Density Analysis of BSDEs

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Financial market model:

- $W := (W_t)_{t \in [0, T]}$ a Brownian motion defined on the probability space $(\Omega, (\mathcal{F}_t)_{t \in [0, T]}, \mathbb{P})$
- Risk-free asset $S^0 := (S^0_t)_{t \in [0,T]}$,

$$dS_t^0 = S_t^0 r \, dt$$

• Asset
$$S := (S_t)_{t \in [0,T]}$$
,

$$dS_t = S_t \Big(\theta_t \, dt + dW_t \Big),$$

where θ is predictable and bounded.

Motivation: pricing and hedging problems in finance

Investing strategy (r = 0): $(x, (\Pi_t)_t)$ such that the associated wealth process denoted $(X_t^{\times,\Pi})_t$ and defined for all $t \in [0, T]$ by:

$$X_t^{\times,\Pi} := x + \int_0^t \Pi_u \frac{dS_u}{S_u} = x + \int_0^t \Pi_u (dW_u + \theta_u du).$$

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- Utility function $U(x) := -e^{-\alpha x}$
- Utility maximisation problem:

$$V(\mathbf{x}) := \sup_{\Pi \in \mathcal{A}} \mathbb{E}[U(X_T^{\mathbf{x},\Pi} - \mathbf{F})],$$

where F is a \mathcal{F}_T measurable variable (the liability of the investor).

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Hu, Imkeller and Müller have showed that it can be reduced to solve a BSDE (Backward Stochastic Differential Equation) of the form:

$$\mathbf{Y}_t = \mathbf{F} + \int_t^T h(s, \mathbf{Y}_s, Z_s) ds - \int_t^T Z_s dW_s, \quad \mathbf{Y}_T = \mathbf{F}$$

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with an explicit formula for the generator h, where (Y, Z) is a pair of adapted processes "*regular enough*".

The value is given by $V(x) = -e^{-\alpha(x-Y_0)}$.

Optimal strategies are characterized by Z_t .

If we are in the Markovian case, we consider the Forward BSDE:

$$\begin{cases} X_t = X_0 + \int_0^t b(s, X_s) ds + \int_0^t \sigma(s, X_s) dW_s \\ Y_t = g(X_T) + \int_t^T h(s, X_s, Y_s, Z_s) ds - \int_t^T Z_s dW_s, \quad t \in [0, T] \end{cases}$$

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<u>Idea</u>: Get the existence of densities for the Y process and for the Z process with estimates of these densities.

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- \bullet Let ${\mathcal C}$ the space of random variables of the form:

$$F = f(W_{t_1}, ..., W_{t_n}), \ (t_1, ..., t_n) \in [0, T]^n, \ f \in \mathcal{C}_b(\mathbb{R}^n).$$

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• The Malliavin derivative *DF* of *F* is the *H*-valued random variable defined as:

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$$DF = \sum_{i=1}^{n} \frac{\partial f}{\partial x_i} (W_{t_1}, \ldots, W_{t_n}) \mathbf{1}_{[0,t_i]}.$$

 We denote by D^{1,2} the closure of C with respect to the Sobolev norm || · ||_{1,2} defined as:

$$\|F\|_{1,2} := \mathbb{E}[|F|^2] + \mathbb{E}\left[\int_0^T |D_t F|^2 dt
ight].$$

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Theorem (Bouleau-Hirsch)

Assume that $\|DF\|_{L^2([0,T])} > 0$ a.s., then F has a probability distribution which is absolutely continuous with respect to the Lebesgue measure on \mathbb{R} , denoted by ρ_F .

