

ECONOPHYSICS: A NEW CHALLENGE FOR FINANCIAL ECONOMICS?

BY

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Financial economics was born in the 1960s. It took less than two decades for the new discipline's main theoretical results to become established, creating what is considered to be mainstream financial economics. Less than thirty years later, a new field of research called "econophysics" was created. This field aims to reinvent modern financial theory and, indirectly, financial economics.

This article proposes to study, by an historical analysis, to what extent econophysics today could constitute one of the major theoretical challenges to financial economics. It shows how these two fields have historical similarities, and analyzes how these similarities call the future evolution of financial theory into question.

I. INTRODUCTION

Financial economics was born in the 1960s. It took less than two decades for the new discipline's main theoretical results (efficient market theory, option pricing model, CAPM, and modern portfolio theory) to become established, creating what is considered to be mainstream financial economics.¹ And, although several later theoretical movements in financial economics (for example, behavioral finance and microstructure of financial markets) have tried to challenge its pre-eminence, the mainstream approach remains dominant in financial economics.² Less than thirty years later, a new field of research called "econophysics" was created. This field was created outside financial economics

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¹On the history of mainstream financial economics, see Bernstein (1992), Jovanovic (2008), Melhring (2005, 2012), Poitras and Jovanovic (2007, 2010), or Whitley (1986a).

²In line with Frickel and Gross (2005, 208), the adjective "dominant" is used here to signify a progressive movement that urges a revival of past ideas to push the field forwards in new directions. Dominance must not be associated with the idea of truth but rather with the ability to provide a progressive evolution of knowledge. In our view, econophysics is not truer than financial economics but, as we will show, by solving old problems observed in finance, it offers a progressive perspective for financial economics.

1 by physicists from statistical physics. It studies economic phenomena and, more
2 specifically, financial markets by using models and concepts imported from condensed
3 matter and statistical physics, a very specific area of physics.³ In other words, econo-
4 physics is characterized by the application to financial markets of models from statistical
5 physics that use stable Lévy processes.⁴ Econophysics, which represents itself as
6 a new paradigm, aims to reinvent modern financial theory and, indirectly, financial
7 economics. This article proposes to study, by an historical analysis, to what extent
8 econophysics today could constitute one of the major theoretical challenges to financial
9 economics.

10 Using hypotheses and mathematical models that financial economists did not or
11 could not develop when their discipline was taking shape in the 1960s, econophysicists
12 are achieving better simulations of stock-price variations and thus providing more
13 accurate forecasts than those obtained from most models used in financial economics
14 (Roehner 2002, McCauley 2004). This situation could pose a considerable challenge
15 to financial economics, which established itself in the 1960s and 1970s, because it
16 claimed, among other things, to provide better forecasts than rival theories at the time.
17 In addition, the mathematical formalisms used by econophysicists today exist in a more
18 general mathematical framework than those currently used in financial economics.
19 These two advantages of econophysics raise the question of whether, and to what
20 extent, it could replace financial economics as the dominant approach in modern financial
21 theory. Were econophysics to become dominant, its assumptions, theoretical framework,
22 and results would form a reference, with other currents positioning themselves and
23 their own work in relation to it.

24 In seeking to address this question, this article is divided into two parts.

25 The first part highlights the historical similarities between the emergence of financial
26 economics in the 1960s and that of econophysics in the 1990s. We begin by describing
27 the birth of financial economics and then the emergence of econophysics. By means of
28 a comparative analysis, we will then show that the factors that led to the development
29 of both financial economics and econophysics are identical: the role played by modern
30 probability theory and by empirical data, and claims on the same scientific criteria.

31 These similarities call the future evolution of financial theory into question. Since
32 econophysics has followed the same development path as financial economics, and
33 since the empirical results and mathematical models of econophysics correspond more
34 closely to empirical data, it is reasonable to ask whether econophysics might take over
35 financial economics, just as financial economics took over from earlier approaches
36 to finance in business schools. This is the question addressed in the second part. We
37

38
39 ³The influence of physics on financial economics is nothing new and has been well documented in the
40 literature. But in spite of the theoretical and historical links between physics and financial economics,
41 econophysics represents a fundamentally new approach that differs from preceding influences. Its practitioners
42 are not economists taking their inspiration from the work of physicists to develop their discipline; this time,
43 it is physicists who are studying various problems brought to light by the methods of the social sciences.
44 On the history of econophysics, see Jovanovic and Schinckus (2013) or Roehner (2002).

45 ⁴Of course, statistical physics cannot be reduced to the use of Lévy processes: it goes back much earlier, to
46 the work of Gibbs and those of Boltzmann. However, econophysics focuses on this very specific class of
47 models. The consequence is that today the literature of econophysics is (not only but) mainly based on the
48 application of the Lévy processes to financial economics—see Gingras and Schinckus (2012) for a
49 bibliometric study of this point.

1 discuss econophysics' two principal theoretical advantages: better forecasting of
 2 stock-market variations, and a more general statistical framework. We then offset these
 3 advantages by looking at the two main hindrances to econophysics' chances of
 4 challenging the dominant approach in financial economics: the poor dialogue between
 5 economists and physicists, and the reluctance of econophysicists to take up the
 6 hypotheses of financial economics. Our analysis of the advantages and disadvantages
 7 provides an historical perspective of the challenges to financial economics raised by
 8 the emergence of econophysics.

11 II. SIMILARITIES BETWEEN TODAY AND THE 1960S

13 This section provides a comparative analysis of the emergence of the two disciplines
 14 that concern us. It shows that the birth of econophysics in the 1990s follows the same
 15 pattern as the birth of financial economics in the 1960s. We will focus particularly on
 16 two major similarities: first, the role played by modern probability theory and by
 17 empirical data; and, second, the manner in which these new scientific communities
 18 were constituted.

20 *Historical Echoes*

22 *The Birth of Financial Economics*

23 As this section reminds us, financial economics owes its institutional birth in the 1960s
 24 to three elements: access to the tools of modern probability theory; the creation of new
 25 empirical data; and the extension of the analysis framework of economics.⁵

27 *On the accessibility of the tools of modern probability theory.* Financial economics is
 28 intimately bound up with modern probability theory, from which its emergence,
 29 main models, and results are inseparable. So close are the links that, further to the
 30 publications of Harrison and Kreps (Harrison and Kreps 1979) and Harrison and
 31 Pliska (1981),⁶ some authors have suggested that economics has been dispossessed
 32 of financial theory, which has since resembled an application of modern probability
 33 theory (MacKenzie 2006, pp. 140–141). Or, as posited by Davis and Etheridge,
 34 Harrison and Pliska's article (1981) "has turned 'financial economics' into 'mathematical
 35 finance'" (Davis and Etheridge 2006, p. 114).

36 Modern probability theory—probability for continuous quantities in continuous
 37 time—emerged in the 1930s (Von Plato 1994) out of a number of works aimed at
 38 renewing traditional probability theory. The development of the modern version of
 39 probability theory was directly based on measure theory (Shafer and Vovk 2001). The
 40 connection was made by Kolmogorov, who proposed the main founding concepts of
 41 this new branch of mathematics.

45 ⁵On the emergence of financial economics, see Jovanovic (2008, 2009b, 2009a), Jovanovic and Schinckus
 46 (2010), MacKenzie (2006), and Whitley (1986a).

47 ⁶These two publications gave a rigorous mathematical framework to definitions, hypotheses, and results
 that constitute the heart of modern financial theory.

1 From these beginnings in the 1930s, modern probability theory developed and became
2 increasingly influential. But it was not until after World War II that Kolmogorov's
3 axioms came to dominate this discipline (Shafer and Vovk 2005, pp. 54–55). It was
4 also after World War II that the American probability school was born, led by Doob⁷
5 and by Feller,⁸ who proved, on the basis of the framework laid down by Kolmogorov,
6 all results obtained prior to the 1950s, thereby enabling them to be accepted and
7 integrated into the discipline's theoretical corpus. These 1950s works led to the creation
8 of a stable corpus that was accessible to non-specialists. From then on, the models and
9 results of modern probability theory were used in the study of financial markets in a
10 more systematic manner, in particular by scholars educated in economics.

11 The first step in this development was the dissemination of mathematical tools
12 enabling the properties of random variables to be used and uncertainty reasoning to be
13 developed. The first two writers to use tools that came out of modern probability theory
14 to study financial markets were Harry Markowitz and A. D. Roy. In 1952, each published
15 an article on the theory of portfolio choice theory.⁹ Both used mathematical properties
16 of random variables to build their model.¹⁰ Their work was to re-prove a result that had
17 long been known (and that was as old as the adage “Don't put all your eggs in one
18 basket”), using a new mathematical language, that of modern probability theory. Their
19 contribution lay not in the result of portfolio diversification, but in the use of this new
20 mathematical language.

21 From the 1960s on, a new stage was embarked upon: authors no longer limited
22 themselves to proving past results using the mathematical formalisms of modern
23 probability theory, but connected mathematical formalism with the main concepts of
24 economics, particularly the concept of equilibrium, to create new theories.

25 The efficient markets theory,¹¹ which can be considered as the first theory built by
26 financial economists, provides a good example. This theory was initially referred
27 to as the “random walk theory.” This term stresses the importance of mathematical
28 formalism in the way issues were tackled before the discipline was constituted. The
29 theory, first formulated by Fama (1965b), made it possible to link the mathematical
30 model of a stochastic process with one of the keystones of economics, the concept of
31 economic equilibrium (Jovanovic 2010). In 1970, Fama based the efficient markets
32 theory on another mathematical concept that came from modern probability theory:
33 the martingale model.¹² For Fama's purposes, the most important attraction of the
34

36 ⁷Doob is without question the American mathematician who has had the greatest influence on modern
37 probability theory in the United States. On Doob, see Bingham (2005).

38 ⁸William Feller emigrated to the United States in 1939. He was one of the first defenders of the axiomatization
39 proposed by Kolmogorov (Shafer et al. 2005). Moreover, Feller's *An Introduction to Probability Theory*
40 *and Its Application* (1950) was, like Doob's 1953 publications, one of the works that most strongly influenced
41 modern probability theory in the United States.

42 ⁹For a retrospective on Markowitz, see Rubinstein (2002) and Markowitz (1999).

43 ¹⁰The mathematical properties of random variables are that the expected value of a weighted sum is the
44 weighted sum of the expected values, while the variance of a weighted sum is not the weighted sum of the
45 variances (because we have to take covariance into account).

46 ¹¹This theory is sometimes called an “hypothesis.” But, from a methodological point of view, it is a fully
47 fledged theory, even if it is used as an hypothesis in some models.

48 ¹²The martingale model had been introduced to model the random character of stock-market prices by
49 Samuelson (1965) and Mandelbrot (1966).

1 martingale formalism was its explicit reference to a set of information.¹³ As such, the
 2 martingale model could be used to test the implication of the efficient markets theory
 3 that, if all available information is used, the expected profit is nil. This idea led to the
 4 definition of an efficient market that is generally used nowadays: “a market in which
 5 prices always ‘fully reflect’ available information is called ‘efficient’” (1970, p. 383).
 6 The part played by economics in the mathematical definition of the martingale model
 7 underlines economics’ key role in the creation of the structure of modern financial
 8 theory.

