

Optimal control of SDEs associated with general Lévy generators

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Based on joint works with Jonathan Bennett

[1] Explicit construction of SDEs associated with polar-decomposed Lévy measures and application to stochastic optimization, *Frontiers of Mathematics in China* **2** (2007), 539–558.

[2] An optimal control problem associated with SDEs driven by Lévy-type processes, *Stochastic Analysis and Applications*, **26** (2008), 471–494.

[3] Stochastic control of SDEs associated with Lévy generators and application to financial optimization, *Frontiers of Mathematics in China* **5** (2010), 89–102.

and [4] JLW, Harry Zheng, On an optimal portfolio-consumption problem associated with Lévy-type generators, in progress.

A fairly large class of Markov processes on \mathbb{R}^d are governed by Lévy generator, either via martingale problem (cf e.g. D W Stroock, “Markov Processes from K. Itô’s Perspectives”, Princeton Univ Press 2003 or V.N. Kolokoltsov, “Markov Processes, Semigroups and Generators”, de Gruyter, 2011) or via Dirichlet form (cf e.g. N Jacob, “Pseudo-Differential Operators and Markov Processes III” Imperial College Press, 2005)

$$\begin{aligned}
 Lf(t, x) &:= \frac{1}{2} a^{i,j}(t, x) \partial_i \partial_j f(t, x) + b^i(t, x) \partial_i f(t, x) \\
 &\quad + \int_{\mathbb{R}^d \setminus \{0\}} \{f(t, x + z) - f(t, x) \\
 &\quad - \frac{z^1 \mathbf{1}_{\{|z| < 1\}} \cdot \nabla f(t, x)}{1 + |z|^2}\} \nu(t, x, dz)
 \end{aligned}$$

where $a(t, x) = (a^{i,j}(t, x))_{d \times d}$ is non-negative definite symmetric and $\nu(t, x, dz)$ is a Lévy kernel, i.e.,

$\forall (t, x) \in [0, \infty) \times \mathbb{R}^d$, $\nu(t, x, \cdot)$ is a σ -finite measure on $(\mathbb{R}^d \setminus \{0\}, \mathcal{B}(\mathbb{R}^d \setminus \{0\}))$ such that

$$\int_{\mathbb{R}^d \setminus \{0\}} \frac{|z|^2}{1 + |z|^2} \nu(t, x, dz) < \infty.$$

For such L , in order to get rid of variable dependence on ν , N El Karoui and J P Lepeltier (Z. Wahr. verw. Geb. 39 (1977)) construct a bimeasurable bijection

$$c : [0, \infty) \times \mathbb{R}^d \times U \rightarrow \mathbb{R}^d \setminus \{0\}$$

such that

$$\int_U 1_A(c(t, x, y)) \lambda(dy) = \int_{\mathbb{R}^d \setminus \{0\}} 1_A(z) \nu(t, x, dz), \quad \forall (t, x)$$

for $A \in \mathcal{B}(\mathbb{R}^d \setminus \{0\})$. Where $(U, \mathcal{B}(U))$ is a Lusin space and λ is a σ -finite measure on it. Actually, we can construct c explicitly in case ν has a polar decomposition (with the stable-like case

as a concrete example). It is well-known (cf e.g. Theorem I.8.1 in N Ikeda and S Watanabe's book): \exists a Poisson random measure

$$N : \mathcal{B}([0, \infty)) \times \mathcal{B}(U) \times \Omega \rightarrow \mathbb{N} \cup \{0\} \cap \{\infty\}$$

on any given probability set-up $(\Omega, \mathcal{F}, P; \{\mathcal{F}_t\}_{t \geq 0})$ with $\mathbf{E}(N(dt, dy, \cdot)) = dt\lambda(dy)$, and

$$\tilde{N}(dt, dy, \omega) := N(dt, dy, \omega) - dt\lambda(dy)$$

being the associated compensating $\{\mathcal{F}_t\}_{t \geq 0}$ -martingale measure.

