

Convexity in Interest rate Theory

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Convexity results of E. Ekström and J. Tysk

N-dimensional considerations

A result for *N*-dimensional Ito-diffusions

Consider a one-dimensional short rate model

$$dr_t = \beta(r_t)dt + \sigma(r_t)dW_t$$

on a filtered stochastic basis (Ω, \mathcal{F}, P) carrying a one-dimensional Brownian motion W .

Under suitable assumptions on β, σ we can guarantee that

$$G(r, t) := E\left(\exp\left(-\int_0^t r_s ds\right)g(r_t)\right)$$

is $C^{2,1}$ -regular for sufficiently regular pay-off functions g .

The Theorem of Ekström/Tysk

Assume sufficient regularity on β, σ, g and assume that g is a **non-negative, convex, decreasing pay-off**. If

$$\beta_{rr}(r) \leq 2$$

for all r , then $r \mapsto G(r, t)$ is convex in r .

J. Ekström and J. Tysk also prove assertions on monotonicity (with respect to changes in volatility and drift), log-convexity and log-concavity..

Questions

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- ▶ How does a multi-dimensional version of the previous theorem look like?
- ▶ Does it hold for jump-diffusions or Lévy-driven processes, too?

Fix $N, d \geq 1$. We consider a stochastic differential equation

$$dr_s = \beta(r_s)ds + \sum_{i=1}^d \sigma_i(r_s)dW_s^i$$

in \mathbb{R}^N and a non-zero linear functional l on \mathbb{R}^N . We assume the coefficients to satisfy C^∞ -boundedness conditions and the initial value r to be deterministic. We **do not assume** ellipticity.

An approach by Semi-group Theory

Consider a Banach space X of pay-off functions and apply criteria for invariant sets in this situation.

- ▶ A semigroup P leaves C invariant, i.e. $P_t C \subset C$ for all $t \geq 0$, if and only if $E(P_\tau x) \in C$ for all bounded random times τ and all $x \in C$ (Resolvent criterion).

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- ▶ Let $C \subset X$ be a closed convex set and assume that for some $n \geq 0$ the set $C_n := D(A^n) \cap C$ has non-empty interior in the space $D(A^n)$ and has closure C in X . A semigroup P leaves C invariant, i.e. $P_t C \subset C$ for all $t \geq 0$ if and only if $l(Ax) \leq 0$ for all $x \in D(A^{n+1}) \cap \partial_n C_n$ and some supporting hyperplane l at x .

Considering the general situation this boils down to checking in an appropriate Banach space of functions an inequality for boundary points of a convex set of functions.

Take $N = d = 1$ and consider the closed convex set of **convex, non-negative and decreasing pay-off functions**. The crucial boundary points in some power domain are those, where the second derivative vanishes and one can check easily the equivalence of the semi-group criterion to the Ekström/Tysk condition.

A Theorem for Jump Diffusions

Consider a diffusion r leaving a certain convex set C of functions invariant, then jump diffusion extensions leave the same convex set of functions C invariant.

In the particular previous case we obtain for

$$dr_t = \beta(r_t)dt + \sigma(r_t)dW_t + \delta(r_t)dN_t$$

that $r \mapsto G(r, t)$ is convex, decreasing and positive if

$$\beta_{rr} \leq 2$$

and $r \mapsto r + \delta(r)$ is linear and increasing.

An analytic approach

Let $g : \mathbb{R}^N \rightarrow \mathbb{R}$ be a **convex, non-negative function** with sufficient regularity such that the following calculations make sense. We consider

$$G(r, t) = E(\exp(-\int_0^t l(r_s) ds) g(r_t))$$

for $t \geq 0$.

First and Second Variation

We denote the derivatives with respect to r by T (tangent map). We shall apply the first and second variation of the stochastic process r with respect to its initial value in a deterministic direction v , which is a stochastic process again,

$$\partial_v r_s = T\beta(r_s)\partial_v r_s ds + \sum_{i=1}^d T\sigma_i(r_s)\partial_v r_s dW_s^i, \quad \partial_v r_0 = v$$

and

$$\begin{aligned} \partial_v^2 r_s = & (T\beta(r_s)\partial_v^2 r_s + T^2\beta(r_s)(\partial_v r_s)^2) ds + \\ & + \sum_{i=1}^d (T\sigma_i(r_s)\partial_v^2 r_s + T^2\sigma_i(r_s)(\partial_v r_s)^2) dW_s^i \end{aligned}$$

with $\partial_v^2 r_0 = 0$.

Taking derivatives

The first derivative of G with respect to r ,

$$\begin{aligned} TG(r, t)v &= -E\left(\exp\left(-\int_0^t l(r_s)ds\right) \int_0^t l(\partial_v r_s)ds g(r_t)\right) + \\ &+ E\left(\exp\left(-\int_0^t l(r_s)ds\right) Tg(r_t)\partial_v r_t\right) \end{aligned}$$

and...

