

[86] 2. Earthquake focal mechanisms should be included as an intrinsic part of the model (see section 2). This would greatly increase the model’s complexity, since we would have to deal with tensor-valued stochastic processes.

[87] 3. The power law distributions that control earth- quakes need to be integrated with the model; that improvement will necessitate using stable statistical distributions: a rapidly developing discipline in mathematical statistics [Zaliapin et al., 2005].

8. Conclusions

[88] We have analyzed the distribution of earthquake size, geometrical scaling of earthquake rupture, and temporal earthquake patterns to derive a possible distribution for average slip on extended faults. The results can be summarized as follows:

[89] 1. Earthquake slip is power law distributed. The power law exponents are derived as they depend on various assumptions.

[90] 2. Formulas are proposed to approximate slip at paleoseismic sites and evaluate uncertainties due to the statistical nature of slip accumulation and its size distribution.

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