Liu Journal of Uncertainty Analysis Applications 2013, **1**:1 http://www.juaa-journal.com/content/1/1/1



REVIEW Open Access

Toward uncertain finance theory

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Abstract

This paper first introduces a paradox of stochastic finance theory that shows the real stock price is impossible to follow any Ito's stochastic differential equation. After a survey on uncertainty theory, uncertain process, uncertain calculus, and uncertain differential equation, this paper discusses some possible applications of uncertain differential equations to financial markets. Finally, it is suggested that a new uncertain finance theory should be developed based on uncertainty theory and uncertain differential equation.

Keywords: Finance, Uncertainty theory, Uncertain process, Uncertain calculus, Uncertain differential equation

Review

When no samples are available to estimate a probability distribution, we have to invite some domain experts to evaluate their belief degree that each event will occur. Perhaps some people think that personal belief degree is subjective probability or fuzzy concept. However, Liu [1] declared that it is inappropriate because both probability theory and fuzzy set theory may lead to counterintuitive results in this case. In order to rationally deal with the belief degree, an uncertainty theory was founded by Liu [2] and subsequently studied by many scholars. Nowadays, uncertainty theory has become a branch of axiomatic mathematics for modeling human uncertainty.

Based on uncertainty theory, the concept of uncertain process was given by Liu [3] as a sequence of uncertain variables indexed by time. Besides, the concept of uncertain integral was also proposed by Liu [3] in order to integrate an uncertain process with respect to a canonical process. Furthermore, Liu [4] recast his work via the fundamental theorem of uncertain calculus and thus produced the techniques of chain rule, change of variables, and integration by parts. Since then, the theory of uncertain calculus was well developed.

After uncertain differential equation was proposed by Liu [3] as a differential equation involving uncertain process, an existence and uniqueness theorem of a solution of uncertain differential equation was proved by Chen and Liu [5] under linear growth condition and Lipschitz continuous condition. The theorem was verified again by Gao [6] under local linear growth condition and local Lipschitz continuous condition. In order to solve uncertain differential equations, Chen and Liu [5] obtained an analytic solution to linear uncertain differential equations. In addition, Liu [7] presented a spectrum of analytic methods to solve some special classes of nonlinear uncertain differential equations. More importantly, Yao and Chen [8] showed that the solution of an uncertain differential



equation can be represented by a family of solutions of ordinary differential equations, thus relating uncertain differential equations and ordinary differential equations. On the basis of the Yao-Chen formula, a numerical method was also designed by Yao and Chen [8] for solving general uncertain differential equations. Furthermore, Yao [9] presented some formulas to calculate the extreme values, first hitting time and integral of solution of uncertain differential equation.

Uncertain differential equations were first introduced into finance by Liu [4] in which an uncertain stock model was proposed and European option price formulas were documented. Besides, Chen [10] derived American option price formulas for this type of uncertain stock model. In addition, Peng and Yao [11] presented a different uncertain stock model and obtained the corresponding option price formulas, and Yu [12] proposed an uncertain stock model with jumps. Uncertain differential equations were also employed to model uncertain currency markets by Liu and Chen [13] in which an uncertain currency model was proposed. Uncertain differential equations were used to simulate interest rate by Chen and Gao [14], and an uncertain interest rate model was presented. On the basis of this model, the price of zero-coupon bond was also produced. Uncertain differential equations were applied to optimal control by Zhu [15] in which Zhu's equation of optimality is proved to be a necessary condition for extremum of uncertain optimal control model.

This paper first introduces a paradox of stochastic finance theory. After a survey on uncertainty theory, uncertain process, uncertain calculus, and uncertain differential equation, this paper shows some possible applications of uncertain differential equations to financial markets. Finally, this paper suggests to develop an uncertain finance theory by using uncertainty theory and uncertain differential equation.

A paradox of stochastic finance theory

The origin of stochastic finance theory can be traced to Louis Bachelier's doctoral dissertation *Théorie de la Speculation* in 1900. However, Bachelier's work had little impact for more than a half century. After Kiyosi Ito invented stochastic calculus [16] and stochastic differential equation [17], stochastic finance theory was well developed among others by Samuelson [18], Black and Scholes [19], and Merton [20] during the 1960s and 1970s.

Traditionally, stochastic finance theory presumes that the stock price (including currency exchange rate and interest rate) follows an Ito's stochastic differential equation. Is it really reasonable? In fact, this widely accepted presumption was continuously challenged by many scholars. Let us assume that the stock price X_t follows the stochastic differential equation

$$dX_t = eX_t dt + \sigma X_t dW_t \tag{1}$$

where e is the log-drift, σ is the log-diffusion, and W_t is a Wiener process. Let us see what will happen with such an assumption. It follows from the stochastic differential equation (1) that X_t is a geometric Wiener process, i.e.,

$$X_t = X_0 \exp((e - \sigma^2/2)t + \sigma W_t)$$
(2)

from which we derive

$$W_{t} = \frac{\ln X_{t} - \ln X_{0} - (e - \sigma^{2}/2)t}{\sigma}$$
(3)

whose increment is

$$\Delta W_t = \frac{\ln X_{t+\Delta t} - \ln X_t - (e - \sigma^2/2)\Delta t}{\sigma}.$$
 (4)