Taking derivatives

...the second one

$$\begin{aligned}
 T^2 G(r, t)(v)^2 &= E(\exp(-\int_0^t l(r_s) ds) (\int_0^t l(\partial_v r_s) ds)^2 g(r_t)) - \\
 &\quad - E(\exp(-\int_0^t l(r_s) ds) \int_0^t l(\partial_v^2 r_s) ds g(r_t)) - \\
 &\quad - E(\exp(-\int_0^t l(r_s) ds) \int_0^t l(\partial_v r_s) ds (Tg(r_t) \partial_v r_t)) - \\
 &\quad - E(\exp(-\int_0^t l(r_s) ds) \int_0^t l(\partial_v r_s) ds (Tg(r_t) \partial_v r_t)) + \\
 &\quad + E(\exp(-\int_0^t l(r_s) ds) (T^2 g(r_t) (\partial_v r_t)^2)) + \\
 &\quad + E(\exp(-\int_0^t l(r_s) ds) (Tg(r_t) \partial_v^2 r_t)).
 \end{aligned}$$

There is a symmetric structure, which is best revealed by the following process

$$C_t = 2 \int_0^t l(\partial_v r_s) ds \partial_v r_t - \partial_v^2 r_t.$$

Hence we can write that

$$\begin{aligned} T^2 G(r, t)(v)^2 = & E(\exp(-\int_0^t l(r_s) ds) g(r_t) \int_0^t l(C_s) ds) - \\ & - E(\exp(-\int_0^t l(r_s) ds) Tg(r_t) C_t) + \\ & + E(\exp(-\int_0^t l(r_s) ds) (T^2 g(r_t) (\partial_v r_t)^2)) \end{aligned}$$

by gathering the respective terms and applying that

$$\int_0^t l(C_s) ds = \left(\int_0^t l(\partial_v r_s) ds \right)^2 - \int_0^t l(\partial_v^2 r_s) ds.$$

The Theorem

Assume first that $l(\partial_v r_t) \geq 0$ and $Tg(r_t)\partial_v r_t \leq 0$ almost surely for all $t \geq 0$, then we obtain that

$$TG(r, t)v \leq 0$$

for $r \in \mathbb{R}^N$ and $t \geq 0$. Next we assume (without granting the first assumption) that

$$2l(v)(TG(r, t)v - TG(r, t)T^2\beta(r)(v)^2) > 0$$

for all $v, r \in \mathbb{R}^N$ with $l(v) > 0$ and $t \geq 0$. Then we conclude that $r \mapsto G(r, t)$ is convex.

Corollary

- ▶ Fix $N, d \geq 1$. Assume that the drift β is linear and that $g = 1$. If $l(\partial_v r_t) \geq 0$ for all $v \in \mathbb{R}^N$ with $l(v) \geq 0$, then

$$r \mapsto E(\exp(\int_0^t l(r_s) ds))$$

is convex with respect to the initial values.

Corollary

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is convex with respect to the initial values.

- ▶ Assume that $N = 1, d \geq 1$ and that g is decreasing. We choose $l(x) = x$ and $v = 1$. Since $\partial_v r_0 = v$ we have that $l(\partial_v r_t) \geq 0$ by continuity of the trajectories and the fact that $\partial_v r_t \neq 0$ almost surely for $t \geq 0$. Consequently we obtain in this case $TG(r, t)v \leq 0$. Supposing finally that

$$2 - \beta_{rr}(r) \geq 0$$

yields convexity such as in Ekström/Tysk.

The Proof

We consider the $5N$ -dimensional Markov process X composed of $(r_t, \partial_v r_t, \partial_v^2 r_t, \int_0^t r_s ds, \int_0^t \partial_v r_s ds)$ for the appropriate initial value $(r, v, 0, 0, 0)$. We interpret the previous expectations as solutions of the respective heat equation for a particular initial function F . We know from Dynkin's formula that

$$E(F(X_t)) = F(X_0) + \int_0^t E(\mathcal{A}F(X_s)) ds,$$

where \mathcal{A} denotes the generator of X .

The Proof

We choose

$$F(x_1, x_2, x_3, x_4, x_5) = \exp(-l(x_4)) Tg(x_1)(2l(x_5)x_2 - x_3),$$

which yields

$$\begin{aligned} \mathcal{A}F(x_1, x_2, x_3, x_4, x_5) &= -l(x_1) \exp(-l(x_4)) Tg(x_1)(2l(x_5)x_2 - x_3) + \\ &+ \exp(-l(x_4)) Tg(x_1)(2l(x_5)x_2 - x_3) + \\ &+ \exp(-l(x_4)) Tg(x_1) \times \\ &\times (2l(x_2)x_2 + 2l(x_5) T\beta(x_1)x_2 - T^2\beta(x_1)(x_2)^2 - T\beta(x_1)x_3). \end{aligned}$$

The Proof

We assume that $T^2G(r, t)(v)^2 = 0$ and want to prove that the derivative at t of this quantity has to be positive. We obtain by the previous consideration and $g = G(r, t)$

$$\begin{aligned}\frac{\partial}{\partial t} \Big|_{t=0} T^2G(r, t)(v)^2 &\geq -\mathcal{A}F(r, v, 0, 0, 0) \\ &= -(2l(v)TG(r, t)v - TG(r, t)T^2\beta(r)(v)^2) > 0.\end{aligned}$$

Discussion

For the proof we need the following structures:

- ▶ A Markov process admitting sufficiently regular solutions of its associated heat equation.

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- ▶ A Markov process admitting first and second variations (see for instance Ph. Protter's book for conditions when semi-martingale-driven SDEs admit stochastic flows).

Discussion

For the proof we need the following structures:

- ▶ A Markov process admitting sufficiently regular solutions of its associated heat equation.
- ▶ A Markov process admitting first and second variations (see for instance Ph. Protter's book for conditions when semi-martingale-driven SDEs admit stochastic flows).
- ▶ The sign of the generator \mathcal{A} on

$$F(x_1, x_2, x_3, x_4, x_5) = \exp(-l(x_4)) Tg(x_1)(2l(x_5)x_2) - x_3$$

at $(r, v, 0, 0, 0)$ is decisive.

- ▶ Erik Ekström, Johan Tysk, *Convexity Theory for the Term structure equation*, preprint (2007).
- ▶ Erik Ekström, Johan Tysk, *Properties of option prices in models with jumps*. to appear in *Math. Finance* (2007).

- ▶ Thank you for your attention!

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- ▶ Further informations on this research @ *Publications* soon on my webpage (google Josef Teichmann).