Self-Organized Criticality Across Thirteen Orders of Magnitude in the Solar-Stellar Connection

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ABSTRACT

The observed size distributions of solar and stellar flares is found to be consistent with the predictions of the fractal-diffusive self-organized criticality (FD-SOC) model, which predicts power law slopes with universal constants of $\alpha_F = (9/5) = 1.80$ for the flux, and $\alpha_E = (5/3) \approx 1.67$ for the fluence, respectively. In this Letter we explore the solar-stellar connection under this aspect, which extends over an unprecedented dynamic range of 13 orders of magnitude between the smallest detected solar nanoflare event ($E_{min} = 10^{24}$ erg) and the largest superflare ($E_{max} = 10^{37}$ erg) on solar-like Gtype stars, observed with the KEPLER mission. The FD-SOC model predicts a scaling law of $L \propto E^{(2/9)}$ for the length scale L as a function of the flare energy E, which limits the largest flare size to $L_{max} \leq 0.14R_{\odot}$ for solar flares, and $L_{stellar} \leq 1.04 R_{\odot}$ for stellar flares on G-type stars. In the overall we conclude that the universality of power laws (and their slopes) is a consequence of SOC properties (fractality, classical diffusion, scale-freeness, volume-flux proportionality), rather than identical physical processes at different wavelengths.

Keywords: methods: statistical — fractal dimension — self-organized criticality —

1. INTRODUCTION

The concept of self-organized criticality (SOC) was originally proposed by Bak, Tang, & Wiesenfeld (1987) and has been used in over 6000 publications since. An interdiscipliary approach of SOC phenomena is sketched in the book by Bak (1996). The SOC paradigm can be understood most easily with the sandpile analogy. If you trickle sand grains on top of a sandpile, the slope (or angle of repose) on the inclines of the sandpile start to steepen until the system reaches a critical angle (which is about 38° in real-world sandpiles). Every additional sand grain dropped onto the critically inclined slope will trigger a small or a large avalanche, which is a highly nonlinear process. Bak et al. (1987) developed the ingenious method of using a n-dimensional lattice grid to simulate next-neighbor interactions with a computer code. The "critical" sandpile is driven by the balance between the input (infalling sand grains) and the output (or avalanches running off the boundaries at the bottom of the sandpile). If the avalanche duration is shorter than the waiting time interval between subsequent avalanches, the system is said to be slowly driven. The sandpile experiments of Bak et al. (1987) produced power law distribution functions with slopes of $\alpha_S = 1.37$ for the avalanche size S, and $\alpha_T = 0.92$ for avalanche durations T, respectively. Extensive literature on this type of numerical SOC simulations can be found in Pruessner (2012).

About two decades later, after the execution of numerous SOC simulations in discretized n-dimensional space, a paradigm shift occurred that has been named the *fractal-diffusive self-organized criticality (FD-SOC)* model (see textbooks by Aschwanden 2011, 2025). The new approach is motivated by the dilemma that microscopic real-world simulations are computationally expensive, if not impossible, such as in the case of bridging microscopic (atomic) scales to macroscopic scales (measurable in a "real-world" scenario). Moreover, next-neighbor interactions have to be simulated with iterative steps, which makes it to a mathematically infinite equation system. Instead, next-neighbor interactions can be approximated by the classical diffusion process, and the spatial inhomogeneity of SOC avalanches can be approximated by fractal dimensions. This way the power law size distributions can be quantified with simple (macroscopic) physical scaling laws, which obey the scale-freeness in their statistical probability distribution functions.

¹ This paper is dedicated to Carolus J. Schrijver (1958-2024) in memoriam.

ASCHWANDEN & SCHRIJVER

It appears that the FD-SOC model is a suitable description for many astrophysical (solar and stellar) phenomena, but has also wide-spead applications in geophysics, biophysics, medical sciences, as well as in financial physics. More detailed descriptions of the FD-SOC model can be found in Aschwanden (2014, 2015), Aschwanden et al. (2016, 2021, 2022), Aschwanden & Gogus (2025), while alternative SOC reviews are given elsewhere (Aschwanden et al. 2013, 2016; Watkins et al. 2016; Sharma et al. 2016; McAteer et al. 2016).