Write

$$A = -\frac{(e - \sigma^2/2)\Delta t}{\sigma}. (5)$$

Note that the real stock price X_t is actually a step function of time with a finite number of jumps although it looks like a curve. During a fixed period, without loss of generality, we assume that X_t is observed to have 100 jumps. Now we divide the period into 10,000 equal intervals. Then we may observe 10,000 samples of X_t . It follows from Equation 4 that ΔW_t has 10,000 samples that consist of 9,900 A's and 100 other numbers:

$$\underbrace{A, A, \cdots, A}_{9,900}, \underbrace{B, C, \cdots, Z}_{100}.$$

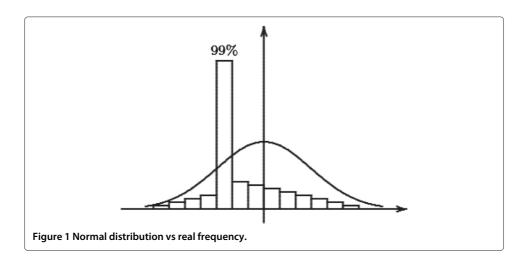
$$(6)$$

Nobody can believe that those 10,000 samples follow a normal probability distribution with expected value 0 and variance Δt . See Figure 1. This fact is in contradiction with the property of Wiener process that the increment ΔW_t is a normal random variable with expected value 0 and variance Δt . Therefore, the stock price X_t does not follow the stochastic differential equation.

Perhaps some people think that the stock price does behave like a geometric Wiener process (or Ornstein-Uhlenbeck process) in macroscopy although they recognize the paradox in microscopy. However, as the very core of stochastic finance theory, Ito's calculus is just built on the microscopic structure (i.e., the differential $\mathrm{d}W_t$) of Wiener process rather than macroscopic structure. More precisely, Ito's calculus is dependent on the presumption that $\mathrm{d}W_t$ is a normal random variable with expected value 0 and variance $\mathrm{d}t$. This unreasonable presumption is what causes the second order term in Ito's formula,

$$dX_t = \frac{\partial h}{\partial t}(t, W_t)dt + \frac{\partial h}{\partial w}(t, W_t)dW_t + \frac{1}{2}\frac{\partial^2 h}{\partial w^2}(t, W_t)dt.$$
 (7)

In fact, the increment of stock price is impossible to follow any continuous probability distribution. On the basis of the above paradox, personally, I do not think Ito's calculus can play the essential tool of finance theory because Ito's stochastic differential equation is impossible to model real stock price.



What is uncertainty theory?

Let Γ be a nonempty set, and \mathcal{L} a σ -algebra over Γ . Each element Λ in \mathcal{L} is called an event. A set function \mathcal{M} from \mathcal{L} to [0,1] is called an *uncertain measure* if it satisfies the following axioms [2]:

Axiom 1. (Normality axiom) $\mathcal{M}\{\Gamma\} = 1$ for the universal set Γ ;

Axiom 2. (Duality axiom) $\mathcal{M}\{\Lambda\} + \mathcal{M}\{\Lambda^c\} = 1$ for any event Λ ;

Axiom 3. (Subadditivity axiom) For every countable sequence of events $\Lambda_1, \Lambda_2, \cdots$, we have

$$\mathcal{M}\left\{\bigcup_{i=1}^{\infty} \Lambda_i\right\} \le \sum_{i=1}^{\infty} \mathcal{M}\{\Lambda_i\}. \tag{8}$$

The triplet $(\Gamma, \mathcal{L}, \mathcal{M})$ is called an *uncertainty space*. In order to obtain an uncertain measure of compound event, a product uncertain measure was defined by Liu [4], thus producing the fourth axiom of uncertainty theory:

Axiom 4. (Product axiom) Let $(\Gamma_k, \mathcal{L}_k, \mathcal{M}_k)$ be uncertainty spaces for $k = 1, 2, \cdots$ The product uncertain measure \mathcal{M} is an uncertain measure satisfying

$$\mathcal{M}\left\{\prod_{k=1}^{\infty}\Lambda_{k}\right\} = \bigwedge_{k=1}^{\infty}\mathcal{M}_{k}\{\Lambda_{k}\}\tag{9}$$

where Λ_k are arbitrarily chosen events from \mathcal{L}_k for $k = 1, 2, \cdots$, respectively.

An *uncertain variable* is defined by Liu [2] as a measurable function ξ from an uncertainty space $(\Gamma, \mathcal{L}, \mathcal{M})$ to the set of real numbers, i.e., for any Borel set B of real numbers, the set

$$\{\xi \in B\} = \{\gamma \in \Gamma \mid \xi(\gamma) \in B\} \tag{10}$$

is an event. In order to describe an uncertain variable in practice, the concept of *uncertainty distribution* is defined by Liu [2] as

$$\Phi(x) = \mathcal{M}\left\{\xi \le x\right\}, \quad \forall x \in \mathcal{R}. \tag{11}$$

Peng and Iwamura [21] proved that a function $\Phi: \mathfrak{R} \to [0,1]$ is an uncertainty distribution if and only if it is a monotone increasing function except $\Phi(x) \equiv 0$ and $\Phi(x) \equiv 1$.

