

# Calculation of the Greeks using Malliavin Calculus

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## Part I

# Greeks and Malliavin Calculus in Finance

# The Model

Let  $B_t = (B_t^1, \dots, B_t^d)$ ,  $t \in [0, T]$ , be a  $d$ -dimensional Brownian motion on the Wiener space  $(\Omega, \mathcal{F}, P)$ .  $\mathcal{F}_t$  denotes the filtration generated by  $B$  and we assume that  $\mathcal{F} = \mathcal{F}_T$ . We model a financial market, in which  $n + 1$  assets are traded,  $n \leq d$ .

- $S_t^0$ ,  $t \in [0, T]$ , is the “bank account” earning a risk free interest rate  $r > 0$  (continuous compounding).
- $S_t = (S_t^1, \dots, S_t^n)$  gives the risky assets (“stocks”).

For bounded, measurable functions  $a : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $\sigma : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}^{n \times d}$  we set

$$dS_t^0 = rS_t^0 dt$$

$$dS_t^i = a^i(t, S_t) S_t^i dt + \sum_{j=1}^d \sigma^{ij}(t, S_t) S_t^i dB_t^j, \quad i = 1, \dots, n.$$

# Strategies

A strategy describes the amount of money invested in each available asset at any given time  $t \in [0, T]$ .

## Definition

A (*self-financing*) strategy is a predictable,  $\mathbb{R}^n$ -valued process  $\pi$  such that “all of the following integrals are well-defined”.

## Remark

- *The strategy is self-financing, because we neither allow consumption nor external money entering the financial market. Thus, strategies are determined by  $n$  coordinates.*
- *All positions are allowed to be negative (short selling).*

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# The Wealth Process of a Portfolio

The wealth process  $X^{x,\pi}$  with initial capital  $x$  associated to the strategy  $\pi$  is defined by  $X_0^{x,\pi} = x$  and its dynamics

$$dX_t^{x,\pi} = \sum_{i=1}^n \frac{\pi_t^i}{S_t^i} dS_t^i + \frac{X_t^{x,\pi} - \sum_{i=1}^n \pi_t^i dS_t^i}{S_t^0} dS_t^0.$$

## Definition

A strategy is *admissible* if the corresponding wealth process is bounded from below by some fixed real number.

The restriction to admissible strategy is economically sensible and disallows “doubling strategies”.



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# No Arbitrage

## Definition

An *arbitrage opportunity* is an admissible strategy  $\pi$  such that  $P(X_T^{0,\pi} \geq 0) = 1$  and  $P(X_T^{0,\pi} > 0) > 0$ .

## Definition

Let  $Q$  be a probability measure on  $(\Omega, \mathcal{F})$  equivalent to  $P$ .  $Q$  is called *equivalent (local) martingale measure* if the discounted price process  $\tilde{S}_t = e^{-rt} S_t$ ,  $t \in [0, T]$ , is a (local) martingale under  $Q$ .

## Remark

*The Fundamental Theorem of Asset Pricing roughly says that the existence of (local) martingale measures is equivalent to the non-existence of arbitrage opportunities.*

## No Arbitrage – 2

### Proposition

Assume there exists a predictable,  $\mathbb{R}^d$ -valued process  $\theta$  such that

- (i)  $a^i(t, S_t) - r = \sum_{j=1}^d \sigma^{ij}(t, S_t) \theta^j(t), \quad i = 1, \dots, n,$
- (ii)  $\int_0^T \|\theta(t)\|^2 dt < \infty \quad a. s.,$
- (iii)  $E\left(\exp\left(-\int_0^T \langle \theta(t), dB_t \rangle - \frac{1}{2} \int_0^T \|\theta(t)\|^2 dt\right)\right) = 1.$

Then our model is free of arbitrage and the probability measure  $Q$  with density  $Z_T = \exp\left(-\int_0^T \langle \theta(t), dB_t \rangle - \frac{1}{2} \int_0^T \|\theta(t)\|^2 dt\right)$  is a martingale measure.

Consequently, the dynamics of the model under  $Q$  are given by

$$dS_t^i = rS_t^i dt + \sum_{j=1}^d \sigma^{ij}(t, S_t) S_t^i dW_t^j, \quad i = 1, \dots, n,$$

where  $W$  denotes a Brownian motion under  $Q$ .

# Complete and Incomplete Markets

## Definition

A *contingent claim* is an  $\mathcal{F}_T$ -measurable,  $Q$ -absolutely integrable random variable. A contingent claim  $Y$  is *attainable* or *replicable* if there is an admissible, self-financing portfolio  $\pi$  and a number  $x$  such that  $X_T^{x,\pi} = Y$  a. s.  $\pi$  is called *replicating portfolio* for  $Y$ .

## Definition

A financial market is *complete*, if every contingent claim is replicable. Otherwise, it is called *incomplete*.

## Proposition

*Our market is complete if and only if  $n = d$  and  $\sigma(t, S_t(\omega))$  is invertible for  $dt \otimes P$  - a. e.  $(t, \omega) \in [0, T] \times \Omega$ .*

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## Complete and Incomplete Markets – 2

### Remark

*Under realistic conditions, completeness is a very strong property. Many experts agree that realistic models are not complete.*

### Assumptions

*From now on, we assume that our model is arbitrage-free and complete.*

# Pricing Contingent Claims

## Definition

$x \in \mathbb{R}$  is an arbitrage-free price of a contingent claim  $Y$  if there is an admissible strategy  $\pi$  such that  $Y = X_T^{x, \pi}$  a. s.

