UTILITY MAXIMIZATION PROBLEM UNDER MODEL UNCERTAINTY INCLUDING JUMPS

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- Bordigoni G., M. A., Schweizer, M. : A Stochastic control approach to a robust utility maximization problem. *Stochastic Analysis and Applications. Proceedings of the Second Abel Symposium, Oslo,* 2005, Springer, 125-151 (2007).
- Faidi, W., M.,A., Mnif, M. : Maximization of recursive utilities : A Dynamic Programming Principle Approach. Preprint (2010).
- Jeanblanc, M., M. A., Ngoupeyou, A. : Robust utility maximization in a discontinuous filtration. Preprint (2010).

We present a problem of utility maximization under model uncertainty :

$$\sup_{\pi} \inf_{\mathbb{Q}} \mathbf{U}(\pi, \mathbb{Q}),$$

where

- π runs through a set of strategies (portfolios, investment decisions, ...)
- \mathbb{Q} runs through a set of models \mathcal{Q} .

If we have a one known model ℙ : in this case, Q = {ℙ} for ℙ a given reference probability measure and U(π, ℙ) has the form of a ℙ-expected utility from terminal wealth and/or consumption, namely

$$\mathsf{U}(\pi,\mathbb{P})=\mathbb{E}ig(U(X^{\pi}_T)ig)$$

where

• X^{π} is the wealth process

and

• *U* is some utility function.

- Schachermayer (2001) (one single model)
- Becherer (2007) (one single model)
- Schied (2007), Schied and Wu (2005)
- Föllmer and Gundel, Gundel (2005)

REFERENCES : BSDE APPROACH

- El Karoui, Quenez and Peng (2001) : Dynamic maximum principle (one single model)
- Hu, Imkeller and Mueller (2001) (one single model)
- Barrieu and El Karoui (2007) : Pricing, Hedging and Designing Derivatives with Risk Measures (one single model)
- Lazrak-Quenez (2003), Quenez (2004), $Q \neq \{\mathbb{P}\}$ but one keep $U(\pi, \mathbb{Q})$ as an expected utility
- Duffie and Epstein (1992), Duffie and Skiadas (1994), Skiadas (2003), Schroder & Skiadas (1999, 2003, 2005) : Stochastic Differential Utility and BSDE.
- Hansen & Sargent : they discuss the problem of robust utility maximization when model uncertainty is penalized by a relative entropy term.

• Let us consider an agent with time-additive expected utility over consumptions paths :

$$\mathbb{E}\big[\int_0^T e^{-\delta t} u(c_t) dt\big].$$

with respect to some model $(\Omega, \mathcal{F}, \mathcal{F}_t, \mathbb{P}, (B_t)_{t \ge 0})$ where $(B_t)_{t \ge 0}$ is Brownian motion under \mathbb{P} .

• Suppose that the agent has some preference to use another model \mathbb{P}^{θ} under which :

$$B_t^ heta = B_t - \int_0^t heta_s ds$$

is a Brownian motion.

EXAMPLE

The agent evaluate the distance between the two models in term of the relative entropy of P^θ with respect to the reference measure P:

$$\mathcal{R}^{\theta} = \mathbb{E}^{\theta} \big[\int_{0}^{T} \boldsymbol{e}^{-\delta t} |\theta_{t}|^{2} dt \big]$$

In this example, our robust control problem will take the form :

$$V_0 := \inf_{\theta} \Big[\mathbb{E}^{\theta} \Big[\int_0^T e^{-\delta t} u(c_t) dt \Big] + \beta \mathcal{R}^{\theta} \Big].$$

• The answer of this problem will be that : $V_0 = Y_0$ where Y is solution of BSDE or recursion equation :

$$Y_t = \mathbb{E}\Big[\int_t^T e^{-\delta(s-t)} \big(u(c_s)ds - \frac{1}{2\beta}d\langle Y \rangle_s\big) \ \Big|\mathcal{F}_t\Big],$$

• This an example of Stochatic differential utility (SDU) introduced by Duffie and Epstein (1992).

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PRELIMINARY AND ASSUMPTIONS

Let us given :

- Final horizon : $T < \infty$
- $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ a filtered probability space where $\mathbb{F} = \{\mathcal{F}_t\}_{0 \le t \le T}$ is a filtration satisfying the usual conditions of right-continuity and \mathbb{P} -completness.
- Possible scenarios given by

 $\mathcal{Q} := \{ \mathbb{Q} \text{ probability measure on } \Omega \text{ such that } \mathbb{Q} \ll \mathbb{P} \text{ on } \mathcal{F}_{\mathcal{T}} \}$

• the density process of $\mathbb{Q} \in \mathcal{Q}$ is the càdlàg *P*-martingale

$$Z_t^{\mathbb{Q}} = \frac{d\mathbb{Q}}{d\mathbb{P}} \Big|_{\mathcal{F}_t} = \mathbb{E} \Big[\frac{d\mathbb{Q}}{d\mathbb{P}} \Big| \mathcal{F}_t \Big]$$

- we may identify $Z^{\mathbb{Q}}$ with \mathbb{Q} .
- Discounting process : S^δ_t := exp(- ∫^t₀ δ_s ds) with a discount rate process δ = {δ_t}_{0≤t≤T}.

• Let $\mathcal{U}_{t,T}^{\delta}(\mathbb{Q})$ be a quantity given by

$$\mathcal{U}_{t,\mathcal{T}}^{\delta}(\mathbb{Q}) = \int_{t}^{\mathcal{T}} e^{-\int_{t}^{s} \delta_{r} \, dr} U_{s} \, ds + e^{-\int_{t}^{\mathcal{T}} \delta_{r} \, dr} \overline{U}_{\mathcal{T}}$$

- where U = (U_t)_{t∈[0,T]} is a utility rate process which comes from consumption and U_T is the terminal utility at time T which corresponds to final wealth.
- Let $\mathcal{R}^{\delta}_{t,T}(\mathbb{Q})$ be a penalty term

$$\mathcal{R}^{\delta}_{t,T}(\mathbb{Q}) = \int_{t}^{T} \delta_{s} e^{-\int_{t}^{s} \delta_{r} dr} \log \frac{Z_{s}^{\mathbb{Q}}}{Z_{t}^{\mathbb{Q}}} ds + e^{-\int_{t}^{T} \delta_{r} dr} \log \frac{Z_{T}^{\mathbb{Q}}}{Z_{t}^{\mathbb{Q}}}.$$

for $\mathbb{Q} \ll \mathbb{P}$ on $\mathcal{F}_{\mathcal{T}}$.

