

Sensitivity Analysis on Long-term Cash flows

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Introduction

Related articles

-  Borovicka, J., Hansen, L.P., Hendricks, M., Scheinkman, J.A.: Risk price dynamics. *Journal of Financial Econometrics* **9**(1), 3-65 (2011)
-  Fournie, E., Lasry J., Lebuchoux, J., Lions P., Touzi, N.: Applications of Mallivin calculus to Monte Carlo methods in finance. *Finance Stoch.* **3**, 391-412 (1999)
-  Hansen, L.P., Scheinkman, J.A.: Long-term risk: An operator approach. *Econometrica* **77**, 177-234 (2009)
-  Hansen, L.P., Scheinkman, J.A.: Pricing growth-rate risk. *Finance Stoch.* **16**(1), 1-15 (2012)

Introduction

Let $W_t = (W_1(t), W_2(t), \dots, W_d(t))^T$ be a standard d -dimensional Brownian motion.

Assumption

An underlying process X_t is a conservative d -dimensional time-homogeneous Markov diffusion process with the following form:

$$dX_t = b(X_t) dt + \sigma(X_t) dW_t, \quad X_0 = \xi.$$

Here, b is a d -dimensional column vector and σ is a $d \times d$ matrix with some technical conditions.

Introduction

In finance, we often encounter the quantity of the form:

$$p_T := \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_t) dt} f(X_T)] .$$

Purpose: to study a **sensitivity analysis** for the quantity p_T with respect to the perturbation of X_t for large T .

This sensitivity is useful for long-term static investors and for long-dated option prices.

Introduction

Let X_t^ϵ be a perturbed process of X_t (with the same initial value $\xi = X_0 = X_0^\epsilon$) of the form:

$$dX_t^\epsilon = b_\epsilon(X_t^\epsilon) dt + \sigma_\epsilon(X_t^\epsilon) dW_t. \quad (1.1)$$

The perturbed quantity is given by

$$p_T^\epsilon := \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_s^\epsilon) ds} f(X_T^\epsilon)].$$

For the sensitivity analysis, we compute

$$\left. \frac{\partial}{\partial \epsilon} \right|_{\epsilon=0} p_T^\epsilon$$

and investigate the behavior of this quantity for large T .

Introduction

For the sensitivity w.r.t. the initial value $X_0 = \xi$, we compute

$$\frac{\partial \rho_T}{\partial \xi}$$

and investigate the behavior of this quantity for large T .

Martingale extraction

Martingale extraction

We denote the infinitesimal generator corresponding to the operator

$$f \mapsto p_T = \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_t) dt} f(X_T)]$$

by \mathcal{L} :

$$\mathcal{L} := \frac{1}{2} \sum_{i,j=1}^d a_{ij}(x) \frac{\partial^2}{\partial x_i \partial x_j} + \sum_{i=1}^d b_i(x) \frac{\partial}{\partial x_i} - r(x)$$

where $a = \sigma\sigma^\top$.

Martingale extraction

Let (λ, ϕ) be an eigenpair of $\mathcal{L}\phi = -\lambda\phi$ with positive function ϕ . It is easily checked that

$$M_t := e^{\lambda t - \int_0^t r(X_s) ds} \phi(X_t) \phi^{-1}(\xi)$$

is a local martingale.

Definition

When the local martingale M_t is a martingale, we say that (X_t, r) admits the *martingale extraction* with respect to (λ, ϕ) .

$$e^{-\int_0^t r(X_s) ds} = M_t e^{-\lambda t} \phi^{-1}(X_t) \phi(\xi)$$

: martingale M_t is extracted from the discount factor

Martingale extraction

When M_t is a martingale, we can define a new measure \mathbb{P} by $\mathbb{P}(A) := \int_A M_T d\mathbb{Q} = \mathbb{E}^{\mathbb{Q}} [\mathbb{I}_A M_T]$ for $A \in \mathcal{F}_T$, that is,

$$M_T = \left. \frac{d\mathbb{P}}{d\mathbb{Q}} \right|_{\mathcal{F}_T}$$

The measure \mathbb{P} is called *the transformed measure* from \mathbb{Q} with respect to (λ, ϕ) .