9
 10 *The creation of new empirical data.* In parallel with the adoption of tools, models, and
 11 concepts from modern probability theory for analyzing financial markets, another
 12 crucial advance occurred in the 1960s: the creation of databases containing long-term
 13 statistical data on the evolution of stock-market prices. These databases allowed
 14 spectacular development of empirical studies used to test models and theories in
 15 finance. The development of these studies was the result of the creation of new
 16 statistical data and the emergence of computers.

17 Beginning in the 1950s, computers gradually found their way into financial institutions
 18 and universities (Sprowls 1963, p. 91). However, owing to the costs of using them and
 19 their limited calculation capacity, “it was during the next two decades, starting in the
 20 early 1960s, as computers began to proliferate and programming languages and
 21 facilities became generally available, that economists more widely became users”
 22 (Renfro 2009, p. 60). The first econometric modeling languages began to be developed
 23 during the 1960s and the 1970s (Renfro 2004, p. 147). From the 1960s on, computer
 24 programs began to appear in increasing numbers of undergraduate, master’s, and
 25 doctoral theses. As computers came into more widespread use, easily accessible
 26 databases were constituted, and stock-market data could be processed in an entirely
 27 new way, thanks to, among other things, financial econometrics (Louçã 2007). Financial
 28 econometrics marked the start of a renewal of investigative studies on empirical data
 29 and the development of econometric tests.

30 A number of financial-econometrics studies were carried out in the 1960s, using
 31 computers. While the first empirical studies of this type date back to 1863 in France
 32 and the early 1930s in the US (Poitras 2006, Jovanovic 2009b), the results were very
 33 limited because all calculations had to be performed by hand. With computers, empirical
 34 study could become more systematic and be conducted on a larger scale. Attempts
 35 were made to test the random nature of stock-market variations in different ways
 36 (Jovanovic 2009b). Markowitz’s hypotheses were used to develop specific computer
 37 programs to assist in making investment decisions.¹⁴

38 Of the databases created during the 1960s, one of the most important was set up by
 39 the Graduate School of Business at the University of Chicago,¹⁵ one of the key institutions

42 ¹³By definition, a martingale model, $E(P_{t+1} | \Phi_t) - P_t = 0$, Φ_t is a filter—that is, to use the terminology of
 43 financial economics, a set of information that increases over time.

44 ¹⁴See, for instance, Cohen and Pogue (1967).

45 ¹⁵In 1960, two University of Chicago professors, James Lorie and Lawrence Fisher, started an ambitious
 46 four-year program of research on security prices. They created the Center for Research in Security Prices
 47 (CRSP), which had an important group of PhD students such as Eugene Fama, Benjamin King, and Arnold
 Moore. Merton Miller joined them one year later, in 1961.

1 in the development of financial economics. The first version of this database, which
2 collected monthly prices from January 1926 through December 1960, greatly facilitated
3 the emergence of empirical studies. Apart from its exhaustiveness, it provided a history
4 of stock-market prices and systematic updates.

5
6 *The institutionalization of financial economics and the challenge to the dominant school*
7 *of thought of the time.* The third element that contributed to the institutional birth of
8 financial economics was the integration of the analysis framework of economics
9 (hypotheses, concepts, method, etc.) into the analysis of financial markets. This integration
10 was the result of the formation in the early 1960s of a community of economists
11 dedicated to the analysis of financial markets.

12 Until the 1960s, finance in the United States was taught mainly in business schools. The
13 textbooks used were very practical, and few of them touched on what became modern
14 financial theory. The research work that formed the basis of modern financial theory was
15 carried out by isolated writers who were trained in economics or were surrounded by
16 economists, such as Working, Cowles, Kendal, Roy, and Markowitz. No university
17 community devoted to the subject existed prior to the 1960s.¹⁶ During the 1960s and
18 1970s, training in American business schools changed radically, becoming more
19 “rigorous.”¹⁷ They began to “academicize” themselves, recruiting increasing numbers
20 of economics professors who taught in university economics departments, such as
21 Miller (Fama 2008). Similarly, prior to offering their own doctoral programs, business
22 schools recruited doctorands who had been trained in university economics departments.

23 The recruitment of economists interested in questions of finance unsettled teaching
24 and research as hitherto practiced in business schools and inside the American Finance
25 Association. The new recruits brought with them their analysis frameworks, methods,
26 hypotheses, and concepts, and also used the new mathematics that arose out of modern
27 probability theory. These changes and their consequences were substantial enough for
28 the American Finance Association to devote part of its annual meeting to them in two
29 consecutive years, 1965 and 1966.

30 At the 1965 annual meeting of the American Finance Association, an entire session
31 was devoted to the necessity to rethink courses in finance curricula. Paul Wendt
32 discussed the development of finance and explained that “a modern concept of technical
33 market analysis is emerging which emphasizes the application of newer analytical
34 techniques and computer technology to test traditional and new theories of stock-price
35 behaviour” (Wendt 1966, pp. 421–422). At the 1966 annual meeting, the new president of
36 the American Finance Association presented a paper on “The State of the Finance Field,”
37 in which he talked of the changes being brought about by “the creators of the New Finance
38 [who] become impatient with the slowness with which traditional materials and teaching
39 techniques move along” (Weston 1967, p. 539).¹⁸ Although these changes elicited many

41 ¹⁶The new research path was not accepted by economists until the 1960s. Milton Friedman’s reaction to
42 Harry Markowitz’s defense of his PhD thesis gives a good illustration. Friedman declared: “It’s not
43 economics, it’s not mathematics, it’s not business administration,” and Jacob Marschak, who supervised
44 Markowitz during his PhD, added: “It’s not literature” (Markowitz 2004).

45 ¹⁷See Mackenzie (2006, pp. 72–73), Whitley (1986a, 1986b), Fourcade and Khurana (2009), and Bernstein
46 (1992).

47 ¹⁸The same issues were raised in training sessions given by Financial Analysts Seminar, one of the leading
professional organizations connected with financial markets (Kennedy 1966).

1 debates (Whitley 1986a, 1986b; MacKenzie 2006; Poitras et al. 2007; Jovanovic 2008;
2 Poitras et al. 2010), none succeeded in challenging the global movement.

3 The antecedents of these new actors were a determining factor in the institutionalization
4 of modern financial theory. Their background in economics allowed them to add
5 theoretical content to the empirical results that had been accumulated since the 1930s
6 and to the mathematical formalisms that had arisen from modern probability theory.
7 In other words, economics brought the theoretical content that was missing.

8 The creation of a new scientific community requires that its new members share
9 common tools, references, and problems. This was precisely the role of textbooks,
10 seminars, and scientific journals. Those in financial economics were developed from
11 the beginning of the 1960s with the arrival of this new generation of professors and
12 students. The two journals that had published articles in finance, the *Journal of Finance*
13 and the *Journal of Business*, changed their editorial policy during the 1960s. Both
14 started publishing articles based on modern probability theory and on modeling
15 (Bernstein 1992, pp. 41–44, 129). They also published several special issues to
16 reinforce the new orientation and results. In 1966, the *Journal of Business* published
17 a special issue on “recent quantitative and formal research on the stock market.” In
18 addition to these two journals, other scientific journals specializing in financial
19 economics were created, such as the *Journal of Financial and Quantitative Analysis* in
20 1965. In 1968, the last-mentioned journal published a special issue on the application
21 of the random walk model to stock prices.

22 It was also during the 1960s that textbooks and collections of articles started to
23 appear.¹⁹ These publications also helped define and stabilize a culture shared by the
24 members of the new community. Several seminars were also organized. The new
25 seminars and the publications contributed to the creation of a truly homogenous
26 community (which shared common problems, common tools, and a common language)
27 and scientific journals and courses in universities. One of the main features of this
28 common culture was the creation of a canonical history of financial economics during
29 the 1960s. This history was created to support theoretical viewpoints—viewpoints
30 that led the mainstream community of scientists to recognize financial economics as a
31 science (Jovanovic 2008).

32 33 *The Birth of Econophysics*

34 Econophysics studies emerged in the 1990s. Their origins—like those of financial
35 economics—lay in modern probability theory, the emergence of new empirical data,
36 and the use of models, hypotheses, and methods taken from a discipline outside the
37 mainstream of the time. The birth of econophysics, therefore, closely resembles that of
38 financial economics in the 1960s.
39

40 *The role of modern probability theory.* From the 1980s on, modern probability theory
41 evolved to some degree, particularly with regard to the definition of stochastic
42 processes known as “Lévy processes” (and, more precisely, as “stable Lévy processes”).
43 Lévy processes include many classes of stochastic processes, such as the Wiener
44

45 ¹⁹For instance, Cootner (1964), Fredrikson (1965), Wu and Zakon (1965), Fredrikson (1971), and Lorie
46 and Brealey (1972) published collections of articles, while Moore (1968), Mao (1969), Jean (1970), and
47 Fama and Miller (1972) published textbooks.

1 process, jump-diffusion processes, and jump stable Lévy processes.²⁰ Jump stable
 2 Lévy processes are characterized by the existence of small jumps in alternation with
 3 big jumps; in general, they have infinite variance. Due to their stable Lévy character,
 4 and unlike jump-diffusion models, jump stable Lévy models have infinite activity
 5 (infinite number of jumps on each time interval) and infinite variation. A specific class
 6 of jump stable Lévy processes considered by econophysicists is composed of stable
 7 Lévy processes.

8 Stable Lévy processes are jump processes, characterized by the stable Lévy
 9 distribution, having an α -stable law type $P(X > x) = x^{-\alpha}$ in which it is possible to
 10 observe constancy of the coefficient α . A stable Lévy distribution with $\alpha = 2$ is a
 11 Gaussian distribution; with $\alpha = 1$, it is a Cauchy distribution; and with $\alpha = 3/2$, it is a
 12 Pareto distribution.²¹ Stable Lévy processes have a distribution with infinite variance,
 13 which is considered the main obstacle to the use of these processes in finance. Indeed,
 14 an infinite variance means that risk can vary considerably, depending on the size of the
 15 sample and the observation scale.

16 At the beginning of the 1980s, stable Lévy processes were the subject of a theoretical
 17 debate in specialist literature on statistics. These processes were essentially seen as
 18 theoretical tools (Zolotarev 1986) or as “monsters” (MacKenzie 2006, p. 108) with no
 19 real practical applications, due to their infinite variance. This theoretical characteristic
 20 added considerably to the complexity of applying these processes to observed phe-
 21 nomena, since the notion of variance very often refers to a well-defined empirical pa-
 22 rameter, which is finite (risk in finance, temperature in thermodynamics, for example).
 23 Considering this situation, physicists developed theoretical solutions whose objective
 24 was to define a class of processes that was compatible with the empirical observations
 25 they had. This theoretical literature led to the development of Lévy processes with
 26 finite variance, called “truncated Lévy processes” (Schinckus 2011b).

