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Makhankov, V. G. and Aguero-Granados, M. A.

Universidad Autonoma del Estado de Mexico, Toluca, Mexico, Bussiness Math, Santa Fe, NM USA, Los Alamos National Laboratory

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Vladimir G. Makhankov[1], Maximo A. Aguero-Granados[2]

[1]BusinessMath, Santa Fe NM, USA

[2]Facultad de Ciencias, Universidad Autonoma del Estado de Mexico, Instituto Literario 100, Toluca 50000, Mexico.

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Address correspondence to V. Makhankov: 4148 Chaparron Pl. Santa Fe, NM 87507 USA, E-mail: vladgala@hotmail.com

[2] Research Sabbatical Stay at Los Alamos National Laboratory, T-Division, NM, USA.

ABSTRACT

We consider mean-reverting stochastic processes and build self-consistent models for forward price dynamics and some applications in power industries. These models are built using the ideas and equations of stochastic differential geometry in order to close the system of equations for the forward prices and their volatility. Some analytical solutions are presented in the one factor case and for specific regular forward price/interest rates volatility. Those models will also play a role of initial conditions for a stochastic process describing forward price and interest rates volatility.

Subsequently, the curved manifold of the internal space i.e. a discrete version of the bond term space (the space of bond maturing) is constructed. The dynamics of the point of this internal space that correspond to a portfolio of different bonds is studied. The analysis of the discount bond forward rate dynamics, for which we employed the Stratonovich approach, permitted us to calculate analytically the regular and the stochastic volatilities. We compare our results with those known from the literature.

I. INTRODUCTION

In this paper we give a self-consistent framework for describing and analyzing (evaluating) two different economic instruments that play a fundamental role in evaluating corresponding derivatives and hence in the risk management. Both have the term structure (mature at a definite time) and the first, discount bond interest rates, is going at finance market, the second, forward prices, at the commodity one. On the basis of stochastic differential geometry (**DSG)** equations that describe spot price/interest rate dynamics and long term dynamics (entire forward prices/interest rates curves) we construct mathematical models which turn out to be very similar.

Our study, at least in its economical part, has been influenced by works of various authors: Cortazar and Schwartz 1994, Schwartz 1997, Hillard and Reis

1998, Clewlow and Strickland 1999 to mention a few. Later we give more detailed references in due places.

The paper is organized as follows: In section II we give a very short introduction to the stochastic differential geometry, i.e. Brownian motion on curved manifolds (e.g., sphere), Makhankov 1995 and 1997, that allows us to take into account stochastic behavior of the forward prices/interest rates volatility and to close the system of equations. Section III presents the study of forward price dynamics and possible solutions for the forward price curve, its mean and volatility. Section IV is devoted to study of interest rates dynamics and some solutions. Finally, section V contains our conclusions.

II. STOCHASTIC DIFFERENTIAL GEOMETRY

Recent development in physics shows that all types of interactions admit geometrization. In the most transparent form it can be seen in the theory of the so-called (1+1) dimensional integrable systems, where in a consistent manner the relationship is established between geometry of the internal (isotopic) space and the type of interaction, see, e.g. Makhankov and Pashaev 1992.

In general discrete form the stochastic equation governing forward prices/rates dynamics (instruments with an internal space) is as follows:

$$dX^i(t) = f^i(t) + \sum_{p=1}^{n} \sigma_p^i(t) dW(t) \qquad \text{(II-1)}$$

where $f^i(t)$ is the price/rate drift, $\sigma_p^i(t)$ is the price/rate volatility and the index i spans the internal space (details will be discussed later on).

From the other hand the equations of stochastic differential geometry that describe the Brownian motion (diffusion) on a curved space (manifold) in terms of Stratonovich differentials (see Stratonovich 1968) read, Kendal 1987, Makhankov 1997:

$$dX^i(t) = \sum_{q=1}^{n} \Sigma_q^i dW^q(t)$$

$$d\Sigma_q^i = -\sum_{j,k=1}^{m} \Gamma_{jk}^i \Sigma_q^j dX^k \qquad \text{(II-2)}$$

Where X^i is an m-dim. vector (a point in an m-dim curved space), Σ_q^i is a matrix of rotating operator that may be constructed out of m vectors that set up

a natural frame on a patch of the bundle and Γ^i_{jk} are conexion coefficients (Christoffel's symbols) through which a curvature of the space is given. The first equation describes an elementary shock a Brownian particle undergoes due to a collision with the stochastic background.

If we assume the state space to be a Riemannian manifold, then the inverse of the Riemannian metric is given by

$$g^{ij} = \sum_q \sigma^i_q \sigma^j_q \qquad (\text{II-3})$$

such that

$$g_{ij} g^{jk} = \delta^k_i$$

Here we face two cases:
1) the conexion is compatible with the metric,
2) the conexion is "arbitrary", and can be determined, e.g. by matching some Ito process governing the system.

In the first case the conexion Γ is expressed in terms of the Riemannian metric g_{ij}, Dubrovin et al, 1984 and we come to a closed system of stochastic differential equations describing a stochastic path on a Riemannian manifold.

In the second case, in order to close our system of equations we match the drift term of the Ito process for the forward price/interest rates (II-1) with the drift term in SDG eqns. (II-2). Then the SDG equations will describe the forward price/interest rates dynamics as a Brownian motion in a curved space with the curvature defined by their drift. Rewriting eq. (II-2) in Ito representation one obtains

$$dX^i = \frac{1}{2}(d\tilde{\sigma}^i_q + \tilde{\sigma}^i_q) \equiv \tilde{f}^i dt + \sum_q \tilde{\sigma}^i_q dW^q \qquad (\text{II-4})$$

with the drift term due to the second eq. (II-2) being

$$\tilde{f}^i = -\frac{1}{2}\sum_{j,k,q} \tilde{\sigma}^k_q \Gamma^i_{kj} \tilde{\sigma}^j_q = -\frac{1}{2}\sum_{k,j} \Gamma^i_{kj} g^{kj} \qquad (\text{II-5})$$

In order to use SDG equations (II-2), equations (II-4) and (II-1) have to be identical, i.e. $\tilde{f}^i = f^i$, $\tilde{\sigma}^i_q = \sigma^i_q$ whence

$$f^i(t) = -\frac{1}{2}\sum_{k,j} \Gamma^i_{kj} g^{kj} \qquad (\text{II-5a})$$

or

$$\Gamma^i_{kj} - \frac{2}{m} g_{kj} f^i(t) \qquad \text{(II-6)}$$

This gives us the relationship between the conexion Γ, the drift $f^i(t)$ of the forward price/interest rate and the volatility structure σ^i_q through the space metric g^{ij}.

If the drift only depends on the geometrical part

$$f^i(t) = f(\sigma^i_q)$$

the system of stochastic differential equations (II-2) becomes closed and self-consistent.

It is well known from differential geometry that one of the most important characteristics of a manifold is its conexion curvature tensor

$$-R^i_{jkl} = \partial_k \Gamma^i_{jl} - \partial_l \Gamma^i_{jk} + \Gamma^i_{pk}\Gamma^p_{jl} - \Gamma^i_{pl}\Gamma^p_{jk}$$

Namely this tensor defines invariant geometrical properties of a manifold and vanishes in the Euclidean case: $R^i_{jkl} = 0$. A nonzero conexion yet does not imply a curved space (it could be related to fictitious-inertial forces) and a nonzero curvature tensor surely does, implying essential drift presence in the system studied. Sometimes instead of the curvature tensor R^i_{jkl} it is easier to calculate the so-called scalar curvature

$$R = \sum_{i,j,l} g^{il} R^i_{jil}$$

that gives us a clear understanding of the state space nature as well. More comprehensive review on these issues is given in Annex 1.

III. FORWARD PRICES AND THEIR MODELING

Definition1. A *forward contract* is a particularly simple derivative and is an agreement to buy (a *long position*) or to sell (a *short one*) an asset at a certain future time (*maturity date T*) for a certain price (the *delivery price K*).
At time the contract is initiated the delivery price should be such that the contract value for both parties is zero. The contract is obligatory.
Definition 2. The *forward price F(t,T)* for a certain contract is defined as the delivery price which would make the contract have zero value.

The forward price and the delivery price are equal at the time the contract is entered into. As time passes, they go apart since pre-specified initially delivery price is constant. Therefore we can think of the forward price as the delivery price at current time t.

There is a very well-known formula relating the forward price and the spot price, Hull 1993

$$F(t,T) = S(t)e^{(r+u-y)(T-t)} \tag{III-1}$$

where r is a risk-less interest rate, u is a storage rate and y a convenience yield. Both the storage rate and especially the convenience yield are usually unknown functions and the convenience yield can be a stochastic process.

Following Cortazar & Schwartz 1994, we will describe the dynamics of the forward price by the equation

$$\frac{dF(t,T)}{F} = \sum_{p=1}^{n} \Sigma_p(t,T) dW^p(t) \tag{III-2}$$

where $\Sigma_p(t,T)$ is the volatility corresponding to a p-th random factor described by the Wiener generator $dW^p(t)$. So model (III-2) describes the n-factor dynamics of the forward curve $F(t,T)$.

We assume that interest rates are deterministic and future prices are equal to forward prices (see, e.g. Hull 1993). In (III-2) we have n independent sources of uncertainty that drive the evolution of the forward curve $F(t,T)$.