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Malliavin calculus and densities estimates (F centered)

Assume that $DF = \Phi_F(W)$ where $\Phi_F : \mathbb{R}^H \to \mathcal{H}$. We set:

$$g_F(x) = \int_0^{+\infty} e^{-u} \mathbb{E}\left[\mathbb{E}^*[\langle \Phi_F(W), \widetilde{\Phi_F^u}(W) \rangle_{L^2([0,T])}]|F=x\right] du$$

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- Where $\widetilde{\Phi_F^u}(W) := \Phi_F(e^{-u}W + \sqrt{1 e^{-2u}}W^*)$
- With W^* an independent copy of W defined on a probability space $(\Omega^*, \mathcal{F}^*, \mathbb{P}^*)$
- Where \mathbb{E}^* is the expectation under \mathbb{P}^* .

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• Where
$$\widetilde{\Phi_F^u}(W) := \Phi_F(e^{-u}W + \sqrt{1 - e^{-2u}}W^*)$$

- With W^* an independent copy of W defined on a probability space $(\Omega^*, \mathcal{F}^*, \mathbb{P}^*)$
- Where \mathbb{E}^* is the expectation under \mathbb{P}^* .

Theorem (Nourdin-Viens)

F has a density ρ_F with respect to the Lebesgue measure if and only if the random variable $g_F(F)$ is positive a.s.. In this case, the support of ρ_F is a closed interval of \mathbb{R} and for all $x \in \text{supp}(\rho_F)$:

$$\rho_F(x) = \frac{\mathbb{E}(|F|)}{2g_F(x)} \exp\left(-\int_0^x \frac{u du}{g_F(u)}\right).$$

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We make the classical assumption:

L) $h: [0, T] \times \mathbb{R}^3 \to \mathbb{R}$ is Lipschitz in (x, y, z) with Lipschitz constants respectively k_x, k_y, k_z , *i.e.* for all $(x, x', y, y', z, z') \in \mathbb{R}^6$:

$$|h(x,y,z) - h(x',y',z')| \le k_x|x-x'| + k_y|y-y'| + k_z|z-z'|.$$

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Theorem (Antonelli-Kohatsu Higa (2005))

Assume that L) holds (plus some conditions on the coefficients b, σ, g and h). We set $K := k_b + k_y + k_\sigma k_z$. Let $t \in (0, T]$. If for some $A \subset \mathbb{R}$ such that $\mathbb{P}(X_T \in A|X_t) > 0$:

$$\begin{cases} \underline{g}e^{-sgn(\underline{g})KT} + \underline{h}(t)\int_{t}^{T}e^{-sgn(\underline{h}(s))Ks}ds \geq 0\\ \underline{g}^{A}e^{-sgn(\underline{g}^{A})KT} + \underline{h}(t)\int_{t}^{T}e^{-sgn(\underline{h}(s))Ks}ds > 0 \end{cases}$$
(1)

or

$$\begin{cases} \overline{g}e^{-sgn(\overline{g})\kappa\tau} + \overline{h}(t)\int_{t}^{T}e^{-sgn(\overline{h}(s))\kappa s}ds &\leq 0\\ \overline{g}^{A}e^{-sgn(\overline{g}^{A})\kappa\tau} + \overline{h}(t)\int_{t}^{T}e^{-sgn(\overline{h}(s))\kappa s}ds &< 0, \end{cases}$$
(2)

is met, then Y_t has a probability distribution that is absolutely continuous with respect to the Lebesgue measure.

Densities existence for BSDEs: previous results

• These conditions are sufficient but not necessary. As shown in the following example.

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Densities existence for BSDEs: previous results

- These conditions are sufficient but not necessary. As shown in the following example.
- Example (M., Possamaï, Réveillac).

Let T = 1, g(x) = x, X = W, h(s, x, y, z) = (s - 2)x.

(2) is not satisfied for any $t \in [0, T]$.

(1) is not satisfied for any $t \in [0, \frac{3-\sqrt{5}}{2})$. However for all $t \in [0, 1]$:

$$\begin{split} Y_t &= \mathbb{E}\left[W_1 + \int_t^1 (s-2)W_s ds \Big| \mathcal{F}_t\right] \\ &= W_t(1 + \int_t^1 (s-2)ds) = W_t(-\frac{1}{2} + 2t - \frac{t^2}{2}), \end{split}$$

admits a density with respect to the Lebesgue measure except when $t = 2 - \sqrt{3}$.