27 From the 1980s onwards, stable Lévy processes were increasingly used in physics,²²
 28 particularly in statistical physics. The latter discipline can be thought of as the contin-
 29 uation of thermodynamics, and the use of stable Lévy processes in this field allowed
 30 more accurate modeling of the phenomenon of turbulence. The first studies on Lévy
 31 processes applied to turbulence phenomena were those of Kolmogorov on the scale
 32 invariance of turbulence in the 1930s. This theme was subsequently addressed by
 33 many physicists and mathematicians, particularly by Mandelbrot in the 1960s when he
 34 defined fractal mathematics²³ and applied it to the phenomenon of turbulence.

38 ²⁰A jump-diffusion process is a process generally compounded of a Wiener process (Brownian motion),
 39 characterized by the Gaussian distribution and another stochastic process (often a Poisson process).
 40 Overall, it is a Brownian motion with jumps at a specific rate dictated by the second process (Poisson
 41 process); the amplitudes of the jumps are characterized by Gaussian, exponential, or other kind of
 42 distributions. These compound processes have distribution with finite variance and a certain leptokurtic
 43 character. Also, the jump-diffusion processes have finite activity (finite number of jumps on each time
 44 interval) and finite variation.

44 ²¹For a statistical presentation of these specific laws, see Schoutens (2003).

45 ²²See Frisch et al. (1994) for an analysis of the influence of Lévy processes in physics.

46 ²³Although modern probability theory was properly created in the 1930s, in particular through the work of
 47 Kolmogorov, it was not until the 1950s that the Kolmogorov's axioms became the mainstream in this
 discipline.

1 Despite the extension of probability theory to thermodynamics, physicists did not
 2 seem disposed to integrate stable Lévy processes into physics (Gupta and Campanha
 3 2002, p. 382). The reason for this methodological position—which mirrors that taken
 4 in financial economics—is that processes with infinite variance were not physically
 5 plausible (i.e., compatible with assumptions of physics).²⁴ As Gupta and Campanha
 6 (1999, p. 232) point out, Lévy processes “have mathematical properties that discourage
 7 a physical approach because they have infinite variance.”²⁵ The use of Lévy processes
 8 in physics necessitated the development of truncated Lévy processes, which allowed
 9 physicists to use these processes to statistically characterize turbulence phenomena
 10 without the problem of infinite variance.²⁶ The first truncated stable Lévy process in
 11 physics was proposed by Zolotarev in 1986. This response of physicists to the indeter-
 12 minate nature of variance paved the way for finance applications to describe the
 13 evolution of financial markets using stable Lévy processes that are not Gaussian. The first
 14 application of this statistical solution to finance was proposed by Mantegna in 1991.

15 Truncated Lévy processes, then, provide an opportunity to solve the problem of
 16 infinite variance observed in the application of Lévy processes in thermodynamics and
 17 finance. However, econophysicists use statistical processes in an instrumentalist
 18 manner. They are uninterested in hypotheses, preferring to focus on prediction. Of
 19 course, this kind of methodology is known in finance, where econometrics is often said
 20 to be based on the instrumentalist Friedmanian methodology (Angrist and Pischke
 21 2008). However, unlike econometrics, econophysics is founded on a physically plausible
 22 analysis in which each statistical parameter must have a physical meaning. From this
 23 perspective, econophysicists have developed more sophisticated stable Lévy processes,
 24 giving them a statistical tool in line with a physically plausible framework (Schinckus
 25 2011b).

26
 27 *The creation of new empirical data.* In parallel with this application of stable Lévy
 28 processes to the study of financial markets, new statistical data were created from the
 29 1990s as a result of the automation of financial centers.

30 Since the 1970s, a number of financial markets have been automated—Toronto,
 31 Paris, Nasdaq, and Euronext, for example. Today, electronic markets flood the financial
 32 sphere with accurate data in real time. The automation of markets has made it possible
 33 to record “intraday” data. Previously, statistical data on financial markets were generally
 34 made up of a single value per day (the average price). Today, by recording “intraday
 35 data,” all prices quoted are conserved. The new data collected in this manner have
 36 stimulated research into distributions of stock-market variations. It is difficult to determine
 37 laws of financial data distribution with any certainty. Mitzenmacher (2004) reminds us
 38

39
 40
 41 ²⁴For example, temperature is assumed to be finite in physics, so to be physically plausible, a statistical
 42 process used in physics must generate a finite temperature.

43 ²⁵In the view of physicists, this property of physical systems is the direct result of the thermodynamic
 44 hypotheses set out by Boltzmann in 1872 when he laid the foundations of contemporary statistical
 45 mechanics.

46 ²⁶The truncation of Lévy distributions consists in normalizing them, using a particular function so that
 47 variance is finite. One can, for example, combine an untruncated Lévy process for the distribution center
 and explain the ends of tails using exponential distributions. On this subject, see Gupta and Campanha
 (2002).

1 of how close these laws are to the so-called exponential laws, and that the two types of
2 law can be distinguished only by means of a large volume of data. The advent of
3 intraday data has made it possible to build sufficiently broad samples to provide firm
4 proof of Mandelbrot's idea that the evolution of financial markets could be characterized
5 using stable Lévy processes such as those used by econophysicists (Kou 2008).

6 In this context, computer technology is presented as a tool that makes it possible to
7 confirm with certainty the hypothesis that the evolution of prices and returns on financial
8 markets can be characterized by a stable Lévy distribution. This growing quantification
9 of financial information takes the form of an accumulation of data stored as temporal
10 series, thereby making market finance "a natural area for physicists" (Gallegati, Keen,
11 et al. 2006, p. 1).

12 Mantegna and Stanley (1999, p. 6), McCauley (2004, p. 7), and Burda, Jurkiewicz,
13 and Nowak (2003, p. 3) also underline the role that computerization played in the
14 emergence of econophysics, and above all the fact that it broadened the perspective
15 of statistical market analysis. Now that financial markets are computerized, tens of
16 thousands of transactions or posted quotes in a single day—time-stamped to the nearest
17 second—can be observed (Engle and Russell 2004). Analyzing these new data sets
18 brings new challenges, and they require new statistical tools to characterize them.
19 More precisely, several phenomena can be detected with intraday data that are not
20 present with monthly or daily data. The latter are generally the last prices quoted
21 during a month or a day, or a mean of the prices quoted during a period, and, therefore,
22 jumps in data are generally smaller and less frequent. The computerization of financial
23 markets has, therefore, contributed to the use of new frameworks such as stable Lévy
24 processes, which are better suited to the modeling of jumps in stock-price variations.

25 Computerization of financial markets and, more generally, of the entire financial
26 sphere has had another consequence that has favored the use of stable Lévy processes
27 with which physicists work. According to Barber and Odean (2001), the computerization
28 of the financial sphere provides an "illusion of knowledge" to online investors, who
29 become excessively self-confident and tend to underestimate risks. This overconfidence
30 of investors leads them to invest more, and in a more speculative way, than they otherwise
31 would.²⁷ Barber and Odean (2001) conclude that, in this way, online trading contributes
32 to an increase in market volatility. This greater volatility has engendered more extreme
33 variations in quotations (Jiang, Tang, et al. 2002), and, therefore, the tails of empirical
34 distributions have become fatter. The normal law, used by almost all models in financial
35 economics, does not allow extreme variations to be taken into account.²⁸ Such variations
36 are, however, perfectly integrated into stable Lévy processes. The increase in the
37 volatility of financial markets implying fatter tails of empirical distributions has, therefore,
38 helped justify the use of statistical tools developed in physics that are suited to the
39 analysis of extreme phenomena.

40
41
42 _____
43 ²⁷For instance, the development of options has increased the volatility (i.e., the price variations) of the
44 underlying asset. Numerous studies on the effects of futures and options listing on the underlying cash-
45 market volatility have been done. For instance, Wei, Poon, and Zee (1997) report an increase in volatility
46 for options on OTC stocks in the USA.

47 ²⁸A device used by financial economists is to combine a normal law and a Poisson law in order to repro-
duce jumps.

1 We can see, therefore, a double contribution of technology to the emergence of
2 econophysics: one is direct, resulting from the computerization of financial markets
3 (better analysis and storage of data); the other, more indirect, results from financial
4 behavior that computerization has engendered.

5
6 *The institutionalization of econophysics.* In less than twenty years, econophysics has
7 earned recognition as a scientific field from academics of the hard sciences.²⁹ To gain
8 this recognition, econophysicists adopted a variety of strategies for spreading their knowledge.
9 Symposia were organized, several specialized journals created, and specific academic courses
10 offered by physics departments to promote the scientific recognition and institutionalization
11 of this new approach. All these strategies played a part not only in disseminating econo-
12 physics but also in creating a shared scientific culture (Nadeau 1995).

13 The first publications date from the 1990s. The founding article by Stanley et al.,
14 published in 1996, strongly influenced physicists and mathematicians who, suddenly,
15 developed a non-Gaussian approach to the study of financial returns (Kutner and Grech
16 2008). Since 1996, sustained growth in the number of articles devoted to econophysics
17 has been observed (Gingras et al. 2012). The increase in the number of articles
18 published each year earned econophysics official recognition as a subdiscipline of
19 physical sciences in 2003—less than ten years after its birth.

20 The first textbooks on econophysics were not far behind, the first being published
21 in 1999 by Mantegna and Stanley (*An Introduction to Econophysics*). The process
22 of institutionalization was reinforced in 2006 with the creation of the Society for
23 Economic Science with Heterogeneous Interacting Agents (ESHIA), whose objective
24 is to promote an interdisciplinary research among economics, physics, and computer
25 science (essentially artificial intelligence). This interdisciplinary project, supported by
26 the creation of new journals,³⁰ is, therefore, aimed at the area covered by econophysics.

27 A further indicator of the emergence and the institutionalization of the new scientific
28 community is the organization of symposia and workshops. The first conference devoted
29 to econophysics was organized in 1997 by the physics department of the University
30 of Budapest. Today, conferences and symposia dedicated to econophysics are very
31 numerous, notable among them being the Nikkei Econophysics Research Workshop
32 and Symposium and Econophysics Colloquium. In addition to the numerous publications
33 about econophysics, all these regular events constitute institutional spaces that are
34 helping to make econophysics a true scientific community.

35 The last major element in the institutionalization of econophysics is university
36 education. Today, the physics departments of the Universities of Fribourg (Switzerland),
37 Ulm (Sweden), Münster (Germany), and Dublin (Ireland) offer courses in econophysics.
38 Since 2002, the Universities of Warsaw and Wrocław (both in Poland) have been
39 offering a bachelor's and a master's degree in econophysics, respectively (Kutner et al.
40 2008). Finally, the University of Houston (Texas, USA) created the first doctoral
41 program in econophysics in 2006,³¹ followed in 2009 by the University of Melbourne

42
43 ²⁹The growing presence of econophysics in the pages of physics journals has probably contributed to the
44 official recognition of the field by the Physics and Astrophysics Classification Scheme (PACS): since 2003,
45 econophysics has been an official subcategory of physics under the code *89.65 Gh*. For further information
46 about the emergence of econophysics, see Jovanovic and Schinckus (2013).