We then can formulate a jump SDE associated with L

$$dS_t = b(t, S_t)dt + \sigma(t, S_t)dW_t + \int_U c(t, S_{t-}, y)\tilde{N}(dt, dy)$$

where $\sigma(t, x)$ is a $d \times m$ -matrix such that

$$\sigma(t, x)\sigma^T(t, x) = a(t, x)$$

and $\{W_t\}_{t \in [0, \infty)}$ is an m -dimensional $\{\mathcal{F}_t\}_{t \geq 0}$ -Brownian motion. We shall consider such equation in the following general formulation

$$\begin{aligned} dS_t &= b(t, S_t)dt + \sigma(t, S_t)dW_t \\ &\quad + \int_{U \setminus U_0} c_1(t, S_{t-}, z) \tilde{N}(dt, dz) \\ &\quad + \int_{U_0} c_2(t, S_{t-}, z) N(dt, dz) \end{aligned}$$

where $U_0 \in \mathcal{B}(U)$ with $\lambda(U_0) < \infty$ is arbitrarily fixed.

Sufficient Maximum Principle

Framstad, Øksendal, Sulem (J Optim Theory Appl 121 (2004))

Øksendal, Sulem (“Applied Stochastic Control of Jump-Diffusions”, Springer, 2005); Math Finance 19 (2009); SIAM J Control Optim 2010; Commun Stoch Anal 4 (2010)

Start with a controlled jump Markov process

$$S_t = S_t^{(u)}, \quad t \in [0, T]$$

for any arbitrarily fixed $T \in (0, \infty)$, by the following

$$\begin{aligned} dS_t &= b(t, S_t, u_t)dt + \sigma(t, S_t, u_t)dW_t \\ &\quad + \int_{U \setminus U_0} c_1(t, S_{t-}, u_{t-}, z) \tilde{N}(dt, dz) \\ &\quad + \int_{U_0} c_2(t, S_{t-}, u_{t-}, z) N(dt, dz) \end{aligned} \tag{1}$$

where the control process $u_t = u(t, \omega)$, taking values in a given Borel set $U \in \mathcal{B}(\mathbb{R}^d)$, is assumed to be $\{\mathcal{F}_t\}$ -predictable and càdlàg.

The performance criterion is

$$J(u) := \mathbf{E} \left(\int_0^T f(t, S_t, u_t) dt + g(S_T) \right), \quad u \in \mathcal{A}$$

for \mathcal{A} the totality of all admissible controls, and for

$$f : [0, T] \times \mathbb{R}^d \times \mathcal{U} \rightarrow \mathbb{R}$$

being continuous, and for $g : \mathbb{R}^d \rightarrow \mathbb{R}$ being concave. The objective is to achieve the following

$$J(\hat{u}) = \sup_{u \in \mathcal{A}} J(u)$$

referring \hat{u} being the optimal control of the system.

Moreover, if $\hat{S}_t = S_t^{(\hat{u})}$ is the solution to the jump type SDE (1) corresponding to \hat{u} , then the pair (\hat{S}, \hat{u}) is called *the optimal pair*.

Now the Hamiltonian is defined

$$H : [0, T] \times \mathbb{R}^d \times \mathcal{U} \times \mathbb{R}^d \times \mathbb{R}^{d \otimes m} \times \mathcal{R} \rightarrow \mathbb{R}$$

via

$$\begin{aligned} & H(t, r, u, p, q, n^{(1)}, n^{(2)}) \\ = & f(t, r, u) + \mu(t, r, u)p + \frac{1}{2}\sigma^T(t, r, u)q \\ & + \int_{U \setminus U_0} n^{(1)}(t, z)c_1(t, r, u, z)\lambda(dz) \\ & + \int_{U_0} [n^{(2)}(t, z)c_2(t, r, u, z) + c_2(t, r, u, z)p]\lambda(dz) \end{aligned}$$

where \mathcal{R} is the collection of all $\mathbb{R}^{d \otimes d}$ -valued processes $n : [0, \infty) \times \Omega \rightarrow \mathbb{R}^{d \otimes d}$ such that the two integrals in the above formulation converge absolutely.

