# Finance, Insurance, and Stochastic Control (II) 

## Jin Ma

USC Department of Mathematics<br>University of Southern California

Spring School on "Stochastic Control in Finance" Roscoff, France, March 7-17, 2010

## Outline

(1) Equity Linked Insurance Pricing
(2) The Indifference Pricing Problem
(3) The UVL Insurance Problem
(4) General Life Insurance Models
(5) The Case of Bereaved Partner
(6) Counter-Party Risk Models
(7) UVL Insurance Problem Once More

## Equity Linked Life Insurance

An Equity-Linked Life insurance is one that

- allows a separate account with cash/investment options
- links the death benefits to the cash/investment performance


## Equity Linked Life Insurance

An Equity-Linked Life insurance is one that

- allows a separate account with cash/investment options
- links the death benefits to the cash/investment performance

Examples of such insurance include

- "ELEPAVG" (Equity-Linked Endowment Policy with Asset Value Guarantee)
- "UVL" (Universal Variable Life)


## Equity Linked Life Insurance

An Equity-Linked Life insurance is one that

- allows a separate account with cash/investment options
- links the death benefits to the cash/investment performance

Examples of such insurance include

- "ELEPAVG" (Equity-Linked Endowment Policy with Asset Value Guarantee)
- "UVL" (Universal Variable Life)


## 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, ...


# Basic elements involved in an UVL insurance 

A Life Model

- Single life
- Multiple life


## Basic elements involved in an UVL insurance

A Life Model

- Single life
- Multiple life

A Market Model

- Tradable assets vs. Non-tradable assets, ...


## Basic elements involved in an UVL insurance

A Life Model

- Single life
- Multiple life

A Market Model

- Tradable assets vs. Non-tradable assets, ...

Benefit Specifications

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


## The Single Life Case

## Basic Elements

- $T(x)$ - Future Life-time r.v., where $x$ is the current age
- $G_{x}(t) \triangleq P\{T(x)>t\} \triangleq{ }_{t} p_{x}, t \geq 0$ - survival function
- ${ }_{h} q_{x+t} \triangleq P\{T(x) \leq t+h \mid T(x)>t\}=1-{ }_{h} p_{x+t}$.
- $\lambda_{x}(t)=\lim _{h \rightarrow 0} \frac{h q_{x+t}}{h}=-\frac{f_{x}(t)}{G_{x}(t)}$ - force of mortality


## The Single Life Case

## Basic Elements

- $T(x)$ - Future Life-time r.v., where $x$ is the current age
- $G_{x}(t) \triangleq P\{T(x)>t\} \triangleq{ }_{t} p_{x}, t \geq 0$ - survival function
- ${ }_{h} q_{x+t} \triangleq P\{T(x) \leq t+h \mid T(x)>t\}=1-{ }_{h} p_{x+t}$.
- $\lambda_{x}(t)=\lim _{h \rightarrow 0} \frac{h q_{x+t}}{h}=-\frac{f_{x}(t)}{G_{x}(t)}$ - force of mortality
- $X_{t} \in\{0,1, \ldots, m\}$ - State Process (finite state Markov, representing "multiple decreements", e.g. short/long term disabilities, withdrawal, retirement, death, etc. $X_{0}=0$, and the state " 1 " is cemetery/absorbing, representing "death".)
- $d S_{t}^{0}=r_{t} S_{t}^{0} ; S_{0}^{0}=s^{0}$-money market
- $d S_{t}=S_{t}\left\{\mu_{t} d t+\sigma_{t} d B_{t}\right\}, S_{0}=s$, - tradable
- $d Z_{t}=Z_{t}^{0}\left\{\mu_{t}^{Z} d t+\sigma_{t}^{Z} d B_{t}+\sigma_{t} d \tilde{B}_{t}\right\}, Z_{0}=z$-non-tradable


## Principle of Equivalent Utility

The original form of "Principle of Equivalent Utility" states that the premium $\Pi$ of a claim $\mathscr{X}$ should be determined by the equation

$$
u(x)=E[u(x+\Pi-\mathscr{X})]
$$

where $u$ is a utility function, and $x$ is the initial wealth.

## Principle of Equivalent Utility

The original form of "Principle of Equivalent Utility" states that the premium $\Pi$ of a claim $\mathscr{X}$ should be determined by the equation

$$
u(x)=E[u(x+\Pi-\mathscr{X})]
$$

where $u$ is a utility function, and $x$ is the initial wealth.

