Applications of complex systems reasoning in economics

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June 23, 2010

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Micro/Macro Fluctuations 1

- The U.S. economy is composed of about 300 million individuals, 100 million households and 25 million firms (http://quickfacts.census.gov/qfd/states/00000.html)
- +In an system in which aggregate fluctuations are the sum of independent fluctuations of components, we would expect very smooth macroeconomic indicators

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Micro/Macro Fluctuations 2

- In reality, as we know, the macroeconmy is subject to significant fluctuations which we call the business cycle
- ◆ Real GDP, employment, price levels and rates of inflation, profit rates, incomes and income distribution have business cycle components

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Micro/Macro Fluctuations 3

- *Statistically business cycle fluctuations reflect correlated movements of corresponding micro-level variables
- One basic puzzle of social science methodology is to explain micro correlations
- Conventional equilibrium theory relies on "external shocks"

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Micro/Macro Fluctuations 4

- *Conventional equilibrium theory sees all fluctuations as giving rise to appropriate negative feedback signals through market-clearing prices
- ★As a result conventional equilibrium systems, though they are adaptive and have many degrees of freedom, are not complex systems in the usual sense

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Micro/Macro Fluctuations 5

- The commodity form of productive organization does give rise to stabilizing negative feedback through price signals
- → If markets are not "complete", the system as a whole will exhibit the characteristic mixture of positive and negative feedbacks of complexity

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Emergent properties 1

- *An important second clue to complex phenomena in economies is the statistics of distributions of wealth and income
- +The original discovery of scaling or power laws characteristic of many complex systems was due to Pareto's research on wealth distributions

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Emergent properties 2

- *Conventional equilibrium theory takes the original ownership distribution of productive resources (land, labor, and capital) as exogenous
- ◆This means either that it is determined arbitrarily outside the system, or that some random process determines ownership distribution

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Emergent properties 3

- *It is thus very hard to rationalize the statistical regularities of income and wealth distributions within the framework of equilibrium theory
- → But these statistical regularities persist
- → Statistical self-organization in scaling distributions is a common characteristic of complex systems

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Emergent properties 4

- There seem to be two self-organizing processes in income and wealth
- ◆Earned income (employee compensation) has an exponential form
- Property income (profits, interest, rent) has a power-law form

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Emergent properties 5

- *Both exponential and power-law distributions are maximum-entropy distributions
- ◆Exponential distributions arise from an additive constraint on the mean
- → Power-law distributions arise from a multiplicative constraint on the logarithm of the mean

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Emergent properties 6

These components of income and wealth distributions reflect the classical distinction between labor mobility tending to equalize levels of wages (additive) and capital mobility tending to equalize profit rates (multiplicative)

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Long-distance correlations 1

- *Complex systems exhibit a slow decay of autocorrelation functions in space and time
- + Shocks to one part of the system propagate with low decay rates throughout the system
- ◆The "butterfly effect" is a dramatic example of this phenomenon

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Long-distance correlations 2

- *Slowly decaying autocorrelations lead to unexpected macro-level causal chains and to unpredictable time series behavior in economic data
- ◆An example of macro links is the complex supply network linking U.S. markets with Chinese and other suppliers

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Long-distance correlations 3

- *Growing financial fragility is another manifestation of these slowly decaying correlations
- Increases in leverage in one part of the financial system propagate through the system, changing its stability
- Sudden crises in liquidity and credit erupt unpredictably

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Heavy tails 1

- Weak decay of autocorrelations also offers a promising explanation of the heavy tails in distributions of price and output fluctuations
- ◆ Conventional statistical methods tend to assume Gaussian distributions which predict a very rapid decay of probability with size of fluctuation

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Heavy tails 2

- The premise on which Gaussian probability rests is that fluctuations are the sum of a large number of weakly correlated shocks
- When autocorrelations decay slowly, shocks become significantly correlated
- Thus large fluctuations are more probable

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Heavy tails 3

- The crisis of conventional economic equilibrium theory is also a crisis of econometrics
- ◆The presumptions of equilibrium theory (external, damped shocks) dovetail with the axioms of econometrics (Gaussian disturbances) to make the economy seem simpler and more stable

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Heavy tails 4

- The complex systems vision of the economic system warns against this error, but does not offer an easy remedy
- There are no general theorems on slowly decaying correlations as widely applicable as the central limit theorem

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Regime change 1

- *Conventional economic equilibrium theory reasons from the axiom of complete and liquid markets
- Under these assumptions, market dynamics are qualitatively homogeneous despite fluctuations
- → Real-world dynamics exhibit striking inhomogeneities over long time scales

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Regime change 2

- In "normal" periods the overnight market for Federal Funds (bank reserves) clears at a relatively high nominal interest rate
- +Adjustment of the supply of reserves promptly and predictably moves the Federal Funds rate

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Regime change 3

- *But in "crisis" periods when financial institutions hoard reserves there is a qualitative change in this market to "liquidity trap" dynamics
- The Federal Funds market produces an interest rate barely above zero
- The provision of more reserves has practically no impact