It is known that the adjoint equation corresponding to an admissible pair (S, u) is the BSDE

$$\begin{aligned} dp(t) = & -\nabla_r H(t, S_t, u_t, p(t), q(t), n^{(1)}(t, \cdot), n^{(2)}(t, \cdot))dt \\ & + q(t)dW_t + \int_{U \setminus U_0} n^{(1)}(t-, z)\tilde{N}(dt, dz) \\ & + \int_{U_0} n^{(2)}(t-, z)N(dt, dz) \end{aligned}$$

with terminal condition

$$p(T) = \nabla g(S_T).$$

Theorem ([3])

Given an admissible pair (\hat{S}, \hat{u}) . Suppose \exists an $\{\mathcal{F}_t\}$ -adapted solution $(\hat{p}(t), \hat{q}(t), \hat{n}(t, z))$ to the BSDE s.t. for $u \in \mathcal{A}$

$$\begin{aligned} & \mathbf{E} \left[\int_0^T (\hat{S}_t - S_t^{(u)})^T \{ \hat{q}(t) \hat{q}(t)^T \right. \\ & \quad \left. + \int_{U_0} [\text{tr}(\hat{n}(t, z) \hat{n}(t, z)^T) \lambda(dz)] \right] \\ & \quad \times (\hat{S}_t - S^{(u)}(t)) dt \Big] < \infty, \end{aligned}$$

$$\begin{aligned} & \mathbf{E} \left[\int_0^T \hat{p}^T(t) \left\{ \int_{U_0} [\text{tr}(c(t, S_{t-}, u_t, z) c^T(t, S_{t-}, u_t, z)) \lambda(dz)] \right. \right. \\ & \quad \left. \left. + \sigma(t, S_t, u_t) \sigma^T(t, S_t, u_t) \right\} \hat{p}(t) dt \right] < \infty, \end{aligned}$$

Theorem (cont'd)

and $\forall t \in [0, T]$

$$H(t, \hat{S}_t, \hat{u}_t, \hat{p}(t), \hat{q}(t), \hat{h}(t, \cdot)) = \sup_{u \in \mathcal{A}} H(t, \hat{S}_t, u, \hat{p}(t), \hat{q}(t), \hat{h}(t, \cdot)). \quad (2)$$

If $\hat{H}(r) := \max_{u \in \mathcal{A}} H(t, r, u, \hat{p}(t), \hat{q}(t), \hat{h}(t, \cdot))$ exists and is a concave function of r , then (\hat{S}, \hat{u}) is an optimal pair.

Remark For (2), it suffices that the function

$$(r, u) \rightarrow H(t, r, u, \hat{p}(t), \hat{q}(t), \hat{h}(t, \cdot))$$

is concave, $\forall t \in [0, T]$.

Optimal control problem

Benth, Karlsen, Reikvam (Finance Stoch 5 (2001); Stochastics
Stochastics Rep 74(2002))

Ishikawa (Appl Math Optim 50 (2004))

Jakobsen, Karlsen (JDE 212 (2005); NoDEA 13 (2006))