- If $x=0$, then it is called Zero Utility Principle.
- If furthermore $u(x)=x$, then is often referred to as "Equivalence Principle".)
- Dynamically, assume that $X_{t}=x+\int_{0}^{t} c_{s} d s-S_{t}, t \geq s \geq 0$, and $\mathscr{X}=S_{T}$, then at any time $t \in[0, T]$ the premium $c_{t}$ can be determined by solving the equation

$$
u(x)=E\left\{u\left(X_{T}\right) \mid X_{t}=x\right\} .
$$

## Principle of Equivalent Utility

- If we use the risk reserve with investment, that is, the dynamic of the risk reserve $X$ follows the following SDE:

$$
\begin{equation*}
X_{t}=x+\int_{0}^{t}\left[r_{s} X_{s}+c_{s}\left(1+\rho_{s}\right)\right] d s+\int_{0}^{t}\left\langle\pi_{s}, \sigma_{s} d B_{s}\right\rangle-S_{t} \tag{1}
\end{equation*}
$$

then we can require that the premium is determined so that the expected utility maximized. In other words, one solves

$$
u(x)=\sup _{\pi \in \mathscr{A}} E\left\{u\left(X_{T}^{\pi}\right) \mid X_{t}=x\right\}
$$

## Principle of Equivalent Utility

- If we use the risk reserve with investment, that is, the dynamic of the risk reserve $X$ follows the following SDE:

$$
\begin{equation*}
X_{t}=x+\int_{0}^{t}\left[r_{s} X_{s}+c_{s}\left(1+\rho_{s}\right)\right] d s+\int_{0}^{t}\left\langle\pi_{s}, \sigma_{s} d B_{s}\right\rangle-S_{t} \tag{1}
\end{equation*}
$$

then we can require that the premium is determined so that the expected utility maximized. In other words, one solves

$$
u(x)=\sup _{\pi \in \mathscr{A}} E\left\{u\left(X_{T}^{\pi}\right) \mid X_{t}=x\right\}
$$

- (Note: This is almost like an optimal control problem for maximizing the expected terminal utility by Merton (1969, 1971). But determing the premium process is rather difficult.)


## Principle of Equivalent Utility

- If we use the risk reserve with investment, that is, the dynamic of the risk reserve $X$ follows the following SDE:

$$
\begin{equation*}
X_{t}=x+\int_{0}^{t}\left[r_{s} X_{s}+c_{s}\left(1+\rho_{s}\right)\right] d s+\int_{0}^{t}\left\langle\pi_{s}, \sigma_{s} d B_{s}\right\rangle-S_{t} \tag{1}
\end{equation*}
$$

then we can require that the premium is determined so that the expected utility maximized. In other words, one solves

$$
u(x)=\sup _{\pi \in \mathscr{A}} E\left\{u\left(X_{T}^{\pi}\right) \mid X_{t}=x\right\}
$$

- (Note: This is almost like an optimal control problem for maximizing the expected terminal utility by Merton (1969, 1971). But determing the premium process is rather difficult.)
- A more practical version of the "premium" is that it is paid as a lump-sum at the time of the contract. Although it is still priced "dynamically", it is paid only once at the initial time $t$.


## A Stochastic Control Point of View

Assume we are in a "risk neutral world". Rewrite (1) as
$X_{t}^{\pi}=X_{0}+p+\int_{0}^{t} r_{s} X_{s}^{\pi} d s+\int_{0}^{t}\left\langle\pi_{s}, \sigma_{s} d B_{s}\right\rangle-Y_{t}=W_{t}-Y_{t}$,
where

- $p$ is the (lump-sum) premium paid at $t=0$,
- $W_{t}^{\pi} \triangleq X_{0}+p+\int_{0}^{t} r_{s} X_{s} d s+\int_{0}^{t}\left\langle\pi_{s}, \sigma_{s} d B_{s}\right\rangle$,
- $Y$ is a general "Loss process" (e.g., $Y_{t}=S_{t}$ )


## Note

If the insurer does not sell the insurance, then $Y=0$, and therefore $p=0$. The utility maximization problem becomes a usual stochastic control problme, and we denote its value function by

$$
\begin{equation*}
V^{0}(x, t) \triangleq \sup _{\pi \in \mathscr{A}} E\left\{u\left(W_{T}^{\pi}\right) \mid W_{t}=x\right\} \tag{2}
\end{equation*}
$$

