

When Mathematical Finance and Actuarial Science Cross

Jin Ma

The logo for the USC Department of Mathematics, featuring the USC crest and the text "Department of Mathematics" and "University of Southern California".

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- 2 Credit Risk and Ruin Problems
- 3 Equity Linked Insurance Pricing
- 4 Reinsurance and Stochastic Control Problems
- 5 Concluding Remarks

Definition (Credit Default Swap (CDS))

A CDS is a contract where

- the “protection buyer” “ A ” pays rates “ R ” at times T_{a+1}, \dots, T_b (the “premium leg”) in exchange for a single protection payment L_{GD} (Loss Given Default, the “protection leg”).
- The buyer receives the protection leg by the protection seller “ B ” at the default time τ of a reference entity “ C ”, provided that $T_a < \tau < T_b$.
- The rates R paid by “ A ” stop in case of default.

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In terms of “Term Life Insurance”:

- Time of death (default) — τ (of the insured “ C ”)
- Death benefit — L_{GD} , payable at the moment of death
- Premium — an annuity (e.g. monthly) at (leveled) rate R
- Coverage period (term) — $[T_a, T_b]$, where $a < b$ are two ages.

Credit Risk vs. Actuarial Problems

	Credit Risk	Actuarial Science
τ	Default time	Ruin time, Future life time ($\tau = T(x)$)
$P\{\tau > t\}$	Survival Proba.	Survival Probability (${}_t p_x = P\{T(x) > t\}$)
$\Lambda(t) = -\ln {}_t p_x$	Hazard Process	Hazard Process
$\lambda(t) = \Lambda'(t)$	Default Intensity	“Force of Mortality” ($\mu(x+t) = -({}_t p_x)' / {}_t p_x$)
	Structure	Ruin Problems
	Reduced form	Life Contingencies

An Example in Risk Management

- Recall that the definition of “Value at Risk” of a r.v. Z :

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- Consider the value process $V_t^\pi = x + Q_t^\pi$ ($Q_0^\pi = 0$) for an investment strategy π . Then one can assess the “risk” associated to this strategy by looking at $\text{VaR}_\alpha(\inf_{t \in [0, T]} Q_t^\pi)$.

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- Define

$$\psi(x, T) = \mathbb{P}\{V_t^\pi < 0 : \exists t \in [0, T]\}. \quad (1)$$

Then

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$$\text{VaR}_\alpha(\inf_{t \geq 0} Q_t^\pi) = \inf\{x : \psi(x, T) \leq \alpha\}.$$

- Assume now that $\psi(x, T) \sim e^{-r^*x}$ for some $r^* \in \mathbb{R}$, then

$$\text{VaR}_\alpha(\inf_{t \geq 0} Q_t^\pi) \sim -\frac{\log \alpha}{r^*}!$$

Note

- In Actuarial Sciences, the quantity $\psi(x, T)$ (or $\psi(x) = \mathbb{P}\{V_t^h < 0 : \exists t > 0\}$) is called “*Ruin Probability*”. The estimate $\psi(x, T) \sim e^{-r^*x}$ is called the *Lundberg bound*, with *Lundberg exponent* r^* .

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- Define the “Average VaR” by

$$\rho(Z) \triangleq AVaR_\alpha(Z) \triangleq \frac{1}{\alpha} \int_0^\alpha VaR_u(Z) du.$$

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- The Lundberg bound also implies that

$$\rho\left(\inf_{t \geq 0} Q_t\right) \sim (1 - \log \alpha)/r^*.$$

(The equality can hold if the Lundberg bound is sharp!)

Basic Insurance Models

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- Add interest income: $X_t = x + \int_0^t [r_s X_s + c_s(1 + \rho_s)] ds - S_t$
- Generalized Cramér-Lundberg Model

$$X_t = x + \int_0^t [r_s X_s + c_s(1 + \rho_s)] ds - \int_0^t \int_{\mathbb{R}_+} f(s, z) \mu(dz ds),$$

where $\mu(dt dz)$ is a Poisson random measure on $(0, \infty) \times \mathbb{R}_+$, with Lévy measure $\nu(dz)$. (In the compound Poisson case, $f(t, z) \equiv z$, $S_t = \sum_{k=1}^{N_t} \Delta S_{T_k}$, where N_t is standard Poisson. $\implies \nu(dz) = \lambda F_{U_1}(dz)$, and $c_t = \mathbb{E}\{\Delta S_t\} = \lambda \mathbb{E}[U_1]$, $t \geq 0$.)

An Exponential Martingale Approach (Gerber, (1973))

- Consider the classical Cremér-Lundberg model: $X_t = x + Q_t$, and $Q_t = ct - S_t$. Define a process $M_t^x \triangleq e^{-r(x+Q_t)} / e^{t\theta(r)}$, $t \geq 0$, $x > 0$, and $r > 0$.