ρ_T can be expressed by

$$\begin{aligned} \rho_T &= \mathbb{E}^{\mathbb{Q}} [e^{-\int_0^T r(X_s) ds} f(X_T)] \\ &= \phi(\xi) e^{-\lambda t} \cdot \mathbb{E}^{\mathbb{Q}} [M_t (\phi^{-1} f)(X_t)] \\ &= \phi(\xi) e^{-\lambda T} \cdot \mathbb{E}^{\mathbb{P}} [(\phi^{-1} f)(X_T)]. \end{aligned}$$

Martingale extraction

We have

$$p_T = \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_s) ds} f(X_T)] = \phi(\xi) e^{-\lambda T} \cdot \mathbb{E}^{\mathbb{P}}[(\phi^{-1} f)(X_T)].$$

This relationship implies that the quantity p_T can be expressed in a relatively more manageable manner. The term $\mathbb{E}^{\mathbb{P}}[(\phi^{-1} f)(X_T)]$ depends on the final value of X_T , whereas $\mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_s) ds} f(X_T)]$ depends on the whole path of X_t at $0 \leq t \leq T$.

This advantage makes it easier to analyze the sensitivity of long-term cash flows.

Martingale extraction

In general, there are infinitely many ways to extract the martingale. We choose a special one.

Definition

Consider a martingale extraction such that $\mathbb{E}^{\mathbb{P}}[(\phi^{-1}f)(X_T)]$ converges to a nonzero constant as $T \rightarrow \infty$. We say this is a martingale extraction *stabilizing* f . In this case,

$$\lim_{T \rightarrow \infty} \frac{1}{T} \ln p_T = -\lambda.$$

For example, if X_t has an invariant distribution μ under \mathbb{P} , then

$$\mathbb{E}^{\mathbb{P}}[(\phi^{-1}f)(X_T)] \rightarrow \int (\phi^{-1}f)(z) d\mu(z)$$

for suitably nice f .

Sensitivity Analysis

Sensitivity Analysis

The rho and the vega

Let X_t^ϵ be a perturbed process of X_t (with the same initial value $\xi = X_0 = X_0^\epsilon$) of the form:

$$dX_t^\epsilon = b_\epsilon(X_t^\epsilon) dt + \sigma_\epsilon(X_t^\epsilon) dW_t. \quad (3.1)$$

The perturbed quantity is given by

$$p_T^\epsilon := \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_s^\epsilon) ds} f(X_T^\epsilon)].$$

For the sensitivity analysis, we compute

$$\left. \frac{\partial}{\partial \epsilon} \right|_{\epsilon=0} \ln p_T^\epsilon$$

and investigate the behavior of this quantity for large T .

Sensitivity Analysis

Assume that (X_t^ϵ, r) also admits the martingale extraction that stabilizes f , then

$$p_T^\epsilon = \phi_\epsilon(\xi) e^{-\lambda(\epsilon)T} \cdot \mathbb{E}^{\mathbb{P}^\epsilon}[(\phi_\epsilon^{-1}f)(X_T^\epsilon)] .$$

Differentiate with respect to ϵ and evaluate at $\epsilon = 0$, then

$$\begin{aligned} \frac{\frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} p_T^\epsilon}{T \cdot p_T} &= \frac{\frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} \phi_\epsilon(\xi)}{T \cdot \phi(\xi)} - \lambda'(0) + \frac{\frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} \mathbb{E}^{\mathbb{P}}[(\phi_\epsilon^{-1}f)(X_T)]}{T \cdot \mathbb{E}^{\mathbb{P}}[(\phi^{-1}f)(X_T)]} \\ &+ \frac{\frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} \mathbb{E}^{\mathbb{P}^\epsilon}[(\phi^{-1}f)(X_T^\epsilon)]}{T \cdot \mathbb{E}^{\mathbb{P}}[(\phi^{-1}f)(X_T)]} . \end{aligned}$$

Here, $\mathbb{E}^{\mathbb{P}}[(\phi^{-1}f)(X_T)] \rightarrow$ (a nonzero constant) as $T \rightarrow \infty$.

