

Lévy Processes in Finance

Summer school Sandbjerg

Jan Kallsen

TU München/CAU Kiel

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Outline

- 1 Mathematical finance
- 2 Stochastic calculus
- 3 Applications to finance

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- 1 Mathematical finance
- 2 Stochastic calculus
- 3 Applications to finance

Price processes and trading strategies

- **securities price process** $S = (S_0(t), \dots, S_d(t))_{t \in [0, T]}$
 - ▶ \mathbb{R}^{d+1} -valued semimartingale on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in [0, T]}, P)$
 - ▶ $S_i(t)$ is price of security i at time t
 - ▶ Example:
 - Bond: $S_0(t) = e^{rt}$
 - Stock: $S_1(t) = S_1(0)e^{L(t)}$ with Lévy process L
- **trading strategy** or **portfolio** $\varphi = (\varphi_0(t), \dots, \varphi_d(t))_{t \in [0, T]}$
 - ▶ \mathbb{R}^{d+1} -valued predictable process
 - ▶ $\varphi_i(t)$ is number of shares of security i at time t
 - ▶ **value process** $V_\varphi(t) := \varphi(t)^\top S(t) = \sum_{i=0}^d \varphi_i(t)^\top S_i(t)$
 - ▶ φ is **self financing** if S -integrable and $V_\varphi = V_\varphi(0) + \varphi \bullet S$

Discounting

for simplified bookkeeping

- Idea: express prices in multiples of **numeraire** security S_0 .
 - ▶ Assume $S_0, (S_0)_- > 0$.
- **discounted price process** $\hat{S} := \frac{1}{S_0} S = (1, \frac{S_1}{S_0}, \dots, \frac{S_d}{S_0})$
- **discounted value process** $\hat{V}_\varphi := \frac{V_\varphi}{S_0} = \varphi^\top \hat{S}$
- Example: for $S_0(t) = e^{rt}$ we have $\hat{S}(t) = e^{-rt} S(t)$ and $\hat{V}_\varphi(t) = e^{-rt} V_\varphi(t)$.
- Lemma:
 φ self financing $\Leftrightarrow \varphi$ \hat{S} -integrable and $\hat{V}_\varphi = \hat{V}_\varphi(0) + \varphi \cdot \hat{S}$
- Lemma: For $(\hat{S}_1, \dots, \hat{S}_d)$ -integrable $(\varphi_1, \dots, \varphi_d)$ and $v \in \mathbb{R}$ ex. unique φ_0 such that $\varphi = (\varphi_0, \dots, \varphi_d)$ self financing with $V_\varphi(0) = v$.
- Consequence: do not worry about self financability and φ_0 .

Arbitrage and consequences

An **admissible** self financing strategy φ is an **arbitrage** if $V_\varphi(0) = 0$, $V_\varphi(T) \geq 0$, $P(V_\varphi(T) > 0) > 0$.

- Without admissibility constraint Black-Scholes model is not arbitrage free.
- We take careless engineering point of view here. Precise statements are difficult.

Rule (law of one price)

Let φ, ψ be admissible strategies in an arbitrage-free market. If $V_\varphi(T) = V_\psi(T)$, then $V_\varphi = V_\psi$.

Rule (fundamental theorem of asset pricing, FTAP)

The market is arbitrage free \iff there exists some **equivalent martingale measure (EMM)** Q (i.e. $Q \sim P$ is probability measure and \hat{S} is Q -martingale).

Derivatives

- **European option**: \mathcal{F}_T -measurable random variable X
 - ▶ X stands for payoff at time T .
 - ▶ Examples:
 - European call $X = (S_1(T) - K)^+$ or put $X = (K - S_1(T))^+$
 - Lookback call $X = (\sup_{t \in [0, T]} S_1(t) - K)^+$
 - ▶ **Discounted payoff** is $\hat{X} := \frac{X}{S_0(T)}$.
- Futures and American options are more involved.
- What is a reasonable price for the option at time $t < T$?
- We distinguish **liquidly traded** and **over-the-counter** options.

Option pricing

for liquidly traded derivatives

- Consider underlyings S_0, S_1 (given processes) and derivative price process S_2 with $S_2(T) = X$ (yet to be determined).
- The FTAP implies:
no arbitrage $\implies \hat{S}_2(t) = E_Q(\hat{X}|\mathcal{F}_t)$ for some EMM Q

Rule

S_2 does not depend on the chosen EMM $Q \iff$

X is **attainable** (i.e. $X = V_\varphi(T)$ for some admissible strategy φ).

Corollary (second fundamental theorem of asset pricing)

An arbitrage free market is **complete** (i.e. any option is attainable)

\iff there exists only one EMM.

- Most models involving non-Gaussian Lévy processes are incomplete.

Problem of incompleteness

for liquidly traded options

- Problem: range of possible prices typically too large
 - ▶ e.g. trivial arbitrage bounds for European call in geometric Lévy model for standard Lévy processes (VG, NIG, ...)
 - ▶ Away from Black-Scholes case arbitrage theory does not yield useful information.
- Approach 1:
consider some derivatives as underlyings, e.g. vanilla options
 - ▶ get fewer EMM's, hence tighter price bounds for other derivatives
 - ▶ Problem: how to get reasonable and arbitrage free model for new underlyings?
 - ▶ Approach seems rather cumbersome.
- Approach 2 (**martingale modelling**):
model prices directly under market's pricing measure
 - ▶ used in practice

Martingale modelling

for liquidly traded options

- Choose parametric class of models for underlyings S_0, S_1 under market's EMM Q .
 - ▶ e.g. $S_0(t) = e^{rt}$, $S_1(t) = S_1(0)e^{L(t)}$ with variance-gamma Lévy process L
- Consider martingale constraint, i.e. \hat{S}^1 Q -martingale.
 - ▶ here: restrict variance-gamma parameters such that $E(e^{L(1)-r}) = 1$
- Determine remaining parameters by **calibration** to observed prices of liquid options at $t = 0$.
 - ▶ choose parameters such that

$$\text{discounted observed price} = E_Q(\text{discounted payoff})$$

for all observed vanilla options

- ▶ use e.g. least squares procedure if more options than parameters
- Use model e.g. for pricing of new non-liquidly traded derivatives.