## The Indifference Pricing Problem

If the insurance is sold, and the liability cannot be traded after its transfer and before the expiration. Then the value function of the insurer should be

$$
\begin{equation*}
U(t, x+p, y)=\sup _{\pi \in \mathscr{A}} E\left\{u\left(W_{T}-Y_{T}\right) \mid W_{t}=x+p, Y_{t}=y\right\} \tag{3}
\end{equation*}
$$

## Definition

Let $y \triangleq Y_{t}$. A premium $p \geq 0$ is said to be " $y$-acceptable" if

$$
\begin{equation*}
V^{0}(t, x) \leq U(t, x+p, y), \quad \forall(t, x) \tag{4}
\end{equation*}
$$

Denote $\mathscr{P}_{y}=\{$ all $y$-acceptable premium $\}$. Define the universal write price, $p^{*}(t, y)$ by $p^{*}(t, y) \triangleq \inf \left\{p \geq 0: V^{0}(t, x) \leq U(t, x+p, y), \forall(t, x)\right\}=\inf \mathscr{P}_{y}$.

## Existence of the Fair Price

## Theorem

Suppose that $\mathscr{P}_{s, z} \neq \emptyset$, and let $p^{*} \triangleq \inf \mathscr{P}_{y}$. Then it holds that

$$
V^{0}(t, x)=U\left(t, x+p^{*}, y\right), \forall(t, x)
$$

## Sketch of the proof

- By Comparison Theorem, $W_{0} \geq \tilde{W}_{0} \Longrightarrow W_{T}^{\pi} \geq \tilde{W}_{T}^{\pi}$ $\Longrightarrow U(t, x+p, y)$ is increasing in $p$.
- Since $Y_{T} \geq 0 \Longrightarrow u\left(W_{T}^{\pi}-Y_{T}\right) \leq u\left(W_{T}^{\pi}\right) \Longrightarrow$

$$
U(t, x, y) \leq V^{0}(t, x) \leq U\left(t, x+p^{*}, y\right)
$$

- If $U(t, \cdot, y)$ is continuous, then $\exists p^{* *} \in\left[0, p^{*}\right]$ s.t.

$$
V^{0}(t, x)=U\left(t, x+p^{* *}, y\right)
$$

- But $p^{* *} \in \mathscr{P}_{s, z} \Longrightarrow p^{*} \leq p^{* *} \Longrightarrow p^{*}=p^{* *}$.


## Indifference Pricing in Finance/Insurance

- First introduced by Hodges and Neuberger (1989), as a pricing principle for contingent claims in an incomplete market.
- The value is within the interval of arbitrage prices

$$
\left[\inf _{Q} E_{Q}\left\{\mathscr{X} e^{-r T}\right\}, \sup _{Q} E_{Q}\left\{\mathscr{X} e^{-r T}\right\}\right],
$$

where $Q$ runs over the set of all EMMs.

## Indifference Pricing in Finance/Insurance

- First introduced by Hodges and Neuberger (1989), as a pricing principle for contingent claims in an incomplete market.
- The value is within the interval of arbitrage prices

$$
\left[\inf _{Q} E_{Q}\left\{\mathscr{X} e^{-r T}\right\}, \sup _{Q} E_{Q}\left\{\mathscr{X} e^{-r T}\right\}\right]
$$

where $Q$ runs over the set of all EMMs.
Existing works for similar problems

- Cvitanić et al.('01), Delbaen et al.('02)... (martingale, duality)
- Rouge \& El Karoui('00) (BSDEs)
- M. Davis ('00), M. Musiela \& Zariphopoulou('02); Young and Zariphopoulou('02) (PDE solutions, power/exponential utility)
- Bielecki, Jeanblanc and Rutkowski ('05) (defaultable claims)


## A Universal Variable Life Insurance Problem

The Universal Variable Life (UVL for short) is an insurance product that offers

- a separate cash account besides a death benefit
- various investment options
- different risk/return relationships (may include money market, bond, common stocks, or even non-tradable equities.)


## Main Features

- The changes in the policy's cash values and death benefits will be related directly to the investment performance of its underlying assets.
- The death benefit will not fall below a minimum amount (usually the initial face amount) even if the invested assets depreciate in value by a substantial amount. Although there is no similar "floor" to protect the cash values.