Antonelli and Kohatsu-Higa have proved an other theorem with upper order conditions on h when it does not depend on z. Let:

$$\tilde{g}(x) := g'(x) + (T-t)h_x(T,x,g(x)),$$

$$egin{aligned} & ilde{h}(s,x,y,z) := -\Big(h_{xt}-hh_{xy}+rac{1}{2}(h_{xxx}+2zh_{xxy}+z^2h_{xxy})\ &+h_yh_x+\sigma_xh_{xx}+z\sigma_xh_{xy})\Big)(s,x,y). \end{aligned}$$

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Theorem (Antonelli-Kohatsu Higa)

Assume that h does not depend on the variable z and suppose that L) holds (plus some conditions on the coefficients). Let $t \in (0, T]$. If for some $A \in \mathbb{R}$ such that $\mathbb{P}(X_T \in A | X_t) > 0$:

$$\begin{cases} \frac{\tilde{g}e^{-sgn(\tilde{g})KT} + \tilde{\underline{h}}(t)\int_{t}^{T}e^{-sgn(\tilde{\underline{h}}(s))Ks}(T-s)ds}{\tilde{g}^{A}e^{-sgn(\tilde{\underline{g}}^{A})KT} + \tilde{\underline{h}}(t)\int_{t}^{T}e^{-sgn(\tilde{\underline{h}}(s))Ks}(T-s)ds} > 0 \end{cases}$$

or

$$\begin{cases} \overline{\tilde{g}} e^{-sgn(\overline{\tilde{g}})KT} + \overline{\tilde{h}}(t) \int_{t}^{T} e^{-sgn(\overline{\tilde{h}}(s))Ks} (T-s) ds &\leq 0 \\ \overline{\tilde{g}}^{A} e^{-sgn(\overline{\tilde{g}}^{A})KT} + \overline{\tilde{h}}(t) \int_{t}^{T} e^{-sgn(\overline{\tilde{h}}(s))Ks} (T-s) ds &< 0 \end{cases}$$

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Densities existence for BSDEs: previous results

We study now the existence of a density for Y and Gaussian estimates of this density in the general case.

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Theorem (Aboura-Bourguin (2012))

Under the above assumptions H1),H2) and H3), Y_t has a density for $t \in (0, T)$ denoted by ρ_{Y_t} satisfying:

$$\frac{\mathbb{E}[|Y_t - \mathbb{E}[Y_t]|]}{2ct} \exp\left(-\frac{(y - \mathbb{E}[Y_t])^2}{2Ct}\right) \le \rho_{Y_t}(y)$$
$$\rho_{Y_t}(y) \le \frac{\mathbb{E}[|Y_t - \mathbb{E}[Y_t]|]}{2Ct} \exp\left(-\frac{(y - \mathbb{E}[Y_t])^2}{2ct}\right).$$

We study the quadratic case under the following assumption:

Q) $h: [0, T] \times \mathbb{R}^3 \to \mathbb{R}$ such that for all $(t, x, y, z) \in ([0, T] \times \mathbb{R})$: $|h(t, x, y, z)| \le K(1 + |y| + |z|^2)$ for some K > 0.

Theorem (M., Possamaï, Réveillac)

Assume that Q) holds with some conditions on the coefficients (but not on the sign of DX_T). Fix $t \in (0, T]$. If for some $A \subset \mathbb{R}$ such that $\mathbb{P}(X_T \in A | X_t) > 0$, $g' \ge 0$, $g'_A > 0$ and $\underline{h}(t) \ge 0$ (resp. $g' \le 0$, $g'_A < 0$ and $\overline{h}(t) \le 0$), then Y_t has a probability distribution which is absolutely continuous with Lebesgue measure.

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Notice that in this theorem we do not need a sign for DX_T .

 \hookrightarrow In the proof we just need to control the norm of DX_T .

