ECONOPHYSICS: BRIDGING THE GAP BETWEEN ECONOMICS AND PHYSICS

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ABSTRACT

Econophysics is multidisciplinary field applies theories and methods from physics to analyze economic phenomena (H. Eugene Stanley, 1995). The article explores the subject nature of Econophysics, historical perspective, models of Econophysics, tools and techniques of Econophysics, applications of Econophysics in economics, focusing on its contributions to various phenomena of the Economics such as market dynamics, wealth distribution, financial risk management, complex systems analysis, agent-based modeling, and forecasting. By applying concepts and tools from physics, econophysics to be instrumental in achieving new insights into complex economic systems, enhancing analysis of market behaviour and ability to manage financial risk and contemporary problems. The paper also discusses the major achievements in the area of Econophysics so far and future potential of Econophysics in economics and its role in shaping economic theory and policy.

Keywords: Econophysics, Physics, Economics, Sociophysics, Complex Systems Interdisciplinary.

Introduction

Econophysics may be appropriately defined as applications of mathematical and statistical approaches and models or methods used in discipline of Physics to examine or analysis problems, concepts and subject nature of Economics. Econophysics is multidisciplinary field applies physics's concepts and methods to study economic problems. It aims to understand economic behaviour and market phenomena using mathematical models and statistical methods inspired by physics. Econophysics evaluates the current and future state in this field, identifying its strengths and potential contributions to economics¹. Econophysics is intersection of heterodox economics and physics of complex systems, with experts engaged in two overlapping but having different methodological programs.² The subject nature of Econophysics, itself treats economic agents or variables as particles interacting within a system, like particles in physics and provide a scope for economic problems and concepts to be considered as problems and methods meant for physical science approaches and methods. These approaches of physical science allow econophysicists to model and analyze complex economic systems in novel ways, often uncovering underlying patterns and principles that traditional economic approaches may overlook. It seeks to provide a new perspective on economic systems,

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¹ Sergio, Da, Silva., Raul, Matsushita (2023), Editorial: Taking stock in Econophysics. Frontiers in Physics, doi: 10.3389/fphy.2023.1159893

² Adrian, K., Yee (2021), Econophysics: making sense of a chimera. European journal for philosophy of science, doi: 10.1007/S13194-021-00413-1

treating them as complex, dynamic systems that exhibit behaviours similar to physical systems. Econophysics targets to reduce gap between economics and physics and offer fresh insights into economic systems.

As depicted in The New Palgrave Dictionary of Economics, the phrase "Econophysics" was coined by H. Eugene Stanley (*Dynamics of Complex Systems Conference in 1995*) in Kolkata, India. The term was used to describe the emerging interdisciplinary field where physicists were tackling issues in economics and finance¹. Mantegna and Stanley (2000) elucidated *"the multidisciplinary domain of Econophysics"* as "a neologism indicating the endeavours of physicists grappling with economic quandaries, endeavouring to assess an array of innovative conceptual methodologies derived from the physical sciences"².

Econophysics is a research area that spans mathematics, economics, physics, and computer science. Initially, it relied heavily on theories and methodologies from nuclear physics and statistical physics.

Historical Perspective

Econophysics, as a field, emerged in the late 20th century as physicists began to apply their methods and theories to analyze economic systems. The field was born out of a desire to understand the complex dynamics of financial markets and economic phenomena using the tools of physics.

As Mirowski (1989) noted, neoclassical economists in the late nineteen century borrowed heavily from the theoretical framework of classical physics to establish the foundations of their discipline.³ The influence is evident in the emphasis of economic theory on describing equilibrium situations, as seen in Pareto's (1906) theoretical framework of efficiency equilibrium.⁴ The historical influence of late 19th century physics on neoclassical economics also led to an emphasis on the maximization of individual utilities in economics. This is worthwhile to mention that classical physics of that era was primarily based on principles of energy minimization, such as the Principle of Least Action (Feynman, 1964).⁵ Although systems whose energy function cannot be defined can still be rigorously analyzed using techniques of nonlinear dynamics, academic disciplines often follow certain paths dictated by available investigative techniques, and economics is no exception.

