

Self-Organized Criticality and Sand Pile Model in Complex Physical Systems

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Abstract: *This study explores the concept of Self-Organized Criticality (SOC) and its application in complex physical systems, with a particular focus on the Sand Pile Model as a key example. SOC describes how systems naturally evolve into a critical state where small, local disturbances can trigger large, unpredictable events, exhibiting power-law distributions. By analyzing the dynamics of the Sand Pile Model, the study investigates how local interactions between elements drive global critical behavior, resulting in cascading avalanches and emergent patterns. The research further examines how these principles can be applied to real-world systems, including earthquakes, neural networks, and financial markets, where similar behaviors are observed. The study uses a qualitative methodology, synthesizing existing literature and case studies to provide insights into the role of local interactions in achieving self-organized critical states and the potential of SOC to explain complex phenomena across different disciplines.*

Keywords: Self-Organized Criticality (SOC), Sand Pile Model, Local Interactions, Global Critical Behavior, Power-Law Distribution, Complex Systems, Emergent Patterns

I. INTRODUCTION

Self-Organized Criticality (SOC) is a concept used to describe the behavior of complex systems that naturally evolve into a critical state through internal processes without requiring precise tuning from external factors. These systems, when in a critical state, exhibit characteristic behaviors such as scale invariance, where the statistical properties of the system remain unchanged across different scales of observation. One of the defining features of SOC is the occurrence of large, often unpredictable events, which are referred to as "avalanches" or "cascades." These events are typically power-law distributed, meaning that while small events occur frequently, large events are rare but impactful. SOC has been used to model various natural and artificial phenomena, including earthquakes, forest fires, and even traffic flow (**Bak, Tang, & Wiesenfeld, 1987; Bak, 1996**).

The Sand Pile Model, developed by **Per Bak, Chao Tang, and Kurt Wiesenfeld in 1987**, is one of the most widely recognized models used to illustrate SOC. In the Sand Pile Model, grains of sand are dropped one at a time onto a pile, and when the pile reaches a critical slope, the addition of a single grain of sand can cause an avalanche. The size of these avalanches can vary, but their frequency follows a power-law distribution, a hallmark of systems in a critical state. This model is simple yet powerful, as it demonstrates how local interactions between grains of sand can lead to large-scale events, even without the need for an external force or fine-tuning. The self-organizing nature of the sandpile reflects the key principle of SOC, where systems spontaneously evolve to a critical state (**Bak et al., 1987; Jensen, 1998**).

The Sand Pile Model provides important insights into the dynamics of complex systems, highlighting the importance of local interactions in determining global behavior. The model suggests that systems can experience large-scale events or transitions, such as avalanches or phase transitions, which occur spontaneously as a result of small disturbances. This property of self-organization challenges the traditional view that external forces or fine-tuning are required to drive critical phenomena. As a result, SOC has been applied to a variety of domains beyond physical systems, including biological processes like neural networks, and social systems like the spread of information or disease (**Paczuski,**

Vespignani, & Bak, 1996). The universality of SOC provides a framework for understanding how complex behaviors can emerge from simple rules governing local interactions, making it a powerful tool in the study of complex systems (**Jensen, 1998; Bak, 1996**).

II. THEORETICAL FRAMEWORK OF STUDY

The theoretical framework of this study is grounded in the concept of Self-Organized Criticality (SOC), which posits that complex systems naturally evolve into a critical state where small perturbations can lead to large, often unpredictable events, such as avalanches. The Sand Pile Model, developed by **Bak, Tang, and Wiesenfeld (1987)**, serves as a primary example of SOC, demonstrating how simple local interactions can give rise to scale-invariant, power-law distributed events. This framework is further supported by research into critical phenomena, where systems display emergent behaviors like phase transitions and the presence of a critical point (Jensen, 1998). The study extends SOC to various complex systems, including ecological networks, economic systems, and neural networks, building on the work of **Paczuski, Vespignani, and Bak (1996)**, which explores the universality of SOC in different domains. By analyzing the dynamics of SOC, this study aims to deepen the understanding of how large-scale behaviors emerge from local interactions and offer insights into the prediction of critical transitions in real-world systems (**Bak, 1996**).