By integrating (III-2) we have (using Ito's lemma)

$$F(t,T) = F(0,T) \exp\left\{ \sum_{i=1}^{n} \left[-\frac{1}{2} \int_0^t \Sigma_i(\tau,T)^2 d\tau + \int_0^t \Sigma_i(\tau,T) dW^i(\tau) \right] \right\} \tag{III-3}$$

Then for the spot price by definition, $S(t) = F(t,t)$, and we have by setting $T = t$

$$S(t) = F(0,t) \exp\left\{ \sum_{i=1}^{n} \left[-\frac{1}{2} \int_0^t \Sigma_i(\tau,t)^2 d\tau + \int_0^t \Sigma_i(\tau,t) dW^i(\tau) \right] \right\} \tag{III-4}$$

It means that the natural logarithm of the spot price (as well as the forward price) is normally distributed at time T given the forward price initially at time zero such that

$$\ln S((T)) \approx N\left[\{\ln(F(0,T)) - \frac{1}{2}\sum_{i=1}^{n}\int_{0}^{T}\Sigma_{i}(\tau,T)^2 d\tau\}, \ \sum_{i=1}^{n}\int_{0}^{T}\Sigma_{i}(\tau,T)^2 d\tau\}\right]$$

By differentiating (III-4) over t we have got the stochastic differential equation for the spot price

$$\frac{dS(t)}{S(t)} = \left\{\frac{\partial \ln F(0,t)}{\partial t} - \sum_{i=1}^{n}\left[\int_{0}^{t}\Sigma_{i}(\tau,t)\frac{\partial\Sigma_{i}(\tau,t)}{\partial t}d\tau - \int_{0}^{t}\frac{\partial\Sigma_{i}(\tau,t)}{\partial t}dW^{i}(\tau)\right]\right\}dt$$
$$+ \sum_{i=1}^{n}\Sigma_{i}(t,t)dW^{i}(t)$$

(III-5)

The term in the curled parentheses can be interpreted as an equivalent to the sum of the deterministic risk-less rate of interest $r(t)$ and a convenience yield $y(t)$ *which in general should be stochastic*. Many well-known models are special cases of this general approach.

Now it's well known that the volatilities in (III-2) are stochastic processes themselves. What kind of stochastic processes they could be? To answer this question we resort to the stochastic differential geometry, Makhankov 1995, 1997 described above. As a result, we obtain a self-consistent model described by the system of stochastic equations. To solve them we have to specify initial conditions: $F(0,T)$, $S(0)$ and $\Sigma_{i}(0,T)$.

Ergo, the dynamics of the forward price logarithm is given by the equation

$$d\ln(F(t,T)) = -\frac{1}{2}\sum_{q=1}^{n}\Sigma^2{}_{q}(t,T)d\,t + \sum_{q=1}^{n}\Sigma_{q}(t,T)d\,W^q(t)$$

(III-6)

in *Ito* differentials. If we assume the internal term space of the model being discrete (what is true in reality) we have got

$$F(t,T) = F(t,kT) = F^{k}(t)$$

Now denoting

$$X^{k}(t) = \ln F^{k}(t)$$

(III-7)

we come to the equation

$$dX^{i}(t) = -\frac{1}{2}\sum_{q=1}^{n}\Sigma^{i}{}_{q}{}^2(t)dt + \sum_{q=1}^{n}\Sigma^{i}{}_{q}dW^{q}(t), \quad i \in (1,...,m)$$

(III-8)

written in *Ito* differentials.

If we wish that the dynamical model of forward price should correspond to the pure Brownian motion in the curved manifold we have to equate eq. (III-7) to the first equation of system (II-1) also written in the *Ito* differentials

$$dX^i(t) = \sum_{q=1}^{n} (\frac{1}{2} d\Sigma_q^i + \Sigma_q^i) dW^q(t) = -\frac{1}{2} \sum_{j,k,q} \Gamma^i_{jk} \Sigma_q^k \Sigma_q^j dt + \sum_{q=1}^{n} \Sigma_q^i dW^q(t)$$

(III-9)

Then we have the equation for self-consistency of the model

$$\sum_{q=1}^{n} \Sigma_q^{i\,2}(t) dt = \sum_{q}^{n} \sum_{j,k}^{m} \Gamma^i_{jk} \Sigma_q^k \Sigma_q^j dt$$

(III-10)

Resolving this equation with respect to

$$A^i_{k,q} = \Gamma^i_{jk} \Sigma_q^j$$

we obtain

$$A^i_{k,q} = \Sigma_q^i \, \delta^i_k$$

Substituting this equation into the second one of (II-1) we come to the equation

$$d\Sigma_p^i(t) = -\Sigma_p^i \sum_{q=1}^{n} \Sigma_q^i dW^q(t)$$

(III-11)

Now our system is closed and self-consistent since the curvature of the term space is defined by the "force term" (the trend) in the equation for the price dynamics.

So we have to solve the following system of equations

$$dX^i(t) = \sum_{q=1}^{n} \Sigma_q^i dW^q(t)$$

$$d\Sigma_p^i(t) = -\Sigma_p^i \sum_{k=1}^{n} \Sigma_q^i dW^q(t)$$

that in *Ito* differentials read

$$dX^i(t) = -\frac{1}{2}\sum_{q=1}^{n}\Sigma_q^{i}{}^2(t)dt + \sum_{q=1}^{n}\Sigma_q^i dW^q(t)$$

$$d\Sigma_q^i = \Sigma_q^i \sum_{p}^{n}(\Sigma_p^i)^2 dt - \Sigma_q^i \sum_{p}^{n}\Sigma_p^i dW^p$$

(III-12)

From the first equation of (III-12) we infer that X has a Gaussian distribution. What about the volatility?

III.1 ONE - FACTOR REDUCTIONS OF THE MODEL.

Let us consider a one-factor reduction of the model. It makes sense since as is well-known some, may be even many power markets as well as financial ones show almost one-factor behavior: principal component analysis gives from 80% to even 90% of the total contribution to the main component according to Clewlow & Strickland (1999), Wilmott (2001).

So, one factor: $dW^p = dW$ and $\Sigma^i = Y$ then

$$dY = Y^3 dt - Y^2 dW$$

(III-13)

and the Fokker-Planck (FP) equation for the transition probability ρ reads

$$\partial_t \rho = \partial_y(-y^3 + \frac{1}{2}\partial_y y^4)\rho$$

(III-14)

Let us consider stationary solutions of (III-14). Then we have

$$\partial_y(-y^3 + \frac{1}{2}\partial_y y^4)\rho = 0$$

with a solution

$$\rho = \frac{cy + a}{y^3}$$

(III-15)

Let us consider the related *Stratonovich* process

$$d\Sigma_q^i = -\Sigma_q^i \sum_{p}^{n}\Sigma_p^i dW^p$$

or for a single-factor, single-term process $Z = \Sigma^i$ we have

$$dZ = -Z^2 dW$$

(III-16)

with the Fokker-Plank equation

$$\partial_t \rho = \frac{1}{2} \partial_z (z^2 \partial_z z^2) \rho$$

and stationary solutions

$$\rho = \frac{a}{z^3} \qquad \text{(III-17)}$$

Then we see that both processes have similar distributions if y is sufficiently small

$$y << \frac{a}{c}$$

It was the *Stratonovich* process. For the *Ito* process we have eq. (III-8). The mean can be estimated from the trend term by the following reasoning: taking average of the equation we have for $\Sigma = <\Sigma> + s$ and $<s> = 0$

$$<d\Sigma> = <\Sigma^3> = <s + <\Sigma>>^3 = 3 <s^2><\Sigma> + <\Sigma>^3$$

Where the variance $<s^2> \approx t <\Sigma>^4$ and the first term in the equation can be neglected. Now since $<d\Sigma> = <\Sigma_{t+1} - \Sigma_t> = d <\Sigma>$ we come to

$$d <\Sigma> = <\Sigma>^3 dt$$

with the solution

$$<\Sigma> = \sigma \sqrt{\frac{1}{1 - t\sigma^2}} \approx \sigma(1 + \frac{1}{2} t \sigma^2)$$

Armed with the above knowledge we can calculate the forward price curve. In order to obtain analytical estimate we restrict ourselves to "short time" horizons:

$$\sigma^2 t << 1 \qquad \text{(III-18)}$$

and consider only first two initial terms in the asymptotic expansion.

Earlier, we consider the statistical properties of the model and its short time horizons. In what follows we study the dynamics of the model in more detail.

Let us consider the single factor version of model (III-2). Then for $F(t,T)$ we have

$$\frac{dF(t,T)}{F(t,T)} = \Sigma(t,T)dW(t) \qquad \text{and} \qquad F(t,t) = S(t) \qquad \text{(III-19)}$$

or using Ito's lemma

$$d \ln F(t,T) = -\frac{1}{2} \Sigma^2(t,T)d\,t + \Sigma(t,T)d\,W(t)$$

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Integrating once we obtain

$$\ln \frac{F(t,T)}{F(0,T)} = -\frac{1}{2}\int_0^t \Sigma^2(u,T)du + \int_0^t \Sigma(u,T)dW(u)$$

or

$$F(t,T) = F(0,T)\exp\{-\frac{1}{2}\int_0^t \Sigma^2(u,T)du + \int_0^t \Sigma(u,T)dW(u)\} \qquad \text{(III-20)}$$

This solution is defined so far accurate to an arbitrary function $F(0,T)$. To restrict this freedom we can specify a random process for the spot price $S(t)$. Now since $S(t)= F(t,t)$, knowing the equation for $S(t)$ gives us the equation for $F(0,T)$ through the parameters involved in the equation for $S(t)$.

Let us consider a mean-reverting process for $S(t)$, viz.

$$\frac{dS}{S} = \alpha(\mu - \ln S)dt + \sigma(t)dW(t) \qquad \text{(III-21)}$$

This kind of processes is very popular in econometrics since, for example, forward prices as well as interest rates appear over time to be pulled back to some long average level. This phenomenon is known as *mean reversion*, see Hull 1993, p388. Also Schwartz's 1997 model for the commodity price dynamics used a single factor mean-reverting process. In fact, equation of type (III-21) appeared in physics long ago and was assumed to describe the so-called Ornstein-Uhlenbeck process, see appendix for more detail.