Martingale modelling

for liquidly traded options (ct'd)

- Question: is the parametric class of models appropriate?
 - ▶ Statistical tests are useless because they concern real world probabilities P .
- Answer: inconsistencies may indicate inappropriate class or model.
 - ▶ e.g. no choice of parameters leads to reasonable fit of observed option prices
 - ▶ or recalibration leads to heavily time-varying models
- However: Entirely different classes may produce reasonable fit, but differ substantially for further e.g. path-dependent options (**model risk**).

Martingale modelling

for liquidly traded options (ct'd)

Two different situations:

- only interested in valuation of new options
 - ▶ modelling under pricing measure Q suffices, physical measure P not needed
- real-world probabilities needed as well (e.g. for value at risk assessments etc.)
 - ▶ model dynamics of S_0, S_1 under both P and Q
 - ▶ choose typically same parametric class under both P and Q
 - ▶ e.g. $S_1(T) = S_1(0)e^{L(t)}$, L VG Lévy process under P and Q
 - ▶ determine P -parameters by estimation using stock price data, Q -parameters by calibration using option prices
 - ▶ OBS! condition $Q \sim P$ leads to constraints on possible parameters

Individual pricing of over-the-counter derivatives

- Situation:
 - ▶ underlyings S_0, S_1 liquidly traded, but not the option
 - ▶ client wants to buy option X from bank at time $t = 0$
 - ▶ reasonable prices π ?
- $\pi \leq \pi_{\text{high}} := \inf\{\pi \in \mathbb{R} : \exists \text{ adm. } \varphi \text{ with } V_0(\varphi) = \pi \text{ and } V_\varphi(T) \geq X\}$
 - ▶ justification: otherwise client can do better by investing in such **superhedge** φ
- $\pi \leq \pi_{\text{low}} := \sup\{\pi \in \mathbb{R} : \exists \text{ adm. } \varphi \text{ with } V_0(\varphi) = \pi \text{ and } V_\varphi(T) \leq X\}$
 - ▶ justification: otherwise bank should sell such **subhedge** φ rather than option
- How to determine $\pi_{\text{high}}, \pi_{\text{low}}$?

Individual pricing

of over-the-counter derivatives (ct'd)

Rule (superhedging)

$$\hat{\pi}_{\text{high}} := \frac{\pi_{\text{high}}}{S_0(0)} = \sup\{E_Q(\hat{X}) : Q \text{ EMM for } S_0, S_1\}$$

$$\hat{\pi}_{\text{low}} := \frac{\pi_{\text{low}}}{S_0(0)} = \inf\{E_Q(\hat{X}) : Q \text{ EMM for } S_0, S_1\}$$

- Price bounds coincide with price interval for liquid options.
- hence same problem as before: range of prices too large
- alternatives:
 - ▶ pricing based on quadratic hedging
 - ▶ utility indifference pricing

Quadratic hedging

and pricing over-the-counter derivatives

- quadratic hedging of option X :
 - ▶ minimize expected squared hedging error

$$\varepsilon^2(\varphi) := E((\hat{V}_\varphi(T) - \hat{X})^2)$$

over all admissible strategies φ

- ▶ optimal strategy φ^* , minimal hedging error $\varepsilon^2 := \varepsilon^2(\varphi^*)$
 - ▶ $\hat{V}_{\varphi^*}(T)$ is orthogonal projection in L^2 of \hat{X} on $\{\hat{V}_\varphi(T) : \varphi \text{ admissible strategy}\}$.
- application to over-the-counter trade
 - ▶ charge e.g. $\pi = V_{\varphi^*}(0) + \lambda\varepsilon^2$ as option's premium
 - ▶ use $V_{\varphi^*}(0)$ to buy hedging strategy φ^* .
 - ▶ take multiple of ε^2 as compensation for unhedgeable risk (parameter λ stands for risk aversion)
 - How to determine φ^* , $V_{\varphi^*}(0)$, ε^2 ?

Quadratic hedging (ct'd)

Theorem (quadratic hedging)

Assumption: \hat{S} is martingale (general case is more involved).

Then

$$V_{\varphi^*}(0) = E(\hat{X})$$

$$\varphi^*(t) = \frac{d\langle \hat{V}, \hat{S}_1 \rangle(t)}{d\langle \hat{S}_1, \hat{S}_1 \rangle(t)}$$

$$\varepsilon^2 = E\left(\langle \hat{V}, \hat{V} \rangle(T) - (\varphi^*)^2 \cdot \langle \hat{S}_1, \hat{S}_1 \rangle(T)\right)$$

with $\hat{V}(t) := E(\hat{X} | \mathcal{F}_t)$.

- How to determine angle brackets $\langle \hat{V}, \hat{S}_1 \rangle$ etc.?