III. REVIEW OF RELATED LITERATURE

Kinouchi, O., & Copelli, M. (2020). Mechanisms of self-organized quasicriticality in neuronal network models. *Frontiers in Physics*, 8, 5227. This review examines the application of self-organized criticality to neuroscience, specifically focusing on how neuronal networks achieve a "quasicritical" state. The authors discuss the biological mechanisms, such as short-term synaptic plasticity, that act as the internal tuning parameters required for the system to settle at the edge of a phase transition. By utilizing models inspired by the Bak-Tang-Wiesenfeld sandpile, the study demonstrates how the brain maintains a balance between excitation and inhibition, allowing for optimal information processing and storage. The findings suggest that SOC is not just a theoretical curiosity but a vital functional state for biological computational systems.

Pradhan, P. (2021). Time-dependent properties of sandpiles. This research focuses on the temporal dynamics of the Abelian sandpile model and its variants. While much of the historical literature focuses on the steady-state spatial distributions, Pradhan explores how the "waiting time" between avalanches and the duration of those avalanches evolve over time. The paper provides a rigorous characterization of the non-linear dynamics that prevent the system from reaching a simple equilibrium. By analyzing the 2-point time transport correlations, the study offers a new mathematical framework for predicting the integrated transport of energy through a complex system, which has direct implications for understanding the predictability of intermittent extreme events in physical systems.

Kaki, B., Farhang, N., & Safari, H. (2022). Evidence of self-organized criticality in time series by the horizontal visibility graph approach. This study utilizes the Horizontal Visibility Graph (HVG) method to detect SOC characteristics in diverse real-world data sets, including financial markets and solar nano-flare emissions. By converting time series data into complex networks, the researchers identify power-law distributions in the network's degree distribution, a hallmark of critical behavior. The results indicate that both the "avalanches" in stock market fluctuations and the energy releases in solar flares follow the same universal scaling laws seen in numerical sandpile models. This reinforces the idea that SOC is a universal governing principle for out-of-equilibrium systems across vastly different scales of magnitude.

ArXiv. (2025). Describing self-organized criticality as a continuous phase transition. This paper addresses a long-standing debate in statistical mechanics: whether SOC can be formally categorized within the framework of continuous phase transitions. By introducing a new control variable termed "drop density," the researchers provide numerical evidence that the Bak-Tang-Wiesenfeld and Manna sandpile models exhibit percolation-like transitions. The study identifies the scaled size of the largest avalanche as the "order parameter" of the system and demonstrates a clear divergence of the correlation length as the critical point is approached. This work provides a more robust mathematical grounding for SOC, linking it more closely to traditional thermodynamics and phase transition theory.

Aschwanden, M. J. (2025). New trends in astrophysical self-organized criticality. This comprehensive review covers the most recent decade of astrophysical research (2015–2025) involving SOC. It highlights how the sandpile paradigm has been successfully applied to model solar flares, magnetospheric substorms, and even lunar cratering. Aschwanden discusses the transition from 2D to 3D geometry in these models, which has helped reconcile differences in power-law slopes observed in various solar phenomena. The review concludes that while "everything that avalanches" is not necessarily SOC, the sandpile remains the most effective heuristic for understanding the scale-invariant energy releases observed in the heliosphere and beyond.

IV. RESEARCH GAP OF THE STUDY

While significant progress has been made in understanding Self-Organized Criticality (SOC) in complex systems, there remains a research gap in fully bridging the theoretical models with real-world applications across diverse fields. While studies demonstrate SOC's relevance in neuroscience and financial systems, the specific mechanisms through which local interactions give rise to global critical behavior in these varied contexts remain under-explored. Additionally, the temporal dynamics and long-term predictability of SOC systems, are still not fully understood, particularly in systems that do not reach equilibrium. More research is needed to better predict intermittent extreme events and improve our understanding of SOC as a universal principle across scales and disciplines.