From the other hand, eqn. (III-20) gives

$$S(t) = F(0,t)\exp\{-\frac{1}{2}\int_0^t \Sigma^2(u,t)du + \int_0^t \Sigma(u,t)dW(u)\} \qquad \text{(III-22)}$$

i.e. $\ln S$ is normally distributed with

the mean = $\ln F - \dfrac{1}{2}\int_0^t \Sigma^2(u,T)du$

the dispersion = $\Sigma(t,t)$

Also from eqn. (III-22) taking the log we have

$$\ln S(t) = \ln F(0,t) - \frac{1}{2}\int_0^t \Sigma^2(u,t)du + \int_0^t \Sigma(u,t)dW(u) \qquad \text{(III-23)}$$

Then by differentiating over t one has

$$d \ln S(t) = [\frac{\partial \ln F(0,t)}{\partial t} - \frac{1}{2}\Sigma^2(t,t) - \int_0^t \Sigma(u,t)\Sigma_t(u,t)d\,u$$

$$+ \int_0^t \Sigma_t(u,t)dW(u)]dt + \Sigma(t,t)dW(t)$$

Easy to check out that from Ito's lemma follows that

$$d \ln S(t) + \frac{1}{2}\Sigma^2(t,t)dt = \frac{dS}{S}$$

Therefore

$$\frac{dS(t)}{S(t)} = [\frac{\partial \ln F(0,t)}{\partial t} - \int_0^t \Sigma(u,t)\Sigma_t(u,t)du + \int_0^t \Sigma_t(u,t)dW(u)]dt$$

$$+ \Sigma(t,t)dW(t)$$

(III-24)

Now if the spot process underlying the forward price dynamics is defined by eqn. (III-21) we have the self-consistent system of equations:

$$\sigma(t) = \Sigma(t,t)$$

(III-25a)

$$\alpha(\mu - \ln S) = \frac{\partial \ln F(0,t)}{\partial t} - \int_0^t \Sigma(u,t)\Sigma_t(u,t)d\,u + \int_0^t \Sigma_t(u,t)d\,W(u)$$

(III-25b)

Rewrite eqn. (III-23) in the form

$$\int_0^t \Sigma(u,t)dW(u) = \{\ln S(t) - \ln F(0,t)\} + \frac{1}{2}\int_0^t \Sigma^2(u,t)du$$

(III-26)

We can easily solve the system of equations (III-24), (III-25) and (III-26) if

1) *The first group of models*

$$\partial_t \Sigma_1(u,t) = -\alpha \Sigma_1(u,t)$$

(III-27)

Or in more general case

2) *The second group of models*

$$\partial_t \Sigma_2(u,t) = -\alpha \Sigma_2(u,t) + f(t)$$

(III-28)

where $f(t)$ is a known function of t. Those conditions allow canceling stochastic integrals in the equations (akin to the risk-less condition) and are necessary for solvability of the whole problem. They look very plausible for they mean that the volatility of the forward price decays from one level another or zero.

For the first group of models if we substitute eqn (III-27) into (III-25) and use (III-26) we come to
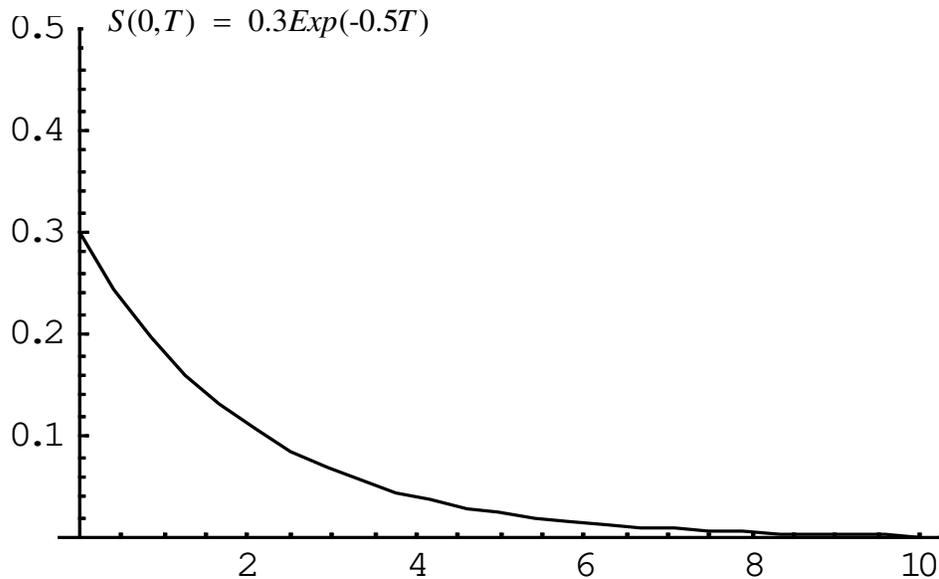
$$\frac{\partial \ln F_1(0,t)}{\partial t} + \alpha \ln F_1(0,t) = \alpha \left(\mu - \frac{1}{2} \int_0^t \Sigma_1^{\ 2}(u,t)du \right) \equiv \Phi_1(t) \qquad \text{(III-29)}$$

In general case (III-28) we have

$$\Phi_2(t) = \alpha \left(\mu - \frac{1}{2} \int_0^t \Sigma_2^{\ 2}(u,t)du \right) + \int_0^t \Sigma_2(u,t)f(u)du - \int_0^t f(u)dW(u) \qquad \text{(III-30)}$$

Equation (III-29) along with (III-30) can be readily solved, provided we know the forward price volatility $\Sigma(t,T)$.

Below we give the graphs of modeled volatility curves and real ones.

$S(0,T) = 0.3 Exp(-0.5T)$
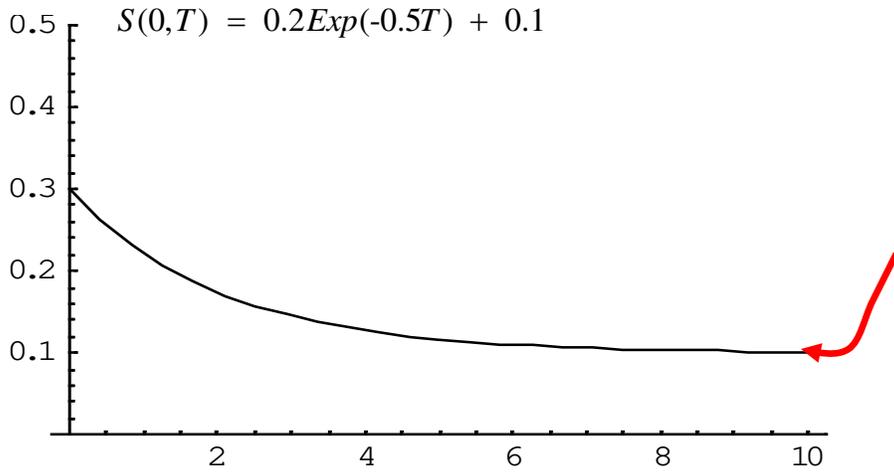
$$S(0,T) \;=\; 0.2\,Exp(-0.5T) \;+\; 0.1$$

Fig 1. Zero maturity asymptotic of the.
Fig. 2. Nonzero maturity asymptotic.

In reality we have the situation very close to the second curve, see Clewlow & Strickland 1999.
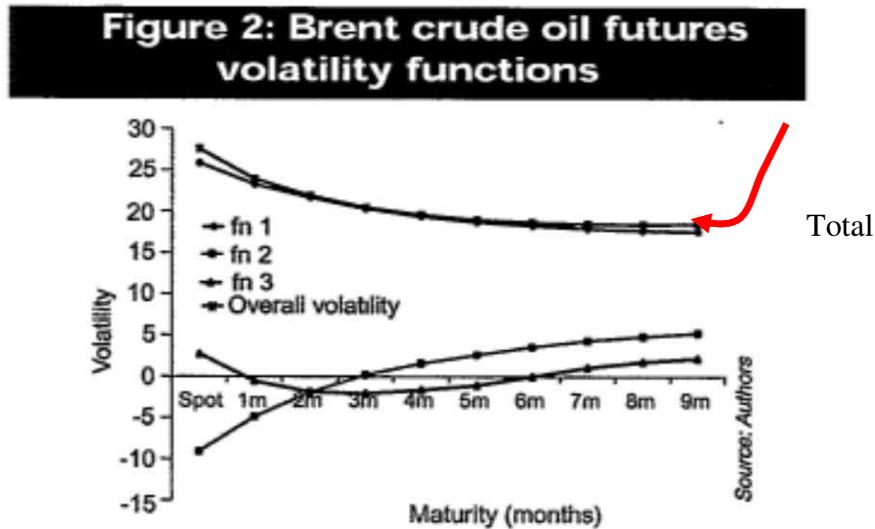

Fig. 3. Principal component analysis for crude oil.