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Example Consider the BSDE:

$$Y_t = W_T + \int_t^T rac{1}{2} |Z_s|^2 ds - \int_t^T Z_s dW_s.$$

Then:

according to the previous theorem, $g' \equiv 1 > 0$ so, Y_t admits a density for all $t \in (0, T]$.

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Indeed, by the uniqueness of the solution to this BSDE: $Y_t = W_t + \frac{1}{2}(T - t), \ Z_t = 1 \text{ and } Y_t \text{ admits a density for all } t \in (0, T].$

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Density existence for the Z process ?

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 \hookrightarrow Aboura and Bourguin proved that Z_t admits a density under convexity and growth conditions for the terminal condition g and for the generator h when $h(x, y, z) = \tilde{f}(x, y) + \alpha z$ where α is constant.

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- \hookrightarrow They have Gaussian estimates of this density when $h \in \mathcal{C}^2_b(\mathbb{R})$ and $g \in \mathcal{C}^2_b(\mathbb{R})$.

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- \hookrightarrow Aboura and Bourguin proved that Z_t admits a density under convexity and growth conditions for the terminal condition g and for the generator h when $h(x, y, z) = \tilde{f}(x, y) + \alpha z$ where α is constant.
- \hookrightarrow They have Gaussian estimates of this density when $h \in \mathcal{C}^2_b(\mathbb{R})$ and $g \in \mathcal{C}^2_b(\mathbb{R})$.
- \hookrightarrow Using the fact that Z_t can be represented by the Clark-Ocone formula and after, taking the Malliavin derivative of Z_t .

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We consider the following FBSDE:

$$\begin{cases} X_t = X_0 + \int_0^t b(s, X_s) ds + \int_0^t \sigma(s, X_s) dW_s \\ Y_t = g(X_T) + \int_t^T (\tilde{f}(s, X_s, Y_s) + h(Z_s)) ds - \int_t^T Z_s dW_s. \end{cases}$$

Theorem (M., Possamaï, Réveillac)

Assume that Q) holds with some conditions which ensure that $DX_T > 0$ and $D^2X_T \ge 0$ and assume that $\tilde{f}_x, \tilde{f}_{xx}, \tilde{f}_{xy}, \tilde{f}_{yy} \ge 0$. Then, if there exists $A \subset \mathbb{R}$ such that $\mathbb{P}(X_T \in A | X_t) > 0$, $g' \ge 0$, $g'' \ge 0$, $g''_A > 0$ and $h'' \ge 0$ then, for all $t \in (0, T]$, Z_t has a density with respect to the Lebesgue measure.

Densities existence: our contribution for Z

Assume now that there exists a function $f \in C^2(\mathbb{R})$ such that for all $t \in [0, T]$: $X_t = f(t, W_t)$.

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Under this assumption, for all $0 \le r, s \le t \le T$: $D_r Y_t = D_s Y_t$ and $D_r Z_t = D_s Z_t$, \mathbb{P} -a.s..

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To simplify assume that $\tilde{f} \equiv 0$ (the generator of the BSDE depends only on z through h).

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To simplify assume that $\tilde{f} \equiv 0$ (the generator of the BSDE depends only on z through h).

Theorem (M., Possamaï, Réveillac)

Assume that Q) and conditions on coefficients hold. Assume that $h'' \ge 0$ and $(g \circ f)'' \ge 0$. Then, if there exists $A \subset \mathbb{R}$ such that $\mathbb{P}(X_T \in A|X_t) > 0$ and $(g \circ f)''_A > 0$, then for all $t \in (0, T]$ Z_t has a density with respect to the Lebesgue measure.