V. STATEMENT OF THE PROBLEM

Self-Organized Criticality (SOC) is a central concept in understanding how complex physical systems evolve into a critical state where small, local disturbances can trigger large, often unpredictable events. Despite its broad applicability, the mechanisms driving SOC, particularly how local interactions within a system lead to global critical behavior, remain poorly understood. The Sand Pile Model, a well-known example of SOC, illustrates how local events, such as the addition of a single grain of sand, can lead to cascading avalanches and large-scale phenomena. However, the dynamics of these systems, especially in real-world scenarios, are not fully characterized. Specifically, the role of local interactions in driving global behavior, and how this translates to systems with non-linear components or heterogeneous structures, remains an area of uncertainty. This study aims to address the gap in understanding the dynamics of SOC, using the Sand Pile Model as a key example, and to explore how local interactions influence global critical behavior across various complex physical systems.

VI. OBJECTIVES OF THE STUDY

The study aims to explore the concept of Self-Organized Criticality (SOC) and its fundamental principles in complex systems, with a particular focus on understanding how these systems naturally evolve into a critical state where small perturbations can lead to large-scale, unpredictable events. A key example of SOC is the Sand Pile Model, which will be analyzed to examine its dynamics and behavior, highlighting how local interactions between individual elements can trigger significant global changes. By investigating these interactions, the study seeks to deepen our understanding of how localized activities in SOC systems contribute to the emergence of critical behavior, offering insights into the underlying mechanisms that govern complex systems in a state of self-organized criticality.

VII. METHODOLOGY OF STUDY

This study adopted a qualitative approach to investigate the concept of Self-Organized Criticality (SOC) and its application in complex physical systems, with a particular focus on the Sand Pile Model as a key example. The research primarily involved an in-depth exploration of the theoretical frameworks and the dynamics of SOC through extensive literature review and conceptual analysis. Through qualitative analysis, the study focused on understanding how local interactions within a system, such as those found in the Sand Pile Model, drive global critical behavior. This involved reviewing examples where SOC has been applied to real-world systems. By examining the qualitative patterns of system behavior, such as the emergence of power-law distributions and the occurrence of avalanches, the research will identify common features and dynamics that define criticality in diverse contexts. This qualitative methodology

allowed for the development of conceptual frameworks that integrated SOC theory with practical examples from both theoretical models and real-world applications.

VIII. ANALYSIS AND INTERPRETATION

I. Concept of Self-Organized Criticality (SOC) in Complex Systems

Self-Organized Criticality (SOC) is a theoretical framework used to describe how complex systems naturally evolve into a critical state, where small perturbations can trigger large-scale, often unpredictable events. These systems reach a critical point through internal processes and local interactions among their components, without requiring any external tuning. The critical state is marked by scale-invariance, meaning that the system's behavior remains statistically similar across different scales, both spatial and temporal. SOC provides a unified explanation for phenomena where systems exhibit sudden transitions or catastrophic events, such as avalanches, earthquakes, and forest fires. The emergence of these events is an inherent property of the system, governed by simple local dynamics that lead to large-scale consequences (Bak, Tang, & Wiesenfeld, 1987).



Figure 1: Sand Pile Model and Complexity

Key Principles of Self-Organized Criticality

One of the core principles of SOC is that these systems spontaneously self-organize into a critical state, without requiring external input or control. This self-organization results in a system that is on the edge of a phase transition, where small disturbances can lead to significant, often unpredictable outcomes. In SOC systems, the events that occur are distributed in such a way that smaller events happen frequently, while larger events are rare but have much more significant impacts. This distribution follows a power-law, which is a hallmark of critical systems. A notable feature of SOC is that it lacks characteristic timescales—events occur at irregular intervals, making it difficult to predict when large events will occur (Jensen, 1998).

The Sand Pile Model

The Sand Pile Model, introduced by Bak, Tang, and Wiesenfeld (1987), is one of the most famous and widely studied examples of SOC. In this model, grains of sand are added to a pile one at a time, and when the pile reaches a critical angle, the addition of a single grain can cause an avalanche. The size of the avalanches varies, but they follow a power-law distribution, demonstrating the system's tendency toward criticality. The model illustrates how local interactions—such as the addition of a grain—can lead to large, global consequences, with the system naturally evolving to a state of criticality where large events (avalanches) occur sporadically and unpredictably. This self-organizing behavior of the sand pile demonstrates the fundamental concept of SOC in a simple, physically intuitive way (Bak et al., 1987).

