A Radical Shift in Managing Risks: Practical Applications of Complexity Theory

Chichilnisky, Graciela

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You can't use complexity theory to predict hurricanes or other climatic events, but if you're smart you can use it to hedge your bets on the future.

Complexity theory is designed to bring order out of a rough-and-tumble world, something close to every insurance professional's or actuary's heart. Whether applied in the laboratory or as part of a mathematical model, it can do wonderful things. But in the real world it's just what its name says—a theory. It may help us understand why things go one way or another, but it can't foretell those things with such certainty that an insurance company could, say, set its rates based on the knowledge that only one hurricane will strike Florida during the coming year.
Yet I see a connection between complexity theory and risk management, which is the essence of insurance. In particular, I've focused on the risk management of complex systems.

Admittedly, the prediction of complex systems is inherently difficult, even impossible, according to some prominent writers (see sidebar on p. 30). But I want to take a position closer to analysts who say that, if you're smart enough, you can sometimes discern patterns, even in chaos. There are situations when you can actually make predictions, providing you have the right instruments. So what's needed for better risk management is something in the nature of a gestalt shift and, in particular, financial innovation.

In short, what's needed is a new method of predicting. And here the function is not to know the future (though people may tell you it is). In finance, analysts are not trying to predict the future (nobody can really do that); instead, they hedge against the risk of a downside future event. Whether we can predict the future is beside the point—it's a misunderstanding of the true nature of the problem. The real challenge is knowing how to hedge effectively against an adverse event. After all, if the future were predictable, there'd be no need for insurance in the first place. All that would be required is a steady accumulation of funds sufficient to pay the resulting losses when a predicted event occurred.

Given that we can't predict events, we must rely on financial innovation. In particular, we must focus on complex systems. They're the most difficult, but they're frequently the best representation of how the world works today. With a bit more sophistication and financial innovation, we can deal with these systems fairly easily. The scheme suggested here won't do a perfect job—nothing can—but it provides a new way of managing and accounting for risk, and of dealing with apparent chaos.

Order from Chaos

The key observation behind all my work comes from mathematics and chaotic systems: Most chaotic systems exhibit order, and that order is given by what are called "chaotic attractors." This means even systems that are difficult to make predictions about, and difficult to observe, will eventually settle into patterns.

Think, for instance, about the weather. Perhaps you've been told that, as a result of the chaotic attractors, there are two possible patterns into which the weather may settle. In one, there will be a high probability of 10 intense and destructive tornadoes. Conversely, in the other pattern, there are clear skies and an absence of intense storms.

Complexity Theory

In essence, there are two corresponding actuarial tables for catastrophic storms, one with 10 such events and another with one. You're not predicting which will come to pass, only that one of two will happen, and that they're equally likely.

Given that you don't know which pattern will occur, the standard principles of insurance aren't really of any help in deciding what steps to take, based on the actuarial tables for the two scenarios. In one case, you decide to be very cautious and assume that 10 storms will occur. Then you have to consider how much capital you'll need to reserve given that assumption. If each of the events eventually costs $1 billion, then you know you'll need $10 billion. Of course, if there's only one catastrophic event, the cost of carrying that collateral has been enormous, since you've been carrying 10 times the amount you needed and the opportunity cost of that capital is not inconsequential.

On the other hand, if you decide there will be only one storm, you'll be under protected. In that situation, the risk you face is financial catastrophe.

What do insurers do in this case? They have several options; all, however, have accompanying downsides. One is to settle on something between the extremes of $1 billion and $10 billion. If both patterns are equally likely, it could seem perfectly logical to divide the difference and choose $5.5 billion as the target. But by doing that you're guaranteed to be wrong 100 percent of the time—50 percent of the time you'll be over-insured and 50 percent of the time you'll be under-insured, by several billion dollars in both cases. If you're too low, and insuring yourself, you're making yourself liable for risks you can't cover. But if you set the target too high, you're forced to set unattractively high rates. You'll likely end up losing your customers. In many instances, that's happening right now.

If you want to insure property in the Caribbean, for example, you must pay a premium that's roughly 33 percent of the
value of the property. You know there isn’t really a one-in-three chance of a catastrophic event. But because the possibility may, in fact, be one in 10, or one in two, insurers have to take something like an average. So we end up with 33 percent. As a result, most of the properties in the Caribbean are uninsured.

Another option is self-insurance. Stanford University, for example, purchases no earthquake insurance. For many years, the university covered itself. Some may think of that as a solution, but it isn’t. In the end, regulators still look at Stanford and say, “Okay—you’re selling insurance to yourself. So all the usual state regulations about insurance apply to the university.” Stanford, as a result, has the same burden of accumulating data, record-keeping, and reserves as any insurance company.

Back to the Weather
Let’s think again about those two weather patterns. What do they really mean? They qualify as chaotic attractors—complex systems. As complex as they are, most of them tend to settle into patterns. The problem is that, using the example I gave, the chaotic behavior could settle into two patterns. One is associated with a certain type of weather; the other is associated with another type. Still, there’s a lot of uncertainty with these two patterns; you can’t make exact predictions—we’re talking about averages here. One pattern is associated with just one catastrophic storm, and the other with 10 storms.