The contents of this Letter includes observations (Chapter 2), data analysis (Chapter 3), discussion (Chapter 4), and conclusions (Chapter 5).

2. OBSERVATIONS

We use multiple data sets to synthesize a joint size distribution, selected by their published power law slopes α_E and energy ranges E_{min} , E_{max} of their energy (in physical units [erg]), not to be confused by their power law slopes of peak fluxes α_F , which are quantified in physical units of [erg/s]. A synthesized size distribution, as composed by 5 individual size distributions, is fitted to a theoretically predicted power law slope, as shown in Fig. (1). While the power law slopes are invariant accross 13 orders of magnitude, the energy definitions of these individual data sets are not homogeneous here. In the following, we subdivide the data sets into three groups according to the used energy definitions and wavelength dependence (Table 1): (i) thermal energies, radiating in soft X-rays (SXR), extremeultraviolet (EUV), and ultraviolet (UV); (ii) non-thermal energies (radiating in hard X-rays and gamma rays), and (iii) bolometric energies, radiating in white light.

2.1. Soft X-Ray Thermal Energies

The smallest solar flare events, also called "nanoflares" or "transient brightenings", were sampled by observations with SXT/Yohkoh (Shimizu 1995), EIT/SOHO (Krucker & Benz 1998; Benz & Krucker 2002; TRACE (Aschwanden et al. 2000; Parnell & Jupp 2000), and GOES (Li et al. 2012).

The physical parameters of nanoflares include electron temperatures of $T_e \lesssim 2$ MK, electron densities of $n_e \approx \lesssim 10^9$ cm⁻³, (loop) length scales of L = 2 - 20 Mm, and thermal energies of $E_{th} = 10^{24} - 10^{26}$ erg (Aschwanden et al. 2000). In comparison, the physical parameters of transient brightenings cover electron temperatures of $T_e \approx 4 - 8$ MK, electron densities of $n_e \approx 2 \times 10^9$ cm⁻³ - 2 × 10¹⁰ cm⁻³, (loop) length scales of $L = 5 \times 10^3$ km to 4×10^4 km, durations of T = 2 - 7 min, and thermal energies of $10^{26} - 10^{29}$ erg (Shimizu 1995).

Other studies yield similar values (Krucker & Benz 1998; Benz & Krucker 2002; Parnell & Jupp 2000), Li et al. 2012, in Fig.7 therein), but cannot be compared with the 3-D standard FD-SOC model here, since they use a 2-D model that results into steeper power law slopes of $\alpha_E \gtrsim 2.3 - 2.6$ for the flare energy. Generally, the thermal energy is defined by

$$E_{th} = 3k_B T_e n_e V = 3k_B T_e \sqrt{EM \times V \times f} , \qquad (1)$$

where k_B is the Boltzmann constant, $EM = n_e^2 V$ is the volumetric emission measure, V is the flare volume, and f is a filling factor. In early studies the volume has been quantified by a flare area A and a constant line-of-sight depth h_0 , i.e., $V = A h_0$, which corresponds to a 2-D (flat-world) model. Moreover, the fractality of the spatial inhomogeneity is ignored with a constant line-of-depth h_0 . Hence, a definition that is more consistent with the FD-SOC model is, after inserting of $n_e = \sqrt{EM \times V}$,

$$E_{th} = 3k_B T_e n_e V = 3k_B T_e \sqrt{EM \times L^{D_V}} , \qquad (2)$$

where D_V is the fractal (Hausdorff) dimension and has a mean value of $D_V = 2.5$ in a 3-D world.