Applications of SOC

SOC has been applied to a wide range of physical systems to explain complex behaviors. In geophysics, the concept of SOC is used to understand earthquakes. Earthquakes occur when stress builds up in the Earth's crust, and when the stress exceeds a certain threshold, it is released suddenly in the form of an earthquake. The frequency and magnitude of earthquakes follow a power-law distribution, similar to the behavior of avalanches in the Sand Pile Model. Similarly, fluid dynamics exhibits SOC, where turbulent flows in fluids can result in sudden, large-scale transitions from laminar flow to chaotic behavior. These transitions occur spontaneously, driven by the internal dynamics of the system rather than external forces (Paczuski, Vespignani, & Bak, 1996).

Power-Law Distributions and Scale-Invariance

A defining feature of SOC is the presence of power-law distributions in the size and frequency of events. In systems exhibiting SOC, small events occur frequently, while large events, although rare, have a disproportionately large impact. This statistical distribution is indicative of critical behavior, where the system operates in a state on the edge of instability. The power-law behavior is observed in various physical systems, such as seismic activity, forest fires, and even traffic flow, where the size and frequency of events follow similar statistical patterns. Power-law distributions suggest that these systems share a common underlying property—criticality—that emerges from the local interactions of their components (Jensen, 1998).

A single mathematical expression that captures the essence of power-law distributions and scale-invariance in Self-Organized Criticality is:

$$P(s) \propto s^{-\tau}$$

Under this assumption $P(s)$ denotes the probability of an event of size s , and τ is the scaling (critical) exponent characteristic of the system. This relation indicates that events of all sizes are possible, with small events occurring frequently and large events occurring rarely but without a characteristic scale. The absence of a dominant length or time scale reflects scale-invariance, a key signature of criticality in SOC systems, and explains why diverse phenomena such as sandpile avalanches, earthquakes, and forest fires exhibit similar statistical behavior (Jensen, 1998).

Self-Organized Criticality provides a framework for understanding complex systems that evolve toward a critical state through internal dynamics. The concept of SOC challenges traditional models of critical phenomena that rely on external tuning and fine-tuning. By exploring examples such as the Sand Pile Model, earthquakes, and fluid dynamics, SOC offers valuable insights into the dynamics of large-scale events in physical systems. The concept's emphasis on local interactions leading to global behavior has made it a powerful tool for explaining complex phenomena across multiple disciplines, from physics to biology and social sciences.

II. Dynamics and Behavior of the Sand Pile Model as a Key Example of SOC

The Sand Pile Model, introduced by Bak, Tang, and Wiesenfeld (1987), serves as a foundational example of Self-Organized Criticality (SOC) in physics. The model demonstrates how local interactions within a system can lead to global, large-scale phenomena that exhibit power-law distributions. By analyzing the dynamics of the Sand Pile Model, we gain insight into the self-organizing nature of critical systems and how small perturbations can result in significant, unpredictable events. The model is a simple yet powerful tool for understanding the emergence of critical behavior in complex systems.