Consider a term life insurance with expiration date $T>0$ and death benefit

$$
\begin{equation*}
b_{t}=g\left(S_{t}^{1}, \cdots, S_{t}^{d}, Z_{t}\right)=g\left(S_{t}, Z_{t}\right) \tag{5}
\end{equation*}
$$

where $g: \mathbb{R}^{d+1} \mapsto(0, \infty)$ is some measurable function.

Consider a term life insurance with expiration date $T>0$ and death benefit

$$
\begin{equation*}
b_{t}=g\left(S_{t}^{1}, \cdots, S_{t}^{d}, Z_{t}\right)=g\left(S_{t}, Z_{t}\right) \tag{5}
\end{equation*}
$$

where $g: \mathbb{R}^{d+1} \mapsto(0, \infty)$ is some measurable function.

## Example

- $g\left(S_{t}, Z_{t}\right)=S_{t}^{i} \vee s^{i}$, for some $i$,
- $g\left(S_{t}, Z_{t}\right)=Z_{t} \vee z$.
- If $Z$ is the retirement fund, one can set $g\left(Z_{t}\right)=Z_{t} \vee e^{\bar{r} t} z$, $t \geq 0$, where $\bar{r}$ is a certain growth rate (such as the interest rate or any contractually pre-determined rate.


## The Death Benefit

Consider a term life insurance with expiration date $T>0$ and death benefit

$$
\begin{equation*}
b_{t}=g\left(S_{t}^{1}, \cdots, S_{t}^{d}, Z_{t}\right)=g\left(S_{t}, Z_{t}\right) \tag{5}
\end{equation*}
$$

where $g: \mathbb{R}^{d+1} \mapsto(0, \infty)$ is some measurable function.

## Example

- $g\left(S_{t}, Z_{t}\right)=S_{t}^{i} \vee s^{i}$, for some $i$,
- $g\left(S_{t}, Z_{t}\right)=Z_{t} \vee z$.
- If $Z$ is the retirement fund, one can set $g\left(Z_{t}\right)=Z_{t} \vee e^{\bar{r} t} z$, $t \geq 0$, where $\bar{r}$ is a certain growth rate (such as the interest rate or any contractually pre-determined rate.


## Note:

In this case the loss process is $Y_{t}=g\left(S_{T}, Z_{T}\right) \mathbf{1}_{\{T(x) \leq t\}}, t \geq 0$.

## Some Optimization Problems

We denote

- $\mathscr{A}=\left\{\pi: E \int_{0}^{T}\left|\pi_{t}\right|^{2} d t<\infty\right\}$
- $E_{t, w, s, z}\{\cdot\}=E\left\{\cdot \mid W_{t}=w, S_{t}=s, Z_{t}=z\right\}$.


## Some Optimization Problems

We denote

- $\mathscr{A}=\left\{\pi: E \int_{0}^{T}\left|\pi_{t}\right|^{2} d t<\infty\right\}$
- $E_{t, w, s, z}\{\cdot\}=E\left\{\cdot \mid W_{t}=w, S_{t}=s, Z_{t}=z\right\}$.
- $J(t, w, s, z ; \pi) \triangleq E_{t, w, s, z}\left\{u\left(W_{T}^{\pi}-Y_{T}\right)\right\}$,
- $J^{0}(t, w ; \pi) \triangleq E_{t, w}\left\{u\left(W_{T}^{\pi}\right)\right\} .\left(T(x)>T, \Longrightarrow Y_{T}=0\right.$.)
- $\widehat{J}(t, w, s ; \pi) \triangleq E_{t, w, s}\left\{u\left(W_{T}^{\pi}-g\left(S_{T}\right) Y_{T}\right)\right\} .\left(g=g\left(S_{T}\right)\right)$


## Some Optimization Problems

We denote

- $\mathscr{A}=\left\{\pi: E \int_{0}^{T}\left|\pi_{t}\right|^{2} d t<\infty\right\}$
- $E_{t, w, s, z}\{\cdot\}=E\left\{\cdot \mid W_{t}=w, S_{t}=s, Z_{t}=z\right\}$.
- $J(t, w, s, z ; \pi) \triangleq E_{t, w, s, z}\left\{u\left(W_{T}^{\pi}-Y_{T}\right)\right\}$,
- $J^{0}(t, w ; \pi) \triangleq E_{t, w}\left\{u\left(W_{T}^{\pi}\right)\right\} .\left(T(x)>T, \Longrightarrow Y_{T}=0\right.$.)
- $\widehat{J}(t, w, s ; \pi) \triangleq E_{t, w, s}\left\{u\left(W_{T}^{\pi}-g\left(S_{T}\right) Y_{T}\right)\right\} .\left(g=g\left(S_{T}\right)\right)$