2.2. Hard X-ray Nonthermal Energies

Size distributions of solar non-thermal flare energies have been sampled from ISEE-3/ICE (Lu et al. 1993), and from HXRBS/SMM data (Crosby et al. 1993). The latter data set contains a number of $N_{\rm HXRBS} = 11,352$ flare events and has been accumulated over an observational epoch of $T_{\rm HXRBS} \approx 11$ [years] (Crosby et al. 1993; Aschwanden et al. 2017). The total energy in electrons (> 25 keV) is found to cover a range of $E_{\rm HXR} \approx 3 \times 10^{28} - 10^{31}$ erg, as computed from the non-thermal electron energy radiated in hard X-ray (HXR) wavelengths at ≥ 25 keV with the HXRBS detectors onboard the SMM spacecraft (Dennis et al. 1985), assuming the thick-target bremsstrahlung model (Brown 1971). The energy in nonthermal electrons for the thick-target model is given by (Brown 1971; Crosby et al. 1993),

$$F(>E_0) = 4.8 \times 10^{24} A E_0^{-\gamma+1} E_m^{\gamma} \gamma(\gamma-1) \beta(\gamma-\frac{1}{2},\frac{1}{2}) \quad [\text{erg s}^{-1}] , \qquad (3)$$

where β is the beta function, and E_0 is the cut-off energy. The power law slopes are found to be essentially invariant during the solar cycle, with best-fit values of $\alpha_E = 1.53 \pm 0.02$, 1.51 ± 0.04 , 1.48 ± 0.02 , and 1.53 ± 0.02 during the year ranges of 1980-1982, 1983-1984, 1985-1987, and 1988-1989, which can be compared with the theoretical FD-SOC model prediction of $\alpha_E = (5/3) \approx 1.67$.

2.3. White-Light Bolometric Energies

Homogeneous searches for stellar flares in every available KEPLER light curve (Borucki et al. 2010) revealed up to 851,168 candidate flare events, which have an average flare energy of $\approx 10^{35}$ erg (Davenport 2016). A KEPLER flare catalog with a sample of 162,262 events that supposedly controls false-positive signals and artifacts of earlier data sets revealed a power law slope of $\alpha_E \approx 2$, from which a superflare with an energy of $\approx 10^{34}$ was estimated to occur on the Sun at least once in 5500 [years] (Yang & Liu 2019). For stellar flares we use the KEPLER data set of G-type stars, which is most similar to the solar flare data set, with the Sun being classified as a G5-type star. This stellar data set comprises $N_{\text{KEPLER}} = 55,269$ stellar flares and is observed during a time period of $T_{\text{KEPLER}} \approx 5$ [years] (Aschwanden & Güdel 2021). E_{bol} represents the energy of the bolometric (white-light) luminosity in the wavelength range of $\lambda \approx 4300 - 8900$ Å with the KEPLER instrument. The wavelength range of KEPLER (4300 - 8900 Å) overlaps with the green line (5548 - 5552 Å), the red line (6682 - 6686 Å), and the blue line (4502 - 4506 Å).

Following Namekata et al. (2017a), there are two differences between solar and stellar observations: time cadences and passbands. HMI/SDO observes the overall Sun with a 45 s cadene and a narrowband filtergram around an Fe λ 6173.3 line, while KEPLER carried out 1-minute cadence observations with 4000–9000 Å broadband filters. The white-light solar flare energy E_{wl} is radiated by a temperature $T_{flare} = 10,000$ K, and is calculated as

$$E_{wl} = \sigma_{\rm SB} T_{flare}^4 \int A_{flare}(t) dt , \qquad (4)$$

with the stellar flare area A_{flare} defined by,

$$A_{flare}(t) = \frac{L_{flare}}{L_{Sun}} \pi R^2 \frac{\int R_{\lambda} B_{\lambda}(5800 \text{ K}) d\lambda}{R_{\lambda} B_{\lambda}(T_{flare}) d\lambda} , \qquad (5)$$

where σ_{SB} is the Stefan-Boltzmann constant, L_{flare}/L_{Sun} is the flare luminosity ratio to the overall luminosity, R is the solar radius, R_{λ} is a response function of HMI/SDO, and $G_{\lambda}(T)$ is the Planck function at a given wavelength λ .

