

Sensitivity analysis and density estimation using the Malliavin calculus

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December 17, 2005

Consider $(F^\zeta)_\zeta$ a family of random variables depending on a parameter ζ .

$$\frac{\partial}{\partial \zeta} \mathbf{E} [f(F^\zeta)] = \begin{cases} \mathbf{E} \left[f'(F^\zeta) \frac{\partial}{\partial \zeta} F^\zeta \right] \\ \simeq \frac{\mathbf{E}[f(F^{\zeta+h})] - \mathbf{E}[f(F^{\zeta-h})]}{2h} \end{cases}$$

Expectations can be computed by the Monte Carlo method:

$$E[F] \simeq \frac{F_1 + \cdots + F_n}{n},$$

where F_1, \dots, F_n is a random sample of F .

Application (1): sensitivity analysis in finance - Greeks

Price process:

$$\frac{dS_t^\zeta}{S_t^\zeta} = r(S_t^\zeta)dt + \sigma(S_t^\zeta)dM_t, \quad S_0^\zeta = x.$$

$F^\zeta = S_T$ and $\zeta \in \{x, r, \sigma, T, \dots\}$.

Payoff function:

- $f(x) = (x - K)^+$



- $f(x) = 1_{[K, \infty)}(x)$.



Option price:

$$E[f(S_T^\zeta)].$$

Greeks:

- $\zeta = x$: Delta
- $\zeta = \sigma$: Vega
- $\zeta = r$: Rho
- $\zeta = T$: Theta.

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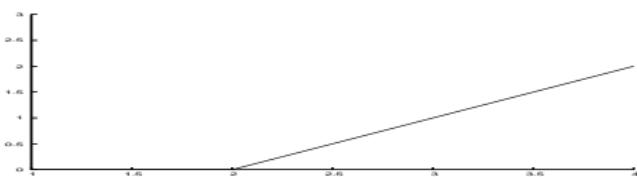
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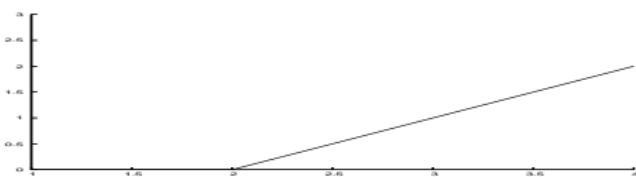
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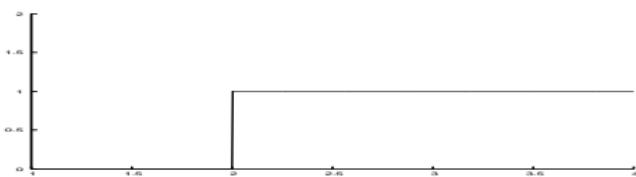
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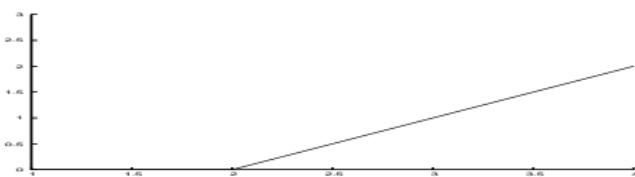
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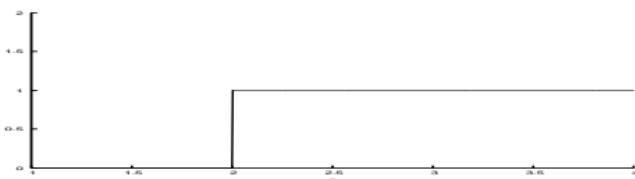
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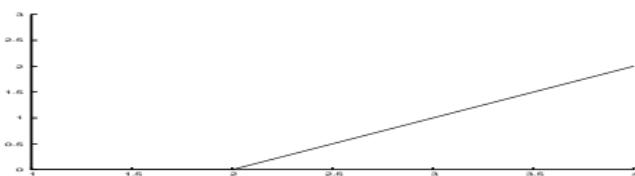
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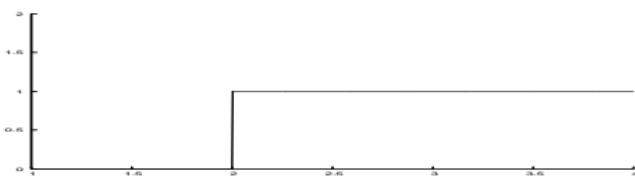
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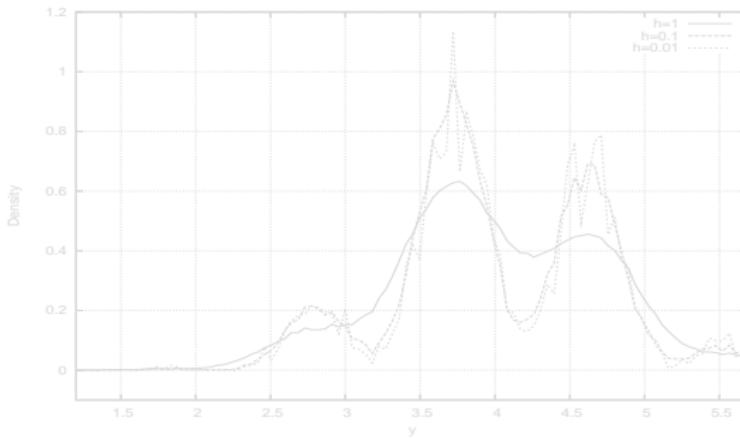
Application (2): density estimation

Let $F^\zeta = F - \zeta$ and $f(x) = 1_{(-\infty,0]}$.