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- Choose $\theta(s) \triangleq \lambda[m(s) - 1] - sc$, where $m(s) = E[e^{sU_1}]$. Then $E[e^{-sQ_t}] = e^{t\theta(s)}$, and $\{M_t^x\}$ is an mgf! (called the *Exponential Mg*, and θ is called the *Adjustment coefficient*)

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- Define $\tau_x \triangleq \inf\{t \geq 0 : X_t < 0\}$. By optional sampling:

$$e^{-rx} = M_0^x = E\{M_{\tau_x}^x\} \geq E\{e^{-\tau_x\theta(r)} \mid \tau_x < \infty\} \psi(x). \quad (2)$$

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- $\psi(x) \leq e^{-rx} \sup_{t \geq 0} e^{t\theta(r)} \leq e^{-r^*t}$ — *Lundberg Bound*
where $r^* \triangleq \sup\{r : \theta(r) \leq 0\}$ is the *Lundberg exponent*.

Lundberg Bounds for Reserve with Investments

Consider a general value process of an portfolio $\pi_t = (\pi_t^1, \dots, \pi_t^k)$ (of an insurance company):

$$X_t = x + \int_0^t \left\{ \underbrace{X_s [r_s + \langle \pi_s, \mu_s - r_s \mathbf{1} \rangle]}_{b(s, X_s)} + \underbrace{c_s (1 + \rho_s)}_{\eta_s} \right\} ds + \int_0^t \underbrace{\langle X_s \pi_s, \sigma_s dW_s \rangle}_{\hat{\sigma}_s} - \int_0^t \int_{\mathbb{R}_+} f(s, z) \mu(ds dz), \quad (3)$$

where W is a Brownian motion, μ is a Poisson random measure, and f is the claim size.

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Question:

Can we find an exponential martingale that leads to the Lundberg bound for the general reserve (3), or even more general Itô-type value processes?

Assume $r = 0$, and S is Compound Poisson. Then

$$X_t = x + \int_0^t [b(s, X_s) + \eta_s] ds + \int_0^t \hat{\sigma}_s dW_s - S_t.$$

For each $\delta \in \mathbb{R}$, let $M_t^\delta = \exp\{-\delta X_t - K_t^\delta\}$, and apply Itô.

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Then one can see that if $K_t^\delta = -V_t^\delta + \frac{1}{2} Y_t^\delta + Z_t^\delta$, where

- $V_t^\delta = \delta \int_0^t [b(s, X_s) + \eta_s] ds;$
- $Y_t^\delta = \delta^2 \int_0^t |\hat{\sigma}_s|^2 ds;$
- $Z_t^\delta \triangleq \int_0^t m_s^f(\delta) ds, m_t^f(\gamma) \triangleq \int_{\mathbb{R}_+} [\exp\{\gamma f(t, z)\} - 1] \nu(dz).$

is a local martingale, and could be a martingale on a good day!

Theorem

Denote $\mathcal{D} = \{\delta \geq 0 : Z_t^\delta < \infty, \mathbb{P}\text{-a.s.}, \forall t \geq 0\}$. Then, $\forall \delta \in \mathcal{D}$,

$$\psi(x) \leq e^{-\delta x} \mathbb{E} \left\{ \sup_{t \geq 0} \exp(K_t^\delta) \right\}.$$

Furthermore, $\delta^* \triangleq \sup \{ \delta \in \mathcal{D} : \mathbb{E} \{ \sup_t \exp(K_t^\delta) \} < \infty \}$ is the Lundberg exponent.

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- **Discounted Risk Reserve** $\pi_t = \rho_t = \mu_t = \sigma_t \equiv 0, r > 0$
 - $K_t^\delta = \int_0^t \left\{ \int_0^\infty [\exp(\delta e^{-rs} x) - 1] \lambda F(dx) - c \delta e^{-rs} \right\} ds$

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 - $K_t^\delta = \int_0^t \{ \int_0^\infty [\exp(\delta e^{-rs} x) - 1] \lambda F(dx) - c \delta e^{-rs} \} ds$
- **Perturbed risk reserve** $\pi_t \equiv 1, \rho_t = r_t = \mu_t \equiv 0, \sigma_t \equiv \varepsilon$
 - $K_t^\delta = t(-c\delta + \frac{1}{2}\delta^2\varepsilon^2 + \int_0^\infty (e^{\delta x} - 1)\lambda F(dx)) \triangleq k(\delta)t$

A General Exponential Martingale Method

In general, one can try to find $I \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R})$ and process K^I so that $M_t^I \triangleq \exp\{-I(t, X_t) - K_t^I\}$ is a martingale. If so then the Lundberg bound could be

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Question:

How to find a rate function?