Start with a Lévy type process

$$\begin{aligned} Z_t = & \mu t + \int_0^t \theta(s) dW_s + \int_0^t \int_{U \setminus U_0} c_1(z) \tilde{N}(ds, dz) \\ & + \int_0^t \int_{U_0} c_2(z) N(ds, dz) \end{aligned}$$

where μ is a constant, $\theta : [0, T] \rightarrow \mathbb{R}$ and $c_1, c_2 : U \rightarrow \mathbb{R}$ are measurable. Here assume that

$$\int_{U_0} (e^{c_2(z)} - 1) \lambda(dz) < \infty.$$

We are concerned with the following 1-dimensional linear SDE

$$\begin{aligned}dS_t &= b(t)S_t dt + \frac{1}{2}\sigma(t)^2 S_t dt + \sigma(t)S_t dW_t \\ &+ S_t \int_U (e^{c_1(z)} - 1 - c_1(z)\mathbf{1}_{\{U \setminus U_0\}}(z))\lambda(dz) dt \\ &+ S_{t-} \int_U (e^{c_1(z)} - 1)\tilde{N}(dt, dz).\end{aligned}$$

Based on the driving processes Z_t and S_t , we construct two processes X_t and Y_t with $X_0 = x$, $Y_0 = y$, via

$$\begin{aligned}
X_t &= x - G_t + \int_0^t \sigma(s) \pi_s X_s dW_s + L_t \\
&+ \int_0^t (r + ([b(s) + \frac{1}{2} \sigma(s)^2 + \int_{U \setminus U_0} (e^{c_1(z)} \\
&\quad - 1 - c_1(z)) \lambda(dz)] - r) \pi_s) X_s ds \\
&+ \int_0^t \pi_{s-} X_{s-} \int_{U \setminus U_0} (e^{c_1(z)} - 1) \tilde{N}(ds, dz) \\
&+ \int_0^t \pi_{s-} X_{s-} \int_{U_0} (e^{c_2(z)} - 1) N(ds, dz)
\end{aligned}$$

and

$$Y_t = ye^{-\beta t} + \beta \int_0^t e^{-\beta(t-s)} dG_s$$

respectively, where

$$G_t := \int_0^t g_s ds$$

with $(g_t)_{t \geq 0}$ being a nondecreasing $\{\mathcal{F}_t\}$ -adapted càdlàg process of finite variation such that $0 \leq \sup_{t \geq 0} g_t < \infty$, L_t is a nondecreasing, nonnegative, and $\{\mathcal{F}_t\}$ -adapted càdlàg process, and $\pi_t \in [0, 1]$ is $\{\mathcal{F}_t\}$ -adapted càdlàg. The triple (G_t, L_t, π_t) is referred as the parameter process.

Remark The background for X_t being the self-financing investment policy according to the portfolio π_t :

$$\frac{dX_t}{X_{t-}} = (1 - \pi_t) \frac{dB_t}{B_t} + \pi_t \frac{dS_t}{S_{t-}}$$

with B_t standing for the riskless bond $dB_t = rB_t dt$.

By Itô formula, the generator A to (X_t, Y_t) is

$$\begin{aligned}
 Av(x, y) = & -\alpha v - \beta y v_y + \sigma(t) \pi x v_{xx} \\
 & + \left\{ (r + \pi([b(t) + \frac{1}{2} \sigma(t)^2 \right. \\
 & + \int_{U \setminus U_0} (e^{c_1(z)} - 1 - c_1(z)) \lambda(dz)] - r)) x v_x \\
 & + \int_{U \setminus U_0} (v(x + \pi x (e^{c_1(z)} - 1), y) \\
 & \quad - v(x, y) - \pi x v_x (e^{c_1(z)} - 1)) \lambda(dz) \\
 & + \left. \int_{U_0} (v(x + \pi x (e^{c_2(z)} - 1), y) - v(x, y)) \lambda(dz) \right\} \\
 & + u(g) - g(v_x - \beta v_y)
 \end{aligned}$$

for any $v \in C^{2,2}(\mathbb{R} \times \mathbb{R})$ and for $\pi \in [0, 1]$, $g \in [0, M_1]$.