46 ³⁰The *Journal of Economic Interaction & Coordination*, and also *Quantitative Finance*.

47 ³¹Information on the program may be found at <http://phys.uh.edu/research/econophysics/index.php>.

1 (Australia).³² All these programs are offered by physics departments, and courses are
2 essentially oriented toward statistical physics and condensed-matter physics.

3 4 *Similarities between the Emergence of Both Disciplines*

5
6 This section underlines the similarities between the emergence of mainstream finan-
7 cial economics and the emergence of econophysics. The similarities suggest that
8 econophysics is, in some respects, a continuation of financial economics.

9 10 *The Role of Probability Theory and Empirical Data*

11 As we saw in the first part, two elements strongly contributed to the emergence of both
12 approaches: the development of modern probability theory, on the one hand; and the
13 evolution of financial markets, which are increasingly quantitative (or digitized), on
14 the other. In each case, these two factors acted as triggers for the emergence of an
15 alternative approach. Let us now look more closely at this point.

16 In the 1960s, as explained earlier, some economists took up random processes
17 at a time when mathematical developments had become newly accessible to non-
18 mathematicians. The use or non-use of these new tools—modern probability theory
19 and work on statistical data—constituted the main element setting the “new approach”
20 against the “traditional approach” of the time: “Mathematical models, careful statistical
21 testing of hypotheses, decision theory, the techniques of operations research, and the
22 new and powerful tool of programming began to be applied to the finance field”
23 (Weston 1967, p. 539).

24 This mathematical evolution went hand in hand with technological developments as
25 the use of computers gradually became widespread. Computers made it possible to
26 perform tests on empirical data (in this case, econometric tests) in order to assess the
27 methods proposed for earning money on financial markets, particularly chartist
28 analysis. In this respect, Rosenfeld (1957, p. 52) proved to be visionary when he
29 suggested using computers for testing theories on a large sample.

30 The development of probability theory combined with finer quantification of
31 financial markets (thanks to developments in computing) were triggering factors in the
32 emergence of econophysics, also. As explained earlier, physicists refined Lévy laws in
33 order to use them in physics. In particular, they developed what are known as truncated
34 Lévy laws, which have finite variance. In this perspective, truncated Lévy laws are to
35 econophysics what the Gaussian framework was to mainstream financial economics³³:
36 statistical justification of its approach and, hence, a justification of its emergence.

37 Once again, in the case of econophysics, computers—again in parallel with
38 mathematical developments—contributed to the emergence of the new approach,
39 because they made possible better quantification of financial operations. Today,
40 electronic markets rule the financial sphere, and allow more accurate study of the
41 evolution of the real-time data they provide (stored in the form of time series). While
42 this type of data has been studied by economists for several decades, the automation of
43

44
45 ³²<http://physics.unimelb.edu.au/Community/Newsroom/News/Econophysics-scholarship-available>

46 ³³The Gaussian framework gave birth to the first studies in financial economics (Jovanovic and Schinckus
47 2010).

1 markets has enabled intraday data providing “three orders of magnitude more data” to
 2 be recorded (Stanley, Amaral, et al. 2000, p. 339). The quantity of data is an important
 3 factor at a statistical level because the larger the sample, the more reliable the identification
 4 of statistical patterns.

5 Computerization of financial centers has led to the recording of huge quantities of
 6 financial data, so much so that econophysicists see finance as truly an “empirical
 7 (rather than axiomatic) science” (Bouchaud 2002). The creation of empirical databases
 8 had played the same role in the 1960s: they stimulated the application of mathematical
 9 models taken from modern probability theory and research into stock-market
 10 variations.

11 Thus, both financial economics and econophysics owe their emergence to the creation
 12 of new mathematics combined with the creation of new statistical data. Morgan and
 13 Morrison (1999) underlined the importance of models in twentieth-century scientific
 14 disciplines, particularly economics. Models have shown themselves to be “mediators”
 15 between theory and reality. In other words, they are neither one nor the other: “It is
 16 precisely because models are partially independent of both theories and world that
 17 they have this autonomous component and so can be used as instruments of exploration
 18 in both domains” (1999, p. 10). Analyses by Cartwright (1983), and Barberousse and
 19 Ludwig (2000), have shown that, in twentieth-century scientific disciplines at least,
 20 scientific models must be interpreted as *fictions*. Although both financial economics
 21 and econophysics owe their emergence to mathematical modeling, they must be
 22 distinguished from one another as regards the place occupied by theory: in the case of
 23 econophysics, there is currently no theoretical explanation to give meaning to the
 24 models used.³⁴

25 26 *The Same Institutionalization Strategy*

27 As regards institutionalization, econophysics is once again following the pattern
 28 observed during the emergence of financial economics: in both cases, a recognized
 29 discipline expanded towards a new field of research whose study had been hitherto
 30 dominated by another discipline. In the 1960s, economics expanded to the study of
 31 financial markets, which, at the time, was dominated by so-called “traditional” financial
 32 theory; in the 1990s, statistical physics expanded to the study of financial markets,
 33 which, at the time, were dominated by financial economics. In both cases, the new
 34 community was made up of scientists trained outside the discipline, and, hence, outside
 35 the mainstream. A kind of colonization of finance has occurred.

36 This colonization can also be detected in the new arrivals’ publication strategy.
 37 They began by publishing in journals of their discipline of origin to make themselves
 38 known and disseminate their results—a sort of takeover of recognized scientific journals
 39 in the discipline of origin.

40 In the 1960s, the newcomers took control of the two main journals specializing in
 41 finance at the time, the *Journal of Business* and the *Journal of Finance*. The aim was
 42 to modify the content of published articles by imposing a more strongly mathematical
 43 content and by using a particular structure: presenting the mathematical model and
 44

45
46 ³⁴This is one of the reasons why we describe econophysics as a scientific field and financial economics as
 47 a scientific discipline.

1 then empirical tests. To reinforce the new orientation, these two journals also pub-
2 lished several special issues. Once control over these journals had been established,
3 the newcomers developed their own journals, such as the *Journal of Financial and*
4 *Quantitative Analysis* created in 1965.

5 Similarly, econophysicists chose to publish and gain acceptance in journals devoted
6 to an existing theoretical field in physics (statistical physics) rather than create new
7 journals outside an existing scientific space and, hence, structure. These journals are
8 among the most prestigious in physics. This editorial strategy is a result not only of the
9 methodology used by econophysicists (deriving from statistical physics) but also of
10 this new community's hope to gain recognition from the existing scientific community
11 quickly, on the one hand, and to reach a larger audience, on the other hand. Then they
12 took control of editorial boards (as in the case of *Physica A* and *The European Journal*
13 *of Physics B*).

14 The new approaches had no alternative to this "colonization strategy," because
15 partisans of the dominant approach (and, hence, of the so-called mainstream journals)
16 rejected these new theoretical developments in which they were not yet proficient.
17 Gradual recognition of the new discipline subsequently allowed new specialist journals
18 to be created, such as the *Journal of Financial and Quantitative Analysis* (1965),
19 *Quantitative Finance* (2001), and the *Journal of Economic Interaction & Coordination*
20 (2006), which are officially indexed under human sciences, making it possible to reach
21 a wider readership (especially in economics).

22 23 *Similar Claims regarding the Discipline's Scientificity*

24 A final similarity is the use of the same discourse to justify the scientificity of the
25 new approach. The emergence of both financial economics and econophysics was
26 accompanied by particularly virulent criticism of the scientificity of existing studies.
27 An analysis of the discourse used reveals three similar justifications: the claim of a
28 more scientific approach, breaking with the past; belittling of the existing approach;
29 and the claim of greater empirical realism.

30
31 *The claim of a more scientific approach.* In each case, proponents of the new approach
32 challenged the traditional approach by asking its adepts to prove that it was scientific. This
33 "confrontational" attitude is founded upon the challengers' contention that the empirical
34 studies, the new mathematics, and methodology they use guarantee a scientificity (i.e.,
35 a way of doing science) absent from the traditional approach.³⁵ The challengers maintain
36 that the scientificity of a theory or a model should determine whether it is adopted or
37 rejected.

38 This confrontational approach was used by the early financial economists in their
39 opposition to the chartists and to financial analysts.³⁶ As an example, James Lorie
40 (Lorie 1965, p. 17) taxed the chartists with not taking into account the tools used in a
41 scientific discipline such as economics. Similarly, Fama (Fama 1965c, p. 59), Fisher
42 and Lorie (1964, pp. 1–2), and Archer (1968, pp. 231–232) presented their results as a
43 "challenge" to chartists and financial analysts. In this debate, financial economists
44 argued that their approach was based on scientific criteria, while chartism was based
45

46 ³⁵See, for instance, Lorie (1966, p. 107).

47 ³⁶These two tendencies represented the traditional approach at the time.

1 on folklore and had no scientific foundation. Consequently, financial economics should
2 supplant previous folkloric practices.

3 Another argument is based on the method used. Consider Fama's three articles
4 (Fama 1965b, 1965c, 1970). All used the same structure: the first part dealt with theo-
5 retical implications of the random walk model and its links with the efficient market
6 hypothesis, while the second part presented empirical results that validate the random
7 walk model. This sequence—theory then empirical results—is today very familiar. It
8 constitutes the hypothetico-deductive method, the scientific method defended in eco-
9 nomics since the middle of the twentieth century.

10 As with financial economics in the 1960s, the main epistemological justification for
11 the emergence of econophysics was the idea that the new approach was more scientific
12 than the old. Econophysicists claim that their approach is more neutral (i.e., not based
13 on an *a priori* model) with regard to the study of chance. They explicitly demonstrate
14 a willingness to develop models that are, on the one hand, more coherent from a
15 physics point of view,³⁷ and, on the other hand, based on “raw” observations of
16 economic systems (Stanley, Gabaix, et al. 2008). This approach is deemed more robust
17 and more scientific than the empirical studies carried out in financial economics
18 (Stanley et al. 2008, p. 3),³⁸ and, in addition, “a claim often made by econophysicists
19 is that their models are more realistic than those offered up by economists and econo-
20 metricians” (Stanley et al. 2008, p. 3).³⁹ By “physically realistic models,” the authors
21 mean that econophysicists need to be able to give a physical meaning to the statistical
22 parameters they use.⁴⁰

23 Bouchaud (2002, p. 238)⁴¹ and McCauley (2004, p. 7) further assert their position
24 by implying that physicists are the class of scientists best placed to deal with economic
25 and financial phenomena.

26
27 *Belittling the existing approach.* Justifying the emergence of a new approach involves
28 systematically calling into question the methodology employed by the mainstream and
29 using a specific vocabulary to denigrate its work.

31 ³⁷That is, that accord with the theoretical principles of modeling in statistical physics—the fact, for
32 example, that in the analysis of stationary physical systems, variance must always be finite, in accordance
33 with the thermodynamic hypotheses (concerning the concept of heat). See Gupta and Campanha (1999).

34 ³⁸In whatever econophysicists write, they seem to be influenced by an empiricist epistemology that can
35 also be seen as an implicit *a priori* about the world, directly inspired by a neo-positivist methodology
36 (Schinckus 2010b).