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Useful techniques:

- Solving ODEs, PDEs, IPDEs
- “Principle of smooth fit” (differential inequalities)
- BSDEs with jumps, and with quadratic growth
- Large deviations (infinite horizon ruin probabilities)

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- “**UVL**” (*Universal Variable Life*)

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Literature:

- Brennan-Schwartz ('76), Boyle-Schwartz ('77), Delbaen ('86), Aase-Persson ('94), Nielson-Sandmann (1995), Kurz ('96), ...
- Also, Young (with Bayraktar, Jaimungal, Ludkovski, Zariphopoulou, ...), Schweizer, Frittelli, Rouge-El Karoui, ...

A Life Model

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Benefit Specifications

- Guaranteed benefit/return
- **“Multiple decrements”** (including death, retirement, long term disability, ...)
-

The Single Life Case

- $T(x)$ — Future Life-time r.v., where x is the current age, with “force of mortality” $\lambda_x(t)$
- $X_t \in \{0, 1, \dots, m\}$ — State Process (Markov, $X_0 = 0$, and the state “1” is cemetery/absorbing, representing “death”).
- $dS_t = S_t\{\mu_t dt + \sigma_t dB_t\}$, $S_0 = s$, — tradable
- $dZ_t = Z_t^0\{\mu_t^Z dt + \sigma_t^Z dB_t + \sigma_t d\tilde{B}_t\}$, $Z_0 = z$ — non-tradable

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Dynamics of general reserve

$$d\hat{W}_t^\pi = [r_t \hat{W}_t^\pi + \pi_t(\mu_t - r_t)]dt + \pi_t \sigma_t dB_t - dA_t,$$

where

- $dA_t = \sum_i l_i(t) a^i(t, S_t, Z_t) dt + \sum_{i \neq j} a^{ij}(t, S_t, Z_t) dN_t^{ij}$
- $l_t^i = \mathbf{1}_{\{X_t=i\}}$, $N_t^{ij} \triangleq \#\{\text{jumps of } X \text{ from } i \text{ to } j \text{ during } [0, t]\}$

An Indifference Pricing Problem

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$$V^0(t, w) \triangleq \sup_{\pi \in \mathcal{A}} E\{u(W_T^\pi) | W_t = w\};$$

$$U^k(t, w, s) \triangleq \sup_{\pi \in \mathcal{A}} E\{u(\hat{W}_T^\pi) | \hat{W}_t = w, S_t = s, X_t = k\}.$$

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Indifference pricing

Find $p > 0$, payable at the $t = 0$, such that, denote $V = U^0$

$$V^0(0, w) = V(0, w + p, s). \quad (4)$$

$(p = \inf\{z > 0 : V(0, w + z, s) \geq V^0(0, w)\})$

An Easier Problem.

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- $g \equiv 1$ (Young-Zariphopoulou, '02)
- $g = g(S_T)$ (My-Yu, '07): Let $c(t, s)$ be the Black-Scholes price of $g(S_T)$. Then the HJB equation for V is:

$$\begin{cases} 0 = V_t + \max_{\pi} \left\{ (\mu - r)\pi V_w + \frac{1}{2}\sigma^2\pi^2 V_{ww} + s\sigma^2\pi V_{ws} \right\} + rwV_w \\ \quad + s\mu V_s + \frac{1}{2}\sigma^2 s^2 V_{ss} + \lambda_x(t)(V^0(w - c, t) - V(w, t, s)), \\ V(T, w, s) = u(w). \end{cases}$$

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Note

In the case of exponential utility ($u(w) = -\frac{1}{\alpha}e^{-\alpha w}$) one has

$$V^0(t, w) = -\frac{1}{\alpha} \exp\left\{-\alpha w e^{r(T-t)} - \frac{(\mu - r)^2}{2\sigma^2}(T - t)\right\} \quad (5)$$

Solution for $g = g(S_T)$

- Assume $V(t, w, s) = V^0(t, w)\Phi(t, s)$, then

$$\Phi_t + rS\Phi_s + \frac{\sigma^2 s^2 \Phi_{ss}}{2} - \frac{s^2 \sigma^2 \Phi_s^2}{2\Phi} + \lambda_x(e^{\{c\alpha e^{r(T-t)}\}} - \Phi) = 0$$

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Remark

The case of $g = g(S_T, Z_T)$ is similar but more complicated, since it no longer has an arbitrage price! (cf. Ma-Yu, SAJ ('07)).