Now we define the value function

$$v(x, y) := \sup_{(\pi, g, L) \in \mathcal{A}} \mathbf{E}^{(X(\pi, g, L), Y(\pi, g, L))} \left[\int_0^\infty e^{-\alpha s} u(g_s) ds \right]$$

where the supremum is taken over all admissible controls and u is a utility function, i.e., u is strictly increasing, differential, and concave on $[0, \infty)$ such that

$$u(0) = u'(\infty) = 0, \quad u(\infty) = u'(0) = \infty.$$

We also denote that

$$\begin{aligned} k(\gamma, \rho) := & \max_{\pi} \left\{ \gamma(r + \pi([b(t) + \frac{1}{2}\sigma(t)^2 \right. \\ & + \int_{U \setminus U_0} (e^{c_1(z)} - 1 - c_1(z))\lambda(dz)] - r)) \\ & + \sigma(t)\pi\rho + \int_{U \setminus U_0} [(1 + \pi(e^{c_1(z)} - 1))^{\gamma} \\ & \quad \left. - 1 - \gamma\pi(e^{c_1(z)} - 1)) \right. \\ & \left. + \int_{U_0} (1 + \pi(e^{c_2(z)} - 1))^{\gamma} - 1] \lambda(dz) \right\}. \end{aligned}$$

Theorem ([2])

[i](Existence) v is well-defined, i.e., there exists an optimal control $(\pi^*, g^*, L^*) \in \mathcal{A}$ such that

$$v(x, y) = \mathbf{E}^{(X(\pi^*, g^*, L^*), Y(\pi^*, g^*, L^*))} \left[\int_0^\infty e^{-\alpha s} u(g_s^*) ds \right].$$

Furthermore, v is a constrained viscosity solution to the following Hamilton-Jacobi-Bellman integro-variational inequality

$$\max \left\{ v_x 1_{\{x \leq 0\}}, \sup_{(\pi, g) \in \mathcal{A}} \{A v\}, (\beta v_y - v_x) 1_{\{x \geq 0\}} \right\} = 0$$

in $D_\beta := \{(x, y) : y > 0, y + \beta x > 0\}$, and

$$v = 0 \quad \text{outside of } D_\beta.$$

Theorem (cont'd)

[ii] (Uniqueness) For $\gamma > 0$ and each $\rho \geq 0$ choose $\alpha > 0$ s.t. $k(\gamma, \rho) < \alpha$. Then the HJB integro-variational inequality admits at most one constrained viscosity solution.

Theorem (JLW and H. Zheng [4])

Assume the following dynamic programming principle hold for the value function v : $\forall t \geq 0$ and for any stopping time τ

$$v(x, y) = \sup_{(\pi, g, L) \in \mathcal{A}} \mathbf{E} \left[\int_0^{t \wedge \tau} e^{-\alpha s} u(g_s^*) ds + e^{-\alpha(t \wedge \tau)} v(X_{t \wedge \tau}^{(\pi, g, L)}, Y_{t \wedge \tau}^{(\pi, g, L)}) \right].$$

Then, v is the unique, constrained (subject to a gradient constraint) viscosity solution of the following integro-differential HJB equation

$$\max \left\{ \beta v_y - v_x, \sup_{(\pi, g) \in \mathcal{A}} \{Av\} \right\} = 0$$

in $D := \{(x, y) : x > 0, y > 0\}$.

Further discussion on the properties of v is under way. 

The case of polar-decomposed Lévy measures

Recall the Lévy generator

$$\begin{aligned} Lf(t, x) &:= \frac{1}{2} a^{i,j}(t, x) \partial_i \partial_j f(t, x) + b^i(t, x) \partial_i f(t, x) \\ &\quad + \int_{\mathbb{R}^d \setminus \{0\}} \left\{ f(t, x + z) - f(t, x) \right. \\ &\quad \left. - \frac{z \mathbf{1}_{\{|z| < 1\}} \cdot \nabla f(t, x)}{1 + |z|^2} \right\} \nu(t, x, dz) \end{aligned}$$

and the associated SDE

$$dS_t = b(t, S_t) dt + \sigma(t, S_t) dW_t + \int_U c(t, S_{t-}, y) \tilde{N}(dt, dy)$$