Sensitivity Analysis

The long-term behavior of the rho and the vega

Under some conditions, the first, third and the last terms go to zero as $T \rightarrow \infty$, thus we have

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} \ln p_T^\epsilon = \lim_{T \rightarrow \infty} \frac{\frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} p_T^\epsilon}{T \cdot p_T} = -\lambda'(0)$$

Sensitivity Analysis

The dynamics under the transformed measure

Let

$$\varphi := \sigma^\top \cdot \frac{\nabla \phi}{\phi} \quad (\text{Girsanov kernel})$$

then

$$B_t := W_t - \int_0^t \varphi(X_s) ds$$

is a Brownian motion under \mathbb{P} . The \mathbb{P} -dynamics of X_t are

$$\begin{aligned} dX_t &= b(X_t) dt + \sigma(X_t) dW_t \\ &= (b(X_t) + \sigma(X_t)\varphi(X_t)) dt + \sigma(X_t) dB_t . \end{aligned}$$

Sensitivity Analysis

The rho

Want to control: $\frac{1}{T} \frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} \mathbb{E}^{\mathbb{P}^\epsilon} [(\phi^{-1}f)(X_T^\epsilon)] \rightarrow 0$ as $T \rightarrow \infty$

The perturbed process X_t^ϵ expressed by

$$\begin{aligned}dX_t^\epsilon &= b_\epsilon(X_t^\epsilon) dt + \sigma(X_t^\epsilon) dW_t \\&= (\sigma^{-1}b_\epsilon + \varphi_\epsilon)(X_t) dt + \sigma(X_t^\epsilon) dB_t^\epsilon \\&= k_\epsilon(X_t^\epsilon) dt + \sigma(X_t^\epsilon) dB_t^\epsilon .\end{aligned}$$

Sensitivity Analysis

The rho

Assume that there exists a function $g : \mathbb{R}^d \rightarrow \mathbb{R}$ with

$$\left| \frac{\partial k_\epsilon(x)}{\partial \epsilon} \right| \leq g(x)$$

on $(\epsilon, x) \in I \times \mathbb{R}^d$ for an open interval I containing 0 such that

(i) there exists a positive number ϵ_0 such that

$$\mathbb{E}^{\mathbb{P}} \left[\exp \left(\epsilon_0 \int_0^T g^2(X_t) dt \right) \right] \leq c e^{aT}$$

for some constants a and c on $0 < T < \infty$.

(ii) for each $T > 0$, there is a positive number ϵ_1 such that

$\mathbb{E}^{\mathbb{P}} \int_0^T g^{2+\epsilon_1}(X_t) dt$ is finite.

(iii) $\frac{1}{T} \cdot \mathbb{E}^{\mathbb{P}} [(\phi^{-1}f)^2(X_T)] \rightarrow 0$ as $T \rightarrow \infty$.

Sensitivity Analysis

Then, $\mathbb{E}^{\mathbb{P}^\epsilon}[(\phi^{-1}f)(X_T^\epsilon)]$ is differentiable at $\epsilon = 0$ and

$$\frac{1}{T} \frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} \mathbb{E}^{\mathbb{P}^\epsilon}[(\phi^{-1}f)(X_T^\epsilon)] \rightarrow 0 .$$

Sensitivity Analysis

The vega

One way:

The method of Fournie et. al. with bounded-derivative coefficients

Fournie et. al.: Applications of Mallivin calculus to Monte Carlo methods in finance. *inance Stoch.* **3**, 391-412 (1999)

The perturbed process X_t^ϵ :

$$dX_t^\epsilon = b(X_t^\epsilon) dt + (\sigma + \epsilon\bar{\sigma})(X_t^\epsilon) dW_t$$

The \mathbb{P}_ϵ -dynamics of X_t^ϵ are

$$dX_t^\epsilon = (b + (\sigma + \epsilon\bar{\sigma})\varphi_\epsilon)(X_t^\epsilon) dt + (\sigma + \epsilon\bar{\sigma})(X_t^\epsilon) dB_t^\epsilon$$

Sensitivity Analysis

The vega

We take apart two perturbations by the chain rule.

$$\begin{aligned}dX_t^\rho &= (b + (\sigma + \rho\bar{\sigma})\varphi_\rho)(X_t^\rho) dt + \sigma(X_t^\rho) dB_t^\rho, \\dX_t^\nu &= (b + \sigma\varphi)(X_t^\nu) dt + (\sigma + \nu\bar{\sigma})(X_t^\nu) dB_t^\nu.\end{aligned}$$