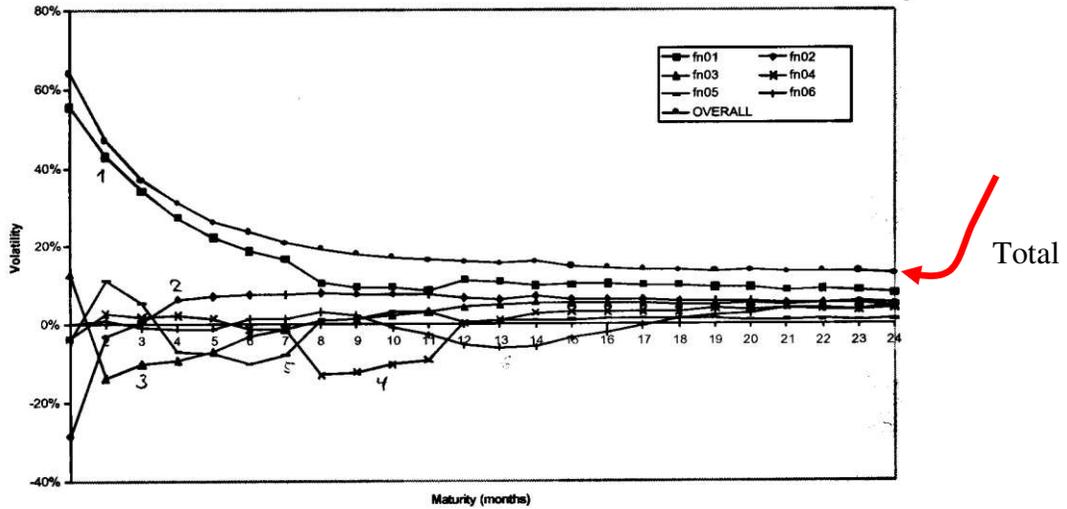## Volatility Functions for **NYMEX** Henry Hub Natural Gas

Fig. 4. Principal component analysis for Natural gas.

The solution of (III-29) (obtained by the variation of constant method) is

$$\ln F_i(0,t) = e^{-\alpha t}\{\int_0^t e^{\alpha u}\Phi_i(u)du + const\}$$

Since at $t = 0$ $\ln F_i(0,0) = \ln S_i(0)$ then

$$\ln F_i(0,t) = e^{-\alpha t}\{\int_0^t e^{\alpha u}\Phi_i(u)du + \ln S_i(0)\} \qquad\qquad \text{(III-31)}$$

We see μ not necessarily be a constant. It can be a function of time.
For example: if the forward price process volatility
1) is a regular function of time, e.g. a first class three-parameter exponential model

$$\Sigma_1(t,T) = \sigma_1 e^{-\alpha(T-t)} \qquad\qquad \text{(III-32a)}$$

and

2) $\mu(t) = \mu_1 e^{\kappa t}$ (a constant κ can be both negative or positive)

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Then the integral is exactly evaluated as

$$F_1(0,T) = \exp\{e^{-\alpha T}\ln S_1(0) + \frac{\alpha\mu_1}{\alpha+\kappa}(e^{\kappa T} - e^{-\alpha T}) - \frac{\sigma_1^2}{4\alpha}(1-e^{-\alpha T})^2\} \qquad \text{(III-33)}$$

Now by setting $\kappa = 0$ we come to the well-known result, Clewlow and Strickland 1997:

$$F_{1R}(0,T) = \exp\left\{e^{-\alpha T}\ln S_1(0) + \mu_1(1-e^{-\alpha T}) - \frac{\sigma_1^2}{4\alpha}(1-e^{-\alpha T})^2\right\} \qquad \text{(III-34)}$$

It's easily seen that in this case the forward price curve is regular with zero volatility.

A bit more tedious calculations are needed to get the forward price curve for the second class four-parameter model:

$$\Sigma_2(t,T) = \sigma_1 e^{-\alpha(T-t)} + \sigma_0 \qquad \text{(III-32b)}$$

In this case we have got

$$<\ln F_{2R}(0,T)> = e^{-\alpha T}S_2(0) + \mu(1-e^{-\alpha T}) - \frac{1}{4\alpha}\left[\sigma_1^2(1-e^{-\alpha T})^2 - 2\sigma_0^2(\alpha T - 1 + e^{-\alpha T})\right] \qquad \text{(III-35)}$$

and

$$Var\left[\ln F_{2R}(0,T)\right] = \frac{\sigma_0^2}{2\alpha}\left[2e^{-\alpha T}(1-\alpha T + \alpha^2 T^2) - e^{-2\alpha T}(2+\alpha^2 T^2)\right] \qquad \text{(III-36)}$$

I.e. in the second case, even for regular volatility of the forward price process $\Sigma(t,T)$ the forward price curve itself $F_{2R}(t,T)$ becomes stochastic with nonzero volatility (III-36).

Even more tedious calculations needed to take into account the stochasticity of the forward price process volatility $\Sigma(t,T)$. For that we should solve the second equation of the stochastic differential geometry (III-12). This equation is self-consistent and can be separately analyzed.

Under the same assumption (single-factor model) for Stratonovich process it reads,

$$d\Sigma_{1St}^{str}(t,T) = -(\Sigma_{1St}^{str})^2(t,T)d\ W(t) \qquad \text{(III-37)}$$

and

$$d\Sigma^i_{1St} = (\Sigma^i_{1St})^3 \, dt - (\Sigma^i_{1St})^2 \, dW \qquad \text{(III-38)}$$

for the Ito process.

From eqns. (III-15) and (III-17) one can see that the solutions to both FP equations are identical if $c \ll a$ or, e.g. when $c = a$. It means that in our case the *Stratonovich* process distributions are a subclass of the more general Ito's distributions.

Since for *Stratonovich* processes we have the conventional calculus, we get

$$\Sigma^{str}_{1St}(\tau,t) = \frac{\sigma(0,t)}{1 + \sigma(0,t)W(\tau)}$$

or

$$\Sigma^{str}_{1St}(t,T) = \frac{\sigma(0,T)}{1 + \sigma(0,T)W(t)} \qquad \text{(III-39)}$$

and

$$\partial_t \Sigma^{str}_{1St}(\tau,t) = \frac{\sigma_t(0,t)}{(1 + \sigma(0,t)W(\tau))^2}$$

In eq. (III-39) $W(t)$ is a standard Wiener process with mean-less and unity-variance Gaussian distribution. So we can express $W(t)$ as a function of Σ^i

$$W(t) = \frac{1}{\Sigma^{str}(t)} - \frac{1}{\Sigma^{str}(0)} \qquad \text{(III-40)}$$

Also it is easy to calculate the mean and variance of $\Sigma^i(t,T)$ for small t (short horizons)

$$E[\Sigma^i(t,T)] \approx \sigma^i(0,T)\{1 + t\,\sigma^i(0,T)^2\}$$
$$Va\ [\Sigma^i(t,T)] \approx t\,\sigma^i(0,T)^4\{1 + 6t\,\sigma^i(0,T)^2\}$$

It should be mentioned that the series over t are asymptotic and in principle are divergent and, strictly speaking, the only point of convergence is $t = 0$ even without its neighborhood. Also as could be expected the results are independent of a sign of the random term.

Now for the Stratonovich process we can go even further. Due to (III-40) we calculate the distribution for $\Sigma^{str}(t,T)$ by means of the formula

$$\rho(w)dW = \rho(1/\Sigma^{str})\frac{dw}{d\Sigma^{str}}d\Sigma^{str} \equiv \rho(\Sigma^{str})d\Sigma^{str}$$

And for $\rho(w)$ is a Gaussian distribution we have

$$\rho(\Sigma^{str}) = \frac{1}{\sqrt{2\pi t}}\exp\left(-\frac{(1/\Sigma^{str}-1/\Sigma_0^{str})^2}{2t}\right)\frac{d\Sigma^{str}}{(\Sigma^{str})^2}$$

a distribution for a reciprocal of W(t).

Let us now continue discussing the dynamic properties of the forward curves.

In the first case, i.e. $\sigma(0,t) = \sigma_1 e^{-\alpha t}$, we have

$$\partial_t\Sigma_{1St}^{str}(\tau,t) = -\alpha\frac{\sigma(0,t)}{(1+\sigma(0,t)W(\tau))^2} = -\frac{\alpha\Sigma_{1St}^{str}(\tau,t)}{(1+\sigma(0,t)W(\tau))} \qquad \text{(III-41)}$$

Assume also that

$$\sigma(0,T)W(T) \ll 1 \qquad \text{(III-42)}$$

The last condition is very essential for the evaluations. This is because

$$\sigma(0,T)W(t) = \sigma_1 e^{-\alpha T}W(t) \Rightarrow^{t\leq T} \sigma_1 e^{-\alpha t}W(t)e^{-\alpha(T-t)} \leq \sigma_1 e^{-\alpha t}W(t) \approx \sigma_1 e^{-\alpha t}\sqrt{t}$$

Here are two cases:

1) $\alpha T \ll 1$ then $\sigma_1^2 T \ll 1$ and
2) $\alpha T \gg 1$ $\sigma_1^2 T$ is arbitrary

Now we have

$$\partial_t\Sigma_{1St} = -\alpha\Sigma_{1St}(u,t) + f_1(u)$$
$$f_1(u) = \alpha W(u)\sigma^2(0,t)[1-2\sigma(0,t)W(u)] \qquad \text{(III-43)}$$

And

$$\frac{1}{\alpha}\Phi_1(u) = \mu + 2\sigma^3(0,u)\int_0^u W(\tau)d\tau - \frac{1}{2}\sigma^2(0,u)W^2(u) - \frac{9}{2}\sigma^4(0,u)\int_0^u W^2(u)d \qquad \text{(III-44)}$$

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Finally, we can solve eqns. (III-29) and (III-44) together to obtain $\ln F(0,T)$ as a stochastic process with the mean and volatility

$$< \ln F_{1St}(0,T) >= \mu(1-e^{-\alpha T}) + e^{-\alpha T}\left\{ \ln S(0) - \frac{1}{2\alpha}\sigma_1^2[1-(1+\alpha T)e^{-\alpha T}] \right.$$

$$\left. -\frac{\sigma_1^4}{6\alpha^2}[1-(1+3\alpha T+\frac{1}{2}(3\alpha T)^2)e^{-3\alpha T}] \right\}$$

(III-45)

and

$$Var[\ln F_{1St}(0,T)] = 4\alpha^2\sigma_1^6 e^{-2\alpha T}\int_0^T e^{-2\alpha(u+z)}du\int_0^T dz\int_0^u < W(\tau)d\tau\int_0^z W(t) > dt$$

$$= \frac{\sigma_1^6}{12\alpha^2}Te^{-4\alpha T}\left\{ -3+2\alpha^2 T^2 + e^{-2\alpha T}(3+6\alpha T+4\alpha^2 T^2) \right\}$$

(III-46)

In what follows we will compare our result with the conventional case

$$\mu = const \text{ and } \Sigma_{1R}(t,T) = \sigma_1 e^{-\alpha(T-t)}$$

with

$$\ln F_{1R}(0,T) = \mu(1-e^{-\alpha T}) + e^{-\alpha T}\ln S_1(0) - \frac{\sigma_1^2}{4\alpha}(1-e^{-2\alpha T})^2$$

for specific values of the parameters involved.