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Densities estimates: linear Feynman-Kac's formula

$$\begin{cases} \partial_t v(t,x) + b(t,x) \cdot Dv(t,x) + \frac{1}{2} Tr \left[\sigma \sigma^T(t,x) D^2 v(t,x)\right] = 0\\ v(T,\cdot) = g(\cdot). \end{cases}$$

$$\begin{cases} dX_s^{t,x} = b(s, X_s^{t,x})ds + \sigma(s, X_s^{t,x})dW_s \\ X_t^{t,x} = x. \end{cases}$$

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$$\mathbf{v}(t,x) = \mathbb{E}[g(X_T^{t,x})] = P_{t,T}g(x), \quad (\mathbf{v} \in \mathcal{C}^{1,2})$$

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Densities estimates: semi-linear Feyman-Kac's formula

$$\begin{cases} \partial_t v(t,x) + b(t,x) \cdot Dv(t,x) + \frac{1}{2}Tr \left[\sigma\sigma^T(t,x)D^2v(t,x)\right] = h(t,\cdot,v,\sigma^T \cdot Dv) \\ v(T,\cdot) = g(\cdot). \end{cases}$$

$$\begin{cases} \varphi \\ dX_s^{t,x} = b(s, X_s^{t,x})ds + \sigma(s, X_s^{t,x})dW_s; \quad X_t^{t,x} = x. \\ dY_s^{t,x} = h(t, X_s^{t,x}, Y_s^{t,x}, Z_s^{t,x})ds - Z_s^{t,x}dW_s; \quad Y_T^{t,x} = g(X_T^{t,x}). \end{cases}$$

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Densities estimates: semi-linear Feyman-Kac's formula

$$\begin{cases} \partial_t v(t,x) + b(t,x) \cdot Dv(t,x) + \frac{1}{2}Tr[\sigma\sigma^T(t,x)D^2v(t,x)] = h(t,\cdot,v,\sigma^T \cdot Dv) \\ v(T,\cdot) = g(\cdot). \end{cases}$$

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Densities estimates: semi-linear Feyman-Kac's formula

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$$\begin{cases} \Rightarrow \\ dX_s^{t,x} = b(s, X_s^{t,x})ds + \sigma(s, X_s^{t,x})dW_s; \quad X_t^{t,x} = x. \\ dY_s^{t,x} = h(t, X_s^{t,x}, Y_s^{t,x}, Z_s^{t,x})ds - Z_s^{t,x}dW_s; \quad Y_T^{t,x} = g(X_T^{t,x}). \end{cases}$$

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$$v(t,x) = Y_t^{t,x}, \quad (v \in \mathcal{C}^{1,2})$$

Rk.: $h \equiv 0 \Longrightarrow Y_s^{t,x} = \mathbb{E}[g(X_T^{t,x})|\mathcal{F}_s].$

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Let
$$u(t, W_t) := v(t, X_t)$$
 where $X_t =: f(t, W_t)$. Then,
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Let $\alpha_u^* := \inf\{\alpha > 0, \ u(t, x) = \mathcal{O}(x^{\alpha})\}.$

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P) Assume that $\alpha^*_{u'} \in (0, +\infty)$ and $\alpha^*_{u''} \in (0, +\infty)$.

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Theorem (M., Possamaï, Réveillac)

Assume Q) and P) hold. Suppose that there exists $\overline{\delta} > 0$ such that $(g \circ f)'' \ge \overline{\delta} > 0$ and $h'' \ge 0$. Then, there exist C > 0, $\delta > 0$ and $\gamma \in (0,1)$ such that for all $t \in (0,T]$ the probability distribution of Z_t has a law which admits a density ρ_{Z_t} such that for all $z \in \mathbb{R}$:

$$\frac{\mathbb{E}[|Z_t - \mathbb{E}[Z_t]|]}{2tC(1+|z|^{2\gamma})}e^{\frac{-z^2}{2t\delta^2}} \le \rho_{Z_t}(z) \le \frac{\mathbb{E}[|Z_t - \mathbb{E}[Z_t]|]}{2t\delta^2}e^{-\frac{1}{2t\delta^2}\int_0^{z-\mathbb{E}[Z_t]}\frac{xdx}{1+|x|^{2\gamma}}}$$

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A same result holds for Y_t (we just use the first derivative of u).