Density of F :

$$\begin{aligned}\phi_F(\xi) &= \frac{\partial}{\partial \zeta} P(F \leq \zeta) = \frac{\partial}{\partial \zeta} \mathbf{E} [1_{(-\infty,0]}(F - \zeta)] = \frac{\partial}{\partial \zeta} \mathbf{E} [f(F^\zeta)] \\ &\approx \frac{\mathbf{E}[f(F - (\zeta + h))] - \mathbf{E}[f(F - (\zeta - h))]}{2h} = \frac{\mathbf{E}[1_{[\zeta-h,\zeta+h]}(F)]}{2h}.\end{aligned}$$

Example: $F = \int_0^T e^{-rt} dN_t$.



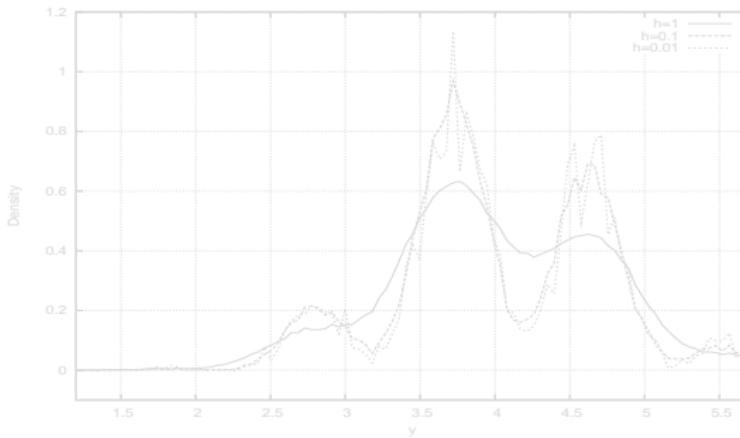
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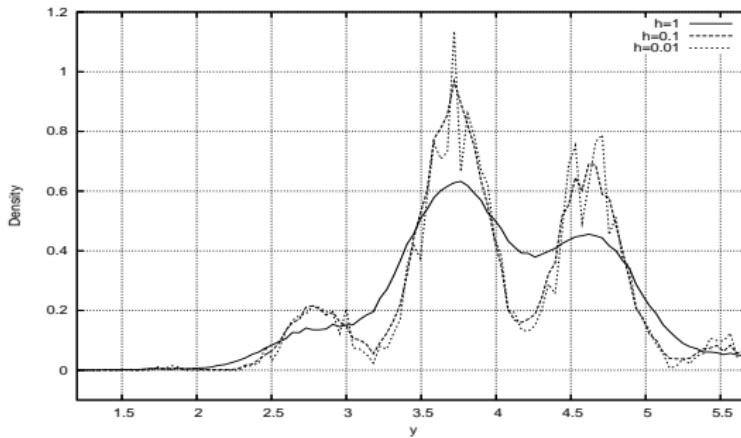
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- ① D_w a derivation operator *acting on random variables*:

$$f'(F^\zeta) = \frac{D_w [f(F^\zeta)]}{D_w F^\zeta}.$$

- ② D_w^* the adjoint of D_w :

$$\langle F, D_w G \rangle = E[FD_w G] = E[GD_w^* F] = \langle G, D_w^* F \rangle.$$

- ③ Main argument:

$$\begin{aligned}\frac{\partial}{\partial \zeta} \mathbf{E} [f(F^\zeta)] &= \mathbf{E} \left[f' \left(F^\zeta \right) \partial_\zeta F^\zeta \right] \\ &= \mathbf{E} \left[\frac{D_w [f(F^\zeta)]}{D_w F^\zeta} \partial_\zeta F^\zeta \right] \\ &= \mathbf{E} \left[f \left(F^\zeta \right) D_w^* \left(\frac{\partial_\zeta F^\zeta}{D_w F^\zeta} \right) \right].\end{aligned}$$

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Proposition

Assume that $D_w F^\zeta \neq 0$, a.s. on $\{\partial_\zeta F^\zeta \neq 0\}$, $\zeta \in (a, b)$. We have

$$\frac{\partial}{\partial \zeta} \mathbf{E} [f(F^\zeta)] = \mathbf{E} \left[f(F^\zeta) D_w^* \left(\frac{\partial_\zeta F^\zeta}{D_w F^\zeta} \right) \right], \quad \zeta \in (a, b). \quad (1)$$