An uncertainty distribution $\Phi(x)$ is said to be regular if it is a continuous and strictly increasing function with respect to x at which $0 < \Phi(x) < 1$, and

$$\lim_{x \to -\infty} \Phi(x) = 0, \quad \lim_{x \to +\infty} \Phi(x) = 1. \tag{12}$$

Let ξ be an uncertain variable with regular uncertainty distribution $\Phi(x)$. Then the inverse function $\Phi^{-1}(\alpha)$ is called the inverse uncertainty distribution of ξ [22].

It is easy to verify that $\Phi^{-1}(\alpha)$ is a continuous and strictly increasing function with respect to $\alpha \in (0,1)$. Conversely, suppose $\Phi^{-1}(\alpha)$ is a continuous and strictly increasing function on (0,1). Define

$$\Phi(x) = \begin{cases} 0, & \text{if } x \le \lim_{\alpha \downarrow 0} \Phi^{-1}(\alpha) \\ \alpha, & \text{if } x = \Phi^{-1}(\alpha) \\ 1, & \text{if } x \ge \lim_{\alpha \uparrow 1} \Phi^{-1}(\alpha). \end{cases}$$

It follows that $\Phi(x)$ is an uncertainty distribution of some uncertain variable ξ . Then for each $\alpha \in (0, 1)$, we have

$$\mathcal{M}\{\xi \le \Phi^{-1}(\alpha)\} = \Phi(\Phi^{-1}(\alpha)) = \alpha.$$

Thus, $\Phi^{-1}(\alpha)$ is just the inverse uncertainty distribution of the uncertain variable ξ . Hence, we have a sufficient and necessary condition of inverse uncertainty distribution: A function $\Phi^{-1}(\alpha):(0,1)\to\mathfrak{R}$ is an inverse uncertainty distribution if and only if it is a continuous and strictly increasing function with respect to α .

The *expected value* of an uncertain variable ξ is defined by Liu [2] as an average value of the uncertain variable in the sense of uncertain measure, i.e.,

$$E[\xi] = \int_0^{+\infty} \mathcal{M}\{\xi \ge r\} dr - \int_{-\infty}^0 \mathcal{M}\{\xi \le r\} dr$$
 (13)

provided that at least one of the two integrals is finite. If ξ has an uncertainty distribution Φ , then the expected value may be calculated by

$$E[\xi] = \int_{0}^{+\infty} (1 - \Phi(x)) dx - \int_{-\infty}^{0} \Phi(x) dx.$$
 (14)

Independence is an extremely important concept in uncertainty theory. The uncertain variables $\xi_1, \xi_2, \dots, \xi_n$ are said to be independent [4] if

$$\mathcal{M}\left\{\bigcap_{i=1}^{n} (\xi_i \in B_i)\right\} = \bigwedge_{i=1}^{n} \mathcal{M}\left\{\xi_i \in B_i\right\} \tag{15}$$

for any Borel sets B_1, B_2, \dots, B_n of real numbers. Equivalently, those uncertain variables are independent if and only if

$$\mathcal{M}\left\{\bigcup_{i=1}^{n} (\xi_i \in B_i)\right\} = \bigvee_{i=1}^{n} \mathcal{M}\left\{\xi_i \in B_i\right\}. \tag{16}$$

Let $\xi_1, \xi_2, \dots, \xi_n$ be independent uncertain variables with uncertainty distributions $\Phi_1, \Phi_2, \dots, \Phi_n$, respectively. If the function $f(x_1, x_2, \dots, x_n)$ is strictly increasing with respect to x_1, x_2, \dots, x_m and strictly decreasing with respect to $x_{m+1}, x_{m+2}, \dots, x_n$, then $\xi = f(\xi_1, \xi_2, \dots, \xi_n)$ is an uncertain variable with inverse uncertainty distribution

$$\Psi^{-1}(\alpha) = f(\Phi_1^{-1}(\alpha), \dots, \Phi_m^{-1}(\alpha), \Phi_{m+1}^{-1}(1-\alpha), \dots, \Phi_n^{-1}(1-\alpha)).$$
 (17)

Then Liu and Ha [23] proved that the uncertain variable $\xi = f(\xi_1, \xi_2, \dots, \xi_n)$ has an expected value

$$E[\xi] = \int_0^1 f(\Phi_1^{-1}(\alpha), \dots, \Phi_m^{-1}(\alpha), \Phi_{m+1}^{-1}(1-\alpha), \dots, \Phi_n^{-1}(1-\alpha)) d\alpha.$$
 (18)

For exploring the details of uncertainty theory, the readers may consult Liu [24].