## Proposition

*In a complete, arbitrage-free model, any claim  $Y$  has a unique arbitrage-free price  $x = E_Q(e^{-rT} Y)$ , where  $Q$  is the unique e. m. m.*

## Remark

- In incomplete markets, there is, in general, no replicating portfolio, only super-replicating ones.*
- Typically, there is an interval of arbitrage-free prices bounded by the super replication prices of the buyer and of the seller.*

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## Black-Scholes PDE

- For simplicity, we pass to the discounted model by setting  $\tilde{S}_t^0 = 1$  and  $\tilde{S}_t^i = e^{-rt} S_t^i$ ,  $i = 1, \dots, n$ .
- For a *European* contingent claim of the form  $Y = f(S_T)$ , let  $C(t, x)$  denote the price of  $Y$  at time  $t$  given  $S_t = x$ , i. e.  $C(t, x) = E_Q(e^{-r(T-t)} f(S_T) | S_t = x)$ .
- By the Feynman-Kac formula,  $C$  satisfies the PDE

$$\frac{\partial}{\partial t} C(t, x) + L_t C(t, x) = rC(t, x), \quad t \in [0, T], x \in \mathbb{R}^n,$$

with  $C(T, x) = f(x)$ . Here, the infinitesimal generator is

$$L_t g(x) = \sum_{i=1}^n r x^i \frac{\partial g}{\partial x^i}(x) + \frac{1}{2} \sum_{i,j=1}^n (\sigma \sigma^T)^{ij}(t, x) x^i x^j \frac{\partial^2 g}{\partial x^i \partial x^j}(x).$$

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# Delta Hedging

## Proposition (Delta Hedging)

The replicating portfolio for the discounted, European claim  $e^{-rT} Y = e^{-rT} f(S_T)$  is given by  $\pi_t^i = \frac{\partial C}{\partial x^i}(t, S_t)$ ,  $i = 1, \dots, n$ :

$$e^{-rT} Y = C(0, S_0) + \int_0^T \left\langle \nabla_x C(t, S_t), d\tilde{S}_t \right\rangle.$$

## Proof.

Apply Itô's formula to the process  $e^{-rt} C(t, S_t)$ . Note that we have to use the risk-neutral dynamics.

In particular,  $\forall t \in [0, T] : X_t^{C(0, S_0), \pi} = e^{-rt} C(t, S_t)$ . □

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## Delta Hedging – 2

- 1 The derivative of the price with respect to the price of the underlying is called the *Delta* of a derivative, e. g.  $\nabla_x C(t, x)$ .
- 2 We can construct a self-financing portfolio as follows.

$t$	$Y$	$\tilde{S}^0$	$\tilde{S}^i$
0	-1	$C(0, S_0)$	0
$t$	-1	$C(t, S_t) - \langle \nabla_x C(t, S_t), \tilde{S}_t \rangle$	$\frac{\partial C}{\partial x^i}(t, S_t)$
$T$	-1	$Y$	0

- 3 The term “Delta hedging” comes from the fact that this portfolio is *Delta neutral*, i. e. the Delta of the portfolio is 0. Note, however, that the portfolio requires continuous trading!

# The Greeks

## Definition

The derivatives of the price of a contingent claim with respect to model parameters are called the *Greeks*.

## Remark

- 1 *The Greeks are generally used to “hedge against risks”.*
- 2 *Some of the risks – like the “risk” of changing prices of the underlying (Delta hedging) – are inherent to the model.*
- 3 *Other risks are inherent to real-life restrictions: e. g., an investor implementing Delta hedging can only trade at discrete times. The corresponding risk can be countered by “Delta-Gamma-hedging”.*
- 4 *There are also risks concerning changes of model parameters.*

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## The Greeks – 2

### Remark

*In reality, hedging is usually very expensive. Therefore, the Greeks are rather used to monitor the development of a portfolio.*

A non-exhaustive list of Greeks:

- Delta: derivative w. r. t. the price of the underlying.
- Gamma: second derivative w. r. t. the price of the underlying.
- Vega: derivative w. r. t. the volatility.
- Rho: derivative w. r. t. the interest rate.
- Theta: derivative w. r. t. time.



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# Finite Differences

- In very few models like the classical Black-Scholes model explicit formulas for the option prices exist. Of course, these formulas can be differentiated to get formulas for the Greeks.
- Numerical differentiation is one method for calculation of the Greeks. Let  $u(\alpha)$  denote the dependence of the price  $u$  of a derivative on some parameter  $\alpha$  and choose  $\epsilon > 0$  small enough. Use

$$\frac{u(\alpha + \epsilon) - u(\alpha)}{\epsilon}$$

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# The Logarithmic Trick

Given a family of random variables  $X^\alpha$ ,  $\alpha \in \mathbb{R}$ , having densities  $p(\alpha, x)$ ,  $x \in \mathbb{R}$ ,  $\alpha \in \mathbb{R}$ ,  $C^1$  in  $\alpha$ , and a bounded, measurable function  $f$ .

$$\begin{aligned}\frac{d}{d\alpha} E(f(X^\alpha)) &= \frac{d}{d\alpha} \int_{\mathbb{R}} f(x) p(\alpha, x) dx \\ &= \int_{\mathbb{R}} f(x) \frac{\frac{\partial p}{\partial \alpha}(\alpha, x)}{p(\alpha, x)} p(\alpha, x) dx \\ &= \int_{\mathbb{R}} f(x) \frac{\partial \log(p(\alpha, x))}{\partial \alpha} p(\alpha, x) dx = E(f(X^\alpha) \pi^\alpha),\end{aligned}$$

where  $\pi^\alpha = \frac{\partial}{\partial \alpha} \log(p(\alpha, X^\alpha))$  does not depend on  $f$ .

# The First Variation Process

Let  $X_t^x$ ,  $x \in \mathbb{R}^n$ ,  $t \in [0, T]$ , be the solution of the SDE

$$dX_t^x = a(X_t^x)dt + \sigma(X_t^x)dB^t \quad (1)$$

with  $X_0^x = x$ . Here,  $a : \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $\sigma : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times d}$  are  $C^1$  functions with linear growth.