3. THEORY OF THE FD-SOC MODEL

In the following we describe briefly a definition of the Fractal-Diffusive Self-Organized Criticality (FD-SOC) model, while more extensive descriptions can be found in textbooks and review papers (Aschwanden 2014, 2015, 2022, 2025; Aschwanden et al. 2016). The FD-SOC model is based on four fundamental assumptions: (i) the 3-D dimensionality, (ii) the mean fractal (Hausdorff) dimension, (iii) the scale-freeness probability distribution function, and (iv) the volume-flux proportionality for incoherent emission processes.

Instead of using the next-neighbor interactions of cellular automata, as defined in the original SOC model of Bak (1987), we quantify the spatial inhomogeneity in terms of the fractal dimension D_d for the Euclidean domains d = 1 (lines), d = 2 (areas), or d = 3 (volumes). Each fractal domain has a maximum fractal dimension of $D_d = d$, a minimum value of $D_d = (d-1)$, and a mean value of $D_V = d - 1/2$,

$$D_V = \frac{(D_{V,\max} + D_{V,\min})}{2} = d - \frac{1}{2}.$$
 (6)

For most applications in the (observed) 3-D world, the dimensional domain d = 3 is appropriate, which implies a fractal dimension $D_V = 2.5$. However, if 2-D areas are observed, the fractal dimension is $D_A = 1.5$ and the dimensionality is d = 2. In this work we will mostly make use of the 3-D fractal domain, while the 2-D domain is discussed elsewhere (e.g., Aschwanden 2022). The fractal volume V is then defined by the standard (Hausdorff) fractal dimension D_V in 3-D and the length scale L (Mandelbrot 1977),

$$V(L) \propto L^{D_V} . \tag{7}$$

We formulate the statistics of SOC avalanches in terms of size distributions (or occurrence frequency distributions) that obey the scale-free probability distribution function (Aschwanden 2014, 2015, 2022), expressed with the power law function

$$N(L) \ dL \propto L^{-d} dL \ , \tag{8}$$

ASCHWANDEN & SCHRIJVER

where d = 1, 2, 3 represent the Euclidean dimensions of the fractal domains and L is the length scale of a SOC avalanche. From this scale-free relationship, the power-law slopes α_x of other SOC parameters x = [A, V, F, E, T] can be derived, such as for the area A, the volume V, the flux F, the fluence or energy E, and the duration T. The resulting power-law slopes α_x can then be obtained mathematically by the method of variable substitution x(L), by inserting the inverse function L(x) and its derivative |dL/dx|,

$$N(x)dx = N[L(x)] \left| \frac{dL}{dx} \right| dL = x^{-\alpha_x} dx , \qquad (9)$$

such as for the flux x = F,

$$\alpha_F = 1 + \frac{(d-1)}{D_V \gamma} = \frac{9}{5} = 1.80 , \qquad (10)$$

or for the fluence x = E,

$$\alpha_E = 1 + \frac{(d-1)}{d\gamma} = \frac{5}{3} = 1.67 .$$
(11)

 γ is the nonlinearity coefficient in the flux-volume relationship,

$$F \propto V^{\gamma} = \left(L^{D_V}\right)^{\gamma} \,, \tag{12}$$

which degenerates to proportionality $F \propto V$ for $\gamma = 1$. Note that the definitions of the peak flux F and fluence E differ only by the fractal dimension D_V (Eq. 10) and the (space-filling) Euclidean dimension d (Eq. 11). In astrophysical high-temperature plasmas, the volume V is approximately proportional to the number of electrons in a region of instability, while the flux F is proportional to the number of emitting photons, which implies that the photon-toelectron ratio is approximately constant and justifies the assumption of the flux-volume proportionality, in the case of incoherent emission mechanisms.

While this brief derivation of Eqs. (5-11) expresses the main assumptions of fractality and linear flux-volume relationship for $\gamma = 1$, an additional assumption needs to be brought in that takes the spatio-temporal evolution into account, which can be approximated with the assumption of (classical) diffusive transport,

$$L \propto T^{\beta/2} = T^{1/2}$$
, (13)

with the transport coefficient $\beta = 1$. We call this model the standard fractal-diffusive self-organized criticality (FD-SOC) model, defined by $[d = 3, \gamma = 1, \beta = 1]$, while the generalized FD-SOC model allows for variable coefficients $[d, \gamma, \beta]$ and alternative dimensionalities (d = 1, 2).