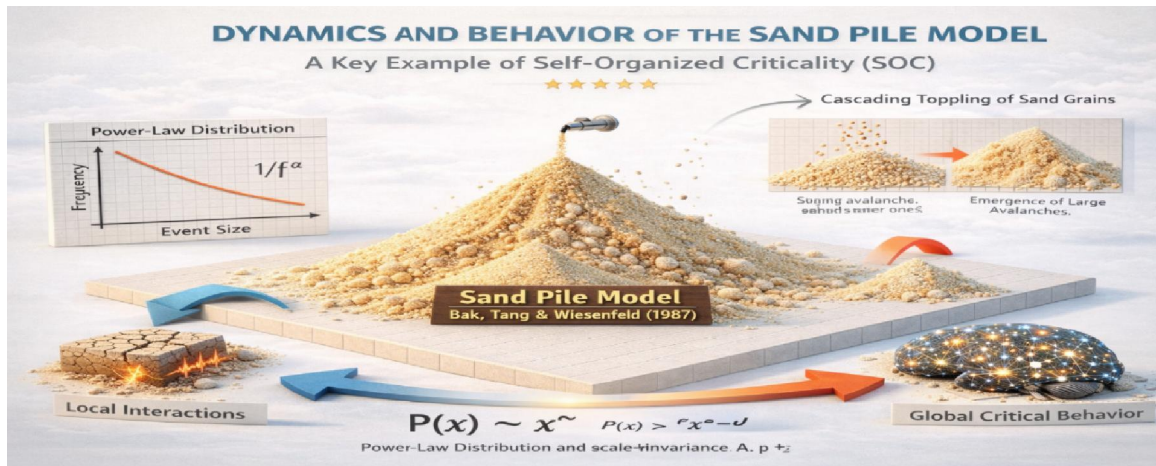


Figure 2: Dynamics of the Sand Pile Model

Model Setup and Basic Dynamics

In the Sand Pile Model, grains of sand are added one at a time to a pile. The pile is made up of discrete elements that are arranged in a grid, with each element able to hold a limited number of grains before becoming unstable. When the number of grains at a particular site exceeds a critical threshold, the pile "topples," sending grains to neighbouring sites. This process is repeated as the sand continues to accumulate, causing further toppling in nearby regions. The toppling of grains continues until the system reaches a stable configuration where no site exceeds the threshold (Bak *et al.*, 1987).

A single formula that succinctly represents the basic dynamics and critical behavior of the Sand Pile Model is the toppling (instability) condition:

$$z_i \geq z_c \Rightarrow z_i \rightarrow z_i - 4, z_j \rightarrow z_j + 1$$

Based on above assumption, z_i denotes the number of grains at site i , z_c is the critical threshold (typically $z_c = 4$ in a two-dimensional grid), and z_j represents the nearest neighbouring sites. This rule governs the redistribution of grains during a toppling event and forms the microscopic mechanism through which local instabilities propagate into avalanches, ultimately driving the system toward a self-organized critical state (Bak *et al.*, 1987).

The key feature of the Sand Pile Model is its inherent ability to reach a critical state. As grains of sand are added, the system eventually becomes "tuned" to a critical point where small disturbances can trigger large-scale avalanches. These avalanches, which are the result of cascading toppling, vary in size and frequency. Importantly, the distribution of avalanche sizes follows a power law, with small avalanches occurring frequently and large avalanches occurring rarely but with greater impact. This power-law behavior is one of the defining characteristics of SOC systems, indicating that the system is operating in a critical state on the verge of instability (Jensen, 1998).

Emergence of Critical Behavior

The self-organizing nature of the Sand Pile Model means that the system evolves toward a critical state without any external control. The model does not require any tuning of parameters to reach criticality; instead, it emerges from the interactions between the components of the system. As sand grains are added, the system undergoes a series of local interactions where sites that exceed the threshold topple and redistribute grains. This redistribution process is local, but it can trigger a chain reaction of topplings throughout the pile, leading to large-scale avalanches. The key insight from this model is that critical behavior emerges naturally through the accumulation of small disturbances, rather than being imposed externally (Bak *et al.*, 1987).

The Sand Pile Model exemplifies the principle of SOC by showing that large, catastrophic events (avalanches) can arise from small, localized changes. The occurrence of these avalanches is not periodic but occurs randomly, with no predictable timing. This randomness is a fundamental feature of SOC systems, where the dynamics of the system are governed by the accumulation of small, often imperceptible, changes that eventually lead to large-scale consequences.

The power-law distribution of avalanche sizes further reinforces the idea that the system is in a critical state, where small events occur frequently, but the likelihood of large events decreases, though they are still significant when they occur (Paczuski, Vespignani, & Bak, 1996).

Power-Law Distribution and Scale Invariance

A central aspect of the Sand Pile Model is the emergence of power-law distributions in the size of avalanches. Power-law distributions are a hallmark of systems exhibiting SOC, and they are characterized by the relationship $P(x) \sim x^{-\alpha}$ where $P(x)$ is the probability of an avalanche of size x , and α is a constant exponent. In the case of the Sand Pile Model, this distribution suggests that while small avalanches are common, larger avalanches become increasingly rare. However, when they do occur, they can have a disproportionately large impact on the system.