37 ³⁹Stylized facts are the simplified presentation of an empirical finding. They often take the form of a gen-
38 eralization of some complicated statistical calculations, which, although essentially true, may have inac-
39 curacies in the detail.

40 ⁴⁰The first reason, as we mentioned above, is the infinite characteristics of variance, which is not physically
41 plausible because the second moment is often related to the system temperature, so infinite variance would
42 imply an infinite temperature. The second reason why physicists rejected stable Lévy processes in physics
43 is that these processes are based on an asymptotical argument. To characterize turbulence phenomena
44 using stable Lévy processes, physicists had then to solve these two theoretical problems in order to fit the
45 statistical tools to the reality they study: on the one hand, they look for a finite variance; and, on the other
46 hand, they want to work in a (locally) non-asymptotical framework. In this perspective, they developed
47 stable Lévy processes. See Schinckus (2011a) for further information.

⁴¹According to Bouchaud (2002, p. 238), finance has become an “empirical (rather than axiomatic)
science.... This means that any statistical model, or theoretical idea, can and must be tested against
available data, as physicists are (probably better than other communities) trained to do.”

1 Cootner's book (1964) was one of the first publications used by the proponents of
2 financial economics to define the discipline. In his introduction, Cootner asserted that:

3 Academic studies have proven to be more sceptical about the folklore of the market
4 place than those of the professional practitioners. To several of the authors represented in
5 this volume the 'patterns' described by some market analysis are mere superstitions.
6 (Cootner 1964, p. 1)
7

8 Cootner (1964) presented the first studies of the financial economists he discussed as
9 the first scientific approach to stock-market variations, which would supplant previous
10 practices, judged to be groundless. The method employed and the empirical test of
11 hypotheses were also presented as a guarantee of the scientificity of the results.

12 Fama (1965c, p. 59) and Lorie (1966, p. 110), other emblematic figures in financial
13 economics, denigrated traditional approaches in a similar manner. Hoffland (1967, pp.
14 85–88) provided a good summary of the situation:

15 Folklore is a body of knowledge incorporating the superstitions, beliefs and practices
16 of the unsophisticated portion of a society.... Folklore is distinguished from scientific
17 knowledge by its lack of rigor.... The Dow Theory is often used as an example of a
18 crudely formulated stock market 'theory'....
19

20 Econophysicists have proceeded in like fashion. In their work, they belittle the
21 methodological framework of financial economics using similar vocabulary. They
22 describe the theoretical developments of financial economics as "inconsistent ... and
23 appalling" (Stanley et al. 1999, p. 288). Despite his being an economist,⁴² Keen (2003,
24 p. 109) discredits financial economics by highlighting the "superficially appealing"
25 character of its key concepts or by comparing it to any "tapestry of beliefs" (Keen
26 2003, p. 108). Marsili and Zhang (1998, p. 51) describe financial economics as "anti-
27 empirical," while McCauley does not shrink from comparing the scientific value of the
28 models of financial economics to that of cartoons: "The multitude of graphs presented
29 without error bars in current economics texts are not better than cartoons, because they are
30 not based on real empirical data, only on falsified neo-classical expectations" (2006, p. 17).

31 The vocabulary used is designed to leave the reader in no doubt: "scientific," "folklore,"
32 "deplorable," "superficial," "sceptical," "superstition," "mystic," "challenge."
33

34 *The claim of greater empirical realism.* Financial economists underlined the importance
35 of the empirical dimension of their research from their very first publications (Lorie 1965,
36 p. 3). They saw the testability of their models and theories as a guarantee of scientificity,
37 and concluded that "The empirical evidence to date provides strong support for the random-
38 walk model" (Fama 1965c, p. 59).

39 The same has happened in the case of the econophysicists, some of whom have
40 insisted that it is precisely because the fundamental concepts of financial economics
41 are "empirically and logically" (Keen 2003, p. 108) erroneous that a new, more "realistic"
42 form of modeling needs to be developed. Here, the term "realistic" must be understood,
43 according to econophysicists, to mean true relationship; that is, the ability to describe
44

46 ⁴²With Rosser (2006, 2008a), Keen is one of the rare breed of economists who have engaged the
47 econophysicists.

1 the true relationship governing changes in financial quotations.⁴³ This realism is,
 2 therefore, essentially *a posteriori* and in no way directed at the nature of hypotheses
 3 formulated *ex ante* (unlike economics, for example). This relationship to empiricism is
 4 very marked in econophysicists, who regularly point out that the empirical dimension
 5 is central in their work. Thus, although the “empirical data” are the same for financial
 6 economists and for physicists (financial quotations in the form of temporal series),
 7 physicists are quick to point to their “direct use of raw data,” thereby criticizing the use
 8 of statistical transformations performed by financial economists to “normalize” data.
 9 Here is Mandelbrot on this point:

10 The Gaussian framework being a statistician’s best friend, often, when he must
 11 process data that are obviously not normal, he begins by “normalizing” them ... in the
 12 same way, it has been very seriously suggested to me that I normalize price changes.
 13 I believe, quite to the contrary, that the long tails of histograms of price changes
 14 contain considerable amounts of information, and that there are a number of cogent
 15 reasons for tackling them head-on. (1997, p. 142)

16
 17 McCauley directly attacks this practice used by financial economists, explaining,

18 We [econophysicists] have no mathematical model in mind *a priori*. We do not
 19 ‘massage’ the data. Data massaging is both dangerous and misleading.... Economists
 20 assume a preconceived model with several unknown parameters and then try to force-fit
 21 the model to a nonstationary time series by a ‘best choice of parameters.’ (2006, p. 8)

22
 23 This methodological position is widespread among econophysicists, who work in
 24 the spirit of experimental physics in contrast to standard methods in economics. This
 25 empirical perspective is also justified, in the view of econophysicists, by the evolution of
 26 financial reality. The computerization of financial markets has led to better quantification
 27 of financial reality, which should now be studied as an “empirical science” (Bouchaud
 28 2002, McCauley 2004). This radical viewpoint espoused by some econophysicists has
 29 an element of naivety. Indeed, in a sense, any sampling method is a massaging of data.
 30 Nevertheless, this viewpoint has led econophysicists to a better consideration of
 31 extreme values, while such values are considered as errors by the majority of financial
 32 economists.⁴⁴ However, econophysicists seem to forget that all statistical data are
 33 embedded in a theory. By developing only physically plausible frameworks, econo-
 34 physicists also appear to have some *a priori* beliefs about the world. Indeed, for econo-
 35 physicists, there are no “abnormal data,” but only data about reality. From this point of
 36 view, whatever econophysicists write, their empiricist methodology can also be seen
 37
 38

39
 40 ⁴³Although they are mainly focused on instrumental prediction, econophysicists often claim they deal with
 41 essential relationships existing in financial phenomena (McCauley 2004).

42 ⁴⁴The way the CRSP database was created provides a good example of apriorism from financial econo-
 43 mists: “Rather than coding and punching all prices twice and then resolving discrepancies manually, we
 44 found a better procedure. We know that the change in the price of a stock during one month is very nearly
 45 independent of its change during the next month. Therefore, if a price changes a large amount from one
 46 date to a second date, and by a similar amount in the opposite direction from the second date to a third,
 47 there is a reason to believe that at the second date the price was misrecorded. A ‘large change’ was rather
 arbitrarily taken to mean a change in magnitude of more than 10 per cent of the previous price plus a
 dollar” (Lorie 1965, p. 7).

1 as an implicit *a priori* assumption about the world directly inspired by a neo-positivist
2 epistemology (Schinckus 2010b, 2011b). In a sense, no way of collecting data can ever
3 be totally neutral, because all data are necessarily the result of a specific process
4 (Schinckus 2010a).

7 III. COULD ECONOPHYSICS DOMINATE TOMORROW'S FINANCIAL 8 THEORY? 9

10 Our first part has shown that the creation of econophysics followed the same path as
11 the creation of financial economics, which subsequently came to dominate financial
12 theory. Furthermore, financial economists in the 1960s and econophysicists in the
13 1990s and 2000s justified the emergence of their approach using the same arguments.
14 The fact remains that financial economics is today a well-established scientific discipline,
15 particularly thanks to its links with economics. Econophysics, on the other hand, still
16 occupies a marginal position in modern financial theory. However, the similarity in the
17 emergence of these two theoretical approaches raises the question of whether econo-
18 physics could emerge as the dominant approach in finance, just as financial economics
19 did several decades earlier.

20 Our point is not to suggest that history repeats itself, but that the way a dominant
21 approach is created leads to the establishment of specific relations with theories,
22 models, and hypotheses (for instance, the discrepancies that will be considered as
23 anomalies). More precisely, defenders of an established approach cannot be indifferent
24 to the emergence of a new community that follows the same path to fight for its place
25 in the scientific community. This second part analyzes this question in detail, examining
26 the strengths and weaknesses of econophysics in its struggle to challenge financial
27 economics.
28

29 *Strengths that Could Help Econophysics Supplant the Mainstream Approach* 30

31 Econophysics has two main strengths to help it become dominant: it explains empirical
32 facts that are not explained by today's mainstream financial economics; and it uses a
33 mathematical framework that represents a continuation of the models used by financial
34 economists, but is more general. These two elements are crucial to the question we are
35 examining, because of the role they played in the emergence of both these scientific
36 approaches.
37

38 *Better Prediction of Empirical Facts* 39

40 Econophysics' first strong point is its ability to explain empirical facts for which financial
41 economics fails to account. As the work of Kuhn (1962) showed, the emergence of a
42 new approach can be justified by its ability to provide answers to the anomalies of
43 the established approach.⁴⁵ In order for a new approach to establish itself as the new
44 reference point for a discipline, it must not only generalize the existing results of the
45

46 ⁴⁵We use the term "anomalies" here to qualify the difference between empirical observations and theoretical
47 predictions.

1 old approach but must also better explain observed empirical facts (including those not
2 explained by the old approach).

3 This question is deeply embedded in the history of financial economics. Indeed, the
4 emergence of new empirical data and new statistical and mathematical models has
5 regularly led financial economists to transform discrepancies between predictions and
6 observations into anomalies. As we have seen, before the 1960s, statistical records of
7 stock prices were monthly data; the CRSP then began collecting daily data. This
8 changed the manner in which observations were treated. In 1978, just a few years after
9 mainstream financial economics created the efficient-market hypothesis, Jensen
10 pointed out that

11 in a manner remarkably similar to that described by Thomas Kuhn in his book, *The*
12 *Structure of Scientific Revolutions*, we seem to be entering a stage where widely scat-
13 tered and as yet incohesive evidence is arising which seems to be inconsistent with the
14 theory. As better data become available (e.g., daily stock-price data) and as our econo-
15 metric sophistication increases, we are beginning to find inconsistencies that our
16 cruder data and techniques missed in the past. It is evidence which we will not be able
17 to ignore. (Jensen 1978, p. 95).
18

19 Econophysics emerged in the wake of similar developments: the creation of new
20 data (intraday data) and the new mathematical tools (truncated Lévy processes in
21 particular) allowed more accurate observation and then transformed several discrep-
22 ancies into anomalies. Consequently, econophysicists highlight the existence of a
23 number of empirical facts that are not explained by mainstream financial economics
24 and constitute anomalies. In view of the fact that prices on financial markets change
25 more frequently and in a more orderly manner than is supposed by the Gaussian frame-
26 work on which financial economics was established, econophysicists use α -stable
27 Lévy processes directly to describe the evolution of financial data. Because econo-
28 physicits had their own theoretical goals (making Lévy processes compatible with a
29 physically plausible approach), they developed a statistical framework outside the
30 traditional approach. Doing so allows them to address a number of empirical facts that
31 the traditional approach of financial economists cannot explain because it uses a
32 Gaussian framework.⁴⁶

33 The main empirical facts to which econophysics proposes answers where financial
34 economics is unable to are “fat tails,” “volatility persistence,” and “volatility clustering.”