Uncertain process

Let T be a totally ordered set (that is usually "time"), and let $(\Gamma, \mathcal{L}, \mathcal{M})$ be an uncertainty space. An *uncertain process* is defined by Liu [3] as a measurable function from $T \times (\Gamma, \mathcal{L}, \mathcal{M})$ to the set of real numbers, i.e., for each $t \in T$ and any Borel set B of real numbers, the set

$$\{X_t \in B\} = \{ \gamma \in \Gamma \mid X_t(\gamma) \in B \} \tag{19}$$

is an event. In other words, an uncertain process is a sequence of uncertain variables indexed by time.

Note that if the index set T becomes a partially ordered set (e.g., time × space, or a surface), then X_t is called an *uncertain field* provided that X_t is an uncertain variable at each point t. That is, an uncertain field is a generalization of an uncertain process.

An uncertain process X_t is said to have an uncertainty distribution $\Phi_t(x)$ if at each time t, the uncertain variable X_t has the uncertainty distribution $\Phi_t(x)$. It is easy to prove that $\Phi_t(x)$ is a monotone increasing function with respect to x and $\Phi_t(x) \not\equiv 0$, $\Phi_t(x) \not\equiv 1$. Conversely, if at each time t, $\Phi_t(x)$ is a monotone increasing function except $\Phi_t(x) \equiv 0$ and $\Phi_t(x) \equiv 1$, it follows that there exists an uncertain variable ξ_t whose uncertainty distribution is just $\Phi_t(x)$. Define

$$X_t = \xi_t, \quad \forall t \in T.$$

Then X_t is an uncertain process and has the uncertainty distribution $\Phi_t(x)$. Thus, a function $\Phi_t(x): T \times \mathfrak{R} \to [0,1]$ is an uncertainty distribution of uncertain process if and only if at each time t, it is a monotone increasing function except $\Phi_t(x) \equiv 0$ and $\Phi_t(x) \equiv 1$.

An uncertainty distribution $\Phi_t(x)$ is said to be regular if at each time t, it is a continuous and strictly increasing function with respect to x at which $0 < \Phi_t(x) < 1$, and

$$\lim_{x \to -\infty} \Phi_t(x) = 0, \quad \lim_{x \to +\infty} \Phi_t(x) = 1. \tag{20}$$

Let X_t be an uncertain process with regular uncertainty distribution $\Phi_t(x)$. Then the inverse function $\Phi_t^{-1}(\alpha)$ is called the inverse uncertainty distribution of X_t . It is easy to prove that $\Phi_t^{-1}(\alpha)$ is a continuous and strictly increasing function with respect to $\alpha \in (0,1)$. Conversely, if $\Phi_t^{-1}(\alpha)$ is a continuous and strictly increasing function with respect to $\alpha \in (0,1)$, it follows that there exists an uncertain variable ξ_t whose inverse uncertainty distribution is just $\Phi_t^{-1}(\alpha)$. Define

$$X_t = \xi_t, \quad \forall t \in T.$$

Then X_t is an uncertain process and has the inverse uncertainty distribution $\Phi_t^{-1}(\alpha)$. Hence, a function $\Phi_t^{-1}(\alpha): T \times (0,1) \to \Re$ is an inverse uncertainty distribution of uncertain process if and only if at each time t, it is a continuous and strictly increasing function with respect to α .

An uncertain process X_t is said to have *independent increments* if

$$X_{t_0}, X_{t_1} - X_{t_0}, X_{t_2} - X_{t_1}, \cdots, X_{t_k} - X_{t_{k-1}}$$
 (21)

are independent uncertain variables where t_0 is the initial time and t_1, t_2, \dots, t_k are any times with $t_0 < t_1 < \dots < t_k$. That is, an independent increment process means that its increments are independent uncertain variables whenever the time intervals do not overlap. Let X_t be a sample-continuous independent increment process with an uncertainty distribution $\Phi_t(x)$ at each time t. When f is a strictly increasing function, Liu [25] proved that the supremum

$$\sup_{0 < t < s} f(X_t) \tag{22}$$

has an uncertainty distribution

$$\Psi(x) = \inf_{0 \le t \le s} \Phi_t(f^{-1}(x)). \tag{23}$$

This result is called the *extreme value theorem* of uncertain process.

An uncertain process X_t is said to have *stationary increments* if its increments are identically distributed uncertain variables whenever the time intervals have the same length, i.e., for any given t > 0, the increments $X_{s+t} - X_s$ are identically distributed uncertain variables for all s > 0.

Let X_t be a stationary independent increment process with a crisp initial value X_0 . Liu [22] showed that there exist two real numbers a and b such that the expected value

$$E[X_t] = a + bt (24)$$

for any time $t \ge 0$. Furthermore, Chen [26] verified that there exists a real number c such that the variance

$$V[X_t] = ct^2 (25)$$

for any time $t \geq 0$.