By looking at (II-22) and (II-24) we see that the **means** differ by the terms proportional to some power of σ_1.

Underline again that the volatility of the spot price process $S(t)$ is defined as $\sigma(t) = \Sigma(t,t)$ therefore

$$\sigma_1(t) = \frac{\sigma_1 e^{-\alpha t}}{1+\sigma_1 e^{-\alpha t}W(t)} \approx \sigma_1 e^{-\alpha t}[1-\sigma_1 e^{-\alpha t}W(t)] \approx \sigma_1 e^{-\alpha t}$$

(III-47)

Then for the second class four-parameter model:

$$\partial_t \Sigma_{2St}(0,t) = -\alpha(\Sigma_{2St}(0,t)+\sigma_0), \quad \text{i.e.} \quad \sigma(0,t) = \sigma_1 e^{-\alpha t} + \sigma_0$$

(III-48)

and

$$\sigma_2(t) \approx \frac{\sigma_0}{1+\sigma_0 W(t)}, \quad \text{at } \alpha T >> 1$$

i.e. the spot price volatility falls down with time that looks plausible for the mean reverting process. Now instead of

$$f_1(u) = \alpha W(u)\sigma^2(0,t)[1 - 2\sigma(0,t)W(u)]$$

we have

$$f_2(u) = \alpha\sigma_0 + \alpha\sigma(0,t)W(u)[\sigma(0,t) - 2\sigma_0 - 2\sigma(0,t)(\sigma(0,t) - \frac{3}{2}\sigma_0)W(u)]$$

and

$$\Phi_2(t) = \alpha\mu - \alpha\frac{\sigma(0,t)}{2}\int_0^t \{(\sigma - 2\sigma_0) - 2\sigma(2\sigma - 3\sigma_0)W(u) + 3\sigma^2(3\sigma - 4\sigma_0)W^2(u)\}du$$

Then up to the lowest orders terms with respect to $W(t)$ we obtain

$$< \ln[F_{2St}(0,T)] > = e^{-\alpha T}\ln S_2(0) + \mu(1 - e^{-\alpha T}) + \frac{1}{2\alpha}\{(\sigma_1^2 - \sigma_0^2)e^{-\alpha T}$$
$$- [\sigma_1^2(1 + \alpha T)e^{-2\alpha T} - \sigma_0^2(1 - \alpha T)]\}$$

(III-49)

and

$$Var[\ln F_{2St}(0,T)] = \frac{\sigma_0^2}{2\alpha}\{e^{-2\alpha T}(4e^{-\alpha T} - 1) + (2\alpha T - 3)\}$$

(III-50)

The term structures of all these models, viz. F_{1R}, F_{1St}, F_{2R} and F_{2St} for specific values of parameters: $\alpha, \mu, \sigma_0, \sigma_1$ and various $S(0)$ are given at Figs. 5 – 8.
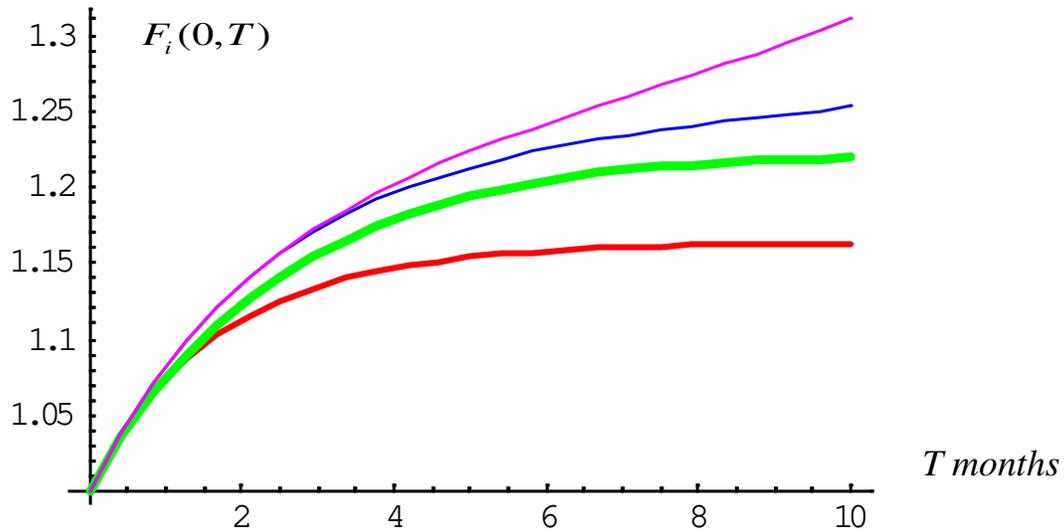
Fig. 5. Plots of $F_i(0,T)$ as functions of the maturity T. The lowest (red) stands for F_{1R}, the next from it up (green) stands for F_{1St}, one more up (blue) for F_{2R}, and the uppers (magenta) for F_{2St}, whereas $S(0)=1$.
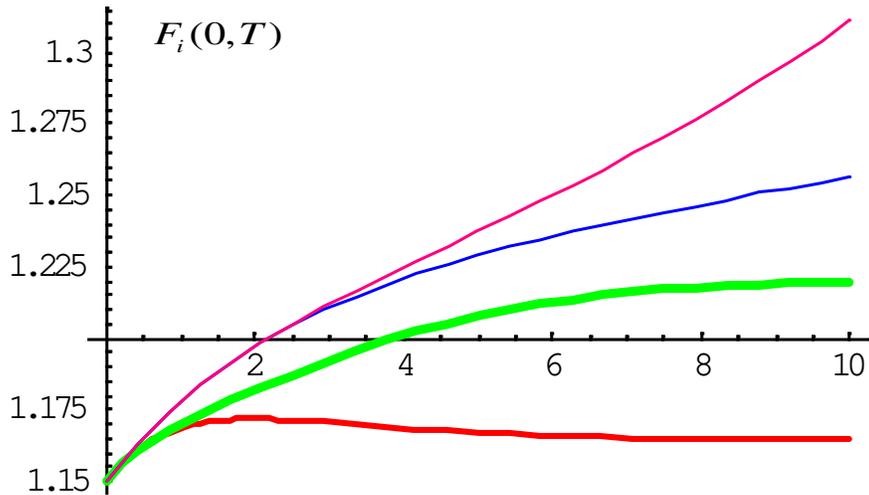

Fig. 6. $F_i(0,T)$ for $S(0)=1.15$.

Fig. 7. $F_i(0,T)$ for $S(0)=2$.
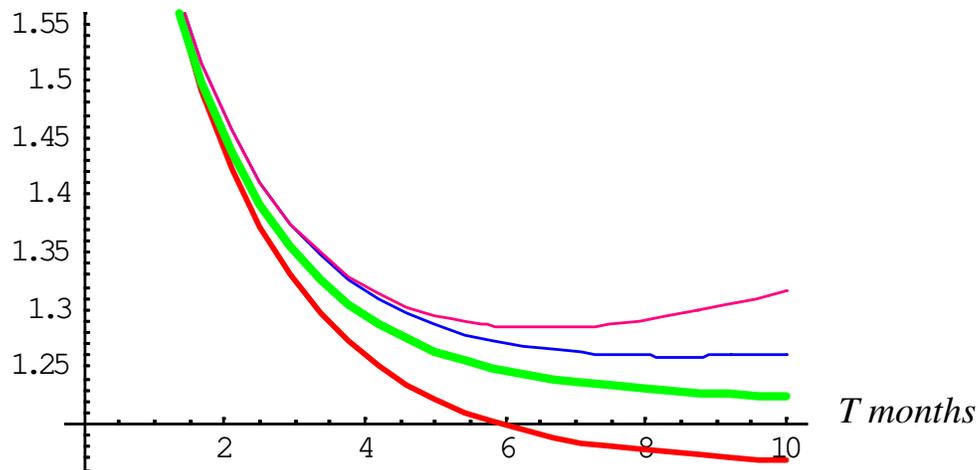
In all calculations we put $\alpha=0.5$, $\mu=0.2$, $\sigma_1=0.14$, $\sigma_0=0.17$. what is very close to the values at the **NYMEX Crude Oil** market.
We see that for the short terms (around during two months) and out of the vicinity of $S(0)=1.15$ all four curves look very close to each other and then they start do disperse about 7-10 % at the end of a year. The vicinity of the point

$S(0) = 1.15$ is the specific one. The function $F_{1R}(T)$ has a humped shape there, and quite essentially differs from the other three curves that have no humps.