Here we consider a special case: ν admits a polar-decomposition.

$$(U, \mathcal{B}(U), \lambda) = (\mathbb{S}^{d-1} \times (0, \infty), \lambda)$$

where λ is σ -finite. Now let

m : a finite Borel measure on \mathbb{S}^{d-1}

$z : \mathbb{R}^d \times \mathbb{S}^{d-1} \times (0, \infty) \rightarrow \mathbb{R}^d \setminus \{0\}$ bimeasurable bijection

$g : \mathbb{R}^d \times \mathbb{S}^{d-1} \times \mathcal{B}((0, \infty)) \rightarrow (0, \infty)$ is a positive kernel

Our ν is then taken the form

$$\nu(x, dz) = \int_{\mathbb{S}^{d-1}} \int_0^\infty 1_{dz}(z(x, \theta, r)) g(x, \theta, dr)$$

M Tsuchiya, Stoch Stoch Reports 38 (1992)
V. Kolkoltsov, Proc London Maths Soc 80 (2000)
V. Kolkoltsov, *Nonlinear Markov Processes and Kinetic Equations*. (CUP, 2010)

Example (Bass, PTRF (1988); Kolkoltsov) Take

$$z(x, \theta, r) = r\theta \quad \text{and} \quad g(x, \theta, dr) = \frac{dr}{r^{1+\alpha(x)}}$$

then

$$\nu(x, dz) = \frac{dr}{r^{1+\alpha(x)}} m(d\theta)$$

Theorem ([1])

(i) For $d \geq 2$, i.e., for the case that the given σ -finite measure space

$$U = (S^{d-1} \times (0, \infty))$$

the coefficient of the jump term in the SDE associated to $\nu(x, dz)$ is given by $c(t, x, (r, \theta)) = r\theta$;

(ii) For the case when $d = 1$, namely, for the case that the given σ -finite measure space

$$(U, \mathcal{B}(U), \lambda) = ((0, \infty), \mathcal{B}((0, \infty)), \lambda)$$

the coefficient of the jump term in the SDE associated to $\nu(x, dz)$ defined by

$$\nu(x, dz) = \frac{dr}{r^{1+\alpha(x)}} \quad \alpha(x) \in (0, 2), \quad x \in \mathbb{R}$$

is given by $|c(t, x, (r, \theta))| = r$.

As an application, we consider a consumption-portfolio optimization problem. The wealth process is modelled via

$$\begin{aligned}dS(t) &= \{\rho_t S(t) + (b(t) - \rho_t)u(t) - w(t)\}dt \\ &\quad + \sigma(t)w(t)dW(t) \\ &\quad + w(t-) \int_{0 < |r| < 1} r\theta \tilde{N}(dt, drd\theta) \\ &\quad + w(t-) \int_{|r| \geq 1} r\theta N(dt, drd\theta).\end{aligned}$$

Our objective is to solve the following consumption-portfolio optimization problem:

$$\sup_{(w,u) \in \mathcal{A}} \mathbf{E} \left[\int_0^T \exp\left(-\int_0^t \delta(s) ds\right) \left[\frac{w(t)^\gamma}{\gamma}\right] dt \right] \quad (3)$$

subject to the terminal wealth constraint

$$S(T) \geq 0 \quad a.s.$$

where \mathcal{A} is the set of predictable consumption-portfolio pairs (w, u) with the control u being tame and the consumption w being nonnegative, such that the above SDE has a strong solution over $[0, T]$.

Theorem ([1])

An optimal control (u^*, w^*) is given by

$$u^*(t, x) = \exp \left\{ \int_0^t \frac{\delta(s)}{\gamma - 1} ds \right\} f(t) \frac{1}{\gamma - 1} x$$

and

$$w^*(t, x) = \hat{\pi} x$$

with $f(t)$ and $\hat{\pi}$ being explicitly constructed.

Thank You!