## The Value Functions

- $V^{0}(t, w)=\sup _{\pi \in \mathscr{A}} J^{0}(t, w ; \pi)$
- $V(t, w, s)=\sup _{\pi \in \mathscr{A}} \widehat{J}(t, w, s ; \pi)$
- $U(t, w, s, z)=\sup _{\pi \in \mathscr{A}} J(t, w, s, z ; \pi)$.


## Solution for $g=g\left(S_{T}\right)$

- First recall the Bellman Principle: for any $h>0$,

$$
\begin{equation*}
V(t, w, s)=\sup _{\pi \in \mathscr{A}} E_{t, w, s}\left\{V\left(t+h, W_{t+h}^{\pi}, S_{t+h}\right)\right\} \tag{6}
\end{equation*}
$$

- Since $g\left(S_{T}\right)$ involves all tradeable assets, and the benefit is paid at a fixed terminal time $T$, one can consider $g\left(S_{T}\right)$ as a contingent claim, and determine its present value by

$$
c(t, s)=E^{Q}\left\{e^{-r(T-t)} g\left(S_{T}\right) \mid S_{t}=s\right\}
$$

- If the death occurs during $[t, t+h]$, then one can set aside the amount of $c\left(t+h, S_{t+h}\right)$ at time $t+h$ to hedge the potential claim lost $g\left(S_{T}\right)$, and consider the remaining optimization problem on $[t+h, T]$ as if there were no insurance involved. Thus,

$$
\begin{aligned}
& E_{t, w, s}\left\{V\left(t+h, W_{t+h}^{\pi}, S_{t+h}\right)\right\} \\
& =E_{t, w, s}\left\{V^{0}\left(t+h, W_{t+h}^{\pi}-c\left(t+h, S_{t+h}\right)\right)\right\}
\end{aligned}
$$

## Solution for $g=g\left(S_{T}\right)$

- Now for any $\pi$ on $[t, t+h]$,

$$
\begin{aligned}
V(t, w, s) \geq & E_{t, w, s}\left\{V\left(t+h, W_{t+h}^{\pi}, S_{t+h}\right)\right\}_{h} p_{x+t} \\
& +E_{t, w, s}\left\{V^{0}\left(t+h, W_{t+h}^{\pi}-c\left(t+h, S_{t+h}\right)\right)\right\}_{h} q_{x+t}
\end{aligned}
$$

- Assume that $c(\cdot, \cdot) \in C^{1,2}$ and satisfies the Black-Scholes PDE, we can apply Itô to both $V\left(W_{t}, t, S_{t}\right)$ and $V^{0}\left(W_{t}-c\left(t, S_{t}\right), t\right)$ from $t$ to $t+h$, and then take conditional expectations and rearrange terms to obtain

$$
\begin{aligned}
& V(w, t, s) \frac{h q_{x+t}}{h} \geq V^{0}(w-c(t, s), t) \frac{h q_{x+t}}{h} \\
& +E\left\{\frac{1}{h} \int_{t}^{t+h}\left\{V_{t}+\mathscr{L}[V]\left(u, W_{u}, S_{u}\right) \mid W_{t}=w\right\} h p_{x+t}\right. \\
& +E\left\{\frac{1}{h} \int_{t}^{t+h}\left\{V_{t}^{0}+\mathscr{L}\left[V^{0}\right]\left(r, W_{u}, S_{u}\right) \mid W_{t}=w\right\}{ }_{h} q_{x+t} .\right.
\end{aligned}
$$