- No differentiability assumption on F , compare with:

$$\frac{\partial}{\partial \zeta} \mathbf{E} [f(F^\zeta)] = \mathbf{E} [f'(F^\zeta) \partial_\zeta F^\zeta].$$

- Independence on the bandwidth parameter h , compare with:

$$\frac{\partial}{\partial \zeta} \mathbf{E} [f(F^\zeta)] \simeq \frac{\mathbf{E}[f(F^{\zeta+h})] - \mathbf{E}[f(F^{\zeta-h})]}{2h}.$$

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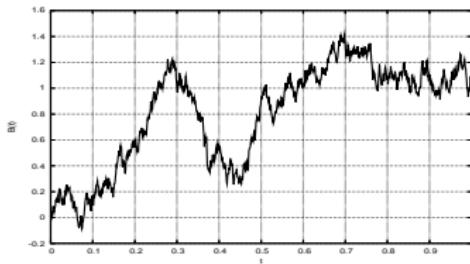
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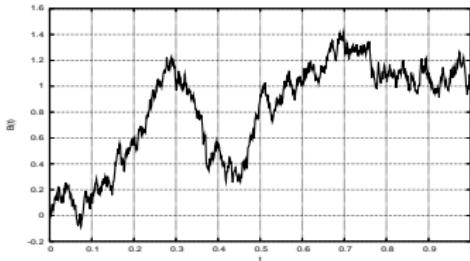
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$$D_w f(B_{t_1}, \dots, B_{t_n}) = \sum_{i=1}^n \int_0^{t_i} w_s ds \frac{\partial f}{\partial x_i}(B_{t_1}, \dots, B_{t_n}).$$

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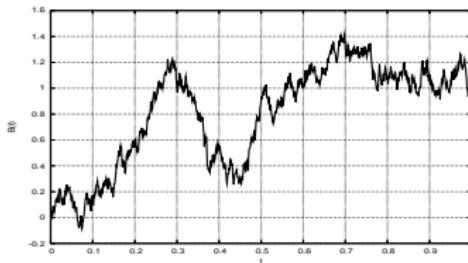
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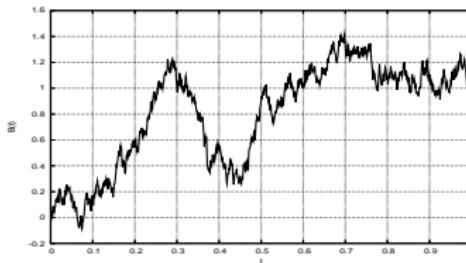
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Delta - first variation process

In [FLL⁺99] the relations

$$f'(F^\zeta) = \frac{D_t [f(F^\zeta)]}{D_t F^\zeta}, \quad 0 \leq t \leq T, \text{ a.s.},$$

and

$$D_t S_T^x = \frac{\partial_x S_T^x}{\partial_x S_t^x} \sigma(S_t^x), \quad 0 \leq t \leq T, \text{ a.s.},$$

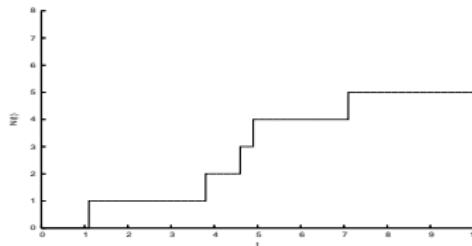
are used, cf. [Nua95]. This gives

$$\partial_x S_T^x f'(S_T^x) = \frac{\partial_x S_T^x}{\sigma(S_t^x)} D_t f(S_T^x), \quad 0 \leq t \leq T, \text{ a.s.}, \quad (2)$$

which implies if $\int_0^T w_s ds = 1$:

$$\begin{aligned} \text{Delta} &= \frac{\partial}{\partial x} \mathbf{E}[f(S_T^x)] = \mathbf{E}[\partial_x S_T^x f'(S_T^x)] = \mathbb{E}\left[\int_0^T w_t \frac{\partial_x S_t^x}{\sigma(S_t^x)} D_t f(S_T^x) dt\right] \\ &= \mathbb{E}\left[f(S_T^x) \delta\left(1_{[0, T]} w \frac{\partial_x S^x}{\sigma(S^x)}\right)\right] = \mathbb{E}\left[f(S_T^x) \int_0^T w_t \frac{\partial_x S_t^x}{\sigma(S_t^x)} dB_t\right] \\ &= \mathbf{E}\left[f(S_T^x) \frac{B_T}{\sigma x T}\right]. \end{aligned}$$

- $M_t = N_t$ is a Poisson process with jump times T_1, T_2, T_3, \dots ,



- $w \in \mathcal{C}_c^1((0, \infty))$, and

$$D_w F = - \sum_{n=1}^{\infty} 1_{\{N_T=n\}} \sum_{i=1}^n \frac{\partial f_n}{\partial x_i}(T_1, \dots, T_n),$$