Utility indifference pricing

of over-the-counter derivatives

- Problem of approach via quadratic hedging:
gains are penalized, hence economically questionable
- Alternative:
increasing utility functions and indifference pricing
- Idea:
sell option only if this is profitable for the bank (i.e. utility rises)

Expected utility maximization

- fix increasing, strictly concave utility function u .
- Examples:
 - ▶ $u(x) = \frac{x^{1-p}}{1-p}$ for $p \in (0, \infty) \setminus \{1\}$
 - ▶ $u(x) = \log(x)$ (behaves like $p = 1$ above)
 - ▶ $u(x) = 1 - e^{-\rho x}$ for $\rho > 0$
- maximize expected utility $E(u(\hat{V}_\varphi(T)))$ over all admissible strategies φ with fixed initial value $V_\varphi(0) = v$
- solution approaches:
 - ▶ dynamic programming
 - ▶ martingale methods (here)

Rule (dual characterization)

φ admissible strategy such that $\frac{dQ}{dP} = \frac{u'(\hat{V}_\varphi(T))}{E(u'(\hat{V}_\varphi(T)))}$ density of EMM
 $\implies \varphi$ is optimal.

Utility indifference pricing

of over-the-counter derivatives

- Question:

- ▶ underlyings S_0, S_1 liquidly traded, but not the option
- ▶ client wants to buy option X from bank at time $t = 0$
- ▶ reasonable prices π for the bank?

- Answer:

- ▶ utility of bank without option trade:

$$U_0 := \sup\{E(u(\hat{V}_\varphi(T))) : \varphi \text{ adm. with } V_0(\varphi) = v\}$$

- ▶ utility of bank with option trade:

$$U_X(\pi) := \sup\{E(u(\hat{V}_\varphi(T) + \hat{\pi} - \hat{X})) : \varphi \text{ adm. with } V_0(\varphi) = v\}$$

- ▶ require $U_X(\pi) > U_0$ (trade raises bank's utility)

- ▶ **utility indifference price**: limiting price with $U_X(\pi) = U_0$

- ▶ reasonable prices are those above indifference price

- Problem: utility indifference prices are hard to compute

- Alternative:

compute approximate solution for small number of options

Approximate utility indifference pricing

of over-the-counter derivatives

- Notation:

- ▶ $\pi(n)$: utility indifference price per unit of X for n options
- ▶ $\varphi(n)$: optimal strategy for n sold options, i.e. for $U_{nX}(n\pi(n))$

- Goal: approximate $\pi(n)$, $\varphi(n)$ for small n

- Expansion:

$$\hat{\pi}(n) = \hat{\pi}(0) + n\delta + o(n)$$

$$\varphi(n) = \varphi^* + n\eta + o(n)$$

- Interpretation:

- ▶ $\pi(0)$: limiting price for very small number of options
- ▶ δ : risk premium per option that is to be sold
- ▶ φ^* : optimal strategy for pure investment problem without options
- ▶ η : hedging strategy per option

- How to determine $\pi(0)$, δ , η ?

Approximate utility indifference pricing

for exponential utility

Rule

Consider $u(x) = 1 - e^{-\rho x}$.

Let φ^* , Q be optimal pair from dual characterisation.

Define $\hat{V}(t) := E_Q(\hat{X} | \mathcal{F}_t)$. Then

$$\hat{\pi}(0) = \hat{V}(0) = E_Q(\hat{X}),$$

$$\delta = \frac{\rho}{2} E_Q(\langle \hat{V}, \hat{V} \rangle^Q(T) - \eta^2 \cdot \langle \hat{S}_1, \hat{S}_1 \rangle^Q(T)),$$

$$\eta(t) = \frac{d\langle \hat{V}, \hat{S}_1 \rangle^Q(t)}{d\langle \hat{S}_1, \hat{S}_1 \rangle^Q(t)}.$$

- Q is the **minimal entropy martingale measure**.
- δ, η solve quadratic hedging problem for X under Q .
- Slightly more involved results hold for power/logarithmic utility with additional numeraire change.

Integral transform methods

for pricing and hedging in general

- Consider e.g. $X = (S_1(T) - K)^+$.

We have encountered objects like

- ▶ $\hat{V}(t) = E(\hat{X} | \mathcal{F}_t),$

- ▶ $\varphi(t) = \frac{d\langle \hat{V}, \hat{S}_1 \rangle(t)}{d\langle \hat{S}_1, \hat{S}_1 \rangle(t)},$

- ▶ $\varepsilon^2 = E\left(\langle \hat{V}, \hat{V} \rangle(T) - \varphi^2 \cdot \langle \hat{S}_1, \hat{S}_1 \rangle(T)\right).$

- How to compute them?

- ▶ typically no closed-form solution available

- ▶ often available: closed-form solutions for $\hat{X} = \hat{S}_1(T)^z, z \in \mathbb{C}.$

- ▶ e.g. $E(\hat{X}) = E(\hat{S}_1(T)^z) = e^{-zrT} E(\exp(zL(T))),$

i.e. extended characteristic function of $L(T)$ for $S_1 = S_1 e^L$

Integral transform methods

for pricing and hedging (ct'd)

- Idea: use integral representation

$$\hat{X} = \int_{R-i\infty}^{R+i\infty} \hat{S}_1(T)^z \ell(z) dz$$

with explicitly given $\ell(z)$, R .