As an important type of uncertain process, a canonical process is a stationary independent increment process whose increments are normal uncertain variables. More precisely, an uncertain process C_t is called a *canonical process* by Liu [4] if (1) $C_0 = 0$ and almost all sample paths are Lipschitz continuous, (2) C_t has stationary and independent increments, and (3) every increment $C_{s+t} - C_s$ is a normal uncertain variable with expected value 0 and variance t^2 .

It is easy to verify that the canonical process C_t is a normal uncertain variable with expected value 0 and variance t^2 and has an uncertainty distribution

$$\Phi(x) = \left(1 + \exp\left(-\frac{\pi x}{\sqrt{3}t}\right)\right)^{-1} \tag{26}$$

at each time t > 0. In addition, for each time t > 0, the ratio C_t/t is a normal uncertain variable with expected value 0 and variance 1. That is,

$$\frac{C_t}{t} \sim \mathcal{N}(0, 1) \tag{27}$$

for any t > 0.

What is the difference between canonical process and the Wiener process? First, canonical process is an uncertain process while the Wiener process is a stochastic process. Second, almost all sample paths of canonical process are Lipschitz continuous functions while almost all sample paths of the Wiener process are continuous but non-Lipschitz functions. Third, canonical process has a variance t^2 while the Wiener process has a variance t at each time t.

Uncertain calculus

Uncertain calculus is a branch of mathematics that deals with differentiation and integration of uncertain processes. The key concept in uncertain calculus is the uncertain integral that allows us to integrate an uncertain process (the integrand) with respect to the canonical process (the integrator). The result of the uncertain integral is another uncertain process.

Let X_t be an uncertain process and let C_t be a canonical process. For any partition of closed interval [a, b] with $a = t_1 < t_2 < \cdots < t_{k+1} = b$, the mesh is written as

$$\Delta = \max_{1 \le i \le k} |t_{i+1} - t_i|. \tag{28}$$

Then the *uncertain integral* of X_t with respect to C_t is defined by Liu [4] as

$$\int_{a}^{b} X_{t} dC_{t} = \lim_{\Delta \to 0} \sum_{i=1}^{k} X_{t_{i}} \cdot (C_{t_{i+1}} - C_{t_{i}})$$
(29)

provided that the limit exists almost surely and is finite. Since X_t and C_t are uncertain variables at each time t, the limit in Equation 29 is also an uncertain variable.

Let Z_t be an uncertain process. If there exist two uncertain processes μ_t and σ_t such that

$$Z_t = Z_0 + \int_0^t \mu_s \mathrm{d}s + \int_0^t \sigma_s \mathrm{d}C_s \tag{30}$$

for any $t \ge 0$, then we say Z_t has an *uncertain differential*

$$dZ_t = \mu_t dt + \sigma_t dC_t. \tag{31}$$

In this case, Z_t is called an uncertain process with drift μ_t and diffusion σ_t . It is clear that uncertain integral and differential are mutually inverse operations. Please also note that an uncertain differential of an uncertain process has two parts, the "dt" part and the "d C_t " part.

Let h(t, c) be a continuously differentiable function. Liu [4] showed that the uncertain process $Z_t = h(t, C_t)$ has an uncertain differential

$$dZ_t = \frac{\partial h}{\partial t}(t, C_t)dt + \frac{\partial h}{\partial c}(t, C_t)dC_t.$$
(32)

This result is called the fundamental theorem of uncertain calculus.

Example 1. Let us calculate the uncertain differential of tC_t . In this case, we have h(t, c) = tc whose partial derivatives are

$$\frac{\partial h}{\partial t}(t,c) = c, \quad \frac{\partial h}{\partial c}(t,c) = t.$$

It follows from the fundamental theorem of uncertain calculus that

$$d(tC_t) = C_t dt + t dC_t. (33)$$

Example 2. Let us calculate the uncertain differential of C_t^2 . In this case, we have $h(t,c) = c^2$ whose partial derivatives are

$$\frac{\partial h}{\partial t}(t,c) = 0, \quad \frac{\partial h}{\partial c}(t,c) = 2c.$$

It follows from the fundamental theorem of uncertain calculus that

$$dC_t^2 = 2C_t dC_t. (34)$$

Example 3. Let f(c) be a continuously differentiable function. Then we have

$$\frac{\partial f}{\partial t}(c) = 0, \quad \frac{\partial f}{\partial c}(c) = f'(c).$$

It follows from the fundamental theorem of uncertain calculus that the uncertain process $f(C_t)$ has an uncertain differential

$$df(C_t) = f'(C_t)dC_t. (35)$$

This formula is also called the *chain rule* of uncertain calculus.

As supplements to uncertain integral, Liu and Yao [27] suggested an uncertain integral with respect to multiple canonical processes. More generally, Chen and Ralescu [28] presented an uncertain integral with respect to the general Liu process.

Uncertain differential equation

The study of uncertain differential equation was pioneered by Liu [3]. Nowadays, uncertain differential equation has achieved fruitful results in both theory and practice. Let f and g be two functions. Then

$$dX_t = f(t, X_t)dt + g(t, X_t)dC_t$$
(36)

is called an *uncertain differential equation*. A solution is an uncertain process X_t that satisfies Equation 36 identically in t.