• We consider the cost functional

$$\boldsymbol{c}(\omega,\mathbb{Q}) := \mathcal{U}_{\boldsymbol{0},\boldsymbol{\mathcal{T}}}^{\delta}(\mathbb{Q}) + \beta \mathcal{R}_{\boldsymbol{0},\boldsymbol{\mathcal{T}}}^{\delta}(\mathbb{Q}) \;.$$

with $\beta > 0$ is a constant which determines the strength of this penalty term.

• Our first goal is to

minimize the functional $\mathbb{Q} \mapsto \Gamma(\mathbb{Q}) := \mathbb{E}^{\mathbb{Q}}[c(.,\mathbb{Q})]$

over a suitable class of probability measures $\mathbb{Q} \ll \mathbb{P}$ on \mathcal{F}_T .

RELATIVE ENTROPY

 Under the reference probability ℙ the cost functional Γ(ℚ) can be written :

$$\begin{split} & \Gamma(\mathbb{Q}) = \mathbb{E}^{\mathbb{P}}\left[Z_{T}^{\mathbb{Q}}\Big(\int_{0}^{T}S_{s}^{\delta}U_{s}\,ds + S_{T}^{\delta}\overline{U}_{T}\Big)\right] \\ & + \beta \mathbb{E}^{\mathbb{P}}\left[\int_{0}^{T}\delta_{s}S_{s}^{\delta}Z_{s}^{\delta}\log Z_{s}^{\mathbb{Q}}\,ds + S_{T}^{\delta}Z_{T}^{\mathbb{Q}}\log Z_{T}^{\mathbb{Q}}\right]. \end{split}$$

• The second term is a discounted relative entropy with both an entropy rate as well a terminal entropy :

$$\mathcal{H}(\mathbb{Q}|\mathbb{P}) := egin{cases} \mathbb{E}^{\mathbb{Q}}\left[\log Z^{\mathbb{Q}}_{\mathcal{T}}
ight], & ext{if } \mathbb{Q} \ \ll \ \mathbb{P} ext{ on } \mathcal{F}_{\mathcal{T}} \ + \infty, & ext{if not} \end{cases}$$

• L^{exp} is the space of all \mathcal{G}_T -measurable random variables X with

$$\mathbb{E}^{\mathbb{P}}\left[\exp\left(\gamma | \pmb{X} |
ight)
ight] < \infty \qquad ext{ for all } \gamma > \mathbf{0}$$

• D_0^{exp} is the space of progressively measurable processes $y = (y_t)$ such that

$$\mathbb{E}^{\mathbb{P}}\Big[\expig(\gamma\, extsf{ess}\, extsf{sup}_{0\leq t\leq \mathcal{T}}|m{y}_t|ig)ig]<\infty,\quad extsf{for all }\gamma>0$$
 .

• D_1^{exp} is the space of progressively measurable processes $y = (y_t)$ such that

$$\mathbb{E}^{\mathbb{P}}\Big[\exp\left(\gamma \int_{0}^{T} |y_{s}| \, ds
ight)\Big] < \infty \quad ext{for all } \gamma > 0 \, .$$

FUNCTIONAL SPACES AND HYPOTHESES (I)

- $\mathcal{M}^{\rho}(\mathbb{P})$ is the space of all \mathbb{P} -martingales $M = (M_t)_{0 \le t \le T}$ such that $\mathbb{E}^{\mathbb{P}}(\sup_{0 \le t \le T} |M_t|^{\rho}) < \infty$.
- Assumption (A) $: 0 \le \delta \le \|\delta\|_{\infty} < \infty, \ U \in D_1^{exp}$ and $\overline{U}_T \in L^{exp}$.
- Denote by Q_f is the space of all probability measures Q on (Ω, G_T) with Q ≪ P on G_T and H(Q|P) < +∞, then :
- For simplicity we will take $\beta = 1$.

THEOREM (BORDIGONI G., M. A., SCHWEIZER, M.)

There exists a unique \mathbb{Q}^* which minimizes $\Gamma(\mathbb{Q})$ over all $\mathbb{Q}\in\mathcal{Q}_f$:

$$\Gamma(\mathbb{Q}^*) = \inf_{\mathbb{Q}\in\mathcal{Q}_f}\Gamma(\mathbb{Q})$$

Furthermore, \mathbb{Q}^* is equivalent to \mathbb{P} .

THE CASE : $\delta = 0$

• The spacial case $\delta = 0$ corresponds to the cost functional

$$\Gamma(Q) = \mathbb{E}^{\mathbb{Q}} \left[\mathcal{U}_{0,T}^{0} \right] + \beta H(\mathbb{Q}|\mathbb{P}) = \beta H(\mathbb{Q}|\mathbb{P}_{\mathcal{U}}) - \beta \log \mathbb{E}^{\mathbb{P}} \left[\exp \left(-\frac{1}{\beta} \mathcal{U}_{0,T}^{0} \right) \right]$$

where
$$\mathbb{P}_{\mathcal{U}} pprox \mathbb{P}$$
 and $rac{d\mathbb{P}_{\mathcal{U}}}{d\mathbb{P}} = c \exp\left(-rac{1}{eta}\mathcal{U}_{0,\mathcal{T}}^0
ight)$.

- Csiszar (1997) have proved the existence and uniqueness of the optimal measure $\mathbb{Q}^* \approx \mathbb{P}_{\mathcal{U}}$ which minimize the relative entropy $H(\mathbb{Q}|\mathbb{P}_{\mathcal{U}})$.
- I. Csiszár : *I*-divergence geometry of probability distributions and minimization problems. *Annals of Probability* **3**, p. 146-158 (1975).

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DYNAMIC STOCHASTIC CONTROL PROBLEM

We embed the minimization of $\Gamma(Q)$ in a stochastic control problem :

• The minimal conditional cost

$$J(au,\mathbb{Q}):=\mathbb{Q}-\mathsf{ess}\,\mathsf{inf}_{\mathbb{Q}'\in\mathcal{D}(\mathbb{Q}, au)}\mathsf{\Gamma}(au,\mathbb{Q}')$$

with $\Gamma(\tau, \mathbb{Q}) := \mathbb{E}_Q[c(\cdot, \mathbb{Q}) | \mathcal{F}_{\tau}],$

- $\mathcal{D}(\mathbb{Q}, \tau) = \{ Z^{\mathbb{Q}'} \mid \mathbb{Q}' \in \mathcal{Q}_f \text{ et } \mathbb{Q}' = \mathbb{Q} \text{ sur } \mathcal{F}_{\tau} \} \text{ and } \tau \in \mathcal{S}.$
- So, we can write our optimization problem as

$$\inf_{\mathbb{Q}\in\mathcal{Q}_f} \Gamma(\mathbb{Q}) = \inf_{\mathbb{Q}\in\mathcal{Q}_f} \mathbb{E}^{\mathbb{Q}} \left[c(\cdot,\mathbb{Q}) \right] = \mathbb{E}^{\mathbb{P}} \left[J(0,\mathbb{Q}) \right].$$

 We obtain the following martingale optimality principle from stochastic control : We have obtained by following El Karoui (1981) :

PROPOSITION (BORDIGONI G., M. A., SCHWEIZER, M.)