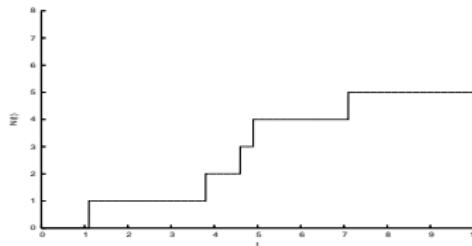
for F of the form

$$F = f_0 1_{\{N_T=0\}} + \sum_{n=1}^{\infty} 1_{\{N_T=n\}} f_n(T_1, \dots, T_n).$$

- Recall that

$$E[F] = f_0 e^{-\lambda T} + e^{-\lambda T} \sum_{n=1}^{\infty} \frac{1}{n!} \int_0^T \cdots \int_0^T f_n(t_1, \dots, t_n) dt_1 \cdots dt_n.$$

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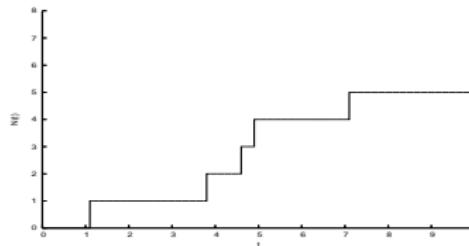
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By standard integration by parts we have, under the boundary condition $w(0) = w(T) = 0$:

$$\begin{aligned}
 E[D_w F] &= -e^{-\lambda T} \sum_{n=1}^m \frac{\lambda^n}{n!} \int_0^T \cdots \int_0^T \sum_{k=1}^n w(t_k) \partial_k f_n(t_1, \dots, t_n) dt_1 \cdots dt_n \\
 &= e^{-\lambda T} \sum_{n=1}^m \frac{\lambda^n}{n!} \int_0^T \cdots \int_0^T f_n(t_1, \dots, t_n) \sum_{k=1}^n \dot{w}(t_k) dt_1 \cdots dt_n \\
 &= E \left[F \sum_{k=1}^{k=N(T)} \dot{w}(T_k) \right] = E \left[F \int_0^T \dot{w}(t) dN(t) \right].
 \end{aligned}$$

Next, letting

$$D_w^* G = G \int_0^T \dot{w}(t) dN(t) - D_w G$$

we get

$$\begin{aligned}
 E[GD_w F] &= E[D_w(FG) - FD_w G] = E \left[F \left(G \int_0^T \dot{w}(t) dN(t) - D_w G \right) \right] \\
 &= E[FD_w^* G].
 \end{aligned}$$

- Price process: $S_t^\zeta = S_0^\zeta e^{rt} (1 + \sigma)^{N_t}$.

- Asian options: $F^\zeta = \frac{1}{T} \int_0^T S_t^\zeta dt$.

$$W_\Delta = \frac{-1}{x\sigma} \left(1 - \frac{\int_0^T S_t^\zeta dt \int_0^T \dot{w}_t d\tilde{N}_t}{\int_0^T w_t S_{t^-}^\zeta dN_t} + \frac{\int_0^T S_t^\zeta dt \int_0^T w_t (\dot{w}_t + rw_t) S_{t^-}^\zeta dN_t}{\left(\int_0^T w_t S_{t^-}^\zeta dN_t\right)^2} \right).$$

- Density of reserve processes: $F = \int_0^T e^{(T-t)r} dN_t$.

$$W_y = \frac{\int_0^T \dot{w}(t) dN(t) + \frac{\int_0^T e^{-rt} w(t)(rw(t) - \dot{w}(t)) dN(t)}{\int_0^T w(t) e^{-rt} dN(t)}}{r \int_0^T w(t) e^{r(T-t)} dX(t)}.$$

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$$W_\Delta = \frac{-1}{x\sigma} \left(1 - \frac{\int_0^T S_t^\zeta dt \int_0^T \dot{w}_t d\tilde{N}_t}{\int_0^T w_t S_{t^-}^\zeta dN_t} + \frac{\int_0^T S_t^\zeta dt \int_0^T w_t (\dot{w}_t + rw_t) S_{t^-}^\zeta dN_t}{\left(\int_0^T w_t S_{t^-}^\zeta dN_t\right)^2} \right).$$

- Density of reserve processes: $F = \int_0^T e^{(T-t)r} dN_t$.