## Solution for $g=g\left(S_{T}\right)$

- Letting $h \rightarrow 0$, noting that

$$
\lim _{h \rightarrow 0} h q_{x+t} / h=\lambda_{x}(t), \lim _{h \rightarrow 0}{ }_{h} p_{x+t}=1, \quad \lim _{h \rightarrow 0} h q_{x+t}=0
$$

and using the fact that $c$ satisfies the Black-Scholes PDE, we obtain the HJB Equation for $V$ :

$$
\left\{\begin{array}{l}
0=V_{t}+\max _{\pi}\left\{(\mu-r) \pi V_{w}+\frac{1}{2} \sigma^{2} \pi^{2} V_{w w}+s \sigma^{2} \pi V_{w s}\right\}+r w V_{w} \\
\quad+s \mu V_{s}+\frac{1}{2} \sigma^{2} s^{2} V_{s s}+\lambda_{x}(t)\left(V^{0}(w-c, t)-V(w, t, s)\right) \\
V(T, w, s)=u(w)
\end{array}\right.
$$

## Solution for $g=g\left(S_{T}\right)$

- Letting $h \rightarrow 0$, noting that

$$
\lim _{h \rightarrow 0} h q_{x+t} / h=\lambda_{x}(t), \lim _{h \rightarrow 0}{ }_{h} p_{x+t}=1, \quad \lim _{h \rightarrow 0} h q_{x+t}=0
$$

and using the fact that $c$ satisfies the Black-Scholes PDE, we obtain the HJB Equation for $V$ :

$$
\left\{\begin{aligned}
& 0= V_{t}+\max _{\pi}\left\{(\mu-r) \pi V_{w}+\frac{1}{2} \sigma^{2} \pi^{2} V_{w w}+s \sigma^{2} \pi V_{w s}\right\}+r w V_{w} \\
&+s \mu V_{s}+\frac{1}{2} \sigma^{2} s^{2} V_{s s}+\lambda_{x}(t)\left(V^{0}(w-c, t)-V(w, t, s)\right) \\
& V(T, w, s)=u(w)
\end{aligned}\right.
$$

Note: In the Black-Scholes world, the HJB equation for $V^{0}$ is

$$
\left\{\begin{array}{l}
V_{t}^{0}+\max _{\pi \in \mathbb{R}_{+}}\left\{\frac{1}{2}|\sigma \pi|^{2} V_{w w}^{0}+\langle\pi, \mu-r\rangle V_{w}^{0}\right\}+r w V_{w}^{0}=0  \tag{7}\\
V^{0}(T, w)=u(w)
\end{array}\right.
$$

## The Case of Exponential Utility

Consider now the case of exponential utility. I.e., $u(w)=-\frac{1}{\alpha} e^{-\alpha w}$.

- $V^{0}$ has the close form solution:

$$
\begin{equation*}
V^{0}(t, w)=-\frac{1}{\alpha} \exp \left\{-\alpha w e^{r(T-t)}-\frac{(\mu-r)^{2}}{2 \sigma^{2}}(T-t)\right\} \tag{8}
\end{equation*}
$$

## The Case of Exponential Utility

Consider now the case of exponential utility. I.e., $u(w)=-\frac{1}{\alpha} e^{-\alpha w}$.

- $V^{0}$ has the close form solution:

$$
\begin{equation*}
V^{0}(t, w)=-\frac{1}{\alpha} \exp \left\{-\alpha w e^{r(T-t)}-\frac{(\mu-r)^{2}}{2 \sigma^{2}}(T-t)\right\} \tag{8}
\end{equation*}
$$

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

$$
\begin{aligned}
& \Phi_{t}+r S \Phi_{s}+\frac{\sigma^{2} s^{2} \Phi_{s s}}{2}-\frac{s^{2} \sigma^{2} \Phi_{s}^{2}}{2 \Phi}+\lambda_{x}\left(e^{\left\{c \alpha e^{r(T-t)}\right\}}-\Phi\right)=0 \\
& \Phi(T, s)=1
\end{aligned}
$$

## The Case of Exponential Utility

Consider now the case of exponential utility. I.e., $u(w)=-\frac{1}{\alpha} e^{-\alpha w}$.