- The family $\{J(\tau, \mathbb{Q}) | \tau \in S, \mathbb{Q} \in Q_f\}$ is a submartingale system;
- Q ∈ Q_f is optimal if and only if {J(τ, Q) | τ ∈ S} is a Q-martingale system;
- For each Q ∈ Q_f, there exists an adapted RCLL process
 J^Q = (J^Q_t)_{0≤t≤T} which is a right closed Q-submartingale such that

$$J^{\mathbb{Q}}_{ au} = J(au, \mathbb{Q})$$

SEMIMARTINGALE DECOMPOSITION OF THE VALUE PROCESS

• We define for all $\mathbb{Q}' \in \mathcal{Q}_f^e$ and $\tau \in S$:

$$ilde{oldsymbol{\mathcal{V}}}(au,\mathbb{Q}'):=\mathbb{E}^{\mathbb{Q}'}\left[\mathcal{U}^{\delta}_{ au,\mathcal{T}}\mid\mathcal{F}_{ au}
ight]+eta\mathbb{E}_{\mathbb{Q}'}\left[\mathcal{R}^{\delta}_{ au,\mathcal{T}}(\mathbb{Q}')\mid\mathcal{F}_{ au}
ight]$$

• The value of the control problem started at time τ instead of 0 is :

$$V(au,\mathbb{Q}) := \mathbb{Q} - \mathsf{ess} \inf_{\mathbb{Q}' \in \mathcal{D}(\mathbb{Q}, au)} \widetilde{V}(au,\mathbb{Q}')$$

So we can equally well take the ess inf under P ≈ Q and over all Q' ∈ Q_f and V(τ) ≡ V(τ, Q') and one proves that V is P-special semimartingale with canonical decomposition

$$V = V_0 + M^V + A^V$$

SEMIMARTINGALE BSDE : CONTINUOUS FILTRATION CASE

- We assume tha $\mathbb{F} = (\mathcal{F}_t)_{t \leq T}$ is continuous.
- Let first consider the following quadratic semimartingale BSDE with :

DEFINITION (BORDIGONI G., M. A., SCHWEIZER, M.)

A solution of the BSDE is a pair of processes (Y, M) such that Y is a \mathbb{P} -semimartingale and M is a locally square-integrable locally martingale with $M_0 = 0$ such that :

$$\begin{cases} -dY_t = (U_t - \delta_t Y_t)dt - \frac{1}{2\beta}d < M >_t - dM_t \\ Y_T = \overline{U}_T \end{cases}$$

 Note that Y is then automatically ℙ-special, and that if M is continuous, so is Y.

Remark

 If F = F^W, for a given Brownian mtotion, then the semimartingale BSDE takes the standard form of quadratique BSDE :

$$\begin{cases} -dY_t = \left(U_t - \delta_t Y_t - \frac{1}{2\beta} |Z_t|^2\right) dt - Z_t \cdot dW_t \\ Y_T = \overline{U}_T \end{cases}$$

- Kobylanski (2000), Lepeltier et San Martin (1998), El Karoui and Hamadène (2003), Briand and Hu (2005, 2007).
- Hu, Imkeler and Mueler (06), Morlais (2008), Mania and Tevzadze (2006), Trevzadze (SPA, 2009)

THEOREM (BORDIGONI G., M. A., SCHWEIZER, M.)

Assume that \mathbb{F} is continuous. Then the couple (V, M^V) is the unique solution in $D_0^{exp} \times \mathcal{M}_{0,loc}(\mathbb{P})$ of the BSDE

$$\begin{cases} -dY_t = (U_t - \delta_t Y_t)dt - \frac{1}{2\beta}d < M >_t - dM_t \\ Y_T = U_T' \end{cases}$$

• Moreover, $\mathcal{E}\left(-\frac{1}{\beta}M^{V}\right) = Z^{\mathbb{Q}^{*}}$ is a \mathbb{P} -martingale such that it's supremum belongs to $L^{1}(\mathbb{P})$ where \mathbb{Q}^{*} is the optimal probability.

• We have also that
$$M^V\in \mathcal{M}^p_0(\mathbb{P})$$
 for every $p\in [0,+\infty[$

LEMMA

Let (Y, M) be a solution of BSDE with M continuous. Assume that $Y \in D_0^{exp}$ or $\mathcal{E}\left(-\frac{1}{\beta}M\right)$ is \mathbb{P} -martingale. For any pair of stopping times $\sigma \leq \tau$, then we have the recursive relation

 $Y_{\sigma} = -\beta \log \mathbb{E}^{\mathbb{P}} \Big[\exp \left(\frac{1}{\beta} \int_{\sigma}^{\tau} \left(\delta_{s} Y_{s} - \alpha U_{s} \right) \, ds - \frac{1}{\beta} Y_{\tau} \right) \, \Big| \, \mathcal{F}_{\sigma} \Big]$

- As a consequence one gets the uniqueness result for the semimartingale BSDE.
- In the case where $\delta = 0$, then this yields to the entropic dynamic risk measure.

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- We consider a filtered probability space (Ω, G, G, P). All the processes are taken G-adapted, and are defined on the time interval [0, T].
- Any special G-semimartingale Y admits a canonical decomposition $Y = Y_0 + A + M^{Y,c} + Y^{Y,d}$ where A is a predictable finite variation process, Y^c is a continuous martingale and $M^{Y,d}$ is a pure discontinuous martingale.
- For each i = 1, ..., n, H^i is a counting process and there exist a positive adapted process λ^i , called the \mathbb{P} intensity of H^i , such that the process N^i with $N^i_t := H^i_t \int_0^t \lambda^i_s ds$ is a martingale.
- We assume that the processes H^i , i = 1, ..., d have no common jumps.

Any discontinuous martingale admits a representation of the

$$dM_t^{Y,d} = \sum_{i=1}^d \hat{Y}_t^i dN_t^i$$

where \hat{Y}^{i} , i = 1, ..., d are predictable processes.