There are instances where investigations into economic phenomena have preceded developments in physics. For example, Bachelier's mathematical theory of random walks, developed in his 1900 thesis on stock price movements analysis, was independently discovered five years later by Einstein to explain Brownian motion.⁶ Bachelier's work, although initially challenged by mathematicians, eventually proved to be correct, demonstrating that stock price returns over very short times follow a distribution with a long tail, known as the "*inverse cubic law*," which converges to a Gaussian distribution at longer timescales (*Mantegna and Stanley, 1999*).

Despite this rich history of idea exchange between the two disciplines, Samuelson (1947) attempted to transform economics into a natural science, basing it on "*operationally meaningful theorems*" subject to empirical verification. However, in the 1950s, economics took a different direction, focusing more on mathematical modelling and axiomatic foundations rather than empirical observations. This shift led to complete separation of theory with reality, culminating in emergence of new subject as econometrics.⁷

The divergence between economics and physics in the 1950s was likely not accidental. By this time, physics had moved far from explaining the observable world, making it difficult to contribute significantly to economics. The quantum mechanics-dominated physics of the time would have seemed alien to anyone interested in explaining economic phenomena. Developments in physics that contributed to the birth of Econophysics, such as nonlinear dynamics or non-equilibrium statistical mechanics, would

¹ Rosser, J.B. (2008). Econophysics. In: Durlauf, S.N., Blume, L.E. (eds) The New Palgrave Dictionary of Economics. Palgrave Macmillan, London. https://doi.org/10.1007/978-1-349-58802-2_443

² Mantegna, Rosario & Stanley, H. (2000), An Introduction to Econophysics: Correlations and Complexity in Finance. 10.1063/1.1341926.

³ Mirowski, P.. (1989).More Heat Than Light: Economics as Social Physics, Physics as Nature's Economics (Cambridge: Cambridge University Press)

⁴ Pareto, V (1906). Manual of Political Economy (trans A S Schwier, 1971) (London: Macmillan)

⁵ Feynman, R P (1964): "The Principle of Least Action", Feynman Lectures in Physics, Vol 2, Chapter 19 (Reading, MA: Addison-Wesley).

⁶ Bachelier, L.. (1900). "Theorie de la speculation", Annales scientifiques de l Ecole Normale Supe-rieure, Ser 3, 17, pp 21-86

⁷ Samuelson, P. A. (1947). Foundations of Economic Analysis, (Cambridge, Massachusetts: Harvard University Press).

Ram Lal Bagaria: Econophysics: Bridging the Gap between Economics and Physics

not emerge until the 1970s and 1980s. The turn towards game theory in the 1950s and 1960s allowed economics to describe human motivations and strategies using mathematical models. However, this movement led to an overemphasis on "individual rationality," promoting a view of economic agents as selfish, paranoid individuals striving to maximize their utilities. This approach fails to account for bounded rationality and the constraints of space, time, and computational resources that govern human decisionmaking.1

Physicists had sporadically delved into economic problems prior to the 1990s. However, a systematic movement only took root in the 1990s, marked by an increasing number of physicists employing their tools to analyze socio-economic phenomena (Farmer et al., 2005).² This shift was partly propelled by the availability of large volumes of high-quality data and the computational power to analyze it using intensive algorithms.

One of the early pioneers of Econophysics was Eugene Stanley who began applying concepts of statistical physics in financial markets in 1980s. Stanley along with his colleagues developed models to describe the movement of stock prices and other financial assets, treating them as stochastic processes akin to those found in physics³. Another influential figure in this area was Per Bak, a physicist who introduced concept of self-organized criticality to explain behaviour of complex systems. Bak's work laid the foundation for applying concepts from complex systems theory to economics. In the 1990s, the field of Econophysics began to gain traction, with researchers from various disciplines contributing to its development. One key area of research was the study of stock price fluctuations, which econophysicists found could be described using models borrowed from physics, such as random walks and fractals⁴.

In the late 1980s, physicist Philip Anderson, alongside Kenneth Arrow, organized a meeting between experts of economics and physics at Santa Fe Institute. This meeting led to early attempts by physicists to apply tools from non-equilibrium statistical mechanics and non-linear dynamics, which had recently been developed, to the economic realm ("The Economy as an Evolving Complex System," 1988). It also sparked the interest of other physicists in this interdisciplinary field. Subsequent developments in the statistical physics group of H. Eugene Stanley at Boston University further solidified Econophysics as a distinct field, with Stanley coining the term in 1995 in Kolkata⁵.