- linearity yields

- ▶ $\hat{V}(t) = \int_{R-i\infty}^{R+i\infty} \hat{V}_z(t) \ell(z) dz$ with $\hat{V}_z(t) := E(\hat{S}_1(T)^z | \mathcal{F}_t)$,

- ▶ $\varphi(t) = \int_{R-i\infty}^{R+i\infty} \varphi_z(t) \ell(z) dz$ with $\varphi_z(t) := \frac{d(\hat{V}_z, \hat{S}_1)(t)}{d(\hat{S}_1, \hat{S}_1)(t)}$,

- ▶ $\varepsilon^2 = \int_{R-i\infty}^{R+i\infty} \int_{R-i\infty}^{R+i\infty} \varepsilon_{yz}^2 \ell(y) \ell(z) dy dz$ with

$$\varepsilon_{yz}^2 = E\left(\langle \hat{V}_y, \hat{V}_z \rangle(T) - \varphi_y \varphi_z \cdot \langle \hat{S}_1, \hat{S}_1 \rangle(T)\right).$$

- determine $\ell(z)$ via inverse Laplace transform

- ▶ for $\hat{X} = (\hat{S}_1(T) - K)^+$ we have $\ell(z) = \frac{K^{1-z}}{z(z-1)}$, $R > 1$

- ▶ for $\hat{X} = (K - \hat{S}_1(T))^+$ we have $\ell(z) = \frac{K^{1-z}}{z(z-1)}$, $R < 0$

Outline

- 1 Mathematical finance
- 2 Stochastic calculus**
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Basic concepts

of the general theory of stochastic processes

filtration $(\mathcal{F}_t)_{t \in \mathbb{R}_+}$

process $X = (X(t))_{t \in \mathbb{R}_+}$

càdlàg: right-continuous with left-hand limits $X_- = (X(t-))_{t \in \mathbb{R}_+}$

of finite variation: difference of two increasing processes

adaptedness: $X(t)$ \mathcal{F}_t -measurable for all t

predictability: slightly stronger than “ $X(t)$ \mathcal{F}_{t-} -measurable for all t ”

stopping time: random time τ with $\{\tau \leq t\} \in \mathcal{F}_t$

stopped process $X^\tau = (X(\tau \wedge t))_{t \in \mathbb{R}_+}$ with stopping time τ

martingale: $E(X(t) | \mathcal{F}_s) = X(s)$ for $s \leq t$

submartingale: $E(X(t) | \mathcal{F}_s) \geq X(s)$ for $s \leq t$

supermartingale: $E(X(t) | \mathcal{F}_s) \leq X(s)$ for $s \leq t$

density process: $Z(t) = E\left(\frac{dQ}{dP} \middle| \mathcal{F}_t\right) = \frac{dQ}{dP} \middle| \mathcal{F}_t$

localized classes \mathcal{C}_{loc} (local martingales etc.): $X^{\tau_n} \in \mathcal{C}$ for all n

special semimartingale: $X = X(0) + M^X + A^X$ with predictable A^X

compensator: finite variation process A^X in above decomposition

semimartingale: $X = X(0) + M + A$ with adapted A

Lévy processes

Definition

- processes of constant growth in stochastic sense
- random counterpart of linear functions
- continuous-time counterpart of random walks

Definition (Lévy process)

adapted càdlàg process X with $X(0) = 0$,
 $X(t) - X(s)$ independent of \mathcal{F}_s (independent increments),
 $\mathcal{L}(X(t) - X(s))$ depends only on $t - s$ (stationary increments)

- Examples:
 - ▶ linear function $X(t) = bt$
 - ▶ Brownian motion with drift $X(t) = \mu t + \sigma W(t)$
 - ▶ Poisson process
 - ▶ compound Poisson process $X(t) = \sum_{i=1}^{N(t)} Y_i$

Lévy processes

Characterization

Theorem (Lévy-Khintchine formula)

Characteristic function is $\varphi_{X(t)}(u) = E(e^{iu^\top X(t)}) = \exp(t\psi(u))$ with *characteristic exponent*

$$\psi(u) = iu^\top b^h - \frac{1}{2}u^\top c u + \int (e^{iu^\top x} - 1 - iu^\top h(x))K(dx)$$

and *Lévy-Khintchine triplet* (b^h, c, K) .

h denotes some *truncation function* as e.g. $h(x) = x1_{\{|x| \leq 1\}}$.

- Interpretation:

- ▶ **Drift coefficient** b stands for linear function.
- ▶ **Diffusion coefficient** c stands for Brownian motion.
- ▶ **Lévy measure** K stands for jumps.

- $b^{\tilde{h}} = b^h + \int (\tilde{h}(x) - h(x))K(dx)$

Lévy processes

Moments

- p -th moments exist if $K|_{\{|x|>1\}}$ has p -th moments.
- $E(X(t)) = b^h t = -i\psi'(0)t$ for $h(x) = x$ (i.e. $h = \text{id}$)
- $\text{Var}(X(t)) = (c + \int x^2 K(dx))t = -\psi''(0)t$ for dimension $d = 1$
- p -th exponential moments exist if $K|_{\{|x|>1\}}$ has p -th exponential moments.
- $E(e^{p^\top X(t)}) = \exp(t(p^\top b^h + \frac{1}{2}p^\top c p + \int (e^{p^\top x} - 1 - p^\top h(x))K(dx)))$, which equals $\exp(t\psi(-ip))$ if ψ has analytic extension.
- X is (local) martingale iff $E(X(1)) = 0$.
- e^X is (local) martingale iff $E(e^{X(1)}) = 1$.