Some analytic methods have been proposed for solving uncertain differential equations. For example, Chen and Liu [5] showed that the linear uncertain differential equation

$$dX_t = (u_{1t}X_t + u_{2t})dt + (v_{1t}X_t + v_{2t})dC_t$$
(37)

has a solution

$$X_{t} = U_{t} \left(X_{0} + \int_{0}^{t} \frac{u_{2s}}{U_{s}} ds + \int_{0}^{t} \frac{v_{2s}}{U_{s}} dC_{s} \right)$$
(38)

where

$$U_t = \exp\left(\int_0^t u_{1s} \mathrm{d}s + \int_0^t v_{1s} \mathrm{d}C_s\right). \tag{39}$$

In addition, Liu [7] verified that the nonlinear uncertain differential equation like

$$dX_t = f(t, X_t)dt + \sigma_t X_t dC_t \tag{40}$$

has a solution

$$X_t = Y_t^{-1} Z_t \tag{41}$$

where

$$Y_t = \exp\left(-\int_0^t \sigma_s \mathrm{d}C_s\right) \tag{42}$$

and Z_t is the solution of uncertain differential equation

$$dZ_t = Y_t f(t, Y_t^{-1} Z_t) dt (43)$$

with initial value $Z_0 = X_0$.

An important contribution to uncertain differential equation is the existence and uniqueness theorem by Chen and Liu [5]. An uncertain differential equation has a unique solution if the coefficients f(t,x) and g(t,x) satisfy linear growth condition

$$|f(t,x)| + |g(t,x)| \le L(1+|x|), \quad \forall x \in \Re, t \ge 0$$
 (44)

and Lipschitz condition

$$|f(t,x) - f(t,y)| + |g(t,x) - g(t,y)| \le L|x - y|, \quad \forall x, y \in \Re, t \ge 0$$
 (45)

for some constant *L*. Moreover, the solution is sample-continuous.

The concept of stability was given by Liu [4]. An uncertain differential equation is said to be *stable* if for any two solutions X_t and Y_t , we have

$$\lim_{|X_0 - Y_0| \to 0} \mathcal{M}\{|X_t - Y_t| > \varepsilon\} = 0, \quad \forall t > 0$$

$$\tag{46}$$

for any given number $\varepsilon > 0$. Yao et al. [29] proved that the uncertain differential equation is stable if the coefficients f(t,x) and g(t,x) satisfy linear growth condition

$$|f(t,x)| + |g(t,x)| < K(1+|x|), \quad \forall x \in \Re, t > 0$$
 (47)

for some constant K and strong Lipschitz condition

$$|f(t,x) - f(t,y)| + |g(t,x) - g(t,y)| \le L(t)|x - y|, \quad \forall x, y \in \Re, t \ge 0$$
 (48)

for some bounded and integrable function L(t) on $[0, +\infty)$.

Uncertain differential equation has been extended by many scholars. For example, uncertain delay differential equation was studied among others by Barbacioru [30], Ge and Zhu [31], and Liu and Fei [32]. In addition, uncertain differential equation with jumps was suggested by Yao [33], and backward uncertain differential equation was discussed by Ge and Zhu [34].

Numerical method

Let α be a number with $0 < \alpha < 1$. An uncertain differential equation

$$dX_t = f(t, X_t)dt + g(t, X_t)dC_t$$
(49)

is said to have an α -path X_t^{α} if it solves the corresponding ordinary differential equation

$$dX_t^{\alpha} = f(t, X_t^{\alpha})dt + |g(t, X_t^{\alpha})|\Phi^{-1}(\alpha)dt$$
(50)

where $\Phi^{-1}(\alpha)$ is the inverse uncertainty distribution of standard normal uncertain variable, i.e.,

$$\Phi^{-1}(\alpha) = \frac{\sqrt{3}}{\pi} \ln \frac{\alpha}{1 - \alpha}.$$
 (51)

Then

$$\mathcal{M}\{X_t \le X_t^{\alpha}, \, \forall t\} = \alpha,\tag{52}$$

$$\mathcal{M}\{X_t > X_t^{\alpha}, \, \forall t\} = 1 - \alpha. \tag{53}$$

This result is called the *Yao-Chen formula* [8]. In addition, at each time t, the solution X_t has an inverse uncertainty distribution

$$\Psi_t^{-1}(\alpha) = X_t^{\alpha}. \tag{54}$$

Furthermore, for any monotone (increasing or decreasing) function *J*, we have

$$E[J(X_t)] = \int_0^1 J(X_t^{\alpha}) d\alpha.$$
 (55)

The Yao-Chen formula relates uncertain differential equations and ordinary differential equations, just like that Feynman-Kac formula relates stochastic differential equations and partial differential equations.

It is almost impossible to find analytic solutions for general uncertain differential equations. This fact provides a motivation to design a numerical method to solve general uncertain differential equation

$$dX_t = f(t, X_t)dt + g(t, X_t)dC_t.$$
(56)

In order to do so, a key point is to obtain an inverse uncertainty distribution $\Psi_t^{-1}(\alpha)$ of its solution X_t at any given time t. For this purpose, Yao and Chen [8] designed the following algorithm:

Step 1. Fix α on (0, 1).