35 In the 1960s, during the creation of mainstream financial economics, Mandelbrot
36 (1963, 1965) and Fama (1965a) drew attention to the high number of extreme events
37 in finance and, hence, to the leptokurticity of empirical distributions, which have fat
38 tails. At the time, these authors proposed describing empirical distributions with stable
39 Lévy processes. The distributions associated with stable Lévy processes are approxi-
40 mately bell-shaped but they assign greater (than Gaussian) probability to events in the
41 center and at the ends of the tails. However, in 1965, Fama (1965a, p. 416) pointed
42 out that the infinite variance of stable Lévy processes is meaningless in financial
43 economics, since this statistical parameter is associated with the concept of risk, and
44

46 ⁴⁶Jovanovic and Schinckus (2013) explain that the Gaussian approach has played a key role in the
47 construction of the mainstream, and, until now, stable Lévy processes have not allowed such results.

1 deplored the fact that no computational definition yet existed for evaluating this
2 parameter. Financial economists did not have the statistical tools that would allow
3 them to describe empirical distributions with a stable Lévy framework. Another reason
4 why fat tails had not been studied lies in the difficulty of clearly identifying statistical
5 processes with weekly or daily data (Mitzenmacher 2004). In this respect, the avail-
6 ability of intraday data since the 1990s has favored the identification of stable Lévy
7 processes and has subsequently stimulated research into these processes. Since the
8 1990s, new statistical tools have been developed with the aim of creating stable Lévy
9 processes with finite variance. Econophysical works are directly in line with these new
10 statistical tools.

11 The second main empirical fact to which econophysics proposes answers is
12 volatility persistence.⁴⁷ In the 1970s, when financial economists created their theoretical
13 framework, they did not have the econometrical tools that could help authors to
14 identify a persistence of volatility. According to this framework, which is based on the
15 Gaussian distribution, there is no memory between stock-market returns. Starting from
16 the 1980s, econometrical tools have been proposed, with the development of
17 ARCH family models (Engle 1982) demonstrating that volatility has slowly decaying
18 autocorrelations showing a dependency between stock-market returns. Later, Schwert
19 (1989) showed this persistence of volatility in a different statistical context.

20 The last main empirical fact is volatility clustering.⁴⁸ In a Gaussian framework, one
21 could expect to see a very uniform time distribution of large and small fluctuations. In
22 1981, an original use of a martingale model led Shiller to observe several periods of
23 large fluctuations and periods of small fluctuations that are not consistent with the
24 Gaussian framework. In other words, periods of intense fluctuations and low fluctuations
25 tend to cluster together (Shiller 1981b, 1981a). Shiller's works generated much debate
26 and favored the emergence of new fields calling neoclassical finance into question—
27 behavioral finance, for example (Schinckus 2009). Econophysicists' tools make it
28 possible to put forward answers to this empirical fact.

29 To answer these anomalies, econophysicists base their analysis on Lévy processes
30 that are better adapted to the empirical data available today. Lévy's α -stable regimes
31 are processes whose accretions are independent and stationary,⁴⁹ and follow an α -stable
32 law type $\Pr \{X > x\} = x^{-\alpha}$ in which it is possible to observe constancy of the parameter α .
33 In their Paretian form, these regimes have $\alpha < 2$, and in these cases, it can be shown that
34 variance is infinite.⁵⁰ As Belkacem (1996, p. 40) emphasized, "from a practical point of
35 view, stable distributions are able to explain the thick distribution tails observed in
36 empirical distributions of asset profitability rates." A similar argument is found in the
37 writings of Tankov (2004, p. 13), who added that these processes are particularly inter-
38 esting in financial economics because they allow discontinuities in the evolution of asset
39

41
42 ⁴⁷Clegg (2006) discusses the integration of this empirical fact by econophysics in greater detail.

43 ⁴⁸See Sornette (2003) for a precise presentation of the response to this anomaly proposed by
44 econophysicists.

45 ⁴⁹The "stationary" character means that the process causing price variations remains the same over time,
46 but it would be erroneous to associate this stationary character with continuity of the process. This is what
47 Mandelbrot pointed out (1997, p. 138) in discussing this link between discontinuity and stationariness.

⁵⁰For $\alpha < 1$, the variance but also the mean are also theoretically infinite (or undefined).

1 returns to be taken into account. This means that these processes are candidates for
2 explaining the leptokurticity of financial data.

3 The analysis of these empirical facts proposed by econophysicists is essentially
4 empirical itself, and points to econophysics' predictive capacity. Above all, the solutions
5 allow a better statistical understanding of the phenomenon of financial data variability.
6 As pointed out by Shiller (1981b), this phenomenon has been underestimated by
7 neoclassical finance's theoretical framework.

8 The integration of empirical facts unexplained by financial economics into the
9 (more generalized) theoretical framework developed by econophysicists is, thus, an
10 argument in favor of a shift in mainstream financial theory.

11 *A More General and More Complete Probabilistic Framework*

12 Econophysics' second strong point in its bid to become the dominant approach is the
13 use of statistical models that generalize those used in financial economics.

14 Financial economists mainly use the Gaussian framework in order to characterize
15 financial uncertainty. Four reasons can be cited to explain the success of the Gaussian
16 framework: the historical development of financial economics (Jovanovic 2008,
17 Jovanovic et al. 2013),⁵¹ simplicity (only two parameters are needed to describe data),
18 the notion of normality (which can refer to the key concept of economic equilibrium),
19 and, above all, statistical justification, which refers to one of the most fundamental
20 theorems⁵²: the central-limit theorem (CLT), which states that the sum
21
22

$$23 Z_n \equiv \sum_{i=1}^n x_i$$

24
25
26
27
28 of n stochastic variables x that are statistically independent, identically distributed,⁵³
29 and with a finite variance converges when $n \rightarrow \infty$ to a Gaussian stochastic process
30 (Feller 1971).

31 For econophysicists, the Gaussian framework is the first step towards describing
32 uncertainty in science. This first step can be generalized to a savage uncertainty "without
33 normality" (Mandelbrot 1997, p. 66). This generalization is based, on one hand, on
34 Lévy's work (1924) on random processes, and, on the other hand, on the generalized
35 central-limit theorem developed by Gnedenko and Kolmogorov (1954). In accordance
36
37

38
39 ⁵¹Jules Regnault, in 1863, was directly influenced by Adolphe Quételet's work on the application of normal
40 distribution to social phenomena (Jovanovic 2001, 2006). Bachelier (1900), whose work was clearly influ-
41 enced by Regnault's (Jovanovic 2000, 2009b, 2012), retained a Gaussian description of the evolution of
42 variation in the price of assets for demonstrating the equivalence between the results obtained in discrete
43 time and in continuous time. Similarly, all the empirical work that emerged from the 1930s onward (Cowles
44 1933, Working 1934, Cowles and Jones 1937, Kendall 1953) used this Gaussian framework because at the
45 time it was impossible to use other kinds of statistical distribution. Indeed, all non-Gaussian observations
46 and "white noise" were characterized through a Gaussian standardization.

47 ⁵²The first fundamental theorem is The Law of Large Numbers (Tijms 2004).

⁵³A generalization of the theorem for variables that are not identically distributed (but always with a finite
variance) was developed by Gnedenko and Kolmogorov (1954).

1 with this generalization, the sum of random variables following Lévy laws, distributed
 2 independently and identically, converge towards the stable Lévy law having the same
 3 parameters. This generalization of the central-limit theorem justifies and provides a
 4 statistical foundation for the use of Lévy laws to characterize complex phenomena.⁵⁴
 5 Lévy's work proposed a generalization of several known distribution laws in the form
 6 of a family of random variable distributions notated as $S_{\alpha,\beta}(\mu,\gamma)$.⁵⁵ By basing their
 7 approach on Lévy processes, econophysicists propose models that encompass all
 8 stochastic processes, such as Gaussian processes and Poisson processes.

9 By making use of stable Lévy processes, econophysics offers a more general
 10 theoretical framework than financial economics, which uses Gaussian distribution.⁵⁶
 11 From this perspective, the models of financial economics are merely particular cases,⁵⁷
 12 and the statistical tools used by econophysicists make possible technical integration of
 13 the Gaussian statistical framework on which mainstream financial economics is based.
 14 As a generalization of the Gaussian framework, the α -stable framework retains the
 15 fundamental properties (fractality and auto-affinity) of the Gaussian framework. There
 16 is, thus, mathematical continuity in the models used by financial economists and those
 17 used by econophysicists. This continuity had been advanced in the 1960s by Mandelbrot
 18 and Fama at the very beginning of the creation of mainstream financial economics.
 19 This attempt was not developed because of the theoretical meaninglessness of the
 20 hypothesis of infinite variance. Several paths have been explored by financial economists
 21 since the 1970s but none dropped the Gaussian framework (Jovanovic et al. 2013).⁵⁸
 22 Today, the evolution of mathematics developed by econophysicists allows the develop-
 23 ment of a Markowitz portfolio theory, a generalized CAPM in an α -stable framework
 24 (Belkacem 1996, Tankov 2004), and the development of a Black and Scholes option-
 25 pricing model in an α -stable framework (Huang and Wu 2004). Such a characteristic
 26 would also make it possible to preserve the concept of market efficiency, even if new
 27 risk-analysis parameters were to be introduced. As Fama pointed out (Fama 1965c),
 28 the concept of efficiency did not necessarily imply the Gaussianity of accretions.

29 The generalization used today by econophysicists is, thus, a strong point. Moreover,
 30 it renders the first attempts of financial economists such as Fama (1963) and Mandelbrot
 31 (1965) viable. These first attempts have been enhanced by the purely mathematical
 32 framework developed by Harrison and Pliska (1981). Indeed, by proposing a non-
 33 economic model in finance, Harrison and Pliska provided an entry point for an evolution
 34 of modern finance less based on economics. As a result, from a mathematical viewpoint,
 35 the models developed by econophysicists continue from where those developed by
 36 financial economists left off. The challenge for the future of econophysics will be to
 37
 38

39
 40 ⁵⁴The central-limit theorem, according to historians of statistics, provided the main argument for the use of
 the Gaussian framework for the study of varied phenomena.