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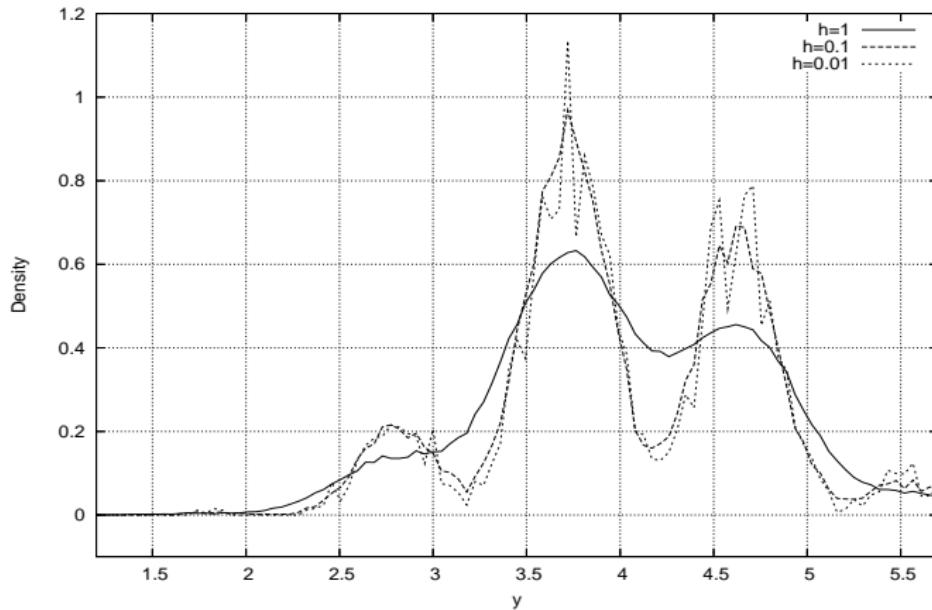
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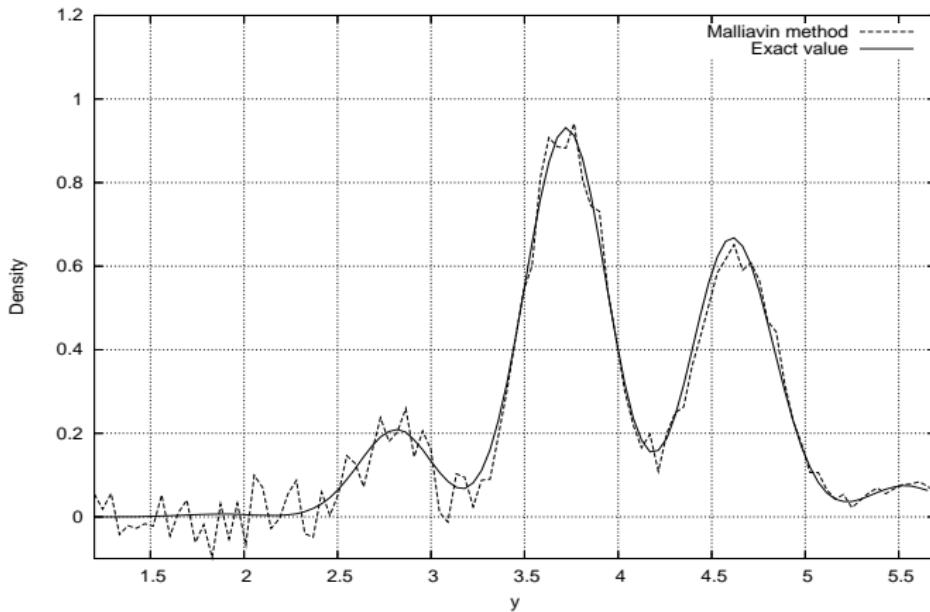
Density estimation - finite differences

$$\phi_F(y) = \frac{\partial}{\partial y} \mathbf{E} [1_{(-\infty, 0]}(F - y)] \simeq \frac{\mathbf{E} [1_{[y-h, y+h]}(F)]}{2h}.$$



Density estimation - Malliavin method

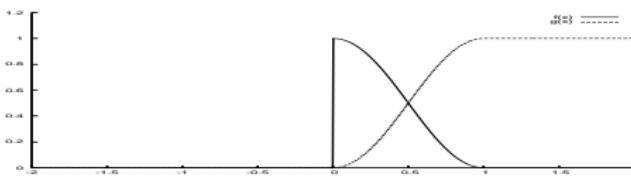
$$\phi_F(y) = \frac{\partial}{\partial y} \mathbf{E} [1_{(-\infty, 0]}(F - y)] = -\mathbf{E} \left[1_{(-\infty, 0]}(F - y) D_w^* \left(\frac{1}{D_w F} \right) \right].$$



Consider the decomposition

$$1_{[0,\infty)} = f + g,$$

where g is \mathcal{C}^1 :