For the first exponential model the variances for both regular and stochastic variants are negligible. For the second model plots of the stochastic variance (yellow or the top curve), the "regular" one (red or the middle curve) and the difference between them (magenta, the bottom curve) are as follows.
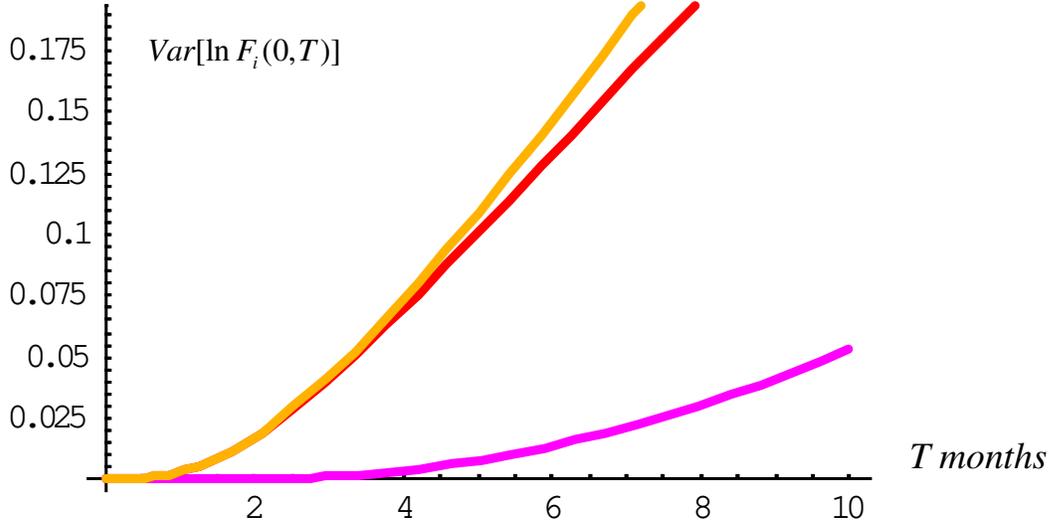

Fig. 8. Variances for the second stochastic and regular models and their difference.

IV. DISCOUNT BOND INTEREST RATES DYNAMICS.

Definitions. $P(t,T)$ *is the zero coupon (discount) bond price* at time t with principle $P(T,T) = \$1$, maturing at $t = T$.

$r(t)$ is the *short term interest rate*.

$v_p(t,T)$ is the *volatility* of $P(t,T)$ corresponding to the p-th component of a n-dimensional vector Wiener process: $d\vec{W} = \{dW^1, ..., dW^n\}$.

$R(t,T)$ is the *rate* of in interest:

$$R(t,T) = -\frac{\ln P(t,T)}{T-t}, \qquad 0 < t \le T \tag{IV-1}$$

such that

$$P(t,T) = e^{-R(t,T)(T-t)} \tag{IV-2}$$

and $R(t_0,T)$ is a *yield curve*.

$F(t,T)$ is the *instantaneous forward rates* $(t \le T)$ defined by the equation

$$F(t,T) = -\frac{\partial \ln P(t,T)}{\partial T}, \qquad 0 < t \leq T \qquad\qquad\text{(IV-3)}$$

and finally

$$F(t,T) = R(t,T) + (T-t)\frac{\partial R(t,T)}{\partial T} \qquad\qquad\text{(IV-4)}$$

with the boundary condition at $t = T$:

$$F(t,t) = R(t,t) = r(t) \qquad\qquad\text{(IV-5)}$$

Due to the weak market efficiency, $P(t,T)$ along with the other variables are supposed to follow certain Markov processes. The most general of them is

$$d\,P(t,T) = P(t,T)[\mu(t,T)d\,t + \sum_{p=1}^{n} v_p(t,T)d\,W^p], \quad p = 1,..\,n, \qquad\qquad\text{(IV-6)}$$

This process is determined by $(n+1)$ unknown functions $\mu(t,T)$ and $v_p(t,T)$. In the arbitrage-free case, Hull 1993 this equation rewritten for $F(t,T)$ was shown to be reduced to a simpler equation only containing the unknown functions $v_p(t,T)$:

$$d\,F(t,T) = \sum_{p=1}^{n}\{-\sigma_p(t,T)v_p(t,T)d\,t + \sigma_p(t,T)d\,W^p\} \qquad\qquad\text{(IV-7)}$$

where

$$\sigma_p(t,T) = -\frac{\partial v_p(t,T)}{\partial T} \qquad\qquad\text{(IV-8)}$$

The **internal space** or the system state space is a discrete version of the bond term space (the space of bond maturing). This space is an N-dimensional vector space with the metric $E[A(t)B(t)]$ and is created by the vectors

$$\vec{X}(t) = \{X_1(t),...,X_N(t)\} \equiv \{F(t,T_1),...,F(t,T_N)\}$$

A point $\vec{X}(t) = \{X_1(t),...,X_N(t)\}$ in this space corresponds to a portfolio (a set) of different bonds, and portfolio dynamics is a movement of the corresponding point in the state space which can be a curved manifold, $R \neq 0$.

This movement is governed by the stochastic differential equations and to solve the pricing problem we have to specify initial and boundary conditions: $F(0,T), F(t,0)$ and $F(t,T_N)$.

So we have the following scheme of logic steps:

1) The weak efficiency of market defines SDE (IV-6) with the freedom defined by unknown functions $\mu(t,T)$ and $v_p(t,T)$ of two variables.

2) The arbitrage freedom reduces this freedom by one function $\mu(t,T)$ and gives the drift term expressed through volatility as

$$f(t,T)d\, t = -\sum_{p=1}^{n}\sigma_p(t,T)v_p(t,T)d\, t \qquad \text{(IV-9)}$$

Such that we have N unknown functions, viz. the volatility vector $\sigma_p(t,T)$.

1) The "fair game" rule (in fact related to the previous) allows one to write down equations of SDE to find the discrete version of $\sigma_p(t,T)$. What is important that the drift term (IV-9), as we have seen, determines the geometric structure of the state space, namely Γ^i_{jk}, see eq. (II-6).

2) The initial and boundary conditions make the problem completely defined, and the freedom left is the two functions of one variable $F(t,0)$ and $F(t,T_N)$, and the set of parameters $F(0,T_i)$.

As a result, we obtain the closed problem to find $F(t,T_i)$ and $\sigma_p(t,T_i)$ with $i = 1,\ldots, N$. Notice that the term axis T is in fact always discrete since bonds are issued by lumps rather than continuously. Also we have to underline that due to the fair game rule the state space structure does not depend on time: specifying $\sigma_p(t,T)$ at some moment t defines this structure, i.e. $\Gamma^i_{jk}, g_{ij}, R_{ijkl}$, and R in the future. This is why the stochastic processes with an internal space, in other words the portfolios, can be classified by conexion curvature of this space:

1. *A zero curvature means a trivial market: no correlations among bonds of different terms. Bonds with various T live independently and their behavior corresponds to pure diffusion on a plane.*
2. *A nonzero curvature implies a rich dynamics and the structure of the bond market is with drifts, correlations and so forth.*

Now the equations of the stochastic differential geometry for the bond forward rates dynamics are

$$d_s X^i(t) = \sum_{p=1}^{n}\sigma^i_p(t)d\, W^p(t) \qquad \text{(IV-10)}$$

$$d_s\sigma^i_q(t) = 2\sum_{p=1}^{q}\sigma^i_p(t)\sum_{r=1}^{n}\sigma^i_r(t)dW^r(t) \qquad \text{(IV-11)}$$

They are written in the Stratonovich form and obtained by equating the Ito drift term of eq. (IV-10) to drift term (II-5a) that related to the structure of the system state space (a set of portfolios). Some solutions to this system of

equations can be derived in a special case of a single factor model and continuous form.

IV.1 REGULAR VOLATILITY

First we study solutions for the system in the case of one factor model and regular volatilities $\sigma_p(t,T)$. Moreover as we did earlier for forward price dynamics we suggest that spot interest rates follow a mean reverting process of the type (III-21)

$$dr(t) = a(b-r)dt + s(t)dW \qquad\qquad\qquad\text{(IV-12)}$$

Again we consider for the volatility in (IV-12) two different approaches (III-27) and (III-28) or more precisely (III-32a) and (III-32b). It turns out that, like in the case of forward price dynamics, equations (III-27) and (III-28) are sufficient in order for stochastic integrals to cancel each other. In the first approach

$$\sigma_1(t,T) = \sigma_1 e^{-a(T-t)}$$

we come to the so-called Vasicek econometric solvable model, 1997 and instead of (IV-10), (IV-11) we have

$$d\,F(t,T) = -\sigma(t,T)v(t,T) + \sigma(t,T)d\,W$$

$$d\sigma(t,T) = -a\sigma(t,T)$$

and

$$\frac{\partial F(t,T)}{\partial t} + aF(t,T) = ab - \int_t^T \sigma^2(\tau,t)d\tau$$

Such that

$$F(t,T) = e^{-a(T-t)}r(t) + b(1 - e^{-a(T-t)}) - \frac{1}{a}\sigma(t,T)v(t,T)$$

Now since

$$P(t,T) = e^{-\int_t^T F(t,\tau)d\tau} \equiv e^{-R(t,T)(T-t)}$$

we finally have Vasicek's result:

$$P(t,T) = A(t,T)e^{-B(t,T)r}$$

with

$$A_1(t,T) = e^{C_1(t,T)} \equiv \exp\left\{\frac{(B(t,T)-(T-t))(a^2 b - \sigma_1^2/2)}{a^2} - \frac{\sigma_1^2 B^2(t,T)}{4a}\right\}$$

$$B(t,T) = \frac{1}{a}(1 - e^{-a(T-t)})\big|_{t\to T} \approx (T-t)$$

More cumbersome calculations allow one to derive analogous but longer formulae in the second case

$$\sigma(t,T) = \sigma_1 e^{-a(T-t)} + \sigma_0$$

In addition to Vasicek's result we now have terms proportional to $\sigma_0\sigma_1$ and σ_1^2. Also a tiny volatility of the forward rate arises as it was in the case of the forward price dynamics, see (III-50).