41 ⁵⁵As Schoutens (2003) makes clear, the family of α -stable distributions identified by Lévy appears to be the
 42 general form of a number of statistical laws known as the normal law $S_{2,0}(\mu,\gamma)$, Cauchy's law $S_{1,0}(\mu,\gamma)$,
 43 Pareto's law $S_{3/2,\beta}(\mu,\gamma)$, and Lévy's law $S_{1/2,1}(\mu,\gamma)$.

44 ⁵⁶To allow comparison of the various models developed by econophysicists, Bucsa, Jovanovic, and
 45 Schinckus (2010) propose a generalized formula that can be used to unify the work of these research fields.

46 ⁵⁷Economists and financiers have long been interested in the leptokurtic nature of price distributions. See
 Louçã (2007, p. 219), Jovanovic (2009b), and Jovanovic et al. (2013).

47 ⁵⁸For instance, Merton's (1976) model on jump processes.

1 show that the models proposed by econophysicists can be integrated into this mathe-
 2 matical framework developed by Harrison and Pliska.⁵⁹

3
 4 *Limitations Hindering Econophysics from Becoming the Dominant Approach in*
 5 *Finance*

6
 7 As we have explained, econophysics was born outside financial economics, and it is
 8 today achieving institutional status with its own prestigious, intellectual actors, awards,
 9 journals, conferences, academic programs, and departments. This last section explains
 10 that in spite of all this, the results, models, and hypotheses of econophysics have so far
 11 failed to break into the mainstream of financial economics, even if, since the 1990s,
 12 financial economists have progressively begun to publish papers related to Lévy
 13 processes.⁶⁰

14 Indeed, although econophysics has undeniable advantages to help it establish itself
 15 as the new dominant approach in finance, two major limitations work against this
 16 eventuality: first, the virtual absence of discussion between financial economists and
 17 econophysicists; and, second, econophysicists' non-use of financial economics and
 18 mathematical-finance hypotheses.

19
 20 *The Virtual Absence of Discussion between Financial Economists and Econophysicists*

21 Although econophysicists explicitly position themselves in relation to financial
 22 economics, it has to be noted that the response of economists to their criticisms has
 23 been, for all intents and purposes, non-existent. Some economists have pointed out the
 24 limitations of the econophysics approach (Gallegati et al. 2006),⁶¹ provoking a virulent
 25 response from one of the leading lights of econophysics (McCauley 2006). Aside from
 26 this brief exchange of communication, real theoretical debate between econophysicists
 27 and economists has so far seemed difficult. The sometimes severe criticisms made
 28 by econophysicists of the models and hypotheses used in financial economics have
 29 apparently failed to convince economists of the need to engage in theoretical discus-
 30 sions. One feature in particular explains the problem: econophysics' framework is not
 31 directly compatible with that of financial economics, and econophysicists do not take
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 34
 35 ⁵⁹As we mentioned earlier, none of the studies by financial economists dealing with Lévy processes
 36 focused on stable Lévy processes or had any connection with the methodology used by econophysicists.

37 ⁶⁰It is essential to point out that Lévy processes include many classes of stochastic processes, such as the
 38 Wiener process, jump-diffusion processes, and jump stable Lévy processes (Variance Gamma Process,
 39 Madan 1990; Generalized Hyperbolic Process, Eberlein 1995; CGMY process, Carr et al. 2000). As a
 40 result, economists and econophysicists may use the same word for quite different models. While econo-
 41 mists use the term "Lévy process" to define (non-stable) jump-diffusion and (non-stable) jump stable Lévy
 42 models, econophysicists use this word to define stable Lévy processes. Moreover, all papers dedicated to
 43 Lévy processes in financial journals have been written by mathematicians or financial economists and not
 44 by physicists. The semantic difference is a source of debates between econophysicists and economists. The
 45 first often claim that they offer a new perspective on finance, while the latter consider that this approach is
 46 an old issue in finance. The point is that these two categories of specialists are not talking about the same
 47 thing: econophysicists use stable Lévy processes, while economists abandoned this kind of processes in the
 1970s (Jovanovic and Schinckus 2013).

⁶¹This critique was published in a physics journal by economists who do not themselves subscribe to the
 mainstream approach of financial economics.

1 this point into account. Indeed, because econophysicists' focus is on mathematical
2 development, very few of them have attempted to make their models compatible with
3 the framework and hypotheses of financial economics.⁶² One who has made the attempt
4 to connect econophysics with financial economics is Bouchaud—see, for instance,
5 Bouchaud and Potters (2003).

6 As a result, the leading economics and finance journals make very few references to
7 the work of econophysicists. Gingras and Schinckus (2012) looked at the ten leading
8 authors in econophysics, between 2002 and 2008, and found 401 references to these
9 authors in the main economics and finance journals as against 2506 in the main physics
10 journals. This disparity highlights the fact that econophysics is not considered the
11 central issue in economic journals.⁶³

12 One might assume that, rather than economists “rejecting” econophysicists, it is
13 more a case of the latter’s having developed their models outside the financial
14 economics’ framework, and, hence, considering that they had no need to publish in
15 mainstream economic journals or adapt their methodology and viewpoints to those of
16 financial economists. With the evolution of the two disciplines as analyzed in the first
17 part of this paper, this is no longer the case. First, we have shown that econophysicists
18 have created new journals (*Quantitative Finance* and *JEIC*) in economics (and not in
19 physics) in order to reach financial economists and an audience outside econophysics.
20 Second, since econophysicists work on the same phenomena as economists, we should
21 expect them to attempt publication in economics journals. To test this hypothesis, we
22 conducted an informal survey, sending a questionnaire to twenty-seven leading econo-
23 physicits (included as source authors in our analysis) about the degree of openness of
24 economics journals to econophysicists. To the question “Have you submitted a paper
25 to a ranked journal in economics?”, a large majority of authors replied “yes.” When
26 authors were asked to give the main reasons⁶⁴ why their paper was rejected, they
27 replied that referees in economic journals often have difficulties with the topic or/and
28 the method used in their paper. Although based on a small sample (but including the
29 central figures of econophysics), these results strongly suggest that economic journals
30 are indeed reluctant to publish papers dedicated to econophysics. The lack of discussion
31 between the two approaches could explain this situation, because no bridges are built
32 between them.

33 Nevertheless, in 2008, the *Journal of Economic Dynamics and Control* published a
34 special issue entitled *Applications of Statistical Physics in Economics and Finance*.
35 Doyne Farmer and Thomas Lux⁶⁵ were guest editors for this special issue, articles
36 for which were written by economists and physicists. This special issue aimed to
37 “overcome the lack of communication between economists and econophysicists”
38

40 ⁶²In addition, as Rosser (2008b, p.10) points out, econophysicists’ models “are only of exchange and do not
41 allow for production, and they often confuse basic concepts such as transactions and income, this latter
42 especially for models of income and wealth distribution.”

43 ⁶³For example, only three references to key writers on econophysics are found in the leading journal of
44 financial economics, the *Journal of Finance* (source: *Web of Science*).

45 ⁶⁴They had to choose between five reasons for their rejection and were invited to comment on their choices:
46 1) the topic of the paper, 2) the assumptions used in the paper, 3) the method used in the paper, 4) the results
47 of the paper, or 5) other reasons.

47 ⁶⁵The first is a physicist and the second an economist; both were among our source authors.

1 (Farmer and Lux 2008, p. 3). As these authors pointed out in their editorial, there is “a
2 lack of communication between physicists and economists. Physicists are perhaps the
3 only group of scientific professionals who are even more arrogant than economists,
4 and in many cases the arrogance and emotions of both sides have been strongly on
5 display” (Farmer and Lux 2008, p. 3). In order to overcome the gap between the two
6 camps, this special issue published twelve articles dedicated to econophysics, written
7 by authors from the fields of economics and physics. Despite this first attempt at
8 dialogue, debate has still not yet really begun.

9 Among the few economics journals to publish articles about econophysics is
10 *Journal of Economic Behavior & Organization*, which recently broadened its editorial
11 line to encompass the problems of complexity in economics (and consequently
12 published several studies dedicated to econophysics). Other journals officially indexed
13 under economics that regularly publish articles on econophysics include *Quantitative*
14 *Finance*, created in 2002, and *Journal of Economic Interaction and Coordination*,
15 created in 2006. The creation of these two journals, both of which have an editorial
16 team made up largely of econophysicists, appears to be a strategy for the dissemination
17 of econophysics, the aim being to make the approach better known to an audience of
18 economists (Gingras et al. 2012).

19 The lack of dialogue can also be connected with the closed nature of economics.⁶⁶
20 Whitley (1986b) portrayed economics as a “reputationally controlled work organization”
21 characterized by a strong and monolithic standardization of research. He explained
22 that economics is a “partitioned bureaucracy” whose segmentation into several
23 subfields allows it to marginalize all anomalies or empirical contradictions of the
24 mainstream. Economics appears as a conservative novelty-producing system since it
25 rewards intellectual innovation only if it is directly in line with the dominant research.
26 All new fields that are not in accordance with the scientific standards used by the
27 mainstream are simply ignored. In this perspective, the conceptual basis of econo-
28 physics could come only from outside the field of economics and be promoted by
29 physicists who saw that the kind of distribution behind economic and financial
30 phenomena, which are a collective response of the interactions of a large number of
31 agents, are analogous to the distributions observed in condensed-matter physics as
32 the result of the collective interactions of a large number of atoms. Starting from that
33 analogy, they applied the methods of statistical mechanics, which explain the emergence
34 of these distributions, to the case of economic and financial behavior. Such a move
35 radically transforms the understanding of economics, as the usual Gaussian framework
36 is replaced by a Lévy framework whose statistical properties are very different and are
37 not necessarily consistent with the conceptual foundation of mainstream economics,
38 based on equilibrium. Moreover, while economic theory is based on an atomistic
39 reductionism in which reality must be explained in terms of rational representative
40 agents, econophysics focuses on the interactions that give rise to complex phenomena

45 ⁶⁶Pieters and Hans (2002) explored intra- and interdisciplinary communication of economics journals by
46 means of citations analysis. They showed that the first tier of economics journals did not cite articles
47 published in journals of management, marketing, anthropology, or psychology between 1995 and 1997.

1 that can be described through interactions between their parts.⁶⁷ These conceptual
2 differences, coupled with the difference in disciplinary training between economists
3 and econophysicists, have contributed to the development of econophysics as a
4 separate scientific culture whose roots stayed in physics instead of developing out of
5 economics, like other new specialties such as behavioral finance and experimental
6 economics.

7 There is, thus, a dynamic of repulsion between economics and econophysics: as
8 Whitley observed, “economics has a strong hierarchy of journals” (Whitley 1986b,
9 p. 192), and researchers who do not conform to the dominant standards are bound
10 to publish outside that core and, therefore, be seen as irrelevant to the core, and conse-
11 quently forced to publish in new journals not recognized by the mainstream of the
12 discipline. Whereas this tendency could have given rise to a new speciality among the
13 social sciences, the fact that the tools of econophysics were imported from physics,
14 and that econophysics was prepared to accept the modeling of social phenomena as
15 legitimate—thus enlarging its scope and possible job market—brought econophysics
16 under the wing of physics as it first grew by using existing physics journals for
17 publication and physics departments as training ground for the new breed of “econo-
18 physicist” (Gingras et al. 2012). We are seeing the emergence of a veritable scientific
19 community independent of financial economics with its own academic courses, its
20 own symposiums, and journals that are recognized in the field of physics.