Then we have

$$\begin{aligned}& \left. \frac{\partial}{\partial \epsilon} \right|_{\epsilon=0} \mathbb{E}^{\mathbb{P}^\epsilon}[(\phi^{-1}f)(X_T^\epsilon)] \\&= \left. \frac{\partial}{\partial \rho} \right|_{\rho=0} \mathbb{E}^{\mathbb{P}^\rho}[(\phi^{-1}f)(X_T^\rho)] + \left. \frac{\partial}{\partial \nu} \right|_{\nu=0} \mathbb{E}^{\mathbb{P}^\nu}[(\phi^{-1}f)(X_T^\nu)].\end{aligned}$$

Sensitivity Analysis

Let Z_t be the variation process given by

$$dZ_t = (b + \sigma\varphi)'(X_t)Z_t dt + \bar{\sigma}(X_t)dB_t + \sum_{i=1}^d \sigma'_i(X_t)Z_t dB_{i,t}, \quad Z_0 = 0_d$$

where σ_i is the i -th column vector of σ and 0_d is the d -dimensional zero column vector.

Theorem

Suppose that $b + \sigma\varphi$ and $\phi^{-1}f$ are continuously differentiable with bounded derivatives. If $\frac{1}{T} \cdot \mathbb{E}^{\mathbb{P}}[|Z_T|] \rightarrow 0$ as $T \rightarrow \infty$, then

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \nu} \Big|_{\nu=0} \mathbb{E}^{\mathbb{P}^\nu}[(\phi^{-1}f)(X_T^\nu)] = 0.$$

Sensitivity Analysis

The vega

One way : the Lamperti transform for univariate processes

The perturbed process is expressed by

$$dX_t^\epsilon = b_\epsilon(X_t^\epsilon) dt + \sigma_\epsilon(X_t^\epsilon) dW_t, \quad X_0^\epsilon = X_0 = \xi, \quad (3.2)$$

Define a function

$$u_\epsilon(x) := \int_\xi^x \sigma_\epsilon^{-1}(y) dy. \quad (3.3)$$

Sensitivity Analysis

The vega

Then we have

$$du_\epsilon(X_t^\epsilon) = (\sigma_\epsilon^{-1} b_\epsilon - \frac{1}{2} \sigma_\epsilon') (X_t^\epsilon) dt + dW_t, \quad u_\epsilon(X_0^\epsilon) = u_\epsilon(\xi) = 0.$$

Set $U_t^\epsilon := u_\epsilon(X_t^\epsilon)$, then

$$dU_t^\epsilon = \delta_\epsilon(U_t^\epsilon) dt + dW_t, \quad U_0^\epsilon = 0, \quad (3.4)$$

where $\delta_\epsilon(\cdot) := ((\sigma_\epsilon^{-1} b_\epsilon - \frac{1}{2} \sigma_\epsilon') \circ v_\epsilon)(\cdot)$.

Sensitivity Analysis

The delta

Let

$$p_T := \mathbb{E}_\xi^{\mathbb{Q}} \left[e^{-\int_0^T r(X_s) ds} f(X_T) \right]$$

The quantity of interest is for large T

$$\nabla_\xi p_T .$$

From the martingale extraction

$$p_T = \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_0^T r(X_s) ds} f(X_T) \right] = \phi(\xi) e^{-\lambda T} \cdot \mathbb{E}^{\mathbb{P}} [(\phi^{-1} f)(X_T)] ,$$

it follows that

$$\frac{\nabla_\xi p_T}{p_T} = \frac{\nabla_\xi \phi}{\phi(\xi)} + \frac{\nabla_\xi \mathbb{E}_\xi^{\mathbb{P}} [(\phi^{-1} f)(X_T)]}{\mathbb{E}_\xi^{\mathbb{P}} [(\phi^{-1} f)(X_T)]} .$$

Sensitivity Analysis

The long-term behavior of the delta

We have

$$\lim_{T \rightarrow \infty} \frac{\nabla_{\xi} p_T}{p_T} = \frac{\nabla_{\xi} \phi}{\phi(\xi)}$$

if

$$|\nabla_{\xi} \mathbb{E}_{\xi}^{\mathbb{P}}[(\phi^{-1} f)(X_T)]| \rightarrow 0 .$$

Sensitivity Analysis

Let Y_t be the first variation process defined by

$$dY_t = (b + \sigma\varphi)'(X_t)Y_t dt + \sum_{i=1}^d \sigma'_i(X_t)Y_t dB_{i,t}, \quad Y_0 = I_d$$

where σ_i is the i -th column vector of σ and I_d is the $d \times d$ identity matrix.