$$A_2(t,T) = e^{C_2(t,T)}$$

$$C_2(t,T) = C_1(t,T) + \frac{\sigma_0\sigma_1}{2a^2}(-2B(t,T) + (T-t)(2e^{-a(T-t)} + a(T-t)) + \frac{\sigma_1^2}{6}(T-t)^3$$

IV.2. STOCHASTIC VOLATILITY

Let us come back to the self-consistent case of stochastic volatility. We can derive some solutions if we proceed to the continual form of (IV-10) and (IV-11), viz.

$$d_s F(t,T) = \sum_{p=1}^{n} \sigma_p(t,T)d\,W^p$$

$$d_s\sigma_q(t,T) = \left(2\int_t^T \sigma_q(t,\tau)d\tau\right)\sum_{r=1}^{n}\sigma_r(t,T)d\,W \tag{IV-12}$$

In a single factor model this system admits exact solutions in terms of stochastic integrals. We put

$$dW^p = dW\delta^{pi}, \quad \sigma_i = \sigma$$

then

$$d\sigma(t,T) = 2\left(\int_t^T \sigma(t,\tau)d\tau\right)\sigma(t,T)d\,W(t)$$

$$dF(t,T) = \sigma(t,T)dW(t) \tag{IV-13}$$

In order to integrate this system of equations we again consider discretization over T in the form:

$$F(t,T) = F(t, j\Delta T) \equiv F_j(t)$$

$$\sigma(t,T) = \sigma(t, j\Delta T) \equiv \sigma_j(t), \qquad t \in (j-1, j)\Delta T$$

Now we substitute the smooth functions with their step-wise approximation.

In the first interval $0 \leq t \leq \Delta T$ we have

$$d\sigma_i(t) = 2\left((\Delta T - y)\sigma_1(t) + \Delta T \sum_{j=2}^{i} \sigma_j(t) \right)\sigma_i(t)dW \tag{IV-14}$$

$$dF_i(t) = \sigma_i(t)dW$$

Wherefrom for $\sigma_1(t)$ we have the equation

$$d\sigma_1(t) = 2(\Delta T - t)\sigma_1^2(t)dW, \qquad t \in (0, \Delta T]$$

with the solution

$$\sigma_1(t) = \frac{\sigma_1(0)}{1 - 2\sigma_1(0)\int_0^t (\Delta T - \tau)dW(\tau)}, \qquad \sigma_1(0) = \sigma_1(t = 0) \tag{IV-15}$$

This solution is defined by the following Stratonovich integral

$$I_1(t) = \int_0^t (\Delta T - \tau)dW(\tau) \tag{IV-16}$$

in the denominator of the r.h.s. of (IV-15). This integral is easy estimated giving

$$E[I_1] = 0$$

$$VAR[I_1] \equiv E[I_1^2] - E^2[I_1] = \int_0^t (\Delta T - \tau)^2 d\tau = \frac{t}{3}(t^2 + 3\Delta T(\Delta T - t))$$

therefore

$$VAR\big|_{\max} = VAR(t = \Delta T) = \frac{1}{3}(\Delta T)^3$$

and the standard deviation

$$SD_{\max} = \frac{1}{\sqrt{3}}(\Delta T)^{3/2} \tag{IV-17}$$

In order for the solution (IV-15) to be finite the denominator should be positive, i.e.

$$d_1 = 1 - 2\sigma_1(0)I_1 > 0$$

or

$$\kappa \frac{2}{\sqrt{3}} \sigma_1(0)(\Delta T)^{3/2} < 1 \tag{IV-18}$$

For estimations, the numerical factor κ in this equation may be taken less than three.

If this condition breaks down the solution becomes singular, and the market loses stability. It is interesting to notice that the stability condition (IV-18) along with the initial volatility $\sigma_1(0)$ contains the term structure step ΔT and the less this step the stable the market.

After the time $t = \Delta T$ the first bond matures and dies, such that the second one becomes the first, the third the second and so on. This process repeats periodically with ΔT.

The equation for $\sigma_2(t)$ is

$$d\sigma_2(t) = 2\{(\Delta T - t)\sigma_1(t) + \Delta T \sigma_2(t)\} \sigma_2(t)dW \tag{IV-19}$$

or

$$\frac{d\sigma_2}{dt} = \alpha_2(t) + \sigma_2^2, \qquad dt = 2\Delta T\, dW$$

with the solution

$$\sigma_2(t) = \frac{\sigma_2(0)\exp\left\{2\Delta T \int_0^t \alpha_2(\tau, W(\tau))dW(\tau)\right\}}{1 - 2\Delta T \sigma_2(0)\int_0^t dW(x)\exp\left\{2\Delta T \int_0^x \alpha_2(\tau, W(\tau))dW(\tau)\right\}}, \qquad 0 < t \le \Delta T \quad \text{(IV-20)}$$

and

$$\alpha_2(t) = \frac{\Delta T - t}{\Delta T} \sigma_1(t)$$

This solution may be easily generalized for the arbitrary $\sigma_i(t)$ by the substitution $2 \to i$ and

$$\alpha_i(t,W) = \frac{\Delta T - t}{\Delta T} \sigma_1(t,W) + \sigma_2(t,W) + \ldots + \sigma_{i-1}(t,W) \tag{IV-21}$$

Solutions (IV-15), (IV-20) and (IV-21) are valid within one time step $0 < t \le \Delta T$, and we have assumed that short term bonds maturing for the time essentially

less than ΔT are absent. Mathematically this means the trivial (zero) boundary condition at the left end of the interval, $t = 0$. Actually, at the market there always are short-term bonds (e.g. overnight bonds) which effect can be modeled by a non-trivial boundary condition at the left end. Then the formula (IV-20) is still valid and eq. (IV-21) is changed as follows:

$$\alpha_i(t,W) = \frac{\Delta T - t}{\Delta T}\sigma_1(t,W) + \sigma_2(t,W) + . . + \sigma_{i-1}(t,W) + \frac{v_0(t)}{\Delta T}$$

where $v_0(t)$ is the specified volatility of the short bond price. In such a way we define the boundary condition at the left end of the bond maturity chain.

Now we can solve the whole problem by first integrating the equation for the forward rates

$$d_s F(t,T) = \sigma(t,T)d\ W$$

and then evaluating the integral

$$R(t,T) = \frac{1}{T-t}\int_t^T F(t,\tau)d\tau$$

Note that the specific behavior of the system now depends on either initial conditions for short range dynamics or boundary conditions for long range epochs and steady states.

To allow for the long term bond effect we assume that at each jump at a moment $j\Delta T$ a new bond with volatility σ_N is born at the right end of the chain, for instance with
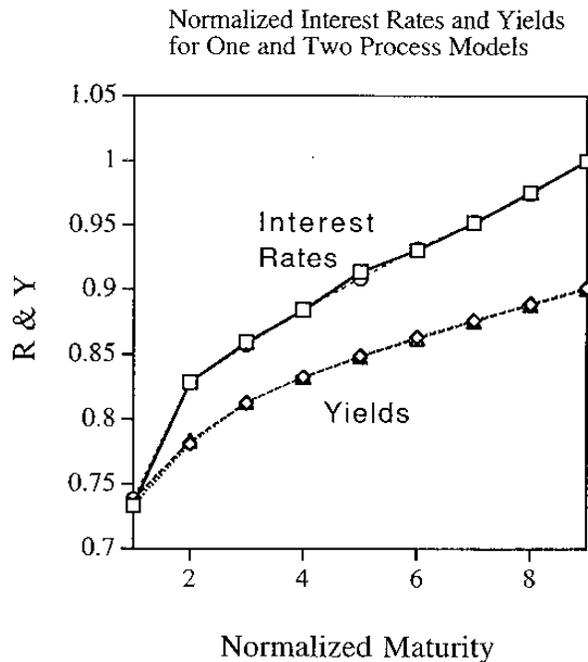
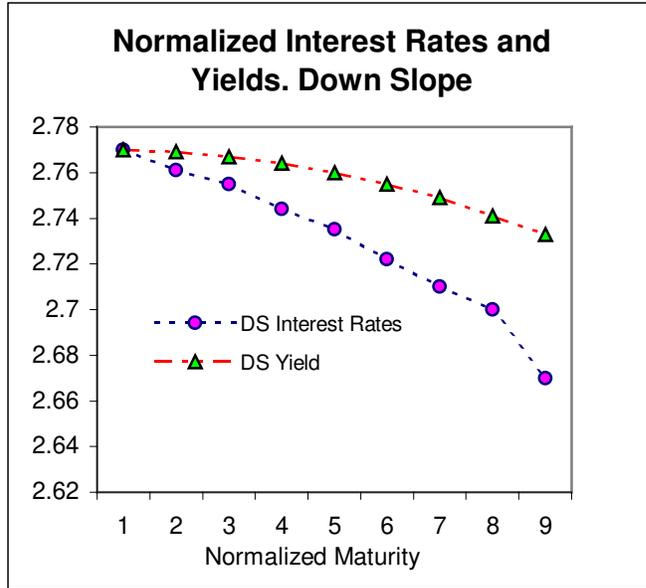$$\sigma_N = const = \sigma$$

V. FINAL COMMENTS AND CONCLUSIONS

The important influence that can cause the boundary conditions to stochastic stationary distributions were formulated in the papers Makhankov et al 1995 and Makhankov 1997 accompanied with numerical studies. Their results and the current ones regarding this work, allow us to make the following conclusions:
1. Boundary conditions and initial conditions must be specified from the market data on the basis of the known methods such as time series, *ARCH* and *GARCH* models and so on.
2. Computer experiments depending on the boundary conditions revealed the following scenarios in the market dynamics:

i) Explosive instability of the solution that leads to unpredictability of the market behavior.

ii) Various types of stationary solutions for the interest rates/yield curve: monotonic upward shape and monotonic downward shape (see Figs. 9 and 10), well known in the literature, Hull 1993. The curve shape is completely determined by the boundary conditions and especially at the left end of the maturity chain and hence by the economy as a whole.

iii) The monotonic upward slope shape of the curve reflects the normal stable market, and the volatility of short term bonds is substantially greater than that of long term ones.

iv) Instability of the market is predicted to occur if the volatility of forward rates exceeds a certain threshold. Usually this instability follows the development of a monotonic downward slope curve as values of volatilities increase at the boundary.

v) The theory is based on the three cornerstones: the weak efficiency of the bond market, arbitrage freedom and the "fair game" rule. Violation of one of them should make the predictions invalid.