- $V^{0}$ has the close form solution:

$$
\begin{equation*}
V^{0}(t, w)=-\frac{1}{\alpha} \exp \left\{-\alpha w e^{r(T-t)}-\frac{(\mu-r)^{2}}{2 \sigma^{2}}(T-t)\right\} \tag{8}
\end{equation*}
$$

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

$$
\begin{aligned}
& \Phi_{t}+r S \Phi_{s}+\frac{\sigma^{2} s^{2} \Phi_{s s}}{2}-\frac{s^{2} \sigma^{2} \Phi_{s}^{2}}{2 \Phi}+\lambda_{x}\left(e^{\left\{c \alpha e^{r(T-t)}\right\}}-\Phi\right)=0 \\
& \Phi(T, s)=1
\end{aligned}
$$

- Define $h(t, s)=c(t, s) \alpha e^{r(T-t)}-\ln \Phi$. Then one shows that

$$
\left\{\begin{array}{l}
h_{t}+s r h_{s}+\frac{1}{2} \sigma^{2} s^{2} h_{s s}-\lambda_{x}(t)\left(e^{h}-1\right)=0  \tag{9}\\
h(T, s)=\alpha g(s)
\end{array}\right.
$$

## The Case of Exponential Utility

- If we change the variable: $v=\log s, \tau=T-t$, (9) becomes:

$$
\left\{\begin{array}{l}
h_{\tau}=\left(r-\frac{1}{2} \sigma^{2}\right) h_{v}+\frac{1}{2} \sigma^{2} h_{v v}-\lambda_{x}(T-\tau)\left(e^{h}-1\right)  \tag{10}\\
h(0, v)=\alpha g\left(e^{v}\right)
\end{array}\right.
$$

Note: The reaction-diffusion PDE (10) has a exponential growth, and we must show that it does not blow-up in finite time!

## The Case of Exponential Utility

- If we change the variable: $v=\log s, \tau=T-t$, (9) becomes:

$$
\left\{\begin{array}{l}
h_{\tau}=\left(r-\frac{1}{2} \sigma^{2}\right) h_{v}+\frac{1}{2} \sigma^{2} h_{v v}-\lambda_{x}(T-\tau)\left(e^{h}-1\right)  \tag{10}\\
h(0, v)=\alpha g\left(e^{v}\right)
\end{array}\right.
$$

Note: The reaction-diffusion PDE (10) has a exponential growth, and we must show that it does not blow-up in finite time!

- Now consider the Initial-Boundary value version of (10) with

$$
h(0, x)=\alpha g(x), \quad h(t, \pm N)=\alpha g( \pm N)
$$

and denote its solution by $h^{N}(t, x)$.

- Define $\tilde{K}=|\alpha|\|g\|_{\infty}$, and let

$$
K \triangleq-\log \left(1-\left(1-e^{-\tilde{K}}\right) e^{\int_{0}^{T} \lambda(u) d u}\right)
$$

## The Case of Exponential Utility

- Consider the function

$$
\beta_{K}(t) \triangleq-\log \left\{1-\left(1-e^{-K}\right) e^{-\int_{0}^{t} \lambda(u) d u}\right\}, t \geq 0
$$

Since $\beta_{K}(t)$ is decreasing in $t$, we have

$$
\tilde{K}=\beta_{K}(T) \leq \beta_{K}(t) \leq \beta_{K}(0)=K, \quad \forall t \in[0, T] .
$$

- It can be easily checked that $h(t, x) \triangleq \beta_{K}(t)$, solves (10) with the Initial-Boundary value:

$$
\begin{equation*}
h(0, x)=K, \quad h(t, \pm N)=\beta_{K}(t) \tag{11}
\end{equation*}
$$

- Thus by Comparison Theorem of PDE $h^{N}(\cdot, \cdot)$ is bounded by $\beta_{\tilde{K}}(\cdot)$.


## The Case of Exponential Utility

- Similarly, denote $v^{N}(\tau, x)=\partial_{x} h^{N}(\tau, x)$, and apply the Comparison Theorem to $v^{N}$ one sees that $v^{N}(\cdot, \cdot)$ is bounded by the function $\tilde{v}(t, x)=K^{\prime} e^{\int_{t}^{T} \lambda(t) d t}$, with $K^{\prime}=|\alpha|\left\|g^{\prime}\right\|_{\infty}$.
- We can now apply the Arzela-Ascoli Theorem to obtain a uniformly bounded solution of the Cauchy problem by letting $N \rightarrow \infty$ !
- The indifference price of the UVL insurance is given by

$$
p=c(0, s)-\frac{h(0, s)}{\alpha} e^{-r T},
$$

Note:
Since $Z$ is non-tradable, this is an "incomplete market" case and the arbitrage free price for the payoff $g\left(S_{T}, Z_{T}\right)$ cannot be determined as in the previous case.

## Note:

Since $Z$ is non-tradable, this is an "incomplete market" case and the arbitrage free price for the payoff $g\left(S_{T}, Z_{T}\right)$ cannot be determined as in the previous case.