THE MODEL : EXAMPLE FROM CREDIT RISK

EXAMPLE (UNDER IMMERSION PROPERTY)

• We assume that \mathbb{G} is the filtration generated by a continuous reference filtration \mathbb{F} and d positive random times τ_1, \dots, τ_d which are the default times of d firms : $\mathbb{G} = (\mathcal{G}_t)_{t \ge 0}$ where

$$\mathcal{G}_t = \bigcap_{\epsilon > 0} \mathcal{F}_{t+\epsilon} \lor \sigma(\tau_1 \land t + \epsilon) \lor \sigma(\tau_2 \land t + \epsilon) \cdots \lor \sigma(\tau_d \land t + \epsilon)$$

where $\sigma(\tau_i \wedge t + \epsilon)$ is the generated σ -fields which is non random before the default times τ_i for each $i = 1, \dots, d$.

- we note $H_t^i = \mathbf{1}_{\{\tau_i \le t\}}$.
- We assume that each τ_i is G-totaly inaccessible and there exists a positive G-adapted process λ^i such that, the process N^i with $N_t^i := H_t^i \int_0^t \lambda_s^i ds$ is a G-martingale.
- Obviously, the process λ^i is null after the default time τ_i .

EXAMPLE

 From Kusuoka, the representation of the discontinuous martingale M^{Y,d} with respect to Nⁱ holds true when the filtration G is generated by a Brownian motion and the default processes.

SEMIMARTINGALE BSDE WITH JUMPS

• Let first consider the following quadratic semimartingale BSDE with jumps :

DEFINITION

A solution of the BSDE is a triple of processes $(Y, M^{Y,c}, \widehat{Y})$ such that Y is a *P*-semimartingale, M is a locally square-integrable locally martingale with $M_0 = 0$ and $\widehat{Y} = (\widehat{Y}^1, \dots, \widehat{Y}^d)$ a \mathbb{R}^d -valued predictable locally bounded process such that :

$$\begin{cases} dY_t = [\sum_{i=1}^d g(\widehat{Y}_t^i)\lambda_t^i - U_t + \delta_t Y_t]dt + \frac{1}{2}d\langle M^{Y,c}\rangle_t + dM_t^{Y,c} + \sum_{i=1}^d \widehat{Y}_t^i dN_t^i \\ Y_T = \overline{U}_T \end{cases}$$

(1)

where $g(x) = e^{-x} + x - 1$.

THEOREM (JEANBLANC, M., M. A., NGOUPEYOU A.)

- There exists a unique triple of process
 (Y, M^{Y,c}, Ŷ) ∈ D^{exp}₀ × M_{0,loc}(P) × L²(λ) solution of the semartingale BSDE with jumps.
- Furthermore, the optimal measure Q* solution of our minimization problem is given :

$$dZ_t^{\mathbb{Q}^*} = Z_{t^-}^{\mathbb{Q}^*} dL_t^{\mathbb{Q}^*}, \quad Z_0^{\mathbb{Q}^*} = 1$$

where

$$dL_t^{\mathbb{Q}^*} = -dM_t^{Y,c} + \sum_{i=1}^d \left(e^{-\widehat{Y}_t^i} - 1\right) dN_t^i.$$

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COMPARISON THEOREM FOR OUR BSDE

THEOREM (JEANBLANC, M., M. A., NGOUPEYOU A.)

Assume that for k = 1, 2, $(Y^k, M^{Y^k,c}, \widehat{Y}^k)$ is solution of the BSDE associated to $(\widetilde{U}^k, \overline{U}^k)$. Then one have

$$Y_t^1 - Y_t^2 \, \leq \, \mathbb{E}^{\mathbb{Q}^{*,2}} \left[\int_t^T rac{S_s^\delta}{S_t^\delta} \left(U_s^1 - U_s^2
ight) ds + rac{S_T^\delta}{S_t^\delta} \left(ar{U}_T^1 - ar{U}_T^2
ight) \left| \mathcal{G}_t
ight]
ight]$$

where $\mathbb{Q}^{*,2}$ the probability measure equivalent to $\mathbb P$ given by

$$\frac{dZ_t^{\mathbb{Q}^{*,2}}}{Z_{t^-}^{\mathbb{Q}^{*,2}}} = -dM_t^{Y^2,c} + \sum_{i=1}^d \left(e^{-\widehat{Y}_t^{i,2}} - 1\right) dN_t^i.$$

In particular, if $U^1 \leq U^2$ and $\bar{U}_T^1 \leq \bar{U}_T^2$, one obtains

$$Y_t^1 \leq Y_t^2$$
, $d\mathbb{P} \otimes dt$ -a.e.

Proof

We denote $\widehat{Y}^{i,12} := \widehat{Y}^{i,1} - \widehat{Y}^{i,2}$ and $M^{12,c} = M^{1,c} - M^{2,c}$. Then :

$$Y_{t}^{12} = \overline{U}_{T}^{12} + \int_{t}^{T} \left(\widetilde{U}_{s}^{12} - \delta_{s} Y_{s}^{12} \right) ds - \sum_{i=1}^{d} \int_{t}^{T} \widehat{Y}_{s}^{i,12} dN_{s}^{i}$$
$$- \sum_{i=1}^{d} \int_{t}^{T} \left[g(\widehat{Y}_{s}^{i,1}) - g(\widehat{Y}_{s}^{i,2}) \right] \lambda_{s}^{i} ds \qquad (2)$$
$$+ \frac{1}{2} \int_{t}^{T} \left(d\langle M^{2,c} \rangle_{s} - d\langle M^{1,c} \rangle_{s} \right) - \int_{t}^{T} dM_{s}^{12,c}$$

IDEA OF THE PROOF (II)

Proof

Note that, for any pair of continuous martingales M^1 , M^2 , denoting $M^{12} = M^1 - M^2$:

$$-\langle M^2, M^{12} \rangle - rac{1}{2} \langle M^2 \rangle + rac{1}{2} \langle M^1 \rangle = rac{1}{2} \langle M^{12} \rangle$$

Using the fact that the process $\langle M^{12} \rangle$ is increasing and that the function g is convex we get :

$$egin{aligned} Y_t^{12} &\leq ar{U}_T^{12} + \int_t^T \left(\widetilde{U}_s^{12} - \delta_s Y_s^{12}
ight) ds \ &+ \sum_{i=1}^d \int_t^T (e^{-\widehat{Y}_s^{i,2}} - 1) \widehat{Y}_s^{i,12} \lambda_s^i ds - \int_t^T d\langle M^{2,c}, M^{12,c}
angle_s \ &- \int_t^T dM_s^{12,c} - \sum_{i=1}^d \int_t^T \widehat{Y}_s^{i,12} dN_s^i. \end{aligned}$$

Proof

Let N^{*} and M^{*,c} be the Q^{*,2}-martingales obtained by Girsanov's transformation from N and M^c, where dQ^{*,2} = Z^{Q^{*,2}}dP.