4. DISCUSSION

Synthesized power law distributions have been constructed earlier, such as from nanoflares to large solar flares (Aschwanden et al. 2000), from thermal to nonthermal energies (Li et al. 2012), for magnetic energies from bright points to large active regions (Parnell et al. 2009), from solar and stellar data (Shibayama et al. 2013), from bolometric energies based on solar, stellar, lunar, and terrestrial records (Schrijver et al. 2011, 2012). What is new here is the application of the FD-SOC model into the solar-stellar flare statistics, which predicts flare energies over an unprecedented scaling range of 13 orders of magnitude.

4.1. Synthesized Solar-Stellar Size Distribution

We carry out a synthesized size distribution (Fig. 1) that can be characterized with a power law function,

$$N(E)dE = N_0[t,T] \left(\frac{E}{E_0}\right)^{-\alpha_E} dE , \qquad (14)$$

where E is the total flare energy (in units of erg), α_E is the power law slope, E_0 is an arbitrary reference value, or a threshold value, $N_0(t,T)$ is a normalization constant that depends on the start time t and duration T of the observations in the data sampling.

The construction of such a synthesized size distribution is shown in Fig.(1), stringed together from 5 different data sets with different flare energy ranges. A total of 15 data sets are listed in Table 1, characterized with their energy

ranges $[E_{min}, E_{max}]$ and power law slopes α_E , from which only 5 data sets are represented in the synthesized size distribution, while the other 10 cases have some inconsistencies as described in the observation Section 2.1.

What Fig.(1) shows is that each of the 5 displayed data sets exhibit a power law distribution that is consistent with the theoretical FD-SOC model with a slope of $\alpha_E = (5/3) \approx 1.67$. All 5 cases show also a turnover at the lower end of the size distributions, which is well-understood in terms of incomplete sampling below the detection threshold. The term "incomplete" refers to the scale-free probability property, which implies a reciprocality between the event frequency N(E) and the SOC parameter E (Eq. 8).

The graphical representation in Fig. (1) serves mostly to illustrate the invariance of power law slopes α_E in each data set, while the flaring frequency (N(E), Eq. 14) is subject to different start times t, observational duration T, and differences in the intrinsic energy ratios $q = E_1/E_2$. For instance, energy closure of flare energies demonstrated ratios of $q = E_{elec}/E_{magn} = 0.51 \pm 0.17$ for electron energies between nonthermal electron energies and dissipated magnetic energies, $q = E_{ions}/E_{magn} = 0.17 \pm 0.17$ for ion energies, $q = E_{CME}/E_{magn} = 0.07 \pm 0.14$ for coronal mass ejection (CME) energies, and $q = E_{heat}/E_{magn} = 0.07 \pm 0.17$ for direct heating processes, which added together fulfill energy closure, $q = E_{sum}/E_{magn} = 0.87 \pm 0.18$ for the sum of all energies. This suggests that a realistic evaluation of flare energies requires the summing of partial flare energies, such as thermal energies, non-thermal energies, and bolometric radiation energies (Aschwanden et al. 2017),

$$E_{tot} = E_{nt} + E_{th} + E_{bol} . aga{15}$$

For solar flares, the total radiative energy $E_{\rm bol}$ in visible wavelengths amount to about 70% of the total flare energy, characterized by a blackbody temperature of $T_{\rm bol} \approx 9000$ K (Woods et al. 2004). Unfortunately, the complete energy budget and flare partition is not known for most flare energetics studies. Once all energy parameters are measured, a complete size distribution $N(E_{tot})$ may be inferred that has the synthesized distribution as shown in Fig. (1).