The *first variation process* is the  $n \times n$ -dimensional process given by

$$dJ_{0 \rightarrow t}(x) = da(X_t^x) \cdot J_{0 \rightarrow t}(x)dt + \sum_{i=1}^d d\sigma^i(X_t^x) \cdot J_{0 \rightarrow t}(x)dB_t^i, \quad (2)$$

and  $J_{0 \rightarrow 0}(x) = I_n$ , where  $\sigma^i$  denotes the  $i$ th column of  $\sigma$ .

# Properties of the First Variation

- 1 If the coefficients of the SDE (1) are  $C^{1+\epsilon}$ , then there is a version of the solution  $X_t^x$  which is differentiable in  $x$ . In this case, the first variation is its Jacobian, i. e.

$$J_{0 \rightarrow t}(x) = d_x X_t^x.$$

- 2 The first variation is almost surely invertible. In fact, it is not difficult to find the SDE for  $J_{0 \rightarrow t}(x)^{-1}$ .

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# The Malliavin Derivative

- Without getting into details, we present the formal set-up of Malliavin calculus.
- The *Malliavin derivative* is a closed operator  $D : \mathcal{D}(D) \subset L^2(\Omega, \mathcal{F}, P) \rightarrow L^2([0, T] \times \Omega, \mathcal{B}([0, T]) \otimes \mathcal{F}, dt \otimes P; \mathbb{R}^d)$ . We write  $D_s F(\omega) = DF(s, \omega)$ ,  $F \in \mathcal{D}(D)$ .
- The dual map  $\delta$  is called *Skorohod stochastic integral*, i. e.  $\delta : \mathcal{D}(\delta) \subset L^2([0, T] \times \Omega, \mathcal{B}([0, T]) \otimes \mathcal{F}, dt \otimes P; \mathbb{R}^d) \rightarrow L^2(\Omega, \mathcal{F}, P)$  satisfies

$$E(F\delta(u)) = E\left(\int_0^T \langle D_s F, u_s \rangle_{\mathbb{R}^d} ds\right), \quad (3)$$

where  $F \in \mathcal{D}(D) \subset L^2(\Omega)$  and  $u \in \mathcal{D}(\delta) \subset L^2([0, T] \times \Omega)$ .



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# The Malliavin Derivative – 2

## Remark

*Usually, calculation of Malliavin derivatives or Skorohod integrals is a hard problem. There are, however, some important special cases.*

- 1 *For the process  $X^x$  solution of (1), the Malliavin derivative is given by*

$$D_s X_t^x = J_{0 \rightarrow t}(x) J_{0 \rightarrow s}(x)^{-1} \sigma(X_s^x) \mathbf{1}_{[0,t]}(s). \quad (4)$$

- 2 *For a predictable process  $u$ , the Skorohod integral coincides with the Itô integral.*
- 3 *Let  $F = (F^1, \dots, F^m)$ ,  $F^i \in \mathcal{D}(D)$ , and  $\varphi \in C^1(\mathbb{R}^m)$ , then  $\varphi(F) \in \mathcal{D}(D)$  and  $D\varphi(F) = \langle \nabla \varphi(F), DF \rangle_{\mathbb{R}^m}$ .*

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# Malliavin Weights

## Definition

For a given stochastic process  $X_t^x$ ,  $t \in [0, T]$ , as in (1) and a fixed time  $t$ , a *Malliavin weight* is a (sufficiently regular) random variable  $\pi$  such that

$$\nabla_x E(f(X_t^x)) = E(f(X_t^x)\pi) \quad (5)$$

for all, say, bounded, measurable  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ .

## Remark

- *By the “logarithmic trick”, Malliavin weights exist for all hypo-elliptic diffusions.*
- *For simplicity, we concentrate on the Delta.*

# Bismut-Elworthy-Li Formula

## Theorem

Assume that  $\sigma(X_t^x(\omega))$  has a right-inverse  $R_t(\omega) \in \mathbb{R}^{d \times n}$  for  $dt \otimes P$  a. e.  $(t, \omega)$  such that  $R_t J_{0 \rightarrow t}(x)^i \in L^2([0, T] \times \Omega)$ ,  $i = 1, \dots, n$ , where  $J_{0 \rightarrow t}(x)^i$  denotes the  $i$ th column of  $J_{0 \rightarrow t}(x)$ . Then for every bounded, measurable  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,

$$\nabla_x E(f(X_T^x)) = E\left(f(X_T^x) \int_0^T \frac{1}{T} R_t J_{0 \rightarrow t}(x) dB_t\right). \quad (6)$$

## Remark

The assumption is satisfied (with  $R_s = \sigma^{-1}(X_s^x)$ ) if  $d = n$ , and  $\sigma$  is uniformly elliptic, i. e.  $\exists \epsilon > 0$  s. t.

$$\xi^T \sigma(x)^T \sigma(x) \xi \geq \epsilon |\xi|^2, \quad \forall x, \xi \in \mathbb{R}^n.$$

# Proof of the Bismut-Elworthy-Li Formula

$$\begin{aligned}\nabla_x E(f(X_T^x)) &= \\ E\left(\frac{1}{T} \int_0^T \nabla f(X_T^x)^T J_{0 \rightarrow T}(x) \underbrace{J_{0 \rightarrow t}(x)^{-1} \sigma(X_t^x) R_t J_{0 \rightarrow t}(x)}_{=I_n} dt\right)\end{aligned}$$