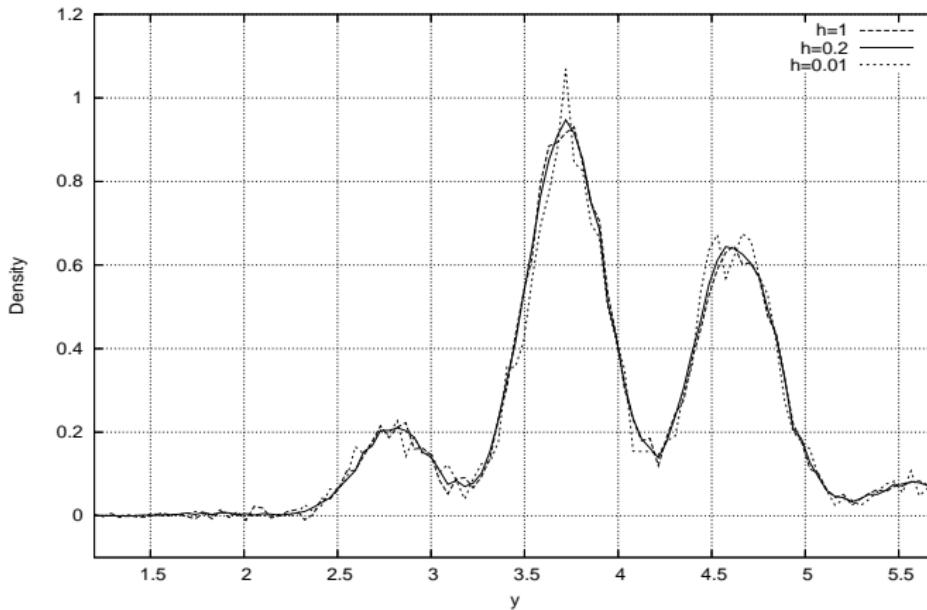


We have

$$\begin{aligned} \frac{d}{dy} E[1_{[0,\infty)}(F - y)] &= \frac{d}{dy} E \left[f \left(\frac{F - y}{h} \right) \right] + \frac{d}{dy} E \left[g \left(\frac{F - y}{h} \right) \right] \\ &= E \left[D_w^* \left(\frac{1}{D_w F} \right) f' \left(\frac{F - y}{h} \right) \right] + \frac{1}{h} E \left[1_{\{F > y\}} f' \left(\frac{F - y}{h} \right) \right]. \end{aligned}$$

Density estimation - localized Malliavin method

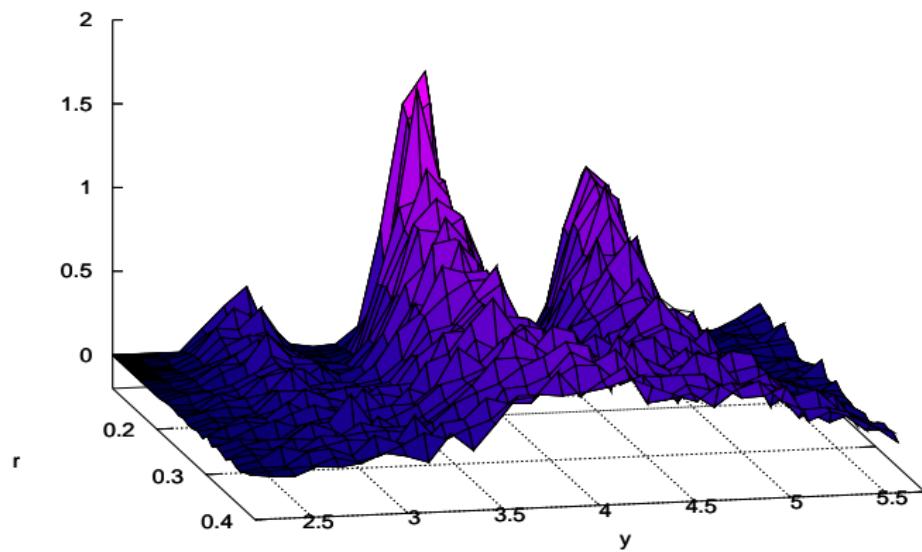
$$\phi_F(y) = -E \left[f \left(\frac{F-y}{h} \right) D_w^* \left(\frac{1}{D_w F} \right) \right] - \frac{1}{h} E \left[1_{\{F>y\}} f' \left(\frac{F-y}{h} \right) \right].$$



Optimization: $f(x) = e^{-x}$, $x \geq 0$, $h = \|W\|_{L^2(\Omega)}^{-1}$.