4.2. Flux and Fluence Limits

The flux is defined in physical units of [energy/time], usually measured at the (background-subtracted) peak flux F at the peak time t_p of an event,

$$F = \max[f(t = t_p)] - f_{BG}$$
, (16)

bound by the time range $t_1 \leq t_p \leq t_2$, where f(t) is the time profile of the flux.

The fluences have the physical unit of [energy], and are measured by the time integral of the time profile f(t). Thus, the fluences (E) are the time-integrated fluxes,

$$E = \int_{t_1}^{t_2} [f(t) - f_{\rm BG}] dt \approx F \times T , \qquad (17)$$

for a time interval (t_1, t_2) with well-defined start times t_1 and end times t_2). Combining the fluence approximation $E \approx F \times T$ (Eq. 17) with the flux-volume relationship $F \propto V$ (Eq. 11), and the fractal dimension relatonship $V \propto L^{D_V}$ (Eq. 6), we obtain the following scaling law for the length scale L as a function of the flare energy E, with $D_V = 5/2$,

$$L(E) = E^{1/(2+D_V)} = E^{2/9} \approx E^{0.22} .$$
(18)

which yields the following limit for an energy range of 13 orders of magnitude,

$$\left(\frac{L_{max}}{L_{min}}\right) = \left(\frac{E_{max}}{E_{min}}\right)^{(2/9)} = \left(10^{13}\right)^{(2/9)} \approx 774 .$$
(19)

The predicted length scales for each order of magnitude from 1 to 13 is enumerated in Table 2. Using the scale of solar nanoflares as an estimation of the smallest SOC event, i.e., $L_{min} \approx 1$ Mm (Aschwanden et al. 2000; Cirtain et al. 2013), we would expect a maximum scale of $L_{max} = L_{min} \times 774$ Mm ≈ 1.04 R_{\odot} for stellar flares. Leaving out the stellar flares we have an energy range of 9 orders of magnitude, which yields a range of $L_{max} = L_{min} \times 100$ Mm ≈ 0.14 R_{\odot} , which limits the maximum flare size to about the largest active region size, which represents a reasonable scaling, given the observed sizes of the largest sunspots and starspots. Schrijver (2011) estimates the size of the largest stellar flares to $L_{max} \approx (0.7 - 1.6)R_{\odot}$. Since flare size distributions allow us to extrapolate and predict the probability of the most extreme events, the magnitude limit of these events depends also on the detection of an exponential drop-off of their size distribution. For solar flares it appears that this finite-size effect occurs at $L_{max} \lesssim 200$ Mm (Aschwanden & Alexander 2001).

ASCHWANDEN & SCHRIJVER

4.3. Constraints on Physical Processes

If a power law size distribution function extends over multiple size distributions, many authors conclude that a single physical process is responsible for the resulting size distributions in multiple data sets. The finding of power law size distributions is then attributed to a single universal physical process. For instance, finding a power law distribution of solar magnetic fields over more than five decades in flux, Parnell et al.(2009) suggest that all surface magnetic features are generated by the same physical mechanism, or that they are dominated by surface processes (such as fragmentation, coalescence, and cancellation) in a way which leads to a scale-free distribution. Because of the similarity of their size distributions, both solar and stellar flares are thought to be produced by a magnetic reconnection process (Pettersen 1989).

From our identification of physical processes in our samples we know that at least 3 different physical processes are at play, namely (i) thermal free-emission, (ii) nonthermal thick-target bremsstrahlung, and (iii) white-light bolometric emission. In addition, conductive and radiative losses are thought to contribute significantly to the dissipated flare energy budget (Shimizu et al. 1995). So, we have evidence for at least 5 physical processes, although the syntheseized size distribution function fits a single power law distribution over 13 order of magnitudes. Thus, the invariant power law range is a necessary but not a sufficient condition for an universal physical process.

How can we remedy this dilemma of energy partition. Theoretically, we can measure the size distributions of energies separately for each energy partition and we can define the avalanche energy in terms of the total (summed) energy dissipated during a flare. Care needs to be exercised to avoid double counting of energies. For instance an electron that gets accelerated in a flare plasma, may also show up in the distribution of electrons that produce free-free bremsstrahlung. Nevertheless, energy closure in solar and stellar flares is feasible (Aschwanden et al. 2017).