Today, physics departments worldwide have groups dedicated to studying economic problems, all over the world. While the issues they tackle vary widely, including questions fluctuations of stock market, models explaining economic inequality, and factors influencing the popularity of certain products over others, a common thread is the observation and explanation of scaling relations. Scaling relations signify the absence of a characteristic scale for the measured variable, indicating universal behaviour that transcends specific system details. The recent influx of physicists into economics has brought a new perspective, seeking invariant patterns across different contexts. This guest for universal behaviour may be seen as the most enduring legacy of Econophysics (Sinha and Raghavendra, 2004)⁶.

Since then, Econophysics has grown into a thriving field with researchers from physics, economics, mathematics, and computer science collaborating to study a wide range of economic phenomena. Econophysicists have applied their methods to study market dynamics, wealth distribution, financial risk management, and other complex economic systems, leading to new insights and models that have enriched analysis of economics. Thus, historical perspective of Econophysics is a story of interdisciplinary collaboration, with researchers from different fields coming together to apply the principles of physics to economics. Through their work, econophysicists have shed new light on dynamics of economic systems, paving way for new approaches to understanding and managing the complexities of the global economy7.

Interplay between Physics and Economics

Econophysics represents a unique interdisciplinary field that bridges gap between physics and economics. It involves applications of physical concepts, theories, and methodologies to economic

Sitabhra Sinha, Bikas K. Chakrabarti - 2012 - Econophysics An Emerging Discipline, retrieved from Sitabhra Sinha, Bikas K. Chakrabarti - 2012 - Econophysics An Emerging Discipline | PDF | Economic Equilibrium | Economics (scribd.com) Farmer, J D, M. Shubik and D. E. Smith (2005). "Is Economics the Next Physical Science?", Physics Today, 58 (9), pp 37-42 Mantegna, R. N., & Stanley, H. E. (1999); An introduction to Econophysics: Correlations and Complexity in Finance. Cambridge

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Ghosh, A. Econophysics research in India in the last two decades, Saha Institute of Nuclear Physics, Kolkata, India.

systems, offering new insights and approaches to understanding complex economic phenomena. *Franck et al.* (2021) discussed influence of physical science concepts on economics and the less studied influence of economics on physics. The interplay between physics and economics in Econophysics is multifaceted and involves several key aspects:

- Complexity and Emergence: Physics provides Econophysics with a framework for understanding complex systems and emergent phenomena. Economic systems, like physical systems, exhibit behaviours that are not predictable from the behaviour of individual components. Concepts from physics, such as self-organization and criticality, are used to study how these complex behaviours emerge from the interactions of individual agents in the economy (*Mantegna & Stanley, 1999*).
- Statistical Mechanics: Statistical mechanics, a branch of physics that deals with the behaviour of large collections of particles, has been instrumental in developing models for understanding economic systems. In Econophysics, statistical mechanics is considered in analysis of behaviour of economic agents and the interactions between them, leading to insights into market dynamics and wealth distribution (*Chakrabarti et al., 2010*).
- Network Theory: Network theory, another area of physics, is considered in analysis of structure and connectivity of economic networks. By treating economic agents or variables equivalent to nodes in a network and their interactions as edges, econophysicists can analyze the flow of information, goods, and money in the economy, revealing important insights into market structure and systemic risk (*Jackson, 2008*).
- Nonlinear Dynamics: Nonlinear dynamics, a field of physics that studies systems whose behaviour is not proportional to their inputs, is used to form the complex dynamics of economic systems. Nonlinear models are being used to study phenomena such as market bubbles, crashes, and oscillations, providing insights into the stability and resilience of economic systems (*Brock & Hommes, 1998*).
- Information Theory: Information theory, a branch of physics that deals with the quantification of information, is used in Econophysics to study how information is transmitted and processed in economic systems. This can be very critical into the efficiency of markets and the impact of information asymmetry on market outcomes (*Cover & Thomas, 2006*).

Econophysics draws heavily on concepts and methodologies from physics to study economic systems. It utilizes tools such as entropy, phase transitions, and scaling laws to analyze complex economic phenomena (*Raine et al., 2006*). Similarly, economics provides Econophysics with a framework for understanding human behaviour, preferences, and decision-making, enriching its analytical capabilities.