Lévy processes

Path properties

- X has differentiable paths $\Leftrightarrow c = 0, K = 0$
 $\Leftrightarrow X$ is linear function.
- X has continuous paths $\Leftrightarrow K = 0$
 $\Leftrightarrow X$ is Brownian motion with drift.
- X is piecewise constant $\Leftrightarrow c = 0, K(\mathbb{R}) < \infty, b^0 = 0$
 $\Leftrightarrow X$ is compound Poisson process.
- X has finitely many jumps on $[0, t]$ $\Leftrightarrow K(\mathbb{R}) < \infty$
- X is of finite variation $\Leftrightarrow c = 0, \int_{|x| \leq 1} |x| K(dx) < \infty$

Lévy processes

Pathwise representation

Theorem (Lévy-Itô decomposition)

Let X have finite expectation and triplet (b^{id}, c, K) .

$$\begin{aligned} X(t) &= b^{id}t + \sqrt{c}W(t) \\ &\quad + \lim_{\varepsilon \rightarrow 0} \left(\sum_{s \leq t} \Delta X(s) 1_{\{|\Delta X(s)| > \varepsilon\}} - \int x 1_{\{|\Delta X(s)| > \varepsilon\}} K(dx) s \right) \\ &= b^{id}t + \sqrt{c}W(t) + \int_{[0, t] \times \mathbb{R}} x (\mu^X - \nu^X)(d(s, x)) \end{aligned}$$

with *random measure of jumps* of X

$$\mu^X([0, t] \times B) := \#\{(s, x) \in [0, t] \times B : \Delta X(s) = x \neq 0\}$$

and *compensated random measure of jumps* of X

$$\nu^X([0, t] \times B) := K(B)t.$$

Lévy processes in finance

for modelling of stock prices etc.

- Brownian motions with drift
 - ▶ most important and the only continuous Lévy processes
 - ▶ density, characteristic exponent, Lévy-Khintchine triplet known in closed form
 - ▶ 2 parameters
 - ▶ all exponential moments exist
- stable Lévy motions
 - ▶ the only self-similar Lévy processes
 - ▶ either Brownian motion or pure jump
 - ▶ characteristic exponent, Lévy-Khintchine triplet known in closed form
 - ▶ 4 parameters
 - ▶ infinite variance (except for Brownian motion)

Lévy processes in finance

for modelling of stock prices etc. (ct'd)

- Merton model

- ▶ Brownian motion + drift + Gaussian jumps
- ▶ finitely many jumps on $[0, t]$
- ▶ characteristic exponent, Lévy-Khintchine triplet known in closed form
- ▶ 5 parameters
- ▶ all exponential moments exist

- Kou model

- ▶ Brownian motion + drift + exponential jumps
- ▶ finitely many jumps on $[0, t]$
- ▶ characteristic exponent, Lévy-Khintchine triplet known in closed form
- ▶ 6 parameters
- ▶ some exponential moments exist

Lévy processes in finance

for modelling of stock prices etc. (ct'd)

- variance gamma (VG) processes
 - ▶ pure jump Lévy process
 - ▶ infinitely many jumps on $[0, t]$ but of finite variation
 - ▶ density, characteristic exponent, Lévy-Khintchine triplet known in closed form
 - ▶ 4 parameters
 - ▶ some exponential moments exist
- normal inverse Gaussian (NIG) processes
 - ▶ pure jump Lévy process
 - ▶ of infinite variation
 - ▶ density, characteristic exponent, Lévy-Khintchine triplet known in closed form
 - ▶ 4 parameters
 - ▶ some exponential moments exist
- etc.

Quadratic variation

and friends

- **quadratic variation** of semimartingale X :

$$[X, X](t) = \lim_{\sup |t_i - t_{i-1}| \rightarrow 0} \sum_{i \geq 1} (X(t_i \wedge t) - X(t_{i-1} \wedge t))^2$$

- ▶ **covariation** of X and Y :

$$[X, Y] = \frac{1}{4}([X + Y, X + Y] - [X - Y, X - Y])$$

- Example: X Lévy process with triplet (b^h, c, K)

- ▶ $[X, X]$ is Lévy process.
- ▶ Triplet of $[X, X]$ is $(c, 0, \tilde{K})$ relative to $\tilde{h} = 0$, where $\tilde{K}(B) = \int 1_B(x^2)K(dx)$.

- **predictable quadratic variation** of X :

$$\langle X, X \rangle = A^{[X, X]} \text{ (compensator of } [X, X])$$

- ▶ **predictable covariation** of X, Y : $\langle X, Y \rangle = A^{[X, Y]}$

- Example: X Lévy process with finite variance

- ▶ $\langle X, X \rangle$ is linear function.
- ▶ $\langle X, X \rangle(t) = \text{Var}(X(t))$

Continuous martingale part

of a semimartingale

- Local martingales X, Y are **orthogonal** if XY local martingale.
- Local martingale X is **purely discontinuous** if orthogonal to all continuous local martingales.
- X local martingale \Rightarrow
there is unique decomposition $X = X(0) + X^c + X^d$ with
 X^c continuous, X^d purely discontinuous, $X^c(0) = 0 = X^d(0)$.
- $X = X(0) + M + A$ semimartingale \Rightarrow
continuous martingale part $X^c := M^c$ does not depend on
decomposition of X
- $[X, Y](t) = \langle X^c, Y^c \rangle(t) + \sum_{s \leq t} \Delta X_s \Delta Y_s$
- X Lévy process with Lévy-Itô decomposition
$$X(t) = b^{\text{id}}t + \sqrt{c}W(t) + \int_{[0,t] \times \mathbb{R}} x(\mu^X - \nu^X)(d(s, x))$$
$$\Rightarrow X^c = \sqrt{c}W$$

Stochastic integral

Heuristics

- Informal “definition”:

- ▶ φ predictable process

- ▶ X semimartingale

- ▶ $\varphi \bullet X(t) := \lim_{\sup |t_i - t_{i-1}| \rightarrow 0} \sum_{i \geq 1} \varphi(t_{i-1})(X(t_i \wedge t) - X(t_{i-1} \wedge t))$