• Then :

$$Y_{t}^{12} \leq \bar{U}_{T}^{12} + \int_{t}^{T} \left(\widetilde{U}_{s}^{12} - \delta_{s} Y_{s}^{12} \right) ds - \sum_{i=1}^{d} \int_{t}^{T} \widehat{Y}_{s}^{i,12} dN_{s}^{i*} - \int_{t}^{T} dM_{s}^{*,c}$$

which implies that

$$Y_t^{12} \leq \mathbb{E}^{\mathbb{Q}^{*,2}} \Big[\int_t^T e^{-\int_t^s \delta_r dr} \widetilde{U}_s^{12} ds + e^{-\int_t^T \delta_r dr} \overline{U}_T^{12} \Big| \mathcal{G}_t \Big]$$

CONCAVITY PROPERTY FOR THE SEMIMARTINGALE BSDE

Theorem

Let define the map $F: D_1^{exp} \times D_0^{exp} \longrightarrow D_0^{exp}$ such that for all $(U, \bar{U}) \in D_1^{exp} \times D_0^{exp}$, we have

$$F(U, \overline{U}) = V$$

where $(V, M^{V,c}, \hat{V})$ is the solution of BSDE associated to (U, \overline{U}) . Then F is concave ,namely,

$$egin{aligned} & F\left(heta U^1+(1- heta)\widetilde{U}^2, heta \overline{U}^1_T+(1- heta)\overline{U}^2_T
ight) \geq heta F(U^1,ar{U}^1_T)+(1- heta)F(U^2,ar{U}^2_T). \end{aligned}$$

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PROBLEM : RECURSIVE UTILITY PROBLEM

- we assume that $U_s = \tilde{U}(c_s)$ and $\bar{U}_T = \bar{U}(\psi)$ where \tilde{U} and \bar{U} are given utility functions, c is a non-negative G-adapted process and ψ a \mathcal{G}_T -measurable non-negative random variable.
- We study the following optimization problem :

$$\sup_{(c,\psi)\in\mathcal{A}(x)} \mathbb{E}^{\mathbb{Q}^*} \left[\int_0^T S_s^{\delta} U(c_s) ds + S_T^{\delta} \overline{U}(\psi) \right] \\ + \mathbb{E}^{\mathbb{Q}^*} \left[\int_0^T \delta_s S_s^{\delta} \ln Z_s^{\mathbb{Q}^*} ds + S_T^{\delta} \ln Z_T^{\mathbb{Q}^*} \right] := \sup_{\substack{(c,\psi)\in\mathcal{A}(x)}} V_0^{x,\psi,c}$$

where V_0 is the value at initial time of the value process V, part of the solution (V, M^V, \hat{V}) of our BSDE, in the case $U_s = U(c_s)$ and $\bar{U}_T = \bar{U}(\psi)$.

PROBLEM : RECURSIVE UTILITY PROBLEM

 The set A(x) is the convex set of controls parameters (c, ψ) ∈ H²([0, T]) × L²(Ω, G_T) such that :

$$\mathbb{E}^{\widetilde{\mathbb{P}}}\big[\int_0^T c_t dt + \psi\big] \leq x,$$

where $\widetilde{\mathbb{P}}$ is a fixed pricing measure, i.e. a probability $\widetilde{\mathbb{P}}$ equivalent to \mathbb{P} with a Radon-Nikodym density \widetilde{Z} with respect to \mathbb{P} given by :

$$d\widetilde{Z}_t = \widetilde{Z}_{t-}(\theta_t dM_t^c + \sum_{i=1}^n (e^{-z_t^i} - 1) dN_t^i), \ \widetilde{Z}_0 = 1 \ .$$

- Here, Q^{*} is the optimal model measure depends on *c*, ψ.
- In a complete market setting, the process *c* can be interpreted as a consumption, ψ as a terminal wealth, with the pricing measure P is the risk neutral probability.

ASSUMPTIONS ON THE UTILITY FUNCTIONS

- The utility functions U and \overline{U} satisfy the usual regular conditions :
- Strictly increasing and concave.
- Continuous differentiable on the set {U > −∞} and {Ū > −∞}, respectively,
- $\ \, {\bf 0} \ \ \, U'(\infty):=\lim_{x\to\infty} U'(x)=0 \ \text{and} \ \ \, \bar U'(\infty):=\lim_{x\to\infty} \bar U'(x)=0,$
- **3** $U'(0) := \lim_{x \to 0} U'(x) = +\infty$ and $U'(0) := \lim_{x \to 0} \overline{U}'(x) = +\infty$,
- S Asymptotic elasticity $AE(U) := \lim \sup_{x \to +\infty} \frac{xU'(x)}{U(x)} < 1.$

PROPOSITION

Let $G : \mathcal{A}(x) \longrightarrow D_0^{exp}$, as $G(c, \psi) = V$ where $(V, M^{V,c}, \widehat{V})$ is the solution of the BSDE associated with $(U(c), \overline{U}(\psi))$. Then

- G is strictly concave with respect to (c, ψ) ,
- 2 Let $G_0(c, \psi)$ be the value at initial time of $G(c, \psi)$, i.e., $G_0(c, \psi) = V_0$. Then $G_0(c, \psi)$ is continuous from above with respect to (c, ψ) ,
- **6** G_0 is upper continuous with respect to (c, ψ) .

REGULARITY RESULT ON THE VALUE FUNCTION

Theorem

- $(V^1, M^{1,c}, \hat{V}^1)$ the solution associated with $(U(c^1), \bar{U}(\psi^1))$ for a given (c^1, ψ^1) .
- Let (V^ε, M^{ε,c}, V^ε) be the solution of the BSDE associated with (U(c¹ + ε(c² − c¹)), Ū(ψ¹ + ε(ψ² − ψ¹))) for a given (c², ψ²).
- Then V^ε is right differentiable in 0 with respect to ε and the triple (∂_εV, ∂_ε M̃^{V,c}, ∂_ε V̂) is the solution of the following BSDE :

$$\begin{cases} d\partial_{\epsilon} V_{t} = \left(\delta_{t}\partial_{\epsilon} V_{t} - U'(c_{t}^{1})(c_{t}^{2} - c_{t}^{1})\right) dt + d\partial_{\epsilon} \widetilde{M}_{t}^{V,c} + \sum_{i=1}^{d} \partial_{\epsilon} \widehat{V}_{t}^{i} d\widetilde{N}_{t}^{i} \\ \partial_{\epsilon} V_{T} = \overline{U}'(\psi^{1})(\psi^{2} - \psi^{1}) \end{cases}$$

where $\widetilde{N}^{i} = N^{i} - \int_{0}^{\cdot} (e^{-v_{t}^{1,i}} - 1)\lambda_{t}^{i} dt$