This power-law behavior is indicative of criticality, where the system is poised at a phase transition. In the context of the Sand Pile Model, the critical state refers to the point at which the system exhibits a balance between stability and instability. The presence of scale-invariance in the system means that the model's behavior remains similar across different scales, whether examining individual grains of sand or the entire pile. This scale-invariance is a defining feature of SOC systems and suggests that the same underlying mechanisms drive the dynamics at all levels of observation (Jensen, 1998).

Implications of SOC

The dynamics of the Sand Pile Model extend to real-world physical systems, where SOC behavior can be observed in various contexts. In geophysics, for example, the occurrence of earthquakes can be modelled as a SOC process. Just as in the Sand Pile Model, stress builds up in the Earth's crust until it reaches a threshold, at which point an earthquake occurs. The frequency and magnitude of earthquakes also follow a power-law distribution, with small tremors being common and large, catastrophic earthquakes being rare but impactful. This pattern is strikingly similar to the behavior observed in the Sand Pile Model, where small avalanches are frequent, and large avalanches are rare but significant (Bak, 1996).

Another example of SOC in physics is in fluid dynamics, where turbulent flows exhibit similar self-organizing behavior. In turbulence, small fluctuations in fluid velocity can grow into large-scale chaotic patterns. These transitions from laminar flow to turbulent flow are not initiated by external factors but emerge spontaneously from the system's dynamics, akin to the avalanches observed in the Sand Pile Model. The study of SOC in these systems helps physicists understand how complex behaviors emerge from relatively simple local interactions, offering a unified framework for studying critical phenomena across different domains (Paczuski *et al.*, 1996).

The Sand Pile Model offers a simple yet profound illustration of Self-Organized Criticality in complex systems. By demonstrating how local interactions between components can lead to large-scale, emergent behaviors, the model provides valuable insights into the nature of critical phenomena. The model's ability to naturally evolve toward a critical state, coupled with the power-law distribution of avalanche sizes, makes it an ideal representation of SOC. Moreover, the Sand Pile Model's applicability to real-world systems like earthquakes and turbulence underscores the relevance of SOC in explaining complex behaviors in nature.

III. The Role of Local Interactions in Driving Global Critical Behavior within SOC Systems

Self-Organized Criticality (SOC) is a powerful concept in physics, where complex systems spontaneously evolve into a critical state through the interplay of local interactions among their components. This local-to-global relationship is crucial in understanding how large-scale phenomena emerge from the simple dynamics of individual elements. The study of SOC systems, such as the Sand Pile Model, provides insights into how local events or disturbances can cascade through the system and lead to global, often unpredictable, outcomes. In this section, researcher examine how local interactions drive global critical behavior in SOC systems, with a focus on the fundamental principles underlying these interactions and their real-world applications.



Figure 3: Local Interactions in Driving Global Critical Behavior within SOC Systems

Local Interactions and Self-Organization

The fundamental idea behind SOC is that complex systems reach a critical state through the accumulation of local interactions, without external tuning. In a typical SOC system, each element or component interacts with its neighbouring elements, and these interactions can trigger events that have a larger-scale impact. For example, in the Sand Pile Model, the addition of a single grain of sand to the pile causes local changes, such as the toppling of individual grains. When the disturbance propagates to neighbouring grains, it may cause additional toppling, resulting in a cascade of events that can affect large portions of the system. These local interactions, despite being simple and individual, collectively drive the system towards a state of criticality where large avalanches can occur, demonstrating how global behavior emerges from local dynamics (**Bak, Tang, & Wiesenfeld, 1987**).

The process of self-organization arises because the system is continually adjusting to the local interactions and disturbances. In other words, SOC systems naturally evolve toward a critical point where they are poised on the edge of instability, a state where even small fluctuations can lead to significant changes. The Sand Pile Model exemplifies this process: as more sand is added, the pile gradually becomes more prone to toppling, and the system self-organizes into a critical state where avalanches of various sizes occur. This behavior is driven entirely by the local interactions between individual grains of sand and their neighbours, with no external control or fine-tuning. As these local interactions accumulate, the system moves toward criticality, exhibiting the global critical behavior of power-law distributions in avalanche sizes (**Jensen, 1998**).