Corollary

Assume that the functions $b + \sigma\varphi$ and σ are continuously differentiable with bounded derivatives. If $\mathbb{E}_\xi^{\mathbb{P}}(\phi^{-1}f)^2(X_T)$ and

$\mathbb{E}_\xi^{\mathbb{P}}\|\sigma^{-1}(X_T)Y_T\|^2$ are bounded on $0 < T < \infty$, then

$\mathbb{E}_\xi^{\mathbb{P}}[(\phi^{-1}f)(X_T)]$ is differentiable by ξ and $|\nabla_\xi \mathbb{E}_\xi^{\mathbb{P}}(\phi^{-1}f)(X_T)| \rightarrow 0$ as $T \rightarrow \infty$. Here, $\|\cdot\|$ is the matrix 2-norm.

Examples of Options

Examples of Options

The CIR short-interest rate model

Under a risk-neutral measure \mathbb{Q} , the interest rate r_t follows

$$dr_t = (\theta - ar_t) dt + \sigma\sqrt{r_t} dW_t$$

with $2\theta > \sigma^2$.

The short-interest rate option price

$$p_T := \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r_t dt} f(r_T)]$$

Want to know the behavior for large T of

$$\frac{\partial p_T}{\partial \theta}, \quad \frac{\partial p_T}{\partial a}, \quad \frac{\partial p_T}{\partial \sigma}$$

Examples

Assume $f(r)$ is a nonnegative continuous function on $r \in [0, \infty)$, which is not identically zero, and that the growth rate at infinity is equal to or less than e^{mr} with $m < \frac{a}{\sigma^2}$.

The associated second-order equation is

$$\mathcal{L}\phi(r) = \frac{1}{2}\sigma^2 r\phi''(r) + (\theta - ar)\phi'(r) - r\phi(r) = -\lambda\phi(r).$$

Set $\kappa := \frac{\sqrt{a^2 + 2\sigma^2} - a}{\sigma^2}$. It can be shown that the martingale extraction with respect to

$$(\lambda, \phi(r)) := (\theta\kappa, e^{-\kappa r})$$

stabilizes f .

Examples

For large T , we have that

$$\lim_{T \rightarrow \infty} \frac{1}{T} \ln p_T = -\theta \kappa ,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial \theta} = -\frac{\sqrt{a^2 + 2\sigma^2} - a}{\sigma^2} ,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial a} = \frac{\theta(\sqrt{a^2 + 2\sigma^2} - a)}{\sigma^2 \sqrt{a^2 + 2\sigma^2}} ,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial \sigma} = \frac{\theta(\sqrt{a^2 + 2\sigma^2} - a)^2}{\sigma^3 \sqrt{a^2 + 2\sigma^2}} ,$$

$$\lim_{T \rightarrow \infty} \frac{\partial}{\partial r_0} \ln p_T = -\frac{\sqrt{a^2 + 2\sigma^2} - a}{\sigma^2} .$$

Examples

The quadratic term structure model

Let X_t be a d -dimensional OU process under the risk-neutral measure \mathbb{Q}

$$dX_t = (b + BX_t) dt + \sigma dW_t$$

where b is a d -dimensional vector, B is a $d \times d$ matrix, and σ is a non-singular $d \times d$ matrix.

The short interest rate : $r(x) = \beta + \langle \alpha, x \rangle + \langle \Gamma x, x \rangle$ where the constant β , vector α and symmetric positive definite Γ are taken to be such that $r(x)$ is non-negative for all $x \in \mathbb{R}^d$.

Examples

Interested in:

$$p_T = \mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(X_t) dt} f(X_T)]$$

with suitably nice payoff function f .