5. CONCLUSIONS

In this study we arrived at the following results and conclusions:

- 1. We construct a synthesized power law distribution function over an energy range of $[E_{min}, E_{max}] = [10^{24}, 10^{37}]$ [erg] that is consistent with the theoretically predicted power law slope of $\alpha_E = (5/3) \approx 1.67$ of the fractaldiffusive self-organized criticality (FD-SOC) model (Fig. 1). The selected data sets (Table 1) encompass nanoflares, microflares, transient brightenings, large solar flares and stellar superflares. A subset of 5 selected data sets (out of 15) is consistent with the 3-D FD-SOC model, which are used in the synthesized size distribution. Most of the remaining 5 data sets use a 2-D model for the thermal flare energy that we consider to be unrealistic, given the numerous stereoscopic observations with the two STEREO spacecraft. This is a case where the Occham's razor philosophy (parsimony law) does not lead to a better model choice.
- 2. The observed 5 size distributions all exhibit a power law tail (AKAS inertial range) at the upper energy side, while the size distribution show a roll-over at the lower energy side, which is readily understood as incomplete sampling in a scale-free size distribution. This roll-over is caused by a threshold value in the detection of SOC avalanches.
- 3. The observed length scales L agree with the theoretical prediction of the FD-SOC model, i.e., $L \propto E^{2/9} \approx E^{0.22}$. This scaling law has a weak energy dependence that produces a small range of 2 orders of magnitude for solar flares length scales L, i.e., 1 Mm $\leq L \leq 100$ Mm for solar flares, and a range of 3 order of magnitudes for stellar flares, i.e., 1 Mm $\leq L \leq 774$ Mm = 1.04 R_{\odot} , while the flare energy extends over 13 order of magnitudes, i.e., 10^{24} [erg] $\leq E \leq 10^{33}$ [erg]. Thus the FD-SOC model allows us a prediction of the SOC avalanche length scale L for any given flare energy E.
- 4. In the overall we conclude that the universality of power laws (and their slopes) is a consequence of SOC properties (fractality, classical diffusion, scale-freeness, volume-flux proportionality), rather than identical physical processes at different wavelengths.

Future studies may focus on the energy partition of solar flare events, so that the total energy can be estimated, since thermal, nonthermal, and white-light bolometric energies appear to represent only lower limits of SOC avalanches only.