Theorem (Yu, '07; M.-Yu, '10)

Under suitable conditions, the value function $U = (U^0, U^1, \dots, U^m)$ is the *unique* viscosity solution to the system of PDDE's:

$$\begin{cases} U_t^k + F_k(t, w, s, DU^k, D^2U^k) + (\mathcal{H}_k U) = 0, \\ U^k(T, w, s) = u(w), \quad k = 0, \dots, m, \end{cases} \quad (7)$$

where

$$\begin{aligned} F_k(\dots) &= \sup_{\pi \in \Pi} \left\{ \pi(\mu_t - r_t)U_w^k + \frac{1}{2}|\sigma_t \pi|^2 U_{ww}^k + \pi \sigma_t^2 s U_{ws}^k \right\} \\ &\quad + \mu_t s U_s^k + \frac{1}{2} \sigma_t^2 s^2 U_{ss}^k + (r_t w - a^k(t, s))U_w^k \\ (\mathcal{H}_k U) &= \sum_{j \neq k} \lambda_t^{kj} (U^j(t, w - a^{kj}(t, s), s) - U^k(t, w, s)). \end{aligned}$$

Main Rationales

- The usual “Multi-Life Contingency” (e.g., pension plans) assumes independent mortality, even for married couples
- Empirical evidence of the **bereaved spouse** (Hu-Goldman ('90) Mariikainen-Valkonen ('96), and Valkonen et al. ('04)) indicated the possible correlated mortality.

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- $T_{x_1}, T_{x_2}, \dots, T_{x_n}$ — future life time random variables,
- $T_m = T_{x_1, \dots, x_n} \triangleq \min\{T_{x_1}, \dots, T_{x_n}\}$ — (Joint-life)
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- $T_M = T_{x_1, \dots, x_n} \triangleq \max\{T_{x_1}, \dots, T_{x_n}\}$ — (Last-survivor)
- If $n = 2$, one has $T_M + T_m = T_{x_1} + T_{x_2}$, $T_M T_m = T_{x_1} T_{x_2}$.
- $F_M(t) + F_m(t) = F_{T_{x_1}}(t) + F_{T_{x_2}}(t)$, $t \geq 0$ where F_T is the distribution function of T .
- If $T_{x_1} \perp T_{x_2}$, then $F_M(t) = F_{T_{x_1}}(t)F_{T_{x_2}}(t) \dots$

The Case of Bereaved Partner

Assume $n = 2$, and that the individual force of mortalities take the form:

$$\begin{cases} \mu_{x_1}(t) = \lambda_{x_1}(t) + \mathbf{1}_{\{T_{x_2} \leq t\}} \gamma_{x_1}(t - T_{x_2}) \\ \mu_{x_2}(t) = \lambda_{x_2}(t) + \mathbf{1}_{\{T_{x_1} \leq t\}} \gamma_{x_2}(t - T_{x_1}), \end{cases} \quad t \geq 0, \quad (8)$$

where λ_{x_i} 's are the (marginal) force of mortality and

$$\gamma_{x_i}(t) = \frac{n_i}{r_i e^t + 1}, \quad i = 1, 2, \quad r_1, r_2, n_1, n_2 > 0.$$

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Note:

This essentially becomes a problem of “*Counter-Party Risk*”, a well-know topic in “Contagion Models” of correlated default!

Existing literature include

- King-Wadhvani, Kodres-Pritsker, Collin-Dufresne, ...
- Jarrow-Yu, Yu (2001, counterparty, two firms)
-

Correlated Default vs. Dependent Mortality

- $\mathcal{F}_t \triangleq \mathcal{F}_t^X \vee \mathcal{F}_t^1 \vee \dots \vee \mathcal{F}_t^I$, where X is some “factor process”, and $\mathcal{F}_t^i = \sigma\{\mathbf{1}_{\{\tau^i \leq s\}} : 0 \leq s \leq t\}$, $\forall i$
- $\mathcal{H}_t^i = \mathcal{F}_t^X \vee \mathcal{F}_t^1 \vee \dots \vee \mathcal{F}_t^{i-1} \vee \mathcal{F}_t^{i+1} \vee \dots \vee \mathcal{F}_t^I$,
 $\implies \mathcal{F}_t = \mathcal{H}_t^i \vee \mathcal{F}_t^i$.
- The following identity is “fundamental”: $\forall Z \in \mathcal{F}$,

$$\mathbf{1}_{\{\tau^i > t\}} \mathbb{E}\{Z | \mathcal{F}_t\} = \mathbf{1}_{\{\tau^i > t\}} \frac{\mathbb{E}\{\mathbf{1}_{\{\tau^i > t\}} Z | \mathcal{H}_t^i\}}{\mathbb{E}\{\mathbf{1}_{\{\tau^i > t\}} | \mathcal{H}_t^i\}}.$$