With chaotic systems, although you’ve identified the attractors, you can’t predict whether you’re going in one direction or another. So in some sense, although chaotic systems are not as chaotic as they appear to be—there’s a method in their madness—you don’t know in which direction they’ll go. The solution is to change our way of looking at them.

The Securities Component
The securities industry has a different approach to this dilemma than the insurance industry. The insurance industry knows that you can’t predict when and where a specific event will hit, so it relies instead on actuarial tables to provide broader (statistically based) predictions inherent in the law of large numbers. This is the principle behind insurance—safety in numbers. You can’t predict precisely who will be hit, but you can predict how many people, overall, will be hit.

Even if a given system is chaotic, you still have enough information to allow you to say something about it. What you say takes the form of one, two, or three possible scenarios and, with them, one, two, or three actuarial tables. Selecting from among them is the problem.

The remedy is to use principles from securities markets. These markets don’t rely on the sort of statistically based information the insurance industry uses. What they do is hedge. They operate by seeking patterns and, armed with that, hedge against probable downside risks. They hedge, for instance, when they use bonds and equities to provide a cushion against interest-rate changes—without looking for the statistical safety in numbers that characterizes insurance procedures. They search for factors that are correlated. Securities houses, then, hedge against correlated events, not by using the law of large numbers or insurance principles, but by using inverse correlation.

The obvious question arises: “What I’ve got isn’t stocks and bonds; it’s two equally probable actuarial tables. What am I supposed to use as a hedge?” The answer is—use the same principles that have proved so successful in the securities markets.

You have to hedge the risks in one of your actuarial tables against the risks in the other. You need to look at the risk not as something you’re trying to predict statistically, but as the development of different scenarios. Then you have to, in effect, insure yourself by using insurance policies within each of the scenarios. In each case you assume the scenario that’s going to occur. Insurance goes as far as possible—because it’s a very efficient method—in dealing with one of the scenarios. But when dealing with multiple scenarios, where you have no way of deciding between them, you have to use a different financial instrument. This instrument is a security contract, one that’s contingent not on interest rates but on actuarial tables instead.

In risk management, the key is to think of the future as a combination of two factors. The first is a statistical factor, which is what you’re familiar with—the actuarial table. But if you’re insecure, as you should be, about what triggers insurable events, your only option is to bet—just like in a casino. If there’s one sort of outcome, and you’re protected for that, you’ll have a shortfall, and if the other scenario happens, and you’re protected for that, you’ll have too much insurance and the carrying cost will be too high. So what you do is buy securities that pay more in one weather pattern and less in another. Thus, by transferring money between the two possible states, you’re protected. In more concrete terms, it’s a combination of a modified reinsurance package, which diversifies the risk and protects the insurer against unknown financial risks, and a security, which captures correlated risks that can’t be diversified in any way.

So the fact that you can’t make predictions becomes irrelevant—as long as you have chaotic attractors, which give rise

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to patterns. While the patterns are not predictable, if you can
distinguish a finite number of patterns you're positioned to
take effective action and hedge the risk inherent in each pattern
by reverse correlation.

So chaotic systems can be used for practical purposes. We
live in a world where we need to know what could happen, hav-
ing used a rational approach to classifying risks. And the risk
may well take the form of one of several patterns, with corre-
spondingly different actuarial tables. Once you've understood
that risk, take action. Take out a hedge against the risk.

Some Pieces in Place

While this concept has yet to be implemented totally, some
pieces of it, interestingly enough, have been coming onto the
market. The Chicago Board of Trade is now selling catastrophe
futures, for example. They're a piece of the total puzzle I've de-
scribed here, since they help you protect against different ac-
ruarial tables governing a given segment of time. But the mar-
ket for these instruments to date has been very thin, because
the same people who buy them—insurers—are the people who
report how much a given instrument should pay, thus poten-
tially creating a conflict of interest.

Meanwhile, during the summer of 1996, Merrill Lynch and
Morgan Stanley floated new instruments, such as environmental
bonds, with exactly the same structure. This is a bond with a
coupon that, in addition to the standard terms, includes a fac-
tor that pays differently depending on the frequency of cata-

trophic events.

In addition, there have been attempts to adapt the funda-
mentals of securities to fit the insurance industry—in a word,
securitization. There are companies that carry insurance con-
tracts as their main asset (but of course, there are also liabili-

ties); they sell shares in the company, thereby accessing the
capital market. Much like Ginnie Mae, you take a large num-
ber of risks, pool them, cut the pool into small portions (shares)
that are put on the market, and let people invest in the com-
pany. This procedure lets companies access capital markets
from an insurance base, and that's good, given the potential for
losses from a major catastrophe. It's nearly impossible to put
aside enough reserves for a "Big One," and the associated cost
of doing so is tremendous.

So different pieces of the theory I've advanced here have
been implemented, but not the whole thing. The bottom line,
then: You can't predict one particular outcome using com-
plexity theory, but you can look for the relevant chaotic at-
tractors. Then, if there turns out to be more than one of these,
you can use securities as hedges between them. The actual fi-
nancial instrument has been created and developed at the Pro-
gram on Information and Resources at Columbia University,
and is in the process of being implemented. It protects against
financial as well as natural catastrophes in the most cost-effi-
cient way.