The chain rule for Malliavin derivatives implies

$$D_t f(X_T^x) = \nabla f(X_T^x)^T D_t X_T^x = \nabla f(X_T^x)^T J_{0 \rightarrow T}(x) J_{0 \rightarrow t}(x)^{-1} \sigma(X_t^x)$$

and we get

$$\begin{aligned}\nabla_x E(f(X_T^x)) &= E\left(\int_0^T \langle D_t f(X_T^x), R_t J_{0 \rightarrow t}(x) / T \rangle dt\right) \\ &= E(f(X_T^x) \delta(t \mapsto R_t J_{0 \rightarrow t}(x) / T))\end{aligned}$$

# Bismut-Elworthy-Li Formula for a Parameter

## Theorem

Let  $a, \sigma$  depend on a real parameter  $\alpha$  and assume that  $\sigma$  is uniformly elliptic (in particular,  $d = n$ ). Then

$$\frac{\partial}{\partial \alpha} E(f(X_T^x)) = E(f(X_T^x) \delta(t \mapsto H(t, T, \alpha))), \quad (7)$$

with  $H(t, T, \alpha) = \frac{1}{T} \sigma(X_t^x, \alpha)^{-1} J_{0 \rightarrow t}(x) J_{0 \rightarrow T}(x)^{-1} \frac{\partial X_T^x}{\partial \alpha}$ .

## Proof.

Proceed similarly to the first proof using

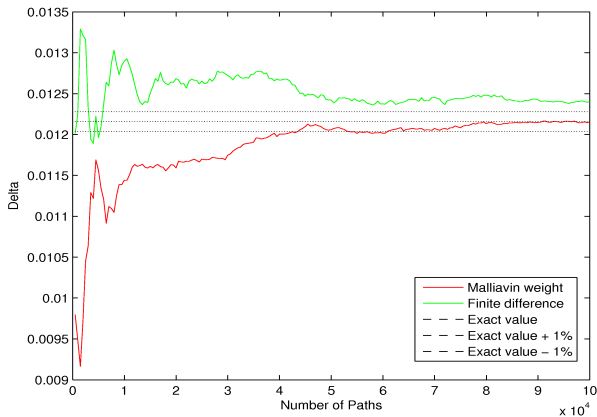
$$\nabla f(X_T^x)^T = D_t f(X_T^x) \sigma(X_t^x, \alpha)^{-1} J_{0 \rightarrow t}(x) J_{0 \rightarrow T}(x)^{-1}. \quad \square$$



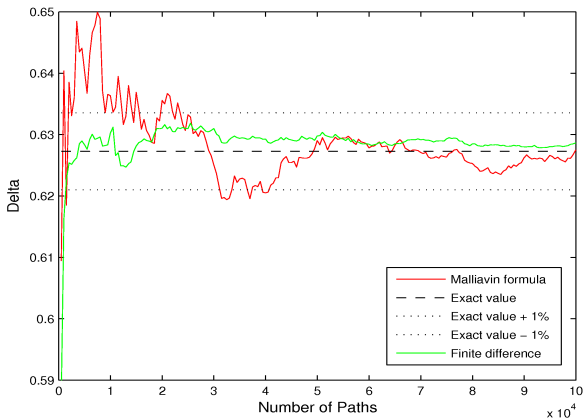
## Remarks on the Bismut-Elworthy-Li Formula

- Similar formulas are also possible in a hypo-elliptic set up.
- In general, the Malliavin weight is defined as Skorohod integral of some process. Note that – unlike the Itô integral – the definition of the Skorohod integral is essentially non-constructive.
- Consequently, formulas for Malliavin weights especially in non-elliptic situations are often not directly usable for computational purposes.

# Example: Delta of a Digital in the Bates Model







# Example: Delta of a Call in the Merton Model



# Summary

- The Greeks, sensitivities of option prices with respect to model parameters, are not only important for computational issues, but also because of their rôle in hedging strategies.
- Calculation using finite differences is often inefficient.
- Malliavin weights provide an alternative, in many situations superior method. In general, they are only given in a non-constructive way.
- Extending the theory to more general models, e. g. models driven by jump-diffusions, is a popular research topic.

## References

-  Jean-Michel Bismut: *Large deviations and the Malliavin calculus*, Birkäuser Verlag, 1984.
-  Eric Fournié, Jean-Michel Lasry, Jérôme Lebuchoux, Pierre-Louis Lions, Nizar Touzi: *Applications of Malliavin calculus to Monte Carlo methods in finance*, Finance and Stochastics 3, 391–412, 1999.
-  Ioannis Karatzas: *Lectures on the mathematics of finance*, CRM Monograph Series 8, AMS, 1996.
-  David Nualart: *The Malliavin calculus and related topics*, Probability and Its Applications, Springer-Verlag, 1995.

## Part II

# Approximation of Malliavin Weights

## Rewriting the Equation

We rewrite equation (1) as

$$dX_t^x = V_0(X_t^x)dt + \sum_{i=1}^d V_i(X_t^x) \circ dB_t^i = \sum_{i=0}^d V_i(X_t^x) \circ dB_t^i, \quad (8)$$

where  $\circ dB_t^i$  denotes the *Stratonovich stochastic integral* and we use the notation " $\circ dB_t^0 = dt$ ".  $V_i(x) = \sigma^i(x)$ ,  $i = 1, \dots, d$  and  $V_0(x) = a(x) - \frac{1}{2} \sum_{i=1}^d dV_i(x) \cdot V_i(x)$ .

### Remark

We understand vector fields  $V : \mathbb{R}^n \rightarrow \mathbb{R}^n$  as functions and as differential operators acting on  $C^\infty(\mathbb{R}^n; \mathbb{R})$  by

$$Vf(x) = df(x) \cdot V(x), \quad f \in C^\infty(\mathbb{R}^n; \mathbb{R}), \quad x \in \mathbb{R}^n.$$

## Assumptions on the Data

- The vector fields  $V_0, \dots, V_d$  are smooth and  $C^\infty$ -bounded, i. e. their derivatives of order greater than 0 are bounded.
- The vector fields satisfy a uniform Hörmander condition, see [Kusuoka 2002] for details. Consequently, the corresponding diffusion  $X$  is hypo-elliptic.