$$P(s) \sim s^{-\tau}$$

Here $P(s)$ is the probability of an avalanche of size s , whereas τ is the critical exponent, which typically falls in the range of $1.5 \leq \tau \leq 21.5$ for many SOC systems. This power-law behavior signifies that smaller avalanches occur frequently, while larger avalanches are rarer but have a significant impact. In addition to the size of avalanches, the duration T of these events also follows a scaling law. The probability $P(T)$ of an avalanche lasting for a time T typically obeys:

$$P(T) \sim T^{-\tau'}$$

In this formulation, τ' is another critical exponent related to the duration of avalanches. The value of τ' can vary but often lies in the range $1.5 \leq \tau' \leq 21.5$

$$S \sim T^z$$

It encodes that z is a dynamical exponent that reflects how the size of an avalanche scales with its duration. This relation indicates that larger events tend to last longer, and this scaling behavior is a characteristic feature of SOC systems.

As the system evolves towards criticality, the rate of activity (such as the number of topplings or events) also exhibits a scaling behavior. The activity rate R of the system near the critical point can be described as:

$$R \sim |\epsilon|^\alpha$$

Based on this assumption ϵ is the deviation from the critical point (a measure of the system's distance from the critical state), α is a critical exponent that determines how the activity rate behaves as the system approaches criticality.

In the specific case of the Sandpile Model (a canonical SOC model), the local dynamics can be described by the "toppling rule," where each grain of sand represents a local site, and toppling occurs when a site exceeds a threshold (usually 4 in a 2D lattice). The evolution of the system can be mathematically modeled by:

The Cascading Effect: From Local to Global Events

$$\frac{dE_i}{dt} = f(E_i, N_i)$$

The above given assumption, states E_i is the energy (or height) of site i , N_i represents the neighbouring sites of i , and $f(E_i, N_i)$ represents the force (interaction) that causes energy to be transferred between neighbouring sites. This formulas describe how local interactions lead to cascading effects that drive the system towards criticality, with global behavior emerging from the interactions of individual components.

The cascading effect is another crucial mechanism by which local interactions drive global critical behavior in SOC systems. In the Sand Pile Model, local disturbances can initiate a chain reaction, where one small event triggers further events in neighbouring regions of the system. This propagation of disturbances is fundamental to understanding how large-scale phenomena emerge. For instance, the toppling of a single grain may destabilize adjacent grains, leading to a larger cascade of toppling. Over time, these small events accumulate and give rise to global avalanches. The size and frequency of these avalanches follow a power-law distribution, which is a hallmark of SOC systems. The self-organizing nature of these systems means that the critical state emerges as a result of the interactions within the system, without the need for any external adjustments (**Bak et al., 1987**).

The cascading effect also explains why large events in SOC systems are unpredictable and irregular. Since the timing and location of disturbances are governed by the local dynamics, it is impossible to predict when a large avalanche will occur. However, the system's behavior remains statistically similar over time, and the power-law distribution of avalanche sizes suggests that the system is poised at a critical threshold. This phenomenon is not limited to sand piles but is also observed in various real-world systems, such as earthquakes and wildfires. In these systems, small local disturbances—such as the accumulation of stress in geological fault lines or the ignition of a single tree in a forest—can trigger larger, system-wide events, highlighting the importance of local interactions in driving global critical behavior (**Paczuski, Vespignani, & Bak, 1996**).

Earthquakes and Turbulence

SOC systems, driven by local interactions, are not confined to theoretical models but are observable in real-world physical phenomena. In geophysics, earthquakes provide a natural example of how local interactions lead to global catastrophic events. Stress builds up locally along fault lines, and when the stress exceeds a threshold, a rupture occurs, releasing energy in the form of an earthquake. The occurrence of these earthquakes follows a power-law distribution, similar to the avalanches in the Sand Pile Model. The local stress interactions between individual fault segments accumulate over time, and when the system reaches a critical state, a large-scale earthquake is triggered. This process demonstrates how small, localized interactions—like the movement of individual tectonic plates—can result in significant, unpredictable global behavior (**Bak, 1996**).