Step 2. Solve $dX_t^{\alpha} = f(t, X_t^{\alpha})dt + |g(t, X_t^{\alpha})|\Phi^{-1}(\alpha)dt$ by any method of ordinary differential equation and obtain the α -path X_t^{α} , for example, by using the recursion formula

$$X_{i+1}^{\alpha} = X_i^{\alpha} + f(t_i, X_i^{\alpha})h + |g(t_i, X_i^{\alpha})|\Phi^{-1}(\alpha)h$$
(57)

where Φ is the standard normal uncertainty distribution and h is the step length.

Step 3. The inverse uncertainty distribution of the solution X_t is determined by

$$\Psi_t^{-1}(\alpha) = X_t^{\alpha}. \tag{58}$$

Uncertain stock model

Uncertain differential equations were first introduced into finance by Liu [4] in which an *uncertain stock model* was proposed,

$$\begin{cases} dX_t = rX_t dt \\ dY_t = eY_t dt + \sigma Y_t dC_t \end{cases}$$
(59)

where X_t is the bond price, Y_t is the stock price, r is the riskless interest rate, e is the log-drift, σ is the log-diffusion, and C_t is a canonical process.

A *European call option* is a contract that gives the holder the right to buy a stock at an expiration time s for a strike price K. The payoff from a European call option is $(Y_s - K)^+$ since the option is rationally exercised if and only if $Y_s > K$. Considering the time value of money resulted from the bond, the present value of this payoff is $\exp(-rs)(Y_s - K)^+$. Hence, the European call option price should be the expected present value of the payoff, i.e.,

$$f_c = \exp(-rs)E[(Y_s - K)^+].$$
 (60)

Liu [4] proved that

$$f_c = \exp(-rs)Y_0 \int_{K/Y_0}^{+\infty} \left(1 + \exp\left(\frac{\pi (\ln y - es)}{\sqrt{3}\sigma s}\right)\right)^{-1} dy.$$
 (61)

A *European put option* is a contract that gives the holder the right to sell a stock at an expiration time s for a strike price K. The payoff from a European put option is $(K - Y_s)^+$ since the option is rationally exercised if and only if $Y_s < K$. Considering the time value of money resulted from the bond, the present value of this payoff is $\exp(-rs)(K - Y_s)^+$. Hence, the European put option price should be the expected present value of the payoff, i.e.,

$$f_p = \exp(-rs)E[(K - Y_s)^+].$$
 (62)

Liu [4] proved that

$$f_p = \exp(-rs)Y_0 \int_0^{K/Y_0} \left(1 + \exp\left(\frac{\pi (es - \ln y)}{\sqrt{3}\sigma s}\right)\right)^{-1} dy.$$
 (63)

An *American call option* is a contract that gives the holder the right to buy a stock at any time prior to an expiration time s for a strike price K. It is clear that the payoff from an American call option is the supremum of $(Y_t - K)^+$ over the time interval [0, s].

Considering the time value of money resulted from the bond, the present value of this payoff is

$$\sup_{0 \le t \le s} \exp(-rt)(Y_t - K)^+. \tag{64}$$

Hence, the American call option price should be the expected present value of the payoff, i.e.,

$$f_c = E \left[\sup_{0 \le t \le s} \exp(-rt)(Y_t - K)^+ \right]. \tag{65}$$

Chen [10] proved that

$$f_c = \exp(-rs)Y_0 \int_{K/Y_0}^{+\infty} \left(1 + \exp\left(\frac{\pi(\ln y - es)}{\sqrt{3}\sigma s}\right)\right)^{-1} dy.$$
 (66)

An *American put option* is a contract that gives the holder the right to sell a stock at any time prior to an expiration time s for a strike price K. It is clear that the payoff from an American put option is the supremum of $(K - Y_t)^+$ over the time interval [0, s]. Considering the time value of money resulted from the bond, the present value of this payoff is

$$\sup_{0 \le t \le s} \exp(-rt)(K - Y_t)^+. \tag{67}$$

Hence, the American put option price should be the expected present value of the payoff, i.e.,

$$f_p = E \left[\sup_{0 \le t \le s} \exp(-rt)(K - Y_t)^+ \right]. \tag{68}$$

Chen [10] proved that

$$f_p = \int_0^{K \exp(-rs)} \sup_{0 \le t \le s} \left(1 + \exp\left(\frac{e}{\sqrt{3}\sigma} + \frac{\pi}{\sqrt{3}\sigma t} \ln \frac{Y_0}{K - y \exp(rt)}\right) \right)^{-1} \mathrm{d}y.$$

It is emphasized that other stock models were also actively investigated by Peng and Yao [11], Yu [12], and Chen et al. [35], among others.