Correlated Default vs. Dependent Mortality

- $\mathcal{F}_t \triangleq \mathcal{F}_t^X \vee \mathcal{F}_t^1 \vee \dots \vee \mathcal{F}_t^l$, where X is some “factor process”, and $\mathcal{F}_t^i = \sigma\{\mathbf{1}_{\{\tau^i \leq s\}} : 0 \leq s \leq t\}$, $\forall i$
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- Let $\lambda^i \in \mathcal{H}^i$ be the “conditional intensity” of τ^i , defined by $\mathbb{P}\{\tau^i > t | \mathcal{H}_t^i\} = e^{-\int_0^t \lambda_s^i ds} \triangleq (\Gamma_t^i)^{-1}$. Then

$$\mathbb{P}\{\tau^i > T | \mathcal{F}_t\} = \mathbf{1}_{\{\tau^i > t\}} (\Gamma_t^i)^{-1} \mathbb{E}\left\{\Gamma_T^i \middle| \mathcal{H}_t^i\right\}$$

- Denote $Z_t^i \triangleq \mathbf{1}_{\{\tau^i > t\}} \Gamma_t^i$, then for $k = 1, \dots, l$, $\prod_{i=1}^k Z_t^i$'s are all non-negative $\{\mathcal{F}_t\}$ -martingales with mean 1.

Representation of Joint Survival Probability

The idea of “change of measure” (Collin-Dufresne et al. 03-04):

Define $\frac{d\mathbb{P}^{1,\dots,k}}{d\mathbb{P}} \Big|_{\mathcal{F}_T} \triangleq \tilde{Z}_T^k = \prod_{i=1}^k Z_T^i$, $k = 1, \dots, l$, then

- $\mathbb{E}\{Z_T^1 Z_T^2 \dots Z_T^k X | \mathcal{F}_t\} = Z_t^1 Z_t^2 \dots Z_t^k \mathbb{E}^{1,\dots,k}\{X | \mathcal{F}_t\}$, $\mathbb{P} - a.s.$

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- If $l = 2$, then for $t_1 \leq t_2$,

$$\begin{aligned}\mathbb{P}\{\tau^1 > t_1, \tau^2 > t_2\} &= \mathbb{E}\left\{\mathbf{1}_{\{\tau^1 > t_1\}} \mathbb{E}\left\{Z_{t_2}^2 (\Gamma_{t_2}^2)^{-1} \Big| \mathcal{F}_{t_1}\right\}\right\} \\ &= \mathbb{E}\left\{Z_{t_1}^1 Z_{t_1}^2 \mathbb{E}^{\mathbb{P}^2}\left\{(\Gamma_{t_1}^1)^{-1} (\Gamma_{t_2}^2)^{-1} \Big| \mathcal{F}_{t_1}\right\}\right\} \\ &= \mathbb{E}^{1,2}\left\{\mathbb{E}^{\mathbb{P}^2}\left\{(\Gamma_{t_1}^1)^{-1} (\Gamma_{t_2}^2)^{-1} \Big| \mathcal{F}_{t_1}\right\}\right\}.\end{aligned}$$

- In particular, if $t_1 = t_2 = t$, then we have

$$\mathbb{P}\{\tau^1 > t, \tau^2 > t\} = \mathbb{E}^{1,2}\left\{e^{-\int_0^t (\lambda_s^1 + \lambda_s^2) ds}\right\} (= \mathbb{P}\{\tau_m > t\}!)$$

Assume that the individual T_{x_i} 's follow the **Gompertz's law (1825)**:

$\lambda_{x_1}(t) = h_1 e^{g_1(x_1+t)}$, $\lambda_{x_2}(t) = h_2 e^{g_2(x_2+t)}$, $h_i > 0$, $g_i > 0$. Then

$$\mathbb{P}\{T_{x_1} > t_1, T_{x_2} > t_2\} = \begin{cases} \frac{c(t_1, t_2)}{(r_2+1)^{n_2}} \sum_{k=0}^{n_2} \binom{n_2}{k} \frac{h_1}{g_1} r_2^{n_2-k} B^1 \left(\tilde{\mathbb{D}}_k^1(t_2) - \tilde{\mathbb{D}}_k^1(t_1) \right) + c(t_2, t_2) & t_1 \leq t_2; \\ \frac{c(t_1, t_2)}{(r_1+1)^{n_1}} \sum_{k=0}^{n_1} \binom{n_1}{k} \frac{h_2}{g_2} r_1^{n_1-k} B^2 \left(\tilde{\mathbb{D}}_k^2(t_1) \right) - \tilde{\mathbb{D}}_k^2(t_2) & t_1 > t_2, \end{cases}$$

where

- $\Delta_k^i(t) = \int_0^t y^{\frac{k}{g_i}} e^{-\frac{h_i}{g_i}y} dy$, $\tilde{\mathbb{D}}_k^i(t) = \mathbb{D}_k^i\left(\frac{\lambda_{x_i}(t)}{h_i}\right)$, $i = 1, 2$,
- $B^1 = e^{-k(t_2+x_1) + \frac{h_1}{g_1}e^{g_1(x_1+t_1)}}$, $B^2 = e^{-k(t_1+x_2) + \frac{h_2}{g_2}e^{g_2(x_2+t_2)}}$,
- $c(t_1, t_2) = \exp\left\{-\frac{h_1}{g_1}[e^{g_1(x_1+t_1)} - e^{g_1x_1}] - \frac{h_2}{g_2}[e^{g_2(x_2+t_2)} - e^{g_2x_2}]\right\}$.