- ▶ alternative notation: $\int_0^t \varphi(s) dX(s) = \varphi \bullet X(t)$

- ▶ $\varphi \bullet X(t) := \sum_{i=1}^d \varphi_i(t) \bullet X_i(t)$ for \mathbb{R}^d -valued φ, X

- Interpretation:

- ▶ φ trading strategy

- ▶ X stock price process

- ▶ $\varphi \bullet X(t) = \int_0^t \varphi(s) dX(s)$ gains due to price changes

Stochastic integral

Properties

- $\varphi \cdot X$ is linear in φ and X
- $\varphi \cdot (\psi \cdot X) = (\varphi\psi) \cdot X$
- $[\varphi \cdot X, Y] = \varphi \cdot [X, Y]$
- $\langle \varphi \cdot X, Y \rangle = \varphi \cdot \langle X, Y \rangle$
- $\Delta(\varphi \cdot X) = \varphi \cdot \Delta X$
- $(\varphi \cdot X)^\tau = \varphi \cdot X^\tau$
- $(\varphi \cdot X)^c = \varphi \cdot X^c$
- X sigma-martingale $\Rightarrow \varphi \cdot X$ sigma-martingale
(slightly more general than local martingale)
- integration by parts:
 $XY = X(0)Y(0) + X_- \cdot Y + Y_- \cdot X + [X, Y]$

Stochastic integral

Properties (ct'd)

Theorem (Itô's formula)

Let X semimartingale in \mathbb{R}^d and $f : \mathbb{R}^d \rightarrow \mathbb{R}$ of class C^2 .

$$\begin{aligned} f(X(t)) &= f(X(0)) + Df(X_-) \cdot X(t) + \frac{1}{2} \sum_{i,j=1}^d D_{ij}f(X_-) \cdot \langle X_i^c, X_j^c \rangle(t) \\ &\quad + \sum_{s \leq t} \left(f(X(s)) - f(X(s-)) - Df(X(s-))^\top \Delta X(s) \right) \end{aligned}$$

For $d = 1$:

$$\begin{aligned} f(X(t)) &= f(X(0)) + f'(X_-) \cdot X(t) + \frac{1}{2} f''(X_-) \cdot \langle X^c, X^c \rangle(t) \\ &\quad + \sum_{s \leq t} \left(f(X(s)) - f(X(s-)) - f'(X(s-)) \Delta X(s) \right). \end{aligned}$$

Random measures

and their integration

- recall **random measure of jumps** of semimartingale X :
 $\mu^X([0, t] \times B) := \#\{(s, x) \in [0, t] \times B : \Delta X(s) = x \neq 0\}$
- **random measure** here means: measure on $\mathbb{R}_+ \times \mathbb{R}^d$ for fixed ω
- **integral process** $\xi * \mu$
 - ▶ $\xi : \Omega \times \mathbb{R}_+ \times \mathbb{R}^d \rightarrow \mathbb{R}$ in some sense **predictable** function
 - ▶ μ random measure
 - ▶ $\xi * \mu(t) := \int_{[0, t] \times \mathbb{R}^d} \xi(s, x) \mu(ds, dx)$ pathwise integral
 - ▶ For $\mu = \mu^X$ we have $\xi * \mu^X(t) = \sum_{s \leq t} \xi(s, \Delta X(s)) 1_{\{\Delta X(s) \neq 0\}}$.
- **compensator** ν^X of μ^X :
the in some sense **predictable** random measure such that
 $\xi * \mu - \xi * \nu$ local martingale for most predictable ξ
- **stochastic integral** $\xi * (\mu^X - \nu^X)$:
extension of $\xi * \mu^X - \xi * \nu^X$ to some ξ where this is not defined

Random measures

Properties

- X Lévy process with triplet $(b^h, c, K) \Rightarrow \nu^X(d(t, x)) = K(dx)dt$
- $\xi * \mu$ and $\xi * (\mu^X - \nu^X)$ are linear in ξ .
- $\varphi \bullet (\xi * \mu) = (\varphi \psi) * \mu$
- $\varphi \bullet (\xi * (\mu^X - \nu^X)) = (\varphi \psi) * (\mu^X - \nu^X)$
- $\xi * (\mu^X - \nu^X)$ is local martingale,
typically (but not always) with jumps
$$\Delta(\xi * (\mu^X - \nu^X))(t) = \xi(t, \Delta X(t)) \mathbf{1}_{\{\Delta X(t) \neq 0\}}$$
- $\langle \eta * (\mu^X - \nu^X), \xi * (\mu^X - \nu^X) \rangle = (\eta \xi) * \nu^X$
- Itô's formula:

$$\begin{aligned} f(X(t)) &= f(X(0)) + f'(X_-) \bullet X(t) + \frac{1}{2} f''(X_-) \bullet \langle X^c, X^c \rangle(t) \\ &\quad + (f(X_- + x) - f(X_-) - f'(X_-)x) * \mu^X(t) \end{aligned}$$

- X special semimartingale:

$$X = X(0) + A^X + X^c + x * (\mu^X - \nu^X)$$

generalizes Lévy-Itô decomposition

Stochastic exponential

as solution to linear stochastic differential equation

- The solution $\mathcal{E}(X)$ to $Z = 1 + Z_- \cdot X$ is called **Doléans exponential** or **stochastic exponential** of X .
- Solution for $\Delta X > -1$:

$$\begin{aligned} \mathcal{E}(X) = \exp & \left(X(t) - X(0) - \frac{1}{2} \langle X^c, X^c \rangle(t) \right. \\ & \left. + \sum_{s \leq t} (\log(1 + \Delta X(s)) - \Delta X(s)) \right) \end{aligned}$$