THEOREM

Moreover, we obtain

$$\partial_{\epsilon} V_t = \mathbb{E}^{\mathbb{P}} \Big[\frac{Z_T^{\mathbb{Q}^{*,1}}}{Z_t^{\mathbb{Q}^{*,1}}} \frac{S_T^{\delta}}{S_t^{\delta}} \bar{U}'(\psi^1)(\psi^2 - \psi^1) + \int_t^T \frac{Z_s^{\mathbb{Q}^{*,1}}}{Z_t^{\mathbb{Q}^{*,1}}} \frac{S_s^{\delta}}{S_t^{\delta}} U'(c_s^1)(c_s^2 - c_s^1) ds \Big| \mathcal{G}_t \Big]$$

 we solve first an equivalent unconstrained problem to the optimization problem : we associate with a pair (c, ψ) ∈ A(x) the quantity

$$X_{0}^{c,\psi} = \mathbb{E}^{ ilde{\mathbb{P}}}\left(\int_{0}^{ au} oldsymbol{c}_{s} oldsymbol{d}s + \psi
ight)$$

- In a complete market setting, X^{c,ψ} is the initial value of the associated wealth.
- Define by

$$u(x) := \sup_{X_0^{c,\psi} \le x} V_0^{(c,\psi)}$$
(3)

where $V_0^{(c,\psi)} = V_0$, $(V, M^{V,c}, \hat{V})$ is the solution of the BSDE associated with $(U(c), \bar{U}(\psi))$.

PROPOSITION

There exists an unique optimal pair (c^0, ψ^0) which solves the unconstrainted optimization problem.

Proof

- The uniqueness is a consequence of the strictly concavity property of V₀.
- We shall prove the existence by using Komlòs theorem.
- We first Step prove that sup_{(c,φ)∈A(x)} V₀^{c,φ} < +∞ : Because ℙ ∈ Q_f^e, we have :

$$\sup_{(c,\phi)\in\mathcal{A}(x)}V_0^{c,\phi}\leq \sup_{(c,\phi)\in\mathcal{A}(x)}\mathbb{E}^{\mathbb{P}}\Big[\bar{U}(\phi)+\int_0^T U(c_s)ds\Big]:=\widetilde{u}(x).$$

Proof

- Using the elasticity assumption on U and \overline{U} , we can prove that $AE(\widetilde{u}) < 1$, which permits to conclude that, for any x > 0, $\widetilde{u}(x) < +\infty$.
- Let $(c^n, \phi^n) \in \mathcal{A}(x)$ be a maximizing sequence such that :

$$\nearrow \lim_{n \to +\infty} V_0^{c^n, \phi^n} = \sup_{(c,\phi) \in \mathcal{A}(x)} V_0^{c,\phi} < +\infty,$$

where the RHS is finite.

• Then conclude by Using Komlòs theorem.

Theorem

• There exists a constant $\nu^* > 0$ such that :

$$u(x) = \sup_{(c,\psi)} \left\{ V_0^{(c,\psi)} + \nu^* \left(x - X^{(c,\psi)} \right) \right\}$$

and if the maximum is attained in the above constraint problem by (c^*, ψ^*) then it is attained in the unconstraint problem by (c^*, ψ^*) with $X^{(c,\psi)} = x$.

Conversely if there exists ν⁰ > 0 and (c⁰, ψ⁰) such that the maximum is attained in

$$\sup_{(c,\psi)}\left\{V_0^{(c,\psi)}+\nu^0\left(x-X_0^{(c,\psi)}\right)\right\}$$

with $X_0^{(c,\psi)} = x$, then the maximum is attained in our constraint problem by (c^0, ψ^0) .

THE MAXIMUM PRINCIPLE (1)

• We now study for a fixed $\nu > 0$ the following optimization problem :

$$\sup_{(c,\psi)} L(c,\psi) \tag{4}$$

where the functional *L* is given by $L(c, \psi) = V_0^{(c,\psi)} - \nu X_0^{(c,\psi)}$

PROPOSITION (JEANBLANC, M., M. A., NGOUPEYOU A.)

The optimal consumption plan (c^0, ψ^0) which solves (4) satisfies the following equations :

$$U'(c_t^0) = \frac{Z_t^{\widetilde{\mathbb{P}}}}{Z_t^{\mathbb{Q}^*}} \frac{\nu}{\alpha S_t^{\delta}} \qquad \bar{U}'(\psi^0) = \frac{Z_T^{\widetilde{\mathbb{P}}}}{Z_T^{\mathbb{Q}^*}} \frac{\nu}{\bar{\alpha} S_T^{\delta}} a.s$$
(5)

where \mathbb{Q}^* is the model measure associated to the optimal consumption (c^0, ψ^0) .

THE MAIN STEPS OF THE PROOF OF THE PROPOSITION (I)

Let consider the optimal consumption plan (c⁰, ψ⁰) which solve (4) and another consumption plan (c, ψ). Consider ε ∈ (0, 1) then :

$$L(\boldsymbol{c}^{0} + \epsilon(\boldsymbol{c} - \boldsymbol{c}^{0}), \psi^{0} + \epsilon(\boldsymbol{c} - \boldsymbol{c}^{0})) \leq L(\boldsymbol{c}^{0}, \psi^{0})$$

Then

$$\frac{1}{\epsilon} \Big[V_0^{(c^0 + \epsilon(c - c^0), \psi^0 + \epsilon(\psi - \psi^0))} - V_0^{(c^0, \psi^0)} \Big] \\ - \nu \frac{1}{\epsilon} \Big[X_0^{(c^0 + \epsilon(c - c^0), \psi^0 + \epsilon(\psi - \psi^0)} - X_0^{(c^0, \psi^0)} \Big] \le 0$$

Because $\left(X_t^{(c,\psi)} + \int_0^t c_s ds\right)_{t\geq 0}$ is a $\widetilde{\mathbb{P}}$ martinagle we obtain : $\frac{1}{\epsilon} \left[X_t^{(c^0+\epsilon(c-c^0),\psi^0+\epsilon(\psi-\psi^0)} - X_t^{(c^0,\psi^0)}\right]$ $= \mathbb{E}^{\widetilde{\mathbb{P}}} \left[\int_t^T (c_s - c_s^0) ds + (\psi - \psi^0) \Big| \mathcal{F}_t\right]$ • Then the wealth process is right differential in 0 with respect to ϵ we define

$$\partial_{\epsilon} X_{t}^{(c^{0},\psi^{0})} = \lim_{\epsilon \to 0} \frac{1}{\epsilon} (X_{t}^{(c^{0}+\epsilon(c-c^{0}),\psi^{0}+\epsilon(c-c^{0}))} - X_{t}^{(c^{0},\psi^{0})})$$