21 In Whitley’s view, the lack of openness displayed by economics is designed to
22 preserve the dominant theoretical framework and to marginalize anomalies that are
23 likely to challenge the discipline: “as long as the theoretical establishment is able to
24 dismiss ‘anomalies’ and difficulties as peripheral and the province of ‘applied’ sub-
25 fields and yet retain control of the assessment of research competence in all areas,
26 fundamental change seems improbable” (Whitley 1986a, p. 204). This closed attitude
27 in economics considerably restricts the scope for econophysics to win over financial
28 economists. Moreover, physics shares the same closed-mindedness: according to the
29 *Science and Engineering Indicators* (2000, p. 103, table 6–54), economics is the most
30 hermetic field of the social sciences, with more than 87% of intra-disciplinary references
31 compared to 50% in sociology. This is even more self-contained than physics, which
32 cite physics journals in about 80% of their references. These data are consistent with
33 Whitley’s (1986b) characterization of economics as a “partitioned bureaucracy”
34 with a strong control over its theoretical core. Consequently, econophysicists have
35 developed a closed attitude about financial economics.

36 *The Refusal of Econophysicists to Incorporate the Framework of Financial Economics*

37 As we have explained, from a statistical point of view, econophysics makes it possible
38 to generalize the models used by the proponents of financial economics. However, up
39 to now, most econophysicists have rejected the framework of financial economics
40 to now, most econophysicists have rejected the framework of financial economics
41

42
43 ⁶⁷Econophysicists consider that particles (individuals) do not interact equally with each other because
44 interactions depend on the distance between the particles. Therefore, as a function of their positions in the
45 system, particles will interact and create different structures (molecules, crystals, etc.). The system will
46 then be self-evolving and complex. Because atoms do not think, econophysicists consider that “market
47 components” (including traders, speculators, and hedgers) obey statistical properties. See Schinckus
(2010a) for a presentation of the main differences between economics and econophysics.

1 because they consider it “too axiomatic and formal to deal with complex systems”
 2 (Challet, Marsili, et al. 2005, p. 14). Even some economists (Keen 2003, p. 110) claim
 3 that all the key concepts (utility, perfect rationality, perfect competition, etc.) used in
 4 financial economics are “nonsense.” They are unobservable terms without an empirical
 5 base. Indeed, despite the fact that experiments exist in economics (Holt and Davis
 6 2005), all key concepts of economics cannot be directly confirmed because, according
 7 to econophysicists, these concepts result from apriorism. In this perspective, the key
 8 notions of economics are considered as “empirically flawed” (Keen 2003, p. 109). For
 9 instance, the existence of equilibrium, which is a keystone of financial economics, has
 10 no foundation for econophysicists. In econophysics, equilibrium is rather considered
 11 as a potential state of the system because “there is no empirical evidence for equilibrium”
 12 seen as a final state of the system (McCauley 2004, p. 6). For econophysicists,
 13 economic equilibrium appears as an *a priori* belief⁶⁸ that provides a “standardized
 14 approach and a standardized language in which to explain each conclusion” (Farmer
 15 and Geanakoplos 2009, p. 17).

16 From a purely econophysics perspective, the financial market, like physical bodies,
 17 moves through different “phases,” which likewise display chaotic or coherent states.
 18 The evolution of phases and phase changes are, therefore, represented by a trajectory
 19 in this space. In accordance with chaos theory, physicists observe that, if one waits
 20 long enough, systems of this type tend to move through states that are neighboring or
 21 comparable to those they have been through in the past. In phase-diagram terms, the
 22 system is said to tend towards what is known as a “strange attractor.” This strange
 23 attractor cannot be assimilated into an equilibrium in the strict meaning of the term;
 24 rather, it is a geometric zone through which the system passes regularly.⁶⁹ The concept
 25 of strange attractor, taken directly from chaos theory, neatly sums up the idea that
 26 econophysicists have of the concept of equilibrium—which, in general, is explicitly
 27 rejected. “There is no empirical proof of the existence of a stable equilibrium,” explains
 28 McCauley (2004, p. 6). In his view, the importance placed on this notion is more the
 29 result of an ideology or a belief than of an observation of reality (McCauley 2004,
 30 p. 295). He goes so far as to use the idea of non-equilibrium to distinguish finance
 31 theory from economics:

32 Standard economic theory and standard finance theory have entirely different origins
 33 and show very little, if any, theoretical overlap. The former, with no empirical basis
 34 for its postulates, is based on idea of equilibrium, whereas finance theory is motivated
 35 by, and deals from the start with, empirical data and modeling via nonequilibrium
 36 stochastic dynamics. (McCauley 2004, p. 6)

38 Mandelbrot (2005, p. 143) adds:

39 Let us return to fluctuations in finance; where do we see an economic equilibrium that is
 40 the equivalent of normal thermodynamic equilibrium? I quickly developed the feeling that
 41 the notion of economic equilibrium is devoid of content and that, to describe price
 42 variation, it will not be sufficient to modify benign chance by incorporating new details.
 43
 44
 45

46 ⁶⁸See Schinckus (2011a) for further information about the importance of equilibrium in econophysics.

47 ⁶⁹This term remains to be defined on the basis of the nature of the phenomenon under study.

1 McCauley (2004, p. 78) explains that the concept of equilibrium used in economics
2 is of Newtonian inspiration (classical mechanics), whereas the idea of non-equilibrium
3 owes more to the logics of statistical mechanics. From this point of view, investors are
4 particles with complex and heterogeneous behavior, operating in a system (the market)
5 whose macroscopic state can be characterized statistically (by Lévy laws) by a set of
6 values (asset prices) that are transitory (non-equilibrium). “This low-level complexity
7 (which is in a way microscopic) [at the individual level] can, under certain conditions,
8 cause surprising and not disorderly effects at the macroscopic (collective) level”
9 (Brandouy 2005, p. 122). This macroscopic perception of non-equilibrium as the sole
10 explanation of the system’s microscopic states is directly inspired by thermodynamic
11 models⁷⁰ and not by economics. This difference between economics and physics is
12 also highlighted by Ruelle (1991, p. 113), and suggests that the notion of non-equilibrium
13 could prove highly useful for the study of certain economic phenomena. Ball
14 (2006, p. 687) adds that “equilibrium is the heart of the dominant economics whereas
15 most models taken from econophysics are explicitly based on the concept of nonequi-
16 librium.” The rejection of the theory and hypotheses of economics does not facilitate
17 theoretical exchanges, and contributes to the absence of debate alluded to in our
18 previous section. The studies put forward by econophysicists are not presented as
19 improvements but rather as new models to replace current economic models purely
20 and simply. Economists, for their part, prefer to remain aloof from this “methodological
21 imperialism” by publishing no (or few) articles on econophysics (Gingras et al. 2012).
22 This rejection of the theoretical framework of financial economics leads econo-
23 physycists to discard the discipline’s main concepts and theories, as McCauley points
24 out: “Econophysicists are safer to ignore the lessons taught in standard economic texts
25 (both micro-macro) than to learn the economists’ production ideas and take them
26 seriously” (2006, p. 608).

27 28 29 IV. CONCLUSION 30

31 In the 1960s, Mandelbrot (1963, 1965) and Fama (1965a) emphasized the leptokurticity of
32 empirical distributions and proposed describing the empirical distribution with stable
33 Lévy processes. Although these processes appeared to be adequate for the statistical
34 description of financial distribution, they were not developed at that time because they
35 have infinite variance, and financial economists did not have the statistical tools to
36 allow them to describe empirical distributions with a stable Lévy framework.

37 Since then, financial economics has attempted to integrate Lévy processes into its
38 modeling (Matacz 2000, Tankov 2004). However, these attempts were founded on
39 non-stable Lévy processes. In the 1990s, new statistical tools were developed with the
40 aim of creating stable Lévy processes with finite variance. Econophysical works
41 are directly in line with these new statistical tools. That is the point explained in this
42 article, which has also shown how the development of a new way of collecting data
43

44 ⁷⁰That is, a dynamic equilibrium in perpetual modification. There have also been attempts to unify the concept of equilibrium in microeconomics and in thermodynamics—see Sousa and Domingos (2006) on this subject. Nevertheless, these analogies are purely “axiomatic and conceptual” (Sousa et al. 2006, p. 162).
45
46
47 They have not, for the time being at least, led to the study of particular economic or financial phenomena.

1 (intraday) has favored the emergence of econophysics. In a sense, this new field can
 2 be seen as the result of the technological and theoretical evolution of finance. This
 3 evolution was initiated by the first attempts of financial economists, but also by the
 4 purely mathematical framework developed by Harrison and Pliska (1981), who
 5 proposed a non-economic model in finance. Harrison and Pliska, in fact, provided an
 6 entry point for a less economics-based evolution of modern finance. As a result, from
 7 a mathematical viewpoint, the models developed by econophysicists continue from
 8 where those developed by financial economists left off. In one respect, econophysics
 9 has benefited from standing outside financial economics: this position means that they
 10 are not subject to the constraints of the theoretical framework, giving them greater
 11 freedom to develop new mathematical tools. And yet, the evolution of financial
 12 economics suggests that econophysicists might successfully infiltrate financial
 13 economics and, consequently, become the next dominant approach in financial economics.
 14 Recent work by financial economists to integrate Lévy processes could build bridges
 15 between the two approaches. These attempts to integrate Lévy processes into the
 16 framework of financial economics are a challenge for that discipline.

17 This paper also shows how econophysics and financial economics have historical
 18 similarities in their emergence process. In the 1960s, chartists and fundamentalists
 19 were marginalized in academic circles by financial economists. Since the 1990s, financial
 20 economists have been challenged by econophysicists, who suggest that they can
 21 supplant them using the same arguments that they themselves used in the 1960s. In
 22 other words, econophysicists are suggesting that their assumptions, their theoretical
 23 framework, and their results could serve as a reference point, with the result that
 24 current mainstream financial economics will position itself in relation to them. This
 25 attempt to supplant the dominant approach by the new theoretical field, combined with
 26 the importance of statistical tools, indicates some historical similarities between the
 27 emergence of both fields.

28 However, despite these similarities, we have highlighted aspects in econophysics'
 29 theoretical framework, results, and hypotheses that are obstacles to its bid to become
 30 the mainstream approach in financial economics.

31 While econophysics is an important challenge for financial economics, we observe
 32 a major difference between financial economics in the 1960s and econophysics today.
 33 Financial economists took over the business schools by marginalizing rival groups.
 34 This is not the case with econophysicists, who do not seem to be able to marginalize
 35 the dominant approach in financial economics. Rather, they appear to be attempting to
 36 carve out a place for themselves inside modern financial theory. However, to strengthen
 37 this position, bridges still need to be built between financial economics and econo-
 38 physics; but that is another story.

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