Let V be the *stabilizing solution* (the eigenvalues of $B - 2aV$ have negative real parts) of

$$2VaV - B^{\top}V - VB - \Gamma = 0,$$

and let $u = (2Va - B^{\top})^{-1}(2Vb + \alpha)$.

The martingale extraction stabilizes f is

$$(\lambda, \phi(x)) = \left(\beta - \frac{1}{2}u^{\top}au + \text{tr}(aV) + u^{\top}b, e^{-\langle u, x \rangle - \langle Vx, x \rangle} \right).$$

Examples

We have that for $1 \leq i, j \leq d$,

$$\lim_{T \rightarrow \infty} \frac{1}{T} \ln p_T = -\beta + \frac{1}{2} u^\top a u - \text{tr}(aV) - u^\top b ,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial b_i} = \frac{\partial \lambda}{\partial b_i} ,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial B_{ij}} = \frac{\partial \lambda}{\partial B_{ij}} ,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial \sigma_i} = \frac{\partial \lambda}{\partial \sigma_i} ,$$

$$\lim_{T \rightarrow \infty} \frac{\nabla_\xi p_T}{p_T} = -u - 2V\xi .$$

Examples of Expected Utilities

Examples

The geometric Brownian motion

Assume that S_t satisfies

$$dS_t = \mu S_t dt + \sigma S_t dW_t$$

with $\mu - \frac{1}{2}\sigma^2 > 0$. Let \mathbb{Q} be the objective measure. Interested in

$$p_T = \mathbb{E}^{\mathbb{Q}}[e^{-rT} S_T^\alpha].$$

The corresponding infinitesimal generator:

$$(\mathcal{L}\phi)(s) = \frac{1}{2}\sigma^2 s^2 \phi''(s) + \mu s \phi'(s) - r\phi(s).$$

The stabilizing martingale extraction:

$$(\lambda, \phi(s)) := (r - \mu\alpha - \frac{1}{2}\sigma^2\alpha(\alpha - 1), s^\alpha)$$

Examples

The geometric Brownian motion

With this (λ, ϕ) , we conclude that

$$\lim_{T \rightarrow \infty} \frac{1}{T} \ln p_T = -r + \mu\alpha + \frac{1}{2}\sigma^2\alpha(\alpha - 1),$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \mu} \ln p_T = -\frac{\partial \lambda}{\partial \mu} = \alpha,$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \sigma} \ln p_T = -\frac{\partial \lambda}{\partial \sigma} = \sigma\alpha(\alpha - 1),$$

$$\lim_{T \rightarrow \infty} \frac{\partial}{\partial S_0} \ln p_T = \frac{\phi'(S_0)}{\phi(S_0)} = \frac{\alpha}{S_0}.$$

Examples

The Heston model

An asset X_t follows

$$\begin{aligned}dX_t &= \mu X_t dt + \sqrt{v_t} X_t dZ_t, \\dv_t &= (\gamma - \beta v_t) dt + \delta \sqrt{v_t} dW_t,\end{aligned}$$

where Z_t and W_t are two standard Brownian motions with $\langle Z, W \rangle_t = \rho t$ for the correlation $-1 \leq \rho \leq 1$.

Interested in:

$$\rho_T := \mathbb{E}^{\mathbb{Q}}[u(X_T)] = \mathbb{E}^{\mathbb{Q}}[X_T^\alpha]$$

Examples

The Heston model

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \mu} \ln p_T = \alpha$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \gamma} \ln p_T$$

$$= -\frac{1}{2} \alpha (1 - \alpha) \cdot \frac{\sqrt{(\beta - \rho \alpha \delta)^2 + \delta^2 \alpha (1 - \alpha)} - \beta + \rho \alpha \delta}{\delta^2}$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \beta} \ln p_T = \frac{\sqrt{(\beta - \rho \alpha \delta)^2 + \delta^2 \alpha (1 - \alpha)} - \beta + \rho \alpha \delta}{\delta^2 \sqrt{(\beta - \rho \alpha \delta)^2 + \delta^2 \alpha (1 - \alpha)}}$$