- **Yor's formula:**

$$\mathcal{E}(X)\mathcal{E}(Y) = \mathcal{E}(X + Y + [X, Y])$$

Geometric Lévy processes

modelling constant relative growth

- Idea: constant *relative* rather than absolute growth
 - ▶ more appropriate e.g. for stock prices
- **geometric Lévy process**:
 - ▶ $\mathbb{R} \setminus \{0\}$ -valued semimartingale with $X(0) = 1$
 - ▶ $\frac{X(t)}{X(s)}$ independent of \mathcal{F}_s
 - ▶ $\mathcal{L}\left(\frac{X(t)}{X(s)}\right)$ depends only on $t - s$
- equivalence for positive X :
 - ▶ X is a geometric Lévy process.
 - ▶ $X = e^L$ for some Lévy process L .
 - ▶ $X = \mathcal{E}(\tilde{L})$ for some Lévy process \tilde{L} .
- moreover:
 - ▶ can relate triplets of L and \tilde{L}
 - ▶ multivariate analogue exists

Semimartingale characteristics

as local Lévy-Khintchine triplets

- local dynamics of functions and processes

	deterministic functions	stochastic processes
constant growth	linear function $X(t) = bt$ slope b	Lévy process $X(t)$ triplet (b^h, c, K)
general case	function $X(t)$ local slope $b(t) = \frac{dX(t)}{dt}$	semimartingale $X(t)$ local triplet $(b^h, c, K)(\omega, t)?$

- Goal: definition of local triplet $(b^h, c, K)(\omega, t)$
- Interpretation:
semimartingale X resembles locally around t
a Lévy process with triplet $(b^h, c, K)(\omega, t)$.

Semimartingale characteristics

Definition

- Lemma: X Lévy process with triplet $(b^h, c, K) \Rightarrow$
 $M(t) := e^{iu^\top X(t)} - \int_0^t e^{iu^\top X(s-)} \psi(u) ds$ is local martingale for any
 $u \in \mathbb{R}^d$, where $\psi(u)$ characteristic exponent of (b^h, c, K) .

Definition (differential characteristics)

A **predictable** triplet $(b^h, c, K)(\omega, t)$ is called **differential characteristics** or **local triplet** of semimartingale X if
 $M(t) := e^{iu^\top X(t)} - \int_0^t e^{iu^\top X(s-)} \psi(s, u) ds$ is a local martingale for any u ,
where $\psi(t, u) := iu^\top b^h(t) - \frac{1}{2}u^\top c(t)u + \int (e^{iu^\top x} - 1 - iu^\top h(x))K(t, dx)$
denotes the characteristic exponent of $(b^h, c, K)(\omega, t)$.

- dependence on truncation function:
 $b^{\tilde{h}}(t) = b^h(t) + \int (\tilde{h}(x) - h(x))K(t, dx)$

Semimartingale characteristics

Alternative characterization

- X semimartingale with local triplet (b^h, c, K) . Define
 - ▶ $X^h := X - (x - h(x)) * \mu^X$
 - ▶ A^{X^h} compensator of X^h
(i.e. $X^h - A^{X^h}$ local martingale)
 - ▶ ν^X compensator of measure of jumps of X
- Then

$$A^{X^h}(t) = \int_0^t b^h(s) ds,$$

$$\langle X_i^c, X_j^c \rangle = \int_0^t c_{ij}(s) ds,$$

$$\nu^X([0, t] \times dx) = \int_0^t K(s, dx) ds.$$

Semimartingale characteristics

Rules

Lemma (Lévy processes)

X is a Lévy process \Leftrightarrow local triplet of X does not depend on (ω, t) .

Lemma (stochastic integrals)

Let (b^h, c, K) be the local triplet of X . Then $\varphi \bullet X$ has local triplet

$$\tilde{b}^h(t) = \varphi(t)^\top b^h(t) + \int \left(\tilde{h}(\varphi(t)x) - \varphi(t)h(x) \right) K(t, dx),$$

$$\tilde{c}(t) = \varphi(t)c(t)\varphi(t)^\top,$$

$$\tilde{K}(t, B) = \int \mathbf{1}_B(\varphi(t)x)K(t, dx).$$

Semimartingale characteristics

Rules (ct'd)

Lemma (C^2 -functions)

Let (b^h, c, K) be the local triplet of X and $f : \mathbb{R}^d \rightarrow \mathbb{R}^n$ of class C^2 . Then $f(X)$ has local triplet

$$\begin{aligned} \tilde{b}_i^h(t) &= Df_i(X(t-))^\top b^h(t) + \frac{1}{2} \sum_{k,l=1}^d D_{kl} f_i(X(t-)) c_{kl}(t) \\ &+ \int \left(\tilde{h}_i(f(X(t-) + x) - f(X(t-))) - Df_i(X(t-))^\top h(x) \right) F(t, dx), \end{aligned}$$

$$\tilde{c}_{ij}(t) = \sum_{k,l=1}^d D_k f_i(X(t-)) c_{kl}(t) D_l f_j(X(t-)),$$

$$\tilde{K}_t(B) = \int 1_B(f(X(t-) + x) - f(X(t-))) K_t(dx).$$

Semimartingale characteristics

Rules (ct'd)

Lemma (change of measure)

Let (b^h, c, K) be the local triplet of X . Suppose that $Q \sim P$ has density process $\mathcal{E}(\varphi \bullet X^c + \psi * (\mu^X - \nu^X))$. Then X has Q -local triplet

$$\tilde{b}^h(t) = b^h(t) + \varphi(t)^\top c(t) + \int h(x) \psi(t, x) K(t, dx),$$

$$\tilde{c}(t) = c(t),$$

$$\tilde{K}(t)(B) = \int 1_B(x) (1 + \psi(t, x)) K(t)(dx).$$

Lemma (predictable covariation)