Models in Econophysics

Econophysics utilizes various models to understand and analyze complex economic systems. These models are often inspired by concepts and techniques from physics, adapted to capture the dynamics of economic phenomena. Some of the key models used in Econophysics include:

- Random Walk Model: The random walk model is a foundational concept in Econophysics, particularly in the study of financial markets. It assumes that the price of a financial asset evolves randomly over time, with each step independent of the previous ones. This model has been used to describe the movement of stock prices and other financial assets *(Mantegna & Stanley, 1999). Gambler (2022)* used random walk model to study various phenomena, including the evolution of reserves in pension funds and contributed to the understanding of financial markets by challenging the assumption of stock price randomness and highlighting the influence of irrationalities among market participants on price changes.
- Agent-Based Models: Agent-based models (ABMs) are computational models that simulate the behaviour of individual agents within a system. In Econophysics, ABMs are used to study how interactions between individual agents can lead to macroscopic market patterns. These models capture complex phenomena such as herding behaviour, market bubbles, and crashes *(LeBaron, 2006).*
- Percolation Models: Percolation models are used to study the emergence of large-scale connectivity in networks. In Econophysics, percolation models have been applied to understand

Ram Lal Bagaria: Econophysics: Bridging the Gap between Economics and Physics

spread of information or financial contagion through networks of economic agents or institutions (Cont et al., 2010).

- Ising Model: The Ising model of statistical physics, has been adapted to study opinion dynamics and decision-making in economics. It models agents as spins on a lattice, where each agent's decision is influenced by the decisions of its neighbours (*Gallegati et al., 2006*).
- Game Theory Models: Game theory is a fundamental tool in economics, and econophysicists often use game theory models to study strategic interactions between agents. These models shed light on phenomena like cooperation, competition, and the evolution of social norms (Sznajd-Weron & Sznajd, 2000).
- Complex Network Models: Complex network theory is used to study structure of economic networks, such as financial markets or trade markets. These models reveal important properties of networks, such as their robustness to shocks or their vulnerability to cascading failures (Newman, 2010).
- Stochastic Models: Stochastic models, which incorporate randomness into their dynamics, are used extensively in Econophysics. These models capture the inherent uncertainty and unpredictability of economic systems, making them useful for analyzing financial markets and other complex systems (*Bouchaud & Potters, 2009*).
- Fractal Models: Fractal geometry, which describes complex, self-similar patterns, has been applied to study the dynamics of economic time series. Fractal models capture the long-range correlations and scaling properties observed in economic data (*Peters, 1994*).

These models are indicative but not limited, several other models like gravity approach, relatively trade theory etc. are being considered by experts in Econophysics, providing a framework for understanding the complex dynamics of economic systems from a physics perspective.

Tools and Techniques in Econophysics

Econophysics employs a variety of tools and techniques to analyze economic systems, drawing on methods from physics, mathematics, and computer science. These tools help econophysicists understand the complex dynamics of economic phenomena and extract meaningful insights from data. Some of the key tools and techniques used in Econophysics include:

- Statistical Physics: Statistical physics provides a powerful framework for understanding the collective behaviour of large-scale systems composed of many interacting components. In Econophysics, statistical physics is employed to model economic systems, considering interactions between individual agents and emergence of macroscopic patterns (Bouchaud & Potters, 2009).
- Network Theory: Network theory is used extensively to analyze structure and dynamics of economic networks, such as financial networks or trade networks. Techniques from network analysis can reveal crucial properties of these networks, including their connectivity, centrality, and susceptibility to shocks (Barabási, 2016).
- Agent-Based Modeling: Agent-based modeling (ABM) is a computational approach used to simulate the behaviour of individual agents within a system. In Econophysics, ABMs are utilized to study how interactions between individual agents can lead to complex market dynamics, such as price fluctuations and market crashes (Tesfatsion, 2002).
- Complex Systems Theory: Complex systems theory offers insights into how simple rules at the individual level can give rise to complex behaviour at the system level. In Econophysics, this theory is used to explore how interactions between individual agents in a market can lead to emergent phenomena like market bubbles or crashes (Holland, 2014).
- Machine Learning: Machine learning techniques are increasingly applied in Econophysics to analyze large datasets and extract patterns and trends. These algorithms are being used to make predictions about future market behaviour or identify hidden patterns in economic data (Lipton, 2018).
- Data Mining: Data mining techniques are utilized in Econophysics to gain insightful information from vast and large datasets. These methods help identify correlations between different variables or trends over time, aiding in the understanding of economic systems (Han et al., 2011).