- Let T_{x_1} and T_{x_2} be two future life time r.v.'s and let K_t be a generic status process, e.g., K could be one of the following:

$$JLI_t = \mathbf{1}_{\{T_{x_1 x_2} \leq t\}}, \quad SLI_t = \mathbf{1}_{\{\overline{T_{x_1 x_2}} \leq t\}}, \quad t \geq 0,$$

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- Define $J(t, w; \pi) \triangleq \mathbb{E}_{t,w} \{u(W_T^\pi - K_T)\}$, where W is the wealth process with investment portfolio π .
- If $K_T \equiv 0$, then denote $J^0(t, w; \pi) \triangleq \mathbb{E}_{t,w} \{u(W_T^\pi)\}$, $\pi \in \mathcal{A}$.
- $U(t, w) \triangleq \sup_{\pi \in \mathcal{A}} J(t, w; \pi)$, $V(t, w) \triangleq \sup_{\pi \in \mathcal{A}} J^0(t, w; \pi)$.

Back to UVL Insurance Pricing

Recall the “separation of variable”: $U(t, w) = V(t, w)\Phi(t, w)$,
where

$$V(t, w) = -\frac{1}{\alpha} \exp\left(-\alpha w e^{r(T-t)} - \frac{(\mu - r)^2}{2\sigma^2}(T - t)\right).$$

Question

What is Φ ?

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Theorem (M.-Yun '10)

- $\Phi(t, w) = \mathbb{E}_{t,w}\{e^{\alpha K_T}\}.$

[Note that $J(t, w; \pi) = J^0(t, w; \pi)\mathbb{E}_{t,w}\{e^{\alpha K_T}\}!$]

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[Note that $J(t, w; \pi) = J^0(t, w; \pi)\mathbb{E}_{t,w}\{e^{\alpha K_T}\}$!]
- The indifference (selling) price is

$$p_t^* = \frac{1}{\alpha} e^{-r(T-t)} \log \Phi(t, w) = \frac{1}{\alpha} e^{-r(T-t)} \log \mathbb{E}_{t,w}[e^{\alpha K_T}].$$

Reinsurance Problem

An insurance company may choose to “cede” some of its risk to a reinsurer by paying a premium. Thus the reserve may look like

$$X_t = x + \int_0^t c_s^h (1 + \rho_s) ds - \int_0^t \int_{\mathbb{R}_+} h(s, x) \mu(dx ds),$$

where h is the “**retention function**”

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Common types of retention functions:

- $h(x) = \alpha x$, $0 \leq \alpha \leq 1$ — Proportional Reinsurance
- $h(x) = \alpha \wedge x$, $\alpha > 0$ — Stop-loss Reinsurance

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Existing Literature

- **Diffusion approximation:** $dX_t = \mu \alpha_t dt + \sigma \alpha_t dW_t$, $X_0 = x$
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Using the so-called “Profit Margin Principle”, one can argue (Liu-M., '09) that a reasonable general reserve dynamics with reinsurance and consumption/investment could look like:

$$dX_t = X_t \left\{ [r_t + \eta_t^\pi + (1 + \rho_t^\alpha)m(t, \alpha)]dt + \langle \pi_t, \sigma_t dW_t \rangle \right\} - D_t dt - \int_{\mathbb{R}_+} \alpha(t, z) f(t, z, \cdot) \mu(dt dz), \quad (10)$$

where $\eta_t^\pi = \langle \pi_t, \mu_t - r_t \mathbf{1} \rangle$, D is the consumption rate, and $m(t, \alpha) = \int_{\mathbb{R}_+} \alpha(t, z) f(t, z) \nu(dz)$ is the reinsurance premium rate.

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Main Difficulty:

For a given $x \geq 0$, find “non-trivial” strategy (π, α, D) , s.t.,

- $\pi \in L^2_{\mathbf{F}}([0, T])$, $0 \leq \alpha_t \leq 1$, $D_t \geq 0$, $t \in [0, T]$;
- The corresponding risk reserve $X^{x, \alpha, \pi, D}$ satisfies $X_0^{x, \alpha, \pi, D} = x$, $X_t^{x, \alpha, \pi, D} \geq 0$, for all $t \in [0, T]$.