# Stochastic Taylor Expansion

We define a degree on the set of multi-indices

$\mathcal{A} = \bigcup_{k \in \mathbb{N}} \{0, \dots, d\}^k$  by  $\deg((i_1, \dots, i_k)) = k + \#\{j \mid i_j = 0\}$ ,  
 i. e. 0s are counted twice.

## Theorem

Fix  $m \in \mathbb{N}$  and  $f \in C^\infty(\mathbb{R}^n)$ . Then

$$f(X_t^x) = \sum_{\substack{\alpha = (i_1, \dots, i_k) \in \mathcal{A} \\ \deg(\alpha) \leq m}} V_{i_1} \cdots V_{i_k} f(x) \int_{0 < t_1 < \dots < t_k < t} \circ dB_{t_1}^{i_1} \circ \dots \circ dB_{t_k}^{i_k} + R_m(f, t, x),$$

with  $\sqrt{E(R_m(f, t, x)^2)} = \mathcal{O}(t^{(m+1)/2})$ .

# Proof of the Stochastic Taylor Expansion

Proof.

By Itô's formula, we get

$$f(X_t^x) = f(x) + \sum_{i=0}^d \int_0^t V_i f(X_s^x) \circ dB_s^i.$$

Now we apply Itô's formula to  $V_i f(X_s^x)$ ,  $i = 0, \dots, d$ , and obtain

$$f(X_t^x) = f(x) + \sum_{i=0}^d \int_0^t \left( V_i f(x) + \sum_{j=0}^d \int_0^s V_j V_i f(X_u^x) \circ dB_u^j \right) \circ dB_s^i.$$

Iterate this procedure and then apply Itô's lemma to get the order estimate for the rest term. □

## Remark

- 1 *The stochastic Taylor expansion is the starting point for many applications in stochastic analysis and for several numerical methods, including*
  - *Stochastic Taylor schemes and*
  - *Cubature on Wiener space, a method based on Terry Lyons's theory of rough paths.*
- 2 *Following the latter concept, we interpret the process  $(B_t^\alpha)_{\alpha \in \mathcal{A}, \deg(\alpha) \leq m}$ ,  $t \in [0, T]$ , as the “probabilistic core” of the solution of the SDE.*
- 3 *Consequently, a better understanding of this process might yield methods to calculate the non-constructive Malliavin-weight formulas presented in the last section.*

## Passing to the Algebraic Framework

- In order to study the stochastic process given by the iterated Stratonovich integrals up to order  $m$ , we embed it into an appropriate algebraic/geometric framework.
- We additionally assume that the Stratonovich drift vanishes, i. e.  $V_0 = 0$ . This assumption simplifies the notation and allows us to use the usual degree  $\deg((i_1, \dots, i_k)) = k$  for  $(i_1, \dots, i_k) \in \{1, \dots, d\}^k$ . Furthermore, it will allow us to omit some subtle restrictions in the following.
- We stress that the results remain *essentially* true for  $V_0 \neq 0$ .

# The Free, Nilpotent Algebra

- Let  $\mathbb{A}_{d,0}^m$  be the space of all non-commutative polynomials in  $e_1, \dots, e_d$  of degree less than or equal to  $m$ .
- Define a non-commutative multiplication on  $\mathbb{A}_{d,0}^m$  by cutting off all monomials of higher degree than  $m$ .  $\mathbb{A}_{d,0}^m$  becomes the *free associative, non-commutative, step- $m$  nilpotent real algebra with unit* in  $d$  generators  $e_1, \dots, e_d$ .
- Let  $W_0$  denote the linear span of the unit element 1 of  $\mathbb{A}_{d,0}^m$ , i. e.  $W_0$  is the space of all polynomials of degree 0. We identify  $W_0 \simeq \mathbb{R}$  and denote by  $x_0$  the projection of  $x \in \mathbb{A}_{d,0}^m$  on  $W_0$ .

## The Free, Nilpotent Algebra – 2

- $\exp : \mathbb{A}_{d,0}^m \rightarrow \mathbb{A}_{d,0}^m$  is defined by  $\exp(x) = 1 + \sum_{i=1}^{\infty} \frac{x^i}{i!}$ .
- The logarithm is defined for  $x_0 \neq 0$  by

$$\begin{aligned} \log(x) &= \log(x_0) + \sum_{i=1}^{\infty} \frac{(-1)^{i-1}}{i} \left( \frac{x - x_0}{x_0} \right)^i \\ &= \log(x_0) + \sum_{i=1}^m \frac{(-1)^{i-1}}{i} \left( \frac{x - x_0}{x_0} \right)^i. \end{aligned}$$

- $\mathbb{A}_{d,0}^m$  equipped with the commutator bracket  $[x, y] = xy - yx$ ,  $x, y \in \mathbb{A}_{d,0}^m$ , is a Lie algebra.