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Regime change 4

- *Regime changes of this kind are met with frequently in complex systems
- → The system evolves with some relatively stable dynamics for a possibly long period and then endogenously undergoes a phase transition to a qualitatively different type of dynamics

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Coordination equilibria 1

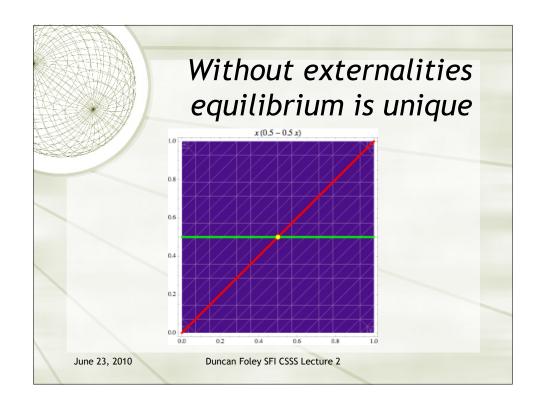
- The complex dynamics of the macroeconomy arise from social interactions that are not completely mediated by the market
- ◆ Examples include the instability of credit provision, and the dynamics of spending and aggregate demand

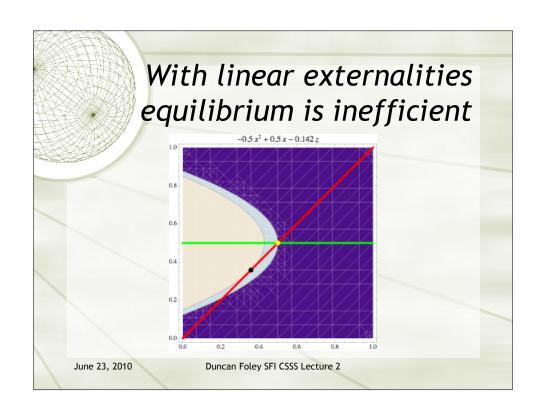
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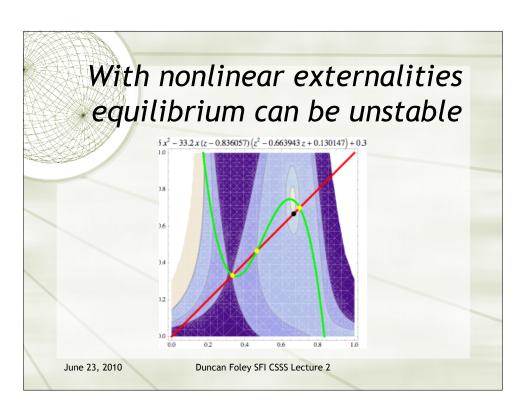
Coordination equilibria 2

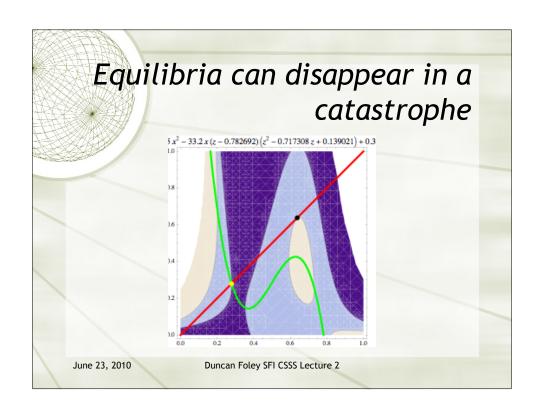
- *A simple model for coordination problems makes the decision of each agent, x, depend on the average decision of all the (identical) agents, <x>
- ◆This is similar to the mean-field approximation that simplifies complex systems analysis

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Nonstationary paths of technical change

- *Complex systems generate nonstationary statistical paths, in contrast to Gaussian perturbation systems
- ◆ Economic technical change is irreversible and path-dependent, giving rise to nonstationary statistics

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Complexity in social and physical systems 1

- *Many of the insights we get from studying complex physical systems shed light on puzzles in social and economic systems
- Much of this illumination simply sweeps away cobwebs created by presumptions that systems have unique, stable equilibria

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Complexity in social and physical systems 2

*Once we are prepared to acknowledge the economy (and more broadly society) as a complex, adaptive system far from equilibrium many phenomena, from crises to unexpected patterns of long-run technical change, become qualitatively understandable

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Complexity in social and physical systems 3

- *In complex biological and physical systems the main shaping force over time is evolutionary selection
- +The system tends to explore all the degrees of freedom available to it
- ◆Those that are self-limiting disappear; those that are self-reinforcing survive

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Complexity in social and physical systems 4

- In human social systems there is an additional level of complexity: the intentional and purposeful behavior of human beings
- ◆ During the 18th century Enlightenment it became a cliche to note the importance of "unintended consequences" of human action

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Complexity in social and physical systems 5

- The fact that human intentions are never realized, however, does not mean they can be neglected in social science reasoning
- → We have legal systems not just from evolutionary selection, but because people have an abstract idea of justice they seek to institutionalize

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Complexity in social and physical systems 6

- *Complex economic systems like energy generation and distribution, transportation, financial services combine planning with blind selection
- The human individual is a paradigmatic complex system; human society is the interaction of an enormous number of human individuals

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