• We take $\lim_{\epsilon \to 0}$ above, we obtain :

$$\partial_{\epsilon} V_0^{(c^0,\psi^0)} - \nu \partial_{\epsilon} X_0^{(c^0,\psi^0)} \leq 0.$$

THE MAIN STEPS OF THE PROOF (III)

• Consider the optimal density $(Z^{\mathbb{Q}_t^{*,1})}_{t\geq 0}$ where its dynamics is given by

$$\frac{dZ_t^{\mathbb{Q}^{*,1}}}{Z_{t^-}^{\mathbb{Q}^{*,1}}} = -dM^{V,c} + \sum_{i=1}^d \left(e^{-\widehat{Y}^{1,i}} - 1\right) dN_t^i$$

then :

$$\partial_{\epsilon} V_t = \mathbb{E}^{\mathbb{Q}^{*,1}} \Big[\frac{S_T^{\delta}}{S_t^{\delta}} \overline{U}'(X_T^1)(X_T^2 - X_T^1) + \int_t^T \frac{S_s^{\delta}}{S_t^{\delta}} U'(c_s^1)(c_s^2 - c_s^1) ds \Big| \mathcal{G}_t \Big].$$

THE MAIN STEPS OF THE PROOF (IV)

From the last result and the explicitly expression of (∂_εX_t^(c⁰,ψ⁰))_{t≥0} we get :

$$egin{aligned} &\partial_\epsilon V^{(c^0,\psi^0)}_0 -
u \partial_\epsilon X^{(c^0,\psi^0)}_0 \ &= \mathbb{E}^{\mathbb{P}}ig[S^{\delta}_T Z^{\mathbb{Q}^{*,1}}_T ar{U}'(\psi^0)(\psi-\psi^0) + \int_0^T S^{\delta}_s Z^{\mathbb{Q}^*}_s U'(c^0_s)(c_s-c^0_s)dsig] \ &-
u \mathbb{E}^{\mathbb{P}}ig[Z^{\widetilde{\mathbb{P}}}(\psi-\psi^0) + \int_0^T Z^{\widetilde{\mathbb{P}}}_s(c_s-c^0_s)dsig] \end{aligned}$$

• Using the equality above we get :

$$\mathbb{E}^{\mathbb{P}}ig[ig(S_T^{\delta}Z_T^{\mathbb{Q}^{st,1}}ar{U}'(\psi^0)-
uZ^{\widetilde{\mathbb{P}}}ig)(\psi-\psi^0) \ +\int_0^Tig(S_s^{\delta}Z_s^{\mathbb{Q}^{st,1}}U'(c_s^0)-
uZ_s^{\widetilde{\mathbb{P}}}ig)(c_s-c_s^0)dsig]\leq 0$$

The main steps of the Proof (V)

• Let define the set $A := \{ (Z^{\mathbb{Q}^*} \overline{U}'(\psi^0) - \nu Z^{\widetilde{\mathbb{P}}})(\psi - \psi^0) > 0 \}$ taking $c = c^0$ and $\psi = \psi^0 + \mathbf{1}_A$ then $\mathbb{P}(A) = 0$ and we get :

$$(Z^{\mathbb{Q}^*} \overline{U}'(\psi^0) - \nu Z^{\widetilde{\mathbb{P}}}) \leq 0$$
 a.s

• Let define for each $\epsilon > 0$

$$\boldsymbol{B} := \{ (\boldsymbol{Z}^{\mathbb{Q}^*} \bar{\boldsymbol{U}}'(\psi^{\boldsymbol{0}}) - \nu \boldsymbol{Z}^{\widetilde{\mathbb{P}}})(\psi - \psi^{\boldsymbol{0}}) < \boldsymbol{0}, \psi^{\boldsymbol{0}} > \epsilon \}$$

 because {ψ⁰ > 0} due to Inada assumption, we can define ψ = ψ⁰ − 1_B then ℙ(B) = 0 and we get

$$(Z^{\mathbb{Q}^*} \overline{U}'(\psi^0) - \nu Z^{\widetilde{\mathbb{P}}}) \ge 0$$
 a.s

We find the optimal consumption with similar arguments.

• we have also :

THEOREM

Let I and \overline{I} the inverse of the functions U' and \overline{U}' . The optimal consumption (c^0, ψ^0) which solve the unconstrained problem is given by :

$$c_t^0 = I(rac{
u^0}{S_t^\delta} rac{Z_t^{\widetilde{\mathbb{P}}}}{Z_t^{\mathbb{Q}^*}}), \quad dt \otimes d\mathbb{P} \ a.s \ , \qquad \psi^0 = \overline{I}(rac{
u^0}{S_T^\delta} rac{Z_T^{\widetilde{\mathbb{P}}}}{Z_T^{\mathbb{Q}^*}}) \ a.s. \ .$$

where $\nu^0 > 0$ satisfies :

$$\mathbb{E}^{\widetilde{\mathbb{P}}}\Big[\int_{0}^{T}I\big(\frac{\nu^{0}}{S_{t}^{\delta}}\frac{Z_{t}^{\widetilde{\mathbb{P}}}}{Z_{t}^{\mathbb{Q}^{*}}}\big)dt+\overline{I}\big(\frac{\nu^{0}}{S_{T}^{\delta}}\frac{Z_{T}^{\widetilde{\mathbb{P}}}}{Z_{T}^{\mathbb{Q}^{*}}}\big)\Big]=x.$$

THE MAIN STEPS OF THE PROOF (1)

- For any initial wealth x ∈ (0, +∞), there exists a unique ν⁰ such that f(ν⁰) = x.
- Let $(c, \psi) \in \mathcal{A}(x)$ and $(V^{(c,\psi)}, M^{V,c}, v)$ (resp. $(V^{(c^0,\psi^0)}, M^{V^0,c}, v^0)$) the solution of the BSDE associated with $(U(c^0), \overline{U}(\psi^0))$ (resp. $(U(c), \overline{U}(\psi))$) then from comparison theorem, we get :

$$egin{aligned} &\mathcal{V}^{(m{c},\psi)}_0 - \mathcal{V}^{(m{c}^0,\psi^0)}_0 \ &\leq \mathbb{E}^{\mathbb{Q}^*}\Big[m{S}^{\delta}_Tig(ar{U}(\psi) - ar{U}(\psi^0)ig) + \int_0^Tm{S}^{\delta}_sig(m{U}(m{c}_s) - m{U}(m{c}^0_sig)ig)m{d}s\Big] \ &\leq \mathbb{E}^{\mathbb{Q}^*}\Big[m{S}^{\delta}_Tar{U}'(\psi^0)(\psi - \psi^0) + \int_0^Tm{S}^{\delta}_sm{U}'(m{c}^0_sig)(m{c}_s - m{c}^0_sig)m{d}s\Big]. \end{aligned}$$