In geophysics, the distribution of earthquake magnitudes follows a power-law, similar to the avalanche size distribution in SOC systems. The relationship between the frequency $N(M)$ of earthquakes with magnitude M is given by the Gutenberg-Richter law:

$$N(M) \sim 10^{-\beta M}$$

Under this assumption $N(M)$ is the number of earthquakes with magnitude M , β is a constant M is the magnitude of the earthquake. This law implies that smaller earthquakes occur much more frequently than larger ones, and the frequency of large earthquakes decreases exponentially with increasing magnitude, exhibiting a typical power-law decay.

The displacement D caused by an earthquake is often related to the amount of stress σ that accumulates along the fault line. The relationship between stress and displacement can be modelled by:

$$D \sim \sigma^\mu$$

Here D is the slip or displacement of the fault, σ is the local stress accumulated along the fault line, and μ is a scaling exponent (typically around 1 for a simple case). This shows how local stress interactions on the fault line, as they accumulate, can lead to large displacements (earthquakes) when the system reaches a critical stress threshold

The size distribution of turbulent structures, such as vortices and eddies, also follows a power-law distribution similar to the size of avalanches in SOC systems. The probability $P(s)$ of a turbulent structure having size s is typically given by:

$$P(s) \sim s^{-\alpha}$$

The assumption expresses $P(s)$ is the probability of a turbulent structure of size s , α is a critical exponent that typically lies between 2 and 3 for many turbulent systems. This relationship shows that large eddies or vortices are much rarer than smaller ones, and the frequency of large turbulent events decreases with increasing size, consistent with the power-law behavior seen in SOC systems.

In the study of turbulence, local interactions between fluid molecules drive the system towards chaotic behavior. When fluid flow reaches a critical threshold, small fluctuations at the microscopic level can trigger large-scale chaotic eddies and vortices. These turbulent flows, once established, exhibit similar power-law distributions in terms of their size and frequency, much like the avalanches in SOC systems. The transition from laminar flow to turbulence is a result of local interactions that cascade into global chaotic behavior, highlighting the role of local dynamics in creating large-scale phenomena (Jensen, 1998).

Local Interactions and Nonlinear Dynamics

Another essential aspect of SOC is the presence of nonlinear dynamics, which are integral to how local interactions lead to global behavior. In nonlinear systems, small changes in local conditions can lead to disproportionate changes in the overall system. This nonlinearity is a characteristic feature of SOC systems, where the relationship between local disturbances and global outcomes is not proportional. In the Sand Pile Model, a single grain of sand can initiate a chain reaction that causes a large avalanche, illustrating the nonlinear nature of the system. Similarly, in biological and ecological systems, small local changes—such as the introduction of a new species or a minor genetic mutation—can have cascading effects on the system's overall structure and behavior. These nonlinear interactions are a key driver of the emergence of criticality in SOC systems (Paczuski *et al.*, 1996).

In conclusion, local interactions play a crucial role in driving global critical behavior in SOC systems. Through the self-organizing dynamics of individual elements and their interactions, complex systems evolve toward a critical state where large, unpredictable events can emerge. The Sand Pile Model provides a clear example of how local events can trigger cascading effects, leading to power-law distributions and global criticality. These principles are observed in a variety of physical systems, from earthquakes to turbulence, where local disturbances cascade into large-scale phenomena. Understanding the role of local interactions in driving global behavior offers valuable insights into the dynamics of complex systems and helps explain the emergence of critical events in nature and society.

IX. CONCLUSION

Self-Organized Criticality (SOC) in complex systems describes how these systems naturally evolve into a critical state where small local disturbances can trigger large-scale, often unpredictable events, exhibiting power-law distributions.

A key example of SOC is the Sand Pile Model, where grains of sand are added to a pile, and when a critical threshold is reached, avalanches occur. The distribution of avalanche sizes follows a power law, showing how local interactions lead to global behavior without external tuning. Local interactions, such as the addition of a grain, can trigger cascading events that spread through the system, driving it towards criticality. These cascading effects highlight the importance of local dynamics in producing global critical behavior, where small perturbations accumulate into large-scale, complex phenomena, characteristic of SOC systems (Bak, Tang, & Wiesenfeld, 1987; Jensen, 1998; Paczuski, Vespignani, & Bak, 1996).

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