Normalized Interest Rates and Yields
for One and Two Process Models

Interest
Rates

Yields

Normalized Maturity

Normalized Interest Rates and Yields. Down Slope

Figs. 9 and 10. Simulated stationary curves for the interest rates and yields:

1) Up-slope curve: 30 years max-maturity, $\sigma(T_N) = -0.7\%$, $v_0 = 52\%$.

2) Down-slope curve: 30 years max-maturity, $\sigma(T_N) = 0.05\%$, $v_0 = 5.2\%$

Figs. 9 and 10 look very plausible, see Hull 1993 Sec 4.1.

Also in the papers cited, the question of what is the dimension of the Wiener process that generates the stochastic behavior at the market was studied. And they stated that a single factor Wiener process may only match short term volatilities of forward rates. If we, following a typical assumption that interest rates volatilities are constant then forward rate dynamics is completely described by two independent Wiener processes. However, in general case long term dynamics requires more than two Wiener processes since, according to the above theory, volatilities change in time and the solution found gives a simple estimate of the time rate for this changing. So in general, the long term dynamics is described by multi-dimensional Wiener process. Evidently, this should count at least three not necessarily equal in contribution. However, due to definite degeneration of the market parameters the minimal number of stochastic processes that operate under various market conditions should be extracted from market data. This picture seams being real due to the principal component analysis, see pictures 3 and 4 and also, e.g. Wilmott 2001, Clewlow and Strickland 1999.

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Annex 1. Equations of Stochastic Differential Geometry.

Let us consider a "pure" Brownian motion on a sphere, and first build a frame bundle on it. Consider a point X_1 on $\mathbf{S}^2$ and a patch of a tangent plane in this point, we denote it as $T_{X_1}\mathbf{S}^2$. It is called a fibre in the point X_1. Then we proceed to a neighboring point X_2 and, doing the same, we get $T_{X_2}\mathbf{S}^2$. In such a way, we can cover all the sphere with these patches sticking them along the lines of their intersections like a soccer ball that gives us a polyhedron. So we have an example of the fibre bundle with the sphere being the base of it and the polyhedron a bundle of fibres (in fact, a bundle of frames in our case). We call this polyhedron a "covering" of the sphere. So finally we have got:
  a) the sphere which is curved (a manifold),
  b) the covering which is a Euclidean space.
  A pure Wiener process (a martingale) satisfying the equations

$$< dW^q dW^p >= \delta^{qp} dt$$

occurs in the Euclidean world. The same takes place for semi-martingales (approximately), for only in a Euclidean space it is possible to represent a stochastic process in the semi-martingale form

$$d\tilde{W}^q = \alpha^q dt + dW^q$$

This means that Wiener processes can only appear on the covering while a particle is moving on the sphere. Now we should adjust both phenomena. Let us consider a covering "boiling" with fluctuating forces (Wiener processes) and a particle in the point X_1 on the sphere. This point also belongs to the fibre $T_{X_1}\mathbf{S}^2$. Hence the particle undergoes a random shock

$$d\vec{X}_1 = \hat{\sigma}_1 d\vec{W}_1$$

jumping into a point X_2 on the sphere. In this point it again undergoes a shock

$$d\vec{X}_2 = \hat{\sigma}_2 d\vec{W}_2$$

and so forth. The matrix $\hat{\sigma}$ defines particle mobility (sensitivity). Here we should emphasize that all differentials considered above are of the Stratonovich type, which allows us to use the standard differential calculus.

Now we assume that all $d\vec{W}_i$ are the same in distribution (pure Brownian motion). And we have to connect $\vec{\sigma}_1$ and $\vec{\sigma}_2$. Note that matrix $\hat{\sigma}$, being in fact a rotating operator, can be constructed out of two vectors $\vec{j}_1$ and $\vec{j}_2$:

$$\hat{\sigma} = \begin{pmatrix} j_1^x & j_2^x \\ j_1^y & j_2^y \end{pmatrix}$$

which gives a natural frame on the patch.

While moving from one patch to another, this frame changes its orientation. So the total change of a vector, $\vec{B}$ due to moving from one point to another consists of two pieces

$$\delta\vec{B} = d\vec{B} + \Gamma\vec{B}d\vec{X} \tag{A1-2}$$

Where the first term is the differential along the path

$$d\vec{B}(t) = \frac{\partial\vec{B}}{\partial X^i}dX^i(t) \qquad .$$

and the second allows for a change of the frame orientation. Since the matrix $\hat{\sigma}$ consists of n vectors it is transformed following the same rule

$$\delta\hat{\sigma} = d\hat{\sigma} + \vec{\vec{\Gamma}}\hat{\sigma}\,d\vec{X}(t) \tag{A1-3}$$

The second assumption is that "Rules must not change from game to game" ("fair game" or no arbitrage opportunity or same action same response) means that the total change of $\hat{\sigma}$ should vanish

$$\delta\hat{\sigma} = 0$$

or

$$d\hat{\sigma} = -\vec{\vec{\Gamma}}\hat{\sigma}\,d\vec{X}(t)$$

which along with the equation for the elementary shock

$$d\vec{X}(t) = \hat{\sigma}\,d\vec{W}$$

gives the equations of Stochastic Differential Geometry on the sphere. Generalization to other curved manifolds is straightforward. Let us stress again that the equations are written in *Stratonovich* differentials.

Annex 2. Ornstein-Uhlenbeck Process.

A unity mass particle moving with the velocity $U(t)$ under the influence of the friction force and a rapidly oscillating random force $\sigma\eta(t)$ is described by the stochastic differential equation

$$dU(t) = -\gamma U\,dt + \sigma\,dW, \quad dW = \eta\,dt \tag{A2-1}$$

Assuming γ and σ being constants (additive noise) we have the following Focker-Plank equation for the above process:

$$\partial_t\rho(u,t\,|\,y,s) = [\gamma\partial_u u + \frac{1}{2}\partial_u^2]\rho(u,t\,|\,y,s)$$

Its solution reads

$$\rho(u,t\,|\,y,s) = [2\pi\,\Sigma(t-s)]^{-1/2}\exp\left\{-\frac{(u - ye^{-\gamma(t-s)})^2}{2\Sigma(t-s)}\right\} \tag{A2-2}$$

with

$$\Sigma(t-s) = \frac{\sigma^2}{2\gamma}(1 - e^{-2\gamma(t-s)})$$

The stationary distribution achieves at $t \to \infty$ or $s \to -\infty$ such that

$$\rho_{stat}(u) = \lim_{t\to\infty}\rho(u,t\,|\,y,s) = \sqrt{\frac{\gamma}{\pi\sigma^2}}\,e^{-\frac{\gamma u^2}{\sigma^2}} \tag{A2-3}$$

The covariance of the process $U(o)$ in the stationary state is as follows

$$E[U(t)U(s)] = \int_{-\infty}^{\infty}du\int_{-\infty}^{\infty}uydy\,\rho(u,t\,|\,t,s)\rho_{stat}(y) = \frac{\sigma^2}{2\gamma}e^{-\gamma|t-s|} \tag{A2-4}$$

Conclusion. This implies the correlation time is determined by the value of $\tau = 1/\gamma$ such that at $|t - s| \gg \tau$ the correlations decay exponentially and at small τ the process $U(o)$ behaves as a white noise. On the contrary, at large τ the process $U(o)$ can be regarded as the Wiener process. What is easily seen from the following non-stationary correlation function

$$E[U(t)U(s)] = \frac{\sigma^2}{2\gamma}(1 - e^{-2\gamma t})e^{-\gamma|t-s|} \to \sigma^2 t$$

A non-zero mean O-U process is

$$dX(t) = \alpha(x_0 - X)dt + \sigma dW$$

where again α, σ, x_0 are constants. Then

$$\partial_t \rho(x,t \mid y,s) = \left[\alpha \partial_x (x - x_0) + \frac{1}{2}\sigma^2 \frac{\partial^2}{\partial x^2}\right]\rho(x,t \mid y,s)$$

The non-stationary solution is

$$\rho(x,t \mid y,s) = [2\pi\Sigma(t-s)]^{-1/2} \exp\left\{\frac{(x - x_0 - ye^{-\alpha(t-s)})^2}{2\Sigma(t-s)}\right\}$$

$$\Sigma(t-s) = \frac{\sigma^2}{2\alpha}(1 - e^{-2\alpha(t-s)})$$

and

$$\rho_{stat}(x) = \sqrt{\frac{\alpha}{\pi\sigma^2}}\; e^{-\frac{\alpha}{\sigma^2}(x-x_0)^2}$$

Now $x = \ln S$, and $x_0 = \mu - \dfrac{\sigma^2}{2}$ for the interest rate case.

Volatility of this distribution as function of time and initial moment are depicted on the following figures.