THE MAIN STEPS OF THE PROOF (2)

It follows from the maximum principle that :

$$\begin{array}{ll} V_0^{(c,\psi)} - V_0^{(c^0,\psi^0)} &\leq & \nu^0 \mathbb{E}^{\mathbb{Q}^*} \left(\frac{Z_T^{\widetilde{\mathbb{P}}}}{Z_T^{\mathbb{Q}^*}} (\psi - \psi^0) + \int_0^T \frac{Z_s^{\widetilde{\mathbb{P}}}}{Z_s^{\mathbb{Q}^*}} (c_s - c_s^0) ds \right) \\ &\leq & \nu^0 \big(\mathbb{E}^{\widetilde{\mathbb{P}}} \Big(\psi + \int_0^T c_s ds \Big) - \mathbb{E}^{\widetilde{\mathbb{P}}} \Big(\psi^0 + \int_0^T c_s^0 ds \Big) \Big) \end{array}$$

- Since $(\boldsymbol{c}, \psi) \in \mathcal{A}(\boldsymbol{x})$, then $\mathbb{E}^{\widetilde{\mathbb{P}}} \left[\psi + \int_{0}^{T} \boldsymbol{c}_{\boldsymbol{s}} d\boldsymbol{s} \right] \leq \boldsymbol{x}$.
- Using that $\mathbb{E}^{\widetilde{\mathbb{P}}}\left[\psi^{0} + \int_{0}^{T} c_{s}^{0} ds\right] = x$, we conclude :

 $V_0^{(c,\psi)} \leq V_0^{(c^0,\psi^0)}$.

1 INTRODUCTION

- 2 THE MINIMIZATION PROBLEM
- **3** A BSDE description for the dynamic value process
- 4 THE DISCONTINUOUS FILTRATION CASE
- **5** Comparison theorem and regularities for the BSDE
- 6 MAXIMIZATION PROBLEM
- **7** THE LOGARITHMIC CASE

LOGARITHMIC CASE (1)

- We assume that δ is deterministic and $U(x) = \ln(x)$ and $\overline{U}(x) = 0$ (hence $I(x) = \frac{1}{x}$ for all $x \in (0, +\infty)$).
- The optimal process $c_t^* = I\left(\frac{\nu}{S_t^o}\frac{\tilde{Z}_t}{Z_t^*}\right) = \frac{S_t^\delta}{\nu}\frac{Z_t^*}{\tilde{Z}_t}$.
- For any deterministic function α such that α(T) = 0, V admits a decomposition as

 $V_t = \alpha(t) \ln(c_t^*) + \gamma_t$

- where γ is a process such that $\gamma_T = 0$.
- Recall that the Radon-Nikodym density Z, and the Radon-Nikodym density of the optimal probability measure Z* satisfy

$$d\widetilde{Z}_{t} = \widetilde{Z}_{t-}(\theta_{t}dM_{t}^{c} + \sum_{i=1}^{n} (e^{-z_{t}^{i}} - 1)dN_{t}^{i}), \ \widetilde{Z}_{0} = 1$$
$$dZ_{t}^{*} = Z_{t-}^{*}(-dM_{t}^{V,c} + \sum_{i=1}^{n} (e^{-y_{t}^{i}} - 1)dN_{t}^{i}), \ Z_{0}^{*} = 1$$

LOGARITHMIC CASE (2)

• In order to obtain a BSDE, we introduce $J_t = \frac{1}{1+\alpha(t)}\beta_t$.

PROPOSITION

(i) The value function V has the form

$$V_t = lpha(t) \ln(c_t^*) + (1 + lpha(t)) J_t$$

where

$$lpha(t) = -\int_t^T e^{\int_t^s \delta(u) du} ds$$

and $(J, \overline{M}^{J,c}, \hat{J})$ is the unique solution of the following Backward Stochastic Differential Equation, where $k(t) = -\frac{\alpha(t)}{1+\alpha(t)}$:

LOGARITHMIC CASE (3)

PROPOSITION

$$dJ_{t} = \left((1 + \delta(t))(1 + k(t))J_{t} - k(t)\delta(t) \right) dt + d\bar{M}_{t}^{J,c} + \frac{1}{2}d\langle\bar{M}^{J,c}\rangle_{t} \\ + \frac{1}{2}k(t)(1 + k(t))\theta_{t}^{2}d\langle M^{c}\rangle_{t} \\ + \sum_{i=1}^{n} j_{t}^{i}d\bar{N}_{t}^{i} + \sum_{i=1}^{n} \left(g(j_{t}^{i})\bar{\lambda}_{t}^{i} + \left(k(t)(e^{-z_{t}^{i}} - 1) + e^{k(t)z_{t}^{i}} - 1 \right) \lambda_{t}^{i} \right) dt$$

• The processes $\bar{M}^{J,c}$ and $d\bar{N}_t^i = dH_t^i - \bar{\lambda}_t^i dt$ are $\bar{\mathbb{P}}$ -martingales where $\frac{d\bar{\mathbb{P}}}{d\bar{\mathbb{P}}}|_{\mathcal{G}_t} = Z_t^{\bar{\mathbb{P}}}$ and $\bar{\lambda}_t^i = e^{k(t)z_t^i}\lambda_t^i$ where

$$dZ_t^{\mathbb{P}} = -Z_{t^-}^{\mathbb{P}}\left(k(t)\theta_t dM_t^c - \sum_{i=1}^d (e^{k(t)z_t^i} - 1)dN_t^i\right)$$

PROPOSITION

ii)

$$dc_t^* = c_{t-}^* \left(-\delta_t dt - dM_t^{V,c} + \theta_t dM_t^c - \theta_t d\langle M^c, M^{V,c} \rangle_t \right. \\ \left. + \sum_{i=1}^d (e^{(y_t^i - z_t^i)} - 1) dN_t^i - \sum_{i=1}^d (g(y_t^i) - g(z_t^i) - g(y_t^i - z_t^i)) \lambda_t^i dt \right)$$

- study more explicit "models" in incomplete market
- Numerical scheme
- replace the entropic penalization by other convex term !!
- consider robustness in the non-dominated case