Optimal Reinsurance/Investment/Consumption Problem

Denote the set of all admissible strategies by $\mathcal{A}(x)$. Consider the following

Utility Maximization Problem

- Cost functional: for $(\pi, \alpha, D) \in \mathcal{A}(x)$

$$J(x; \pi, \alpha, D) \triangleq E \left\{ \int_0^T U_1(t, D_t) dt + U_2 \left(X_T^{x, \alpha, \pi, D} \right) \right\};$$

- Value function:

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Need to do:

- Show that $\mathcal{A}(x) \neq \emptyset$
- Find the optimal $(\pi^*, \alpha^*, D^*) \in \mathcal{A}(x)$.

Wider-sense Strategies " $\mathcal{A}^w(x)$ ":

$$(\pi, \alpha, D) \in \mathcal{A}^w(x) \quad \text{if} \quad \alpha \in L^2_{\mathbb{F}}([0, T] \times \mathbb{R}, dt \otimes d\nu).$$

Theorem

For any consumption process D and $B \in \mathcal{F}_T$ satisfying $B \geq 0$ and $E^Q \left\{ \int_0^T \beta_s D_s ds + \beta_T B \right\} = x$ (budget constraint), there exists a (π, α) , such that $(\pi, \alpha, D) \in \mathcal{A}^w(x)$ and the corresponding reserve processes $X^{x, \pi, \alpha, D}$ satisfies

$$X_t^{x, \pi, \alpha, D} > 0, \quad 0 \leq t \leq T; \quad \text{and} \quad X_T^{x, \pi, \alpha, D} = B, \quad P\text{-a.s.}$$

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Problems:

- In general $0 \leq \alpha \leq 1$ does not hold in the above.
- How can one check whether a wider-sense optimal strategy is actually a true optimal strategy?

Some Not-So-Easy Problems

Idea

“Fictitious Market” \oplus “Duality Method” (Karatzas, Shreve, Cvitanic...).

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- Define $\delta(x) \triangleq -x\mathbf{1}_{(-\infty,0)}(x)$ (*support function* of $[0, 1]$) and $\mathcal{D} \triangleq \{v \in L^2_{\mathbb{F}}(dt \otimes d\nu) : \sup_{t \in [0, R]} \int_{\mathbb{R}^+} |v(t, z)| \nu(dz) < C_R, \forall R > 0\}$.

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- For $v \in \mathcal{D}$, $\alpha \in L^2$, consider a market with *fictitious* interest rate, appreciation rate, and expense loading:

$$r_t^{\alpha, v} \triangleq r_t + m(t, \alpha v + \delta(v)), \mu_t^v \triangleq \mu_t + m(t, \delta(v)), \rho_s^v \triangleq \rho_s + v x.$$

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- Then the *fictitious* reserve equation under EMM Q is

$$dX_t^v = r_t^{\alpha, v} X_t^v dt + X_t^v \langle \pi_t, \sigma_t dW_t^0 \rangle + dN_t^\alpha - D_t dt, X_0^v = x.$$

Theorem

Assume that

- $B \in L^\infty_{\mathcal{F}_T}(\Omega)$, $D \in L^\infty_{\mathbf{F}}([0, T] \times \Omega; \mathbb{R}_+)$, and
- $B \geq 0$ with $E(B) > 0$.

Suppose that for some $u^* \in \mathcal{D}$ the associated strategy $(\pi^*, \alpha^*, D) \in \mathcal{A}^w(x)$ satisfies, for all $v \in \mathcal{D}$,

$$E\left\{H_T^{*,v} B + \int_0^T H_s^{*,v} D_s ds\right\} \leq E\left\{H_T^{*,u^*} B + \int_0^T H_s^{*,u^*} D_s ds\right\} = x,$$

where $H^{*,v}$ is the “State-Price-Density” process with “fictitious” interest rate $r^{\alpha^*,v}$. Then $(\pi^*, \alpha^*, D) \in \mathcal{A}(x)$. Further, the corresponding reserve satisfies $X_T^* = B$, a.s.

Theorem (Liu-M., '09)

The solvability of the utility optimization problem is equivalent to the solvability of the following “*Forward-Backward SDE*”:

$$\left\{ \begin{array}{l} H_t = 1 + \int_0^t H_s [r_s + m(s, \delta(v) + \alpha v)] ds - \int_0^t H_s \langle \theta_s, dW_s \rangle \\ \quad + \int_0^t \int_{\mathbb{R}^+} H_s - \rho_s \tilde{N}_p(dsdz); \\ X_t = l_2(yH_T) - \int_t^T \left\{ X_s [r_s + m(s, \delta(v) + \alpha v) + \langle \pi_s, \sigma_s \theta_s \rangle] \right. \\ \quad \left. + (1 + \rho_s) m(s, \alpha) \right\} ds + \int_t^T l_1(s, yH_s) ds \\ \quad - \int_t^T X_s \langle \pi_s, \sigma_s dW_s \rangle + \int_t^T \int_{\mathbb{R}^+} [\alpha f](s, z) N_p(dsdz), \end{array} \right.$$

where $l_1(s, \cdot) = [U'_1(s, \cdot)]^{-1}$, and $l_2(\cdot) = [U'_2(\cdot)]^{-1}$.