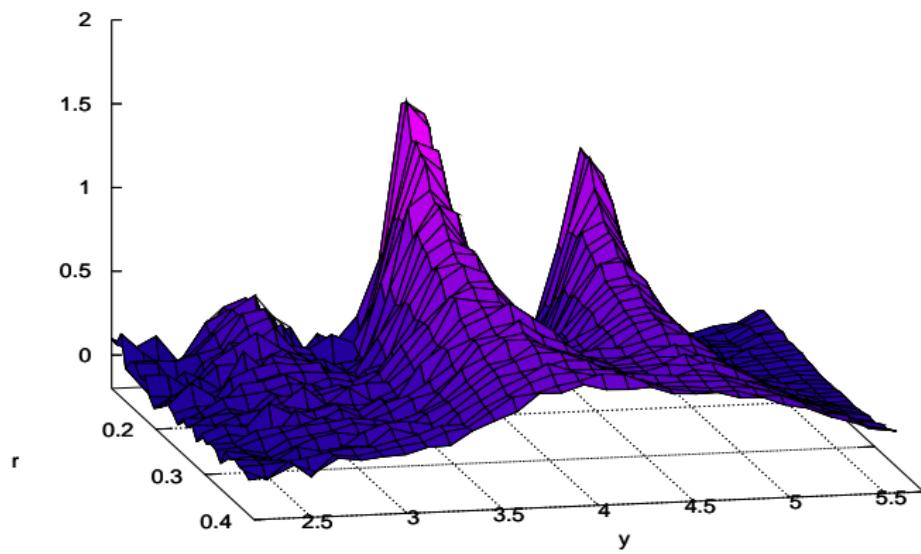
Density estimation - finite differences

$$F = \int_0^T e^{-rt} dN_t.$$



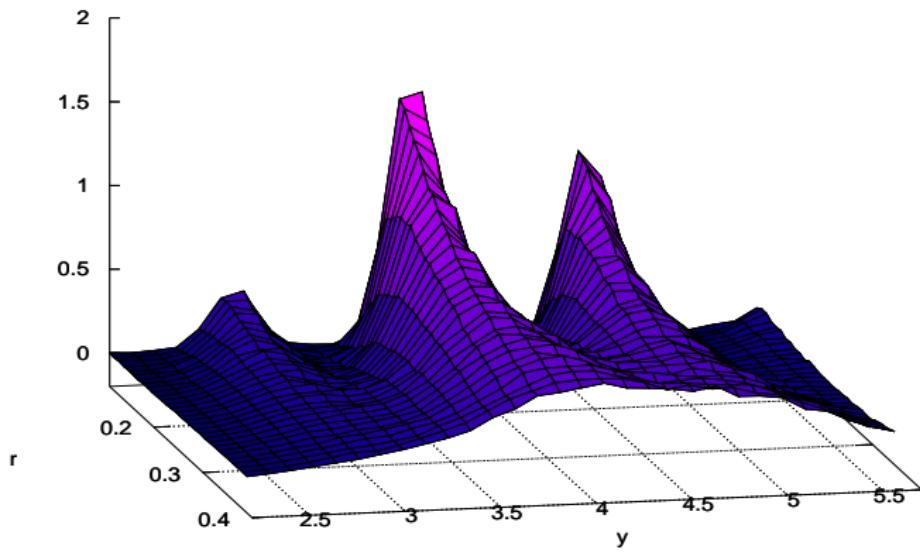
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Density estimation - localized Malliavin method

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- $(N_t)_{t \in \mathbb{R}_+}$ is a Poisson process with intensity $\lambda > 0$,
- $(Z_k)_{k \geq 1}$ an i.i.d. sequence of random variables with probability distribution $\nu(dx)$,
- $(X_t)_{t \in \mathbb{R}_+}$ a compound Poisson process with Lévy measure $\mu(dy) = \lambda\nu(dy)$. and finite intensity λ :

$$X_t = \sum_{k=1}^{N_t} Z_k, \quad t \in \mathbb{R}_+, \tag{3}$$

- $(S_t^x)_{t \in \mathbb{R}_+}$ a jump-diffusion price process:

$$\begin{cases} \frac{dS_t^x}{S_t^x} = r(S_t^x)dt + \sigma_1(S_t^x)dB_t + \sigma_2(S_{t^-}^x)dX_t, \\ S_0^x = x. \end{cases}$$

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$$D_w f(B_{t_1}, \dots, B_{t_n}, T_1, \dots, T_n) = \sum_{i=1}^n \int_0^{t_i} w_s ds \frac{\partial f}{\partial x_i}(B_{t_1}, \dots, B_{t_n}, T_1, \dots, T_n).$$

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$$\frac{\partial}{\partial \sigma_2} \mathbf{E}[f(S_T^{\sigma_2})] = \mathbf{E}\left[f(S_T^{\sigma_2}) \frac{B_T}{\sigma_1 T} \left(\frac{N_T}{1+\sigma_2} - \lambda T\right)\right].$$

- Derivation with respect to absolutely continuous jump amplitudes [BMM].