# The Free, Nilpotent Lie Group

- Let  $\mathfrak{g}_{d,0}^m$  denote the *free, step- $m$  nilpotent Lie algebra* generated by  $\{e_1, \dots, e_d\}$ , i. e.

$$\mathfrak{g}_{d,0}^m = \langle \{e_i, [e_i, e_j], [e_i, [e_j, e_k]], \dots \mid i, j, k = 1, \dots, d\} \rangle.$$

- We define the *step- $m$  nilpotent free Lie group* as the exponential image of  $\mathfrak{g}_{d,0}^m$ , i. e.  $G_{d,0}^m = \exp(\mathfrak{g}_{d,0}^m)$ .
- $G_{d,0}^m$  is a Lie group and  $\mathfrak{g}_{d,0}^m$  is its Lie algebra, which can be seen by the Campbell-Baker-Hausdorff formula

$$\exp(x) \exp(y) = \exp\left(x + y + \frac{1}{2}[x, y] + \frac{1}{12}([x, [x, y]] - [y, [y, x]]) + \dots\right).$$

Note that  $\exp : \mathfrak{g}_{d,0}^m \rightarrow G_{d,0}^m$  and  $\log : G_{d,0}^m \rightarrow \mathfrak{g}_{d,0}^m$  define a global chart of the Lie group.

## Example: The Heisenberg Group

- The Heisenberg group is the group  $G_{2,0}^2$ , a 3-dimensional submanifold of  $\mathbb{A}_{2,0}^2 \simeq \mathbb{R}^7$ .
- It allows the following representation as a matrix group:

$$G_{2,0}^2 = \left\{ \begin{pmatrix} 1 & a & c \\ 0 & 1 & b \\ 0 & 0 & 1 \end{pmatrix} \mid a, b, c \in \mathbb{R} \right\}.$$

The Lie algebra – as tangent space at  $I_3 \in G_{2,0}^2$  – is

$$\mathfrak{g}_{2,0}^2 = \left\{ \begin{pmatrix} 0 & x & z \\ 0 & 0 & y \\ 0 & 0 & 0 \end{pmatrix} \mid x, y, z \in \mathbb{R} \right\}.$$

- An isomorphism is given by  $e_1 = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ ,  $e_2 = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ .



# Encoding Iterated Stratonovich Integrals

## Definition

For  $y \in \mathbb{A}_{d,0}^m$  we define a stochastic process  $Y_t^y$ ,  $t \in [0, T]$ , by

$$Y_t^y = y \left( \sum_{\alpha \in \mathcal{A}, \deg(\alpha) \leq m} B_t^\alpha e_\alpha \right),$$

where  $e_\alpha = e_{i_1} \cdots e_{i_k}$  for  $\alpha = (i_1, \dots, i_k)$ ,  $e_\emptyset = 1$ , and  $B_t^\alpha$  denotes the corresponding iterated Stratonovich integral, i. e.

$$B_t^\alpha = \int_{0 < t_1 < \dots < t_k < t} \circ dB_{t_1}^{i_1} \cdots \circ dB_{t_k}^{i_k}.$$

## Remark

$Y_t^1$  is the vector of the iterated Stratonovich integrals written in the canonical basis of  $\mathbb{A}_{d,0}^m$ .

# The Geometry of the Iterated Integrals

## Theorem

Define vector fields  $D_i(x) = xe_i$ ,  $x \in \mathbb{A}_{d,0}^m$ ,  $i = 1, \dots, d$ .

- ①  $Y_t^y$  is solution of the SDE

$$dY_t^y = \sum_{i=1}^d D_i(Y_t^y) \circ dB_t^i, \quad Y_0^y = y.$$

- ② Given  $y \in G_{d,0}^m$ , we have  $Y_t^y \in G_{d,0}^m$  a. s. for all  $t \in [0, T]$ .  
 ③ By the Feynman-Kac formula,

$$E(Y_t^y) = y \exp\left(\frac{t}{2} \sum_{i=1}^d e_i^2\right).$$

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 ③ By the Feynman-Kac formula,

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## The Geometry of the Iterated Integrals – 2

### Remark

- ① *The formula (3) allows efficient calculation of all moments of iterated Stratonovich integrals. Observe that  $E(Y_t^1) \notin G_{d,0}^m$ .*
- ② *By Statement 2, we can carry out all relevant calculations for iterated Stratonovich integrals in the vector space  $\mathfrak{g}_{d,0}^m$  by passing to the process  $Z_t = \log(Y_t^1)$ .*

### Example

For  $m = d = 2$ , we get  $Z_t = B_t^1 e_1 + B_t^2 e_2 + A_t[e_1, e_2]$ , where  $A_t$  denotes Lévy's area

$$A_t = \frac{1}{2} \int_0^t B_s^1 \circ dB_s^2 - \frac{1}{2} \int_0^t B_s^2 \circ dB_s^1.$$

# Malliavin Weights on $G_{d,0}^m$

## Proposition

Fix  $w \in \mathfrak{g}_{d,0}^m$  and  $t \in [0, T]$ . There is a non-adapted, Skorohod integrable,  $\mathbb{R}^d$ -valued process  $a_s$ ,  $0 \leq s \leq T$  such that for any bounded, measurable function  $f : G_{d,0}^m \rightarrow \mathbb{R}$  and any  $y \in G_{d,1}^m$

$$\frac{\partial}{\partial \epsilon} \Big|_{\epsilon=0} E(f(Y_t^{y+\epsilon w})) = E(f(Y_t^y) \pi_{d,0}^m),$$

where  $\pi_{d,0}^m = \delta(a)$ .