- Computational Modeling: Computational modeling involves using computer simulations to study the behaviour of economic systems. In Econophysics, these models are used to test hypotheses, explore complex interactions, and predict future market behaviour (LeBaron, 2006).
- Time Series Analysis: Time series analysis is crucial for studying the behaviour of economic variables over time. This technique helps uncover patterns and trends in economic data, such as seasonality, trends, and cycles, providing valuable insights into economic systems (*Wei, 2006*).

The aforementioned tools and techniques form the foundation of Econophysics, providing researchers with the means to analyze and understand the complex dynamics of economic systems.

Applications of Econophysics in Economics

Econophysics, as an interdisciplinary field, has a wide range of applications in economics. By applying concepts and methods from physics to economic systems, econophysicists have been able to gain new insights into complex economic phenomena. Some key applications of Econophysics in economics include:

- Market Dynamics: Econophysics has played a crucial role in understanding financial markets dynamics. By applying principles from statistical physics, researchers have come up with models that capture the complex behaviour of stock prices, trading volumes, and market volatility (Bouchaud & Potters, 2009). These models, such as the random walk model, have provided insights into the seemingly unpredictable nature of financial markets.
- Wealth Distribution: The analysis of wealth distribution is another area in which Econophysics has provided significant contributions. Researchers have used concepts from statistical mechanics to model wealth as a complex system, leading to the observation of power-law distributions in wealth distribution patterns (Chatterjee et al., 2007). These findings have implications for understanding income inequality and social welfare policies.
- Financial Risk Management: Econophysics has also contributed towards new approaches in financial risk management. By analyzing the interconnectedness of financial markets using network theory, researchers have identified systemic risks and developed strategies to mitigate these (Battiston et al., 2016). This has led to improvements in financial regulation and risk management practices.
- Complex Systems Analysis: Econophysics has provided valuable insights into economic systems as complex, adaptive systems. By modeling interactions between individual agents in a market, researchers have studied the emergence of collective behaviours such as market bubbles and crashes (Haldane, 2011). These studies have deepened our understanding of market dynamics and instability.
- Agent-Based Modeling: Agent-based models (ABMs) have been a key tool in Econophysics for simulating the behaviour of individual agents in an economy. ABMs allow researchers to study how interactions between agents lead to complex market dynamics, such as herding behaviour and the emergence of market trends (LeBaron, 2006).
- Forecasting and Prediction: Econophysics has also contributed to the development of new methods for forecasting and predicting economic trends. By applying machine learning and data mining techniques to large datasets, researchers have identified patterns and trends that can be used to make more accurate predictions about future market behaviour (*Moore & Shannon, 2017*). These methods have improved our ability to anticipate market movements and make informed investment decisions.

Econophysics has a noteworthy impact on economics, providing new tools, methods, and insights that have helped to advance our understanding of complex economic systems.

Future Potential

Econophysics may provide a more comprehensive understanding of economic systems as complex, dynamic systems. The subject may lead to better models and policies that account for inherent complexity of economic interactions. Econophysics can continue to contribute to development of new risk management strategies. By analyzing dynamics of financial markets, researchers may identify early warning signs of potential crises and develop strategies to mitigate risks. Econophysics may improve economic policy by providing insights into the effects of different policy interventions. By modeling impact of policy changes on economic systems, policymakers can make more informed decisions. Econophysics

Ram Lal Bagaria: Econophysics: Bridging the Gap between Economics and Physics

can benefit from increased collaboration with other disciplines, such as computer science and mathematics. The interdisciplinary approach has potential to lead to new insights and methodologies for studying economic systems.

Conclusion

Econophysics offers a unique approach to studying economic systems, drawing on concepts and methods from physics to analyze complex economic phenomena. By bridging the gap between economics and physics, Econophysics provides new insights into the behaviour of markets and economies, with the potential to inform economic policy and decision-making in the future.

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- 34
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