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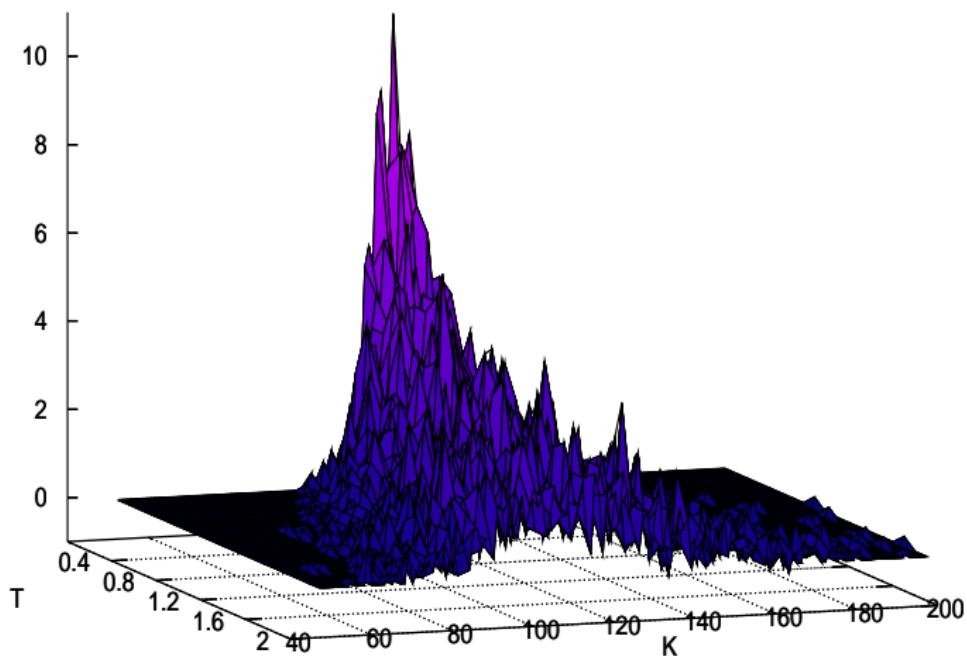
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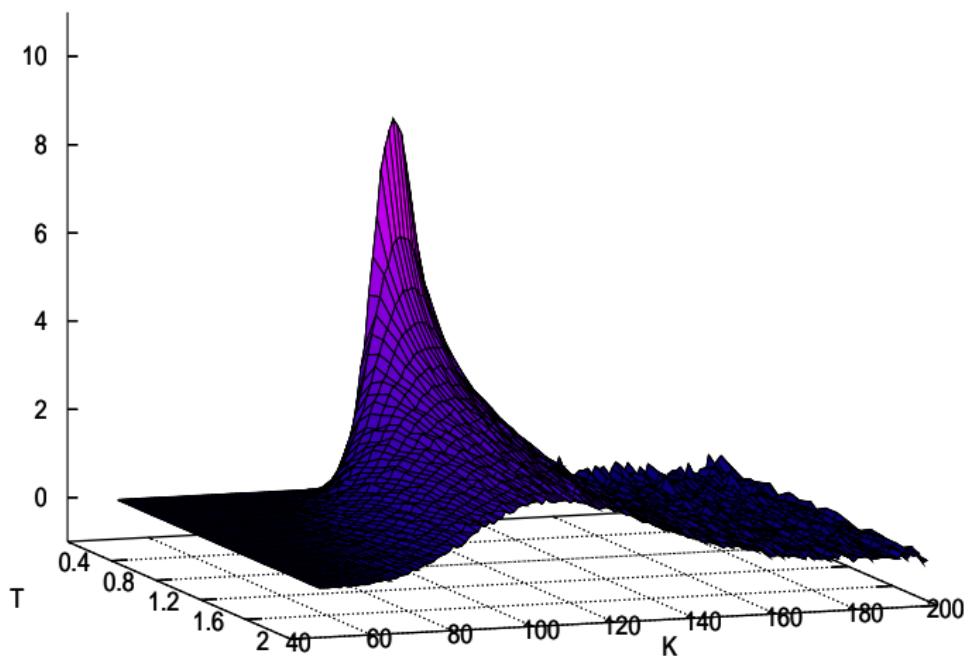
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E. Benhamou.

Smart Monte Carlo: various tricks using Malliavin calculus.
Quant. Finance, 2(5):329–336, 2002.



M.-P. Bavouzet-Morel and M. Messaoud.

Computation of Greeks using Malliavin's calculus in jump type market models.
Preprint, 2004.



M.H.A. Davis and M.P. Johansson.

Malliavin Monte Carlo Greeks for jump diffusions.
Stochastic Processes and their Applications.



V. Debelley and N. Privault.

Sensitivity analysis of European options in jump diffusion models via the Malliavin calculus on Wiener space.
Preprint, 2004.



E. Fournié, J.M. Lasry, J. Lebuchoux, P.L. Lions, and N. Touzi.

Applications of Malliavin calculus to Monte Carlo methods in finance.
Finance and Stochastics, 3(4):391–412, 1999.



E. Fournié, J.M. Lasry, J. Lebuchoux, and P.L. Lions.

Applications of Malliavin calculus to Monte-Carlo methods in finance. II.
Finance and Stochastics, 5(2):201–236, 2001.



A. Kohatsu-Higa and R. Pettersson.

Variance reduction methods for simulation of densities on Wiener space.
SIAM J. Numer. Anal., 40(2):431–450, 2002.



Y. El Khatib and N. Privault.

Computations of Greeks in markets with jumps via the Malliavin calculus.
Finance and Stochastics, 4(2):161–179, 2004.



N. Privault and X. Wei.

A Malliavin calculus approach to sensitivity analysis in insurance.
Insurance Math. Econom., 35(3):679–690, 2004.