## Proof.

The proof is very similar to the proof for the existence of Malliavin weights in a general, hypo-elliptic setting. See [Teichmann 06].  $\square$

# Malliavin Weights on $G_{d,0}^m$ – 2

## Remark

- 1 *The strategy  $a$  is universal in the sense that it does only depend on  $m$ ,  $d$ ,  $t$  and – in a linear way –  $w$ .*
- 2 *Yet again, the Malliavin weight is given in an essentially non-constructive way due to the Skorohod integration. Note that even calculation of the strategy  $a$  is difficult. However, approximation of the weight  $\pi_{d,0}^m$  turns out to be possible.*

# Approximation with Universal Weights

## Theorem

Fix  $x, v \in \mathbb{R}^n$ ,  $t \in [0, T]$  and  $m \geq 1$  such that  $v$  can be written as

$$v = \sum_{\alpha \in \mathcal{A} \setminus \{\emptyset\}, \deg(\alpha) \leq m-1} w_\alpha [V_{i_1}, [V_{i_2}, [\dots, V_{i_k}] \dots]](x)$$

for some  $w_\alpha \in \mathbb{R}$ . Define  $w = \sum w_\alpha [e_{i_1}, [e_{i_2}, [\dots, e_{i_k}] \dots]] \in \mathfrak{g}_{d,0}^m$  and let  $\pi_{d,0}^m$  denote the corresponding universal Malliavin weight. Then for any  $C^\infty$ -bounded function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  we have

$$\left. \frac{\partial}{\partial \epsilon} \right|_{\epsilon=0} E(f(X_t^{x+\epsilon v})) = E(f(X_t^x) \pi_{d,0}^m) + \mathcal{O}(t^{(m+1)/2}).$$

We note that the constant in the leading order term in the error estimate depends only on the first derivative of  $f$ .



# Cubature

## Definition

Given a finite Borel measure  $\mu$  on  $\mathbb{R}^n$  with finite moments of order up to  $m \in \mathbb{N}$ . A *cubature formula* of degree  $m$  is a collection of weights  $\lambda_1, \dots, \lambda_k > 0$  and points  $x_1, \dots, x_k \in \text{supp}(\mu) \subset \mathbb{R}^n$  such that the following equality holds for any polynomial  $f$  on  $\mathbb{R}^n$  of degree less or equal  $m$ :

$$\int_{\mathbb{R}^n} f(x) \mu(dx) = \sum_{i=1}^k \lambda_i f(x_i).$$

# Chakalov's Theorem

## Theorem

*For any Borel measure  $\mu$  on  $\mathbb{R}^n$  with finite moments of order up to  $m$  there is a cubature formula with size  $k \leq \dim \text{Pol}_m(\mathbb{R}^n)$ , the space of polynomials on  $\mathbb{R}^n$  of order up to  $m$ .*

## Remark

- *Chakalov's Theorem is non-constructive, and construction of efficient cubature formulas remains a non-trivial problem in higher dimensions.*
- *The bound on the size is a consequence of Caratheodory's Theorem.*

# Cubature Formulas for the Universal Weight

## Theorem

Fix  $t \in [0, T]$  and  $w \in \mathfrak{g}_{d,0}^m$ . There are points  $x_1, \dots, x_r \in G_{d,0}^m$  and weights  $\rho_1, \dots, \rho_r \neq 0$  such that

$$E(Y_t^1 \pi_{d,0}^m) = w \exp\left(\frac{t}{2} \sum_{i=1}^d e_i^2\right) = \sum_{j=1}^r \rho_j x_j.$$

Furthermore, we may choose  $r \leq 2 \dim \mathbb{A}_{d,0}^m + 2$ .

# Proof of Cubature for the Universal Weight

Proof.

The first equality is an easy consequence of

$$E(Y_t^y) = y \exp\left(\frac{t}{2} \sum_{i=1}^d e_i^2\right), \quad y \in \mathbb{A}_{d,0}^m.$$

Now define two positive measures on the Wiener space by  $\frac{dQ_+}{dP} = (\pi_{d,0}^m)_+$  and  $\frac{dQ_-}{dP} = (\pi_{d,0}^m)_-$ . By absolute continuity of  $Q_{\pm}$  w. r. t.  $P$ ,  $Y_t^1 \in G_{d,0}^m$   $Q_{\pm}$ -a. s. Chakalov's Theorem applied to the laws of  $Y_t^1$  under  $Q_+$  and  $Q_-$  yields

$$E_P(Y_t^1 \pi_{d,0}^m) = E_{Q_+}(Y_t^1) - E_{Q_-}(Y_t^1) = \sum_{j=1}^r \rho_j X_j. \quad \square$$

# A Reformulation of Cubature

- A theorem from sub-Riemannian geometry (Chow's Theorem) implies that any  $x \in G_{d,0}^m$  can be joined to 1 by the the solution of the ODE

$$\dot{x}_t = \sum_{i=1}^d D_i(x_t) \dot{\omega}_i(t) = \sum_{i=1}^d x_t e_i \dot{\omega}_i(t)$$

for some paths of bounded variation  $\omega_i : [0, T] \rightarrow \mathbb{R}$ ,  $i = 1, \dots, d$ , i. e.  $x_0 = 1$  and  $x_t = x$ .

- We denote the solution to the above ODE for some path  $\omega$  of bounded variation by  $Y_s^1(\omega)$ ,  $s \in [0, T]$ .
- Consequently, there are  $\omega_j \in C_{bv}([0, T]; \mathbb{R}^d)$ ,  $j = 1, \dots, r$ , s. t.

$$E(Y_t^1 \pi_{d,0}^m) = \sum_{j=1}^r \rho_j Y_t^1(\omega_j).$$

# Malliavin Weights by Cubature

## Theorem

Fix  $x, v \in \mathbb{R}^n$ ,  $t \in [0, T]$  and  $m \geq 1$  such that  $v$  can be written as

$$v = \sum_{\alpha \in \mathcal{A} \setminus \{\emptyset\}, \deg(\alpha) \leq m-1} w_\alpha [V_{i_1}, [V_{i_2}, [\dots, V_{i_k}] \dots]](x)$$

for some  $w_\alpha \in \mathbb{R}$ . Define  $w = \sum w_\alpha [e_{i_1}, [e_{i_2}, [\dots, e_{i_k}] \dots]] \in \mathfrak{g}_{d,0}^m$  and let  $\rho_j, \omega_j$ ,  $j = 1, \dots, r$ , denote the cubature formula for the corresponding weight  $\pi_{d,0}^m$ . Then

$$\left. \frac{\partial}{\partial \epsilon} \right|_{\epsilon=0} E(f(X_t^{x+\epsilon v})) = \sum_{j=1}^r \rho_j f(X_t^x(\omega_j)) + \mathcal{O}(t^{(m+1)/2}),$$

where  $\frac{dX_t^x(\omega_j)}{dt} = \sum_{i=1}^d V_i(X_t^x(\omega_j)) \dot{\omega}_j^i(t)$  with  $X_0^x(\omega_j) = x$ .