Examples

The Heston model

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \delta} \ln p_T = -\rho\alpha \cdot \frac{\sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)} - \beta + \rho\alpha\delta}{\delta^2 \sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)}} + \frac{(\sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)} - \beta + \rho\alpha\delta)^2}{\delta^3 \sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)}}$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \frac{\partial}{\partial \rho} \ln p_T = -\frac{\alpha \sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)} - \alpha\beta + \rho\alpha^2\delta}{\delta \sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)}}$$

$$\lim_{T \rightarrow \infty} \frac{\partial}{\partial X_0} \ln p_T = \frac{\alpha}{X_0}$$

$$\begin{aligned} \lim_{T \rightarrow \infty} \frac{\partial}{\partial v_0} \ln p_T \\ = -\frac{1}{2}\alpha(1 - \alpha) \cdot \frac{\sqrt{(\beta - \rho\alpha\delta)^2 + \delta^2\alpha(1 - \alpha)} - \beta + \rho\alpha\delta}{\delta^2} . \end{aligned}$$

Examples of LEFTs

Examples of LEFTs

The sensitivity analysis of the expected utility and the return of an exchange-traded fund (ETF) is explored.

The leveraged ETF (LEFT) L_t can be written by

$$\frac{L_t}{L_0} = \left(\frac{X_t}{X_0} \right)^\beta e^{-r(\beta-1)t - \frac{\beta(\beta-1)}{2} \int_0^t \sigma^2(X_u) / X_u^2 du} .$$

We consider a power utility function of the form

$$u(c) = c^\alpha, \quad 0 < \alpha \leq 1 .$$

Interested in the sensitivity analysis of

$$\rho_T := \mathbb{E}^{\mathbb{Q}}[u(L_T)]$$

Examples of LEFTs

The 3/2 model

The dynamics of X_t follows the 3/2 model

$$dX_t = (\theta - aX_t)X_t dt + \sigma X_t^{3/2} dW_t$$

with $\theta, a, \sigma > 0$ under the objective measure \mathbb{Q} .

The expected utility and the return of LETF is

$$p_T := \mathbb{E}^{\mathbb{Q}}[u(L_T)] = \mathbb{E}^{\mathbb{Q}}\left[e^{-\frac{\alpha\beta(\beta-1)\sigma^2}{2} \int_0^T X_u du} X_T^{\alpha\beta}\right] \cdot e^{-r\alpha(\beta-1)T} .$$

Interested in the behavior for large T of

$$\frac{\partial p_T}{\partial \theta} , \quad \frac{\partial p_T}{\partial a} , \quad \frac{\partial p_T}{\partial \sigma}$$

Examples of LEFTs

The 3/2 model

The corresponding infinitesimal operator is

$$\frac{1}{2}\sigma^2 x^3 \phi''(x) + (\theta - ax)x\phi'(x) - \frac{1}{2}\alpha\beta(\beta - 1)\sigma^2 x\phi(x) = -\lambda\phi(x).$$

Set

$$\ell := \sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)} - \left(\frac{1}{2} + \frac{a}{\sigma^2}\right).$$

It can be shown that the martingale extraction with respect to

$$(\lambda, \phi(x)) := (\theta\ell, x^{-\ell})$$

stabilizes $f(x) := x^{\alpha\beta}$ when $|\beta| \leq 3$

Examples of LEFTs

The 3/2 model

If $\frac{a}{\sigma^2} + 1 - \alpha\beta > 0$, then

$$\lim_{T \rightarrow \infty} \frac{1}{T} \ln p_T = -\theta \left(\sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)} - \left(\frac{1}{2} + \frac{a}{\sigma^2}\right) \right),$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial \theta} = -\sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)} - \left(\frac{1}{2} + \frac{a}{\sigma^2}\right),$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial a} = \frac{\theta \left(\sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)} - \left(\frac{a}{\sigma^2} + \frac{1}{2}\right) \right)}{\sigma^2 \sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)}},$$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \cdot \frac{\partial \ln p_T}{\partial \sigma} = \frac{2a\theta \left(\sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)} - \left(\frac{1}{2} + \frac{a}{\sigma^2}\right) \right)}{\sigma^3 \sqrt{\left(\frac{1}{2} + \frac{a}{\sigma^2}\right)^2 + \alpha\beta(\beta - 1)}}.$$

Conclusion

Thank you !