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Object, Instrument	Energy range	Power law	References	
	(erg)	slope		
	$[E_{min}, E_{max}]$	α_E		
Thermal energy E_{Th} :				
Solar flares, SXT/Yohkoh	$[10^{25} - 10^{29}]$	1.5 - 1.6	Shimizu (1995), Fig.11	
Solar flares, EIT/SOHO	$[10^{24.9} - 10^{26.3}]$	2.3 - 2.6	Krucker & Benz (1998), Fig.2	
Solar EUV nanoflares TRACE	$[10^{24} - 10^{26}]$	1.79	Aschwanden et al.(2000), Fig.10	
Solar flares, TRACE	$[10^{23} - 10^{26}]$	2.4 - 2.6	Parnell & Jupp (2000), Figs.1,4,5	
Solar EUV nanoflares, EIT/SOHO	$[10^{25} - 10^{27}]$	2.31	Benz & Krucker (2002), Figs.4,5,8	
Solar flares, GOES	$[10^{24} - 10^{32}]$	2.38	Li et al. (2012), Fig.7	
Non-thermal energy E_{NT} :				
Solar flares, HXRBS/SMM	$[10^{25} - 10^{32}]$	$1.53{\pm}0.02$	Crosby et al. (1993) , Fig.6	
Solar flares, ISEE-3/ICE	$[10^{27} - 10^{33}]$	$1.47{\pm}0.04$	Lu et al.(1993), Fig.8	
Solar flares, ISEE-3/ICE	$[10^{29} - 10^{30}]$	1.67 - 1.74	Bromund et al. (1995) , Table 1	
Solar microflares, RHESSI	$[10^{24} - 10^{30}]$	2.0	Hannah et al. (2008) , Fig. 18)	
White-light radiative energy E_{wl} :				
Stellar flares KEPLER KIC	$[10^{33} - 10^{36}]$	1.56, 1.98	Davenport (2016), Fig.6	
Stellar flares KEPLER G-type stars	$[10^{32} - 10^{37}]$	$1.96{\pm}0.04$	Yang & Liu (2019), Fig.3	
Stellar flares GOES class X,M,C	$[10^{28} - 10^{32}]$		Cai et al.(2024), Fig.9	
Stellar flares L dwarfs	$[10^{33} - 10^{34}]$		Paudel et al. (2024) , Fig.4 3	
Stellar flares solar-type	$[10^{29} - 10^{36}]$		Name kate et al. $(2017a)$, Figs. $6,8,12$	
Synthesized energy distributions:				
EUV + SXR + HXR	$[10^{24} - 10^{32}]$	1.80	Aschwanden et al.(2000), Fig.10	
EUV + SXR	$[10^{24} - 10^{30}]$	$1.54{\pm}0.03$	Aschwanden & Parnell (2002), Fig.10	
TB + QS + AR	$[10^{24} - 10^{30}]$	$1.54{\pm}0.05$	Aschwanden (2004) , Fig. 9.27	
QS + AR	$[10^{17} - 10^{23} \text{ Mx}]$	$1.85{\pm}0.14$	Parnell et al. (2009), Fig.5	
EUV + SXR + HXR + SF	$[10^{24} - 10^{35}]$	$1.87{\pm}0.10$	Schrijver et al. (2011), Fig.3	
EUV + SXR + HXR + SF	$[10^{24} - 10^{37}]$	2.30	Schrijver et al. (2012), Fig.3	
EUV + SXR + HXR	$[10^{24} - 10^{32}]$	2.0 - 2.3	Li et al. (2012), Fig.7	
EUV + SXR + HXR + SF	$[10^{24} - 10^{36}]$	1.80	Shibayama et al. (2013), Fig. 13.12	

Table 1. The energy ranges $[E_{min}, E_{max}]$ and power law slopes α_E of solar and stellar flare energies.

 $\label{eq:extreme ultraviiolet, SXR=soft X-rays, HXR=hard X-rays, TB=Transient brightenings, QS=Quiet Sun, AR=Active regions, SF=Stellar flares.$

Relative	Absolute	Relative	Predicted	Nomenclature
energy	energy	energy	length scale	
logarithm	logarithm	factor	L(E) [Mm]	
0	24	1	1.00 Mm	Nanoflares
1	25	10	$1.67~\mathrm{Mm}$	
2	26	100	$2.78 \ \mathrm{Mm}$	
3	27	1,000	$4.64 \ \mathrm{Mm}$	Microflares
4	28	10,000	$7.74~\mathrm{Mm}$	
5	29	100,000	$12.9~\mathrm{Mm}$	
6	30	1,000,000	$21.5~\mathrm{Mm}$	Large solar flares
7	31	10,000,000	$35.9~\mathrm{Mm}$	
8	32	100,000,000	$59.9~\mathrm{Mm}$	
9	33	1,000,000,000	$100 {\rm Mm}$	Stellar flares
10	34	10,000,000,000	$167 \ \mathrm{Mm}$	
11	35	100,000,000,000	$278~{\rm Mm}$	
12	36	1,000,000,000,000	$464~{\rm Mm}$	
13	37	10,000,000,000,000	$774~{\rm Mm}$	Stellar superflares

Table 2. Predicted scaling law of length scales L(E) as a function of the solar and stellar flare energies E.



Figure 1. Energy size distributions synthesized from solar observations with HXRBS/SMM, ISEE-3/ICE, SXT/Yohkon, TRACE, and from stellar observations with KEPLER. The theoretical prediction of the power law slope for the flare energy distribution is $\alpha_E = 1.67$ in the FD-SOC model. The flare frequencies are aligned to the predicted power law size distribution.