## Remarks

- The above formula is some kind of finite difference formula! Instead of solving the PDE problem for different starting points, we solve the SDE problem for different trajectories  $\omega$  of the Brownian motion.
- The same procedure *without*  $\pi_{d,0}^m$  gives a method for calculation of  $E(f(X_t^x)) \approx \sum_{j=1}^l \lambda_j f(X_t^x(\tilde{\omega}_j))$ . This method – introduced by T. Lyons and N. Victoir – is known as “Cubature on Wiener space”.
- For actual computations, it is necessary to iterate the procedure along a partition  $0 = t_0 < t_1 < \dots < t_k = t$  of  $[0, t]$ . This is possible using cubature on Wiener space, yielding, at least in theory, a method of order  $\frac{m-1}{2}$  for approximation of Greeks.

# The Heat Kernel on $G_{d,0}^m$

## Definition

The density  $p_t(x)$ ,  $x \in G_{d,0}^m$ , of the process  $Y_t^1$  with respect to the Haar measure on the Lie group is called *heat kernel*.

## Remark

- 1 *The name comes from the fact that  $p_t$  is the fundamental solution of the heat equation with respect to the sub-Laplacian  $L = \frac{1}{2} \sum_{i=1}^d D_i^2$ , the infinitesimal generator of  $Y_t^1$ .*
- 2 *We can equivalently study the density of  $Z_t = \log(Y_t^1)$  with respect to the Lebesgue measure on  $\mathfrak{g}_{d,0}^m$ .*
- 3 *In the setting without drift, the heat kernel always exists as a Schwartz function. With non-vanishing drift, we need to factor out the direction in  $\mathfrak{g}_{d,1}^m$  corresponding to  $t$ .*



# Approximation of the Heat Kernel

- We want to find polynomial approximations of the heat kernel  $p_t$  or the Malliavin weight  $d \log p_t \circ Y_t^1$ . In the following we concentrate on the former problem and tacitly switch to  $\mathfrak{g}_{d,0}^m$  using the same notation  $p_t$  for the density of  $Z_t$  on  $\mathfrak{g}_{d,0}^m$ .
- Choose a suitable (Gaussian) measure  $Q_t$  with density  $r_t$  on  $\mathfrak{g}_{d,0}^m$  and let  $h_\alpha(t, \cdot)$  denote the corresponding family of orthonormal (Hermite) polynomials.
- We need to calculate the integral

$$\int_{\mathfrak{g}_{d,0}^m} \frac{p_t(z)}{r_t(z)} h_\alpha(t, z) Q_t(dz) = \int_{\mathfrak{g}_{d,0}^m} p_t(z) h_\alpha(t, z) dz = E(h_\alpha(t, Z_t)).$$

## Approximation of the Heat Kernel – 2

- This procedure will give us an approximation

$$\frac{p_t(z)}{r_t(z)} = \sum_{\alpha} a_t^{\alpha} h_{\alpha}(t, z)$$

valid in  $L^2(\mathfrak{g}_{d,0}^m, Q_t)$ , where  $a_t^{\alpha} = E(h_{\alpha}(t, Z_t))$ .

- Note that  $h_{\alpha}(t, Z_t)$  is a (time dependent) *polynomial* in the iterated Stratonovich integrals of order up to  $m$ . Calculation of  $a_t^{\alpha}$  is enabled by

$$E(\tilde{Y}_t^1) = \exp_{\tilde{m}}\left(\frac{t}{2} \sum_{i=1}^d e_i^2\right),$$

where  $\tilde{Y}$  denotes the process of iterated integrals in  $\mathbb{A}_{d,0}^{\tilde{m}}$  with  $\tilde{m} > m$  large enough.

- The use of  $Q_t$  instead of  $dz$  is preferable because polynomials are not  $dz$ -square integrable.





# Applications of Approximate Heat Kernels

- Approximate heat kernels can be used to make higher order Taylor schemes for approximation of SDEs feasible.
- In the context of Malliavin weights, they provide approximations to the universal Malliavin weights defined before.
- Finally, the subject of heat kernels on Lie groups is a well-established subject of mathematical research in its own right.

# Summary

- By using Stochastic Taylor expansion, we constructed universal Malliavin weights, which yield approximations to the Greeks for a very general class of ( $d$ -dimensional) problems.
- Cubature formulas for Greeks – in combination with cubature on Wiener space – yield high-order methods for the calculation of the Greeks, even in situations, where no direct formula for the Greeks is possible.
- There are still many open problems regarding actual usability of this theory for computations.

## References

-  Shigeo Kusuoka: *Approximation of expectation of diffusion processes based on Lie algebra and Malliavin calculus*, Advances in mathematical economics 6, 69-83, 2004.
-  Terry Lyons: *Differential equations driven by rough signals*, Rev. Mat. Iberoam. 14, No.2, 215-310, 1998.
-  Terry Lyons and Nicolas Victoir: *Cubature on Wiener space*, Proceedings of the Royal Society London A 460, 169-198, 2004.
-  Josef Teichmann: *Calculating the Greeks by Cubature formulas*, Proceedings of the Royal Society London A 462, 647-670, 2006.