Let (b^h, c, K) be the local triplet of X . Then

$$d\langle X_i, X_j \rangle(t) = (c_{ij}(t) + \int x_i x_j K(t, dx)) dt.$$

Outline

- 1 Mathematical finance
- 2 Stochastic calculus
- 3 Applications to finance**

Option pricing in geometric Lévy model

Integral transform method

- Model:

- ▶ $S_0(t) = e^{rt}$, $S_1(t) = S_1(0)e^{L(t)}$ with Lévy process L
- ▶ ψ characteristic exponent of L under pricing measure Q
- ▶ \hat{S}_1 Q -martingale (i.e. $\psi(-i) = r$)
- ▶ recall that we need $\hat{V}_z(t)$

- Solution:

$$\hat{V}_z(t) = \hat{S}_1(t)^z \exp((T-t)(\psi(-iz) - rz))$$

- Price of European call at t :

$$S_1(t) \int_{R-i\infty}^{R+i\infty} \left(\frac{K}{S_1(t)} \right)^{1-z} \frac{e^{(T-t)(\psi(-iz)-r)}}{2\pi iz(z-1)} dz$$

Utility maximization in geometric Lévy model

Power and logarithmic utility

- Model:

- ▶ $S_0(t) = e^{rt}$, $S_1(t) = S_1(0)e^{L(t)}$ with Lévy process L
- ▶ initial endowment v
- ▶ utility function $u(x) = \frac{x^{1-p}}{1-p}$ or $u(x) = \log(x)$
- ▶ maximize $E(u(V_\varphi(T)))$ over admissible φ

- Solution:

- ▶ $\varphi^*(t) = \gamma \frac{V_{\varphi^*}(t-)}{S_1(t-)}$ with $\gamma \in \mathbb{R}$
- ▶ $\hat{V}_{\varphi^*}(t) = v e^{\mathcal{O}(\gamma(\hat{S}_1)_-^{-1} \cdot \hat{S}_1)}$
- ▶ Density process of EMM Q from dual characterization is
$$Z(t) = e^{-\alpha(T-t)} \frac{u'(\hat{V}_{\varphi^*}(t))}{E(u'(\hat{V}_{\varphi^*}(t)))}$$
 with $\alpha \in \mathbb{R}$.
- ▶ γ is determined as root from an equation.
- ▶ α is known in terms of γ .

- Properties of the solution:

- ▶ Fixed fraction of current wealth is invested in the stock.
- ▶ Value of the optimal strategy is geometric Lévy process.
- ▶ Density process of Q is geometric Lévy process.
- ▶ S_1 is geometric Lévy process under Q as well.

Utility maximization in geometric Lévy model

Exponential utility

- Model:

- ▶ $S_0(t) = e^{rt}$, $S_1(t) = S_1(0)e^{L(t)}$ with Lévy process L
- ▶ initial endowment v
- ▶ utility function $u(x) = 1 - e^{-\rho x}$
- ▶ maximize $E(u(\hat{V}_\varphi(T)))$ over admissible φ

- Solution:

- ▶ $\varphi^*(t) = \frac{\gamma}{\hat{S}_1(t-)}$ with $\gamma \in \mathbb{R}$
- ▶ $\hat{V}_{\varphi^*}(t) = v + \gamma(\hat{S}_1)^{-1} \cdot \hat{S}_1$
- ▶ Density process of EMM Q from dual characterization is
$$Z(t) = e^{-\alpha(T-t)} \frac{u'(\hat{V}_{\varphi^*}(t))}{E(u'(\hat{V}_{\varphi^*}(t)))}$$
 with $\alpha \in \mathbb{R}$.
- ▶ γ is determined as root from an equation.
- ▶ α is known in terms of γ .

- Properties of the solution:

- ▶ Fixed discounted amount of money is invested in the stock.
- ▶ Value of the optimal strategy is Lévy process + v .
- ▶ Density process of Q is geometric Lévy process.
- ▶ S_1 is geometric Lévy process under Q as well.

Quadratic hedging in geometric Lévy model

Martingale case

- Model:

- ▶ $S_0(t) = e^{rt}$, $S_1(t) = S_1(0)e^{L(t)}$ with Lévy process L
- ▶ ψ characteristic exponent of L
- ▶ \hat{S}_1 martingale (i.e. $\psi(-i) = r$)
- ▶ recall that we need $\hat{V}_z(t)$, $\varphi_z(t)$, ε_{yz}^2

- Solution:

- ▶ $\hat{V}_z(t) = \hat{S}_1(t)^z \exp((T-t)(\psi(-iz) - rz))$
- ▶ $\varphi_z(t) = \frac{\hat{V}_z(t-) \psi(-i(z+1)) - \psi(-iz) - r}{\hat{S}_1(t-) \psi(-2i) - 2r}$
- ▶

$$\begin{aligned} \varepsilon_{yz}^2 &= S_1(0)^{y+z} (e^{\kappa(y+z)T} - e^{(\kappa(y)+\kappa(z))T}) \\ &\quad \times \left(1 - \frac{(\kappa(y+1) - \kappa(y))(\kappa(z+1) - \kappa(z))}{\kappa(2)(\kappa(y+z) - \kappa(y) - \kappa(z))} \right) \end{aligned}$$

with $\kappa(z) := \psi(-iz) - rz$

- can now apply results to approximate indifference pricing