Uncertain currency model

Liu and Chen [13] assumed that the exchange rate follows an uncertain differential equation and then proposed an *uncertain currency model*,

$$\begin{cases} dX_t = uX_t dt & \text{(Domestic currency)} \\ dY_t = vY_t dt & \text{(Foreign currency)} \end{cases}$$

$$dZ_t = eZ_t dt + \sigma Z_t dC_t & \text{(Exchange rate)}$$
(69)

where X_t represents the domestic currency with domestic interest rate u, Y_t represents the foreign currency with foreign interest rate v, and Z_t represents the exchange rate, that is, the domestic currency price of one unit of foreign currency at time t.

A *currency option* is a contract that gives the holder the right to exchange one unit of foreign currency at an expiration time s for K units of domestic currency. Suppose that the price of this contract is f in domestic currency. Then the investor pays f for buying

the contract at time 0 and receives $(Z_s - K)^+$ in domestic currency at the expiration time s. Thus, the expected return of the investor is

$$-f + \exp(-us)E[(Z_s - K)^+].$$
 (70)

On the other hand, the bank receives f for selling the contract at time 0 and pays $(1-K/Z_s)^+$ in foreign currency at the expiration time s. Thus, the expected return of the bank is

$$f - Z_0 \exp(-\nu s) E[(1 - K/Z_s)^+].$$
 (71)

The fair price of this contract should make the investor and the bank have an identical expected return, i.e.,

$$-f + \exp(-us)E[(Z_s - K)^+] = f - Z_0 \exp(-vs)E[(1 - K/Z_s)^+]. \tag{72}$$

Thus, the currency option price is

$$f = \frac{1}{2} \exp(-us)E[(Z_s - K)^+] + \frac{1}{2} \exp(-vs)Z_0E[(1 - K/Z_s)^+].$$
 (73)

Liu and Chen [13] proved that

$$f = \frac{1}{2} \exp(-us) Z_0 \int_{K/Z_0}^{+\infty} \left(1 + \exp\left(\frac{\pi (\ln y - es)}{\sqrt{3}\sigma s}\right) \right)^{-1} dy$$

$$+\frac{1}{2}\exp(-\nu s)Z_0\int_0^1\left(1+\exp\left(\frac{\pi(\ln(K/Z_0)-\ln y-es)}{\sqrt{3}\sigma s}\right)\right)^{-1}\mathrm{d}y.$$

Uncertain interest rate model

Real interest rates do not remain unchanged. Chen and Gao [14] assumed that the interest rate X_t follows an uncertain differential equation,

$$dX_t = (m - aX_t)dt + \sigma dC_t \tag{74}$$

where m, a, and σ are positive numbers, and C_t is a canonical process.

A *zero-coupon bond* is a bond bought at a price lower than its *face value*, that is, the amount it promises to pay at the maturity date. For simplicity, we assume that the face value is always US\$1. Then the price of a zero-coupon bond with a maturity date *s* is

$$f = E\left[\exp\left(-\int_0^s X_t dt\right)\right]. \tag{75}$$

Chen and Gao [14] proved that

$$f = \frac{\sqrt{3}\sigma}{a}(s-g)\exp\left(-\frac{ms}{2a} - \left(r_0 - \frac{m}{a}\right)g\right)\csc\left(\frac{\sqrt{3}\sigma}{a}(s-g)\right)$$
(76)

where

$$g = \frac{1}{a} (1 - \exp(-as)). \tag{77}$$

Uncertain finance theory

At the beginning of this paper, a paradox was proposed to show that the real stock price is impossible to follow an Ito's stochastic differential equation. It follows from Figure 1 that the increments behave like an uncertain variable rather than a random variable. This fact motives us to model stock prices by uncertain differential equations. Personally, I think uncertain calculus may play a potential mathematical foundation of finance theory.

If we say that the classical finance theory is a methodology dealing with financial markets by using probability theory, then uncertain finance theory is a methodology dealing with financial markets by using uncertainty theory. In addition to the uncertain stock models shown above, we may also accept other uncertain differential equations, for example,

$$dX_t = (m - aX_t)dt + \sigma X_t dC_t, \tag{78}$$

$$dX_t = (m - aX_t)dt + \sigma\sqrt{X_t}dC_t, \tag{79}$$

$$dX_t = (m - aX_t)dt + \sigma\sqrt{b + X_t}dC_t.$$
(80)

Conclusions

At first, a paradox of stochastic finance theory was introduced in this paper. After a survey on uncertainty theory, uncertain process, uncertain calculus, and uncertain differential equation, this paper summarized uncertain stock model, uncertain currency model, and uncertain interest model by using the tool of uncertain differential equation. Finally, it was suggested that an uncertain finance theory should be developed based on uncertainty theory.

Competing interests

The author declares that he has no competing interests.

Acknowledgements

This work was supported by the National Natural Science Foundation of China, grant no. 61273044.

Received: 12 February 2013 Accepted: 18 February 2013 Published: 24 April 2013

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doi:10.1186/2195-5468-1-1

Cite this article as: Liu: Toward uncertain finance theory. Journal of Uncertainty Analysis Applications 2013 1:1.

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