A New Type of Stochastic Control Problem

Recall the general form of a reserve equation with reinsurance:

$$dX_t = b(X_t, \alpha_t, \pi_t)dt + \sigma(\pi_t)dB_t - \int_{\mathbb{R}_+} [\alpha f](t, x) \tilde{\mu}(dxdt),$$
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Suppose that the random field α is such that there exists some predictable pair (β, u) so that the martingale

$$M_t^u \triangleq \int_0^t \beta_s dB_s + \int_0^t \int_{\mathbb{R}_+} [\alpha f](s, x)\tilde{\mu}(dxds)$$

satisfies the following property:

$$d[M^u]_t = dt + u_t dM_t^u, \quad t \geq 0. \quad (11)$$

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Note: Since $\Delta[M^u]_t = (\Delta M_t^u)^2 = u_t \Delta M_t^u$, u exactly controls the jumps of the reserve, that is, the claim size!

Such decomposition is possible for all bounded, predictable u , provided

- ν is “atom-free” with $\nu^c([-1, 1]) = +\infty$,
- The probability space is “nice” (for example, Wiener-Poisson — Buckdhan-Ma-Rainer, '08)

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Now note that the BM B itself satisfies (11) with $u \equiv 0$, we can rewrite the reserve dynamics in the following new form:

$$\begin{aligned}dX_t &= b(X_t, u_t, \pi_t)dt + \tilde{\sigma}(\pi_t)dM_t^u, \\X_0 &= x,\end{aligned}$$

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Note:

This controlled dynamics is **NEW**: The driving martingale varies with u in the class of the “*normal martingales*”!

Normal Martingales Revisited

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- $u_t = -2M_{t-}$ — “*Parabolic martingale*” (Protter-Sharpe (1979)).

The Stochastic Control Problem

The set of “admissible controls at time t ” — $\mathcal{U}(t)$:

- $(\Omega, \mathcal{F}, P; \mathbf{F}^t = \{\mathcal{F}_s\}_{s \geq t})$ — Wiener-Poisson space
- (π, u) — $U_1 \times U$ -valued, \mathbf{F}^t -predictable on $[t, T]$,
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Basic Elements

- Given $\mu = (\Omega, \mathcal{F}, P, \mathbf{F}^t, \pi, u, X^u) \in \mathcal{U}(t)$, $t \geq 0$, the controlled dynamics is

$$dY_s = b(Y_s, \pi_s, u_s)ds + \sigma(Y_{s-}, \pi_s, u_s)dX_s^u, \quad Y_t = y, \quad s \geq t.$$

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- Value function:

$$V(t, y) = \inf_{\mu \in \mathcal{U}(t)} E\{g(Y_T^{t,y}(\mu))\}, \quad (t, y) \in [0, T] \times \mathbb{R}^m.$$

- Prove the Dynamic Programming Principle

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$$\begin{cases} V_t(t, y) + \inf_{(\pi, u) \in \bar{U}} \mathcal{L}_{\pi, u}[V](t, y) = 0, \\ V(T, y) = g(y), \end{cases} \quad (12)$$

where

$$\begin{aligned} \mathcal{L}_{\pi, u}[\varphi](t, y) = & \nabla_y \varphi b(y, \pi, u) + \sum_{i=1}^d \left\{ \mathbf{1}_{\{u^i=0\}} \frac{1}{2} (D_{yy}^2 \varphi \sigma^i, \sigma^i) \right. \\ & \left. + \mathbf{1}_{\{u^i \neq 0\}} \frac{\varphi(t, y + u^i \sigma^i) - \varphi(t, y) - u^i \nabla_y \varphi(t, y) \sigma^i}{(u^i)^2} \right\}. \end{aligned}$$

Main Results

- Prove the Dynamic Programming Principle
- Derive the HJB equation

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- Prove that the value function $V(\cdot, \cdot)$ is the **unique viscosity solution** to the HJB eq. (12). (Buckdahn-Ma-Rainer ('08))

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“Ask not only what finance can do for insurance. Ask also what insurance can do for finance.”

— Hans Bühlmann (1987)

References: <http://www-rcf.usc.edu/~jinma/>

THANK YOU VERY MUCH!