## A Dynamic Strategy

We consider the following more aggressive (or adventurous) strategy:

- Assuming that the death of the insured occurs before $t+h$
- Instead of putting aside a certain amount of money at the $t+h$ to hedge the future claim, the insurer simply continue to invest all of his current wealth freely, but knowing that he is liable to pay $g\left(S_{T}, Z_{T}\right)$ at time $T$.
- Consider an auxiliary control problem assuming death happens before $T$

$$
\tilde{J}(t, x, s, z ; \pi) \triangleq E_{t, x, s, z}\left\{u\left(X_{T}^{\pi}\right)-g\left(S_{T}, Z_{T}\right)\right\}
$$

with the corresponding value function $\tilde{U}(t, x, s, z)$.

- Then $U$ satisfies a HJB equation: (assuming $\mu=r$ )

$$
\left\{\begin{aligned}
& 0= U_{t}+\max _{\pi}\left\{\frac{1}{2} \sigma^{2} \pi^{2} U_{w w}+\left(U_{w s} S \sigma^{2}+U_{w z} Z \sigma^{Z} \sigma\right) \pi\right\} \\
&+r w U_{w}+U_{s} S \mu+U_{z} Z \mu^{Z}+\frac{1}{2} \sigma^{2} U_{s s} S^{2} \\
&+\frac{1}{2} U_{z z} Z^{2}\left(\tilde{\sigma}^{2}+\sigma^{Z^{2}}\right)+U_{s z} S Z \sigma \sigma^{Z}+\lambda_{x}(t)(\tilde{U}-U) \\
& U(w, T, s, z)=u(w)
\end{aligned}\right.
$$

where $\tilde{U}$ satisfies a similar HJB equation with $\lambda_{x} \equiv 0$.

Using the similar techniques as before, modulo the technicalities of showing the no blow-ups, we can derive the indifference price in this case:

- The premium $p(t, s, z)=\frac{1}{\alpha} e^{-r(T-t)} h(T-t, \log s, \log z)$,
- $h$ is a bounded, classical solution to the PDE

$$
\left\{\begin{array}{l}
h_{\tau}-\frac{1}{2} \tilde{\sigma}^{2} h_{y_{2}}^{2}-\frac{1}{2} \sigma^{2} h_{y_{1} y_{1}}-\frac{1}{2}\left(\tilde{\sigma}^{2}+\sigma^{z 2}\right) h_{y_{2} y_{2}}-\sigma \sigma^{z} h_{y_{1} y_{2}} \\
-\left(r-\frac{1}{2} \sigma^{2}\right) h_{y_{1}}-\left(\mu^{z}-\frac{\mu-r}{\sigma} \sigma^{z}-\frac{\tilde{\sigma}^{2}+\sigma^{22}}{2}\right) h_{y_{2}} \\
-\lambda_{x}(T-\tau)\left(e^{\tilde{h}-h}-1\right)=0 \\
h\left(0, y_{1}, y_{2}\right)=0
\end{array}\right.
$$

and $\tilde{h}$ is a bounded, classical solution to a similar PDE as above, with $\lambda_{x} \equiv 0$, and $\tilde{h}\left(0, y_{1}, y_{2}\right)=\alpha g\left(e^{y_{1}}, e^{y_{2}}\right)$.

## Multiple-decrement Case

## Main Features

- Allowing "multiple decrement": such as short/long term disabilities, withdrawl, retirement, death, etc.
- benefit payable at a random time, e.g., "moment of death".
- the payments may depend on the different status as well as the transitions between them.


## Multiple-decrement Case

## Main Features

- Allowing "multiple decrement": such as short/long term disabilities, withdrawl, retirement, death, etc.
- benefit payable at a random time, e.g., "moment of death".
- the payments may depend on the different status as well as the transitions between them.


## The State/Status Process $\left\{X_{t}\right\}_{t \geq 0}$

- A Markov chain with finite state space $\{0,1, \ldots, m\}$, representing the numerical code of the "status".
- $i=1$ to be the "cemetary state" (death), and $X_{0}=0$
- denote $I_{t}^{i}=\mathbf{1}_{\left\{X_{t}=i\right\}}$ to be the "status indicator" and define the counting process
$N_{t}^{i j} \triangleq \#\{$ transitions of $X$ from state $i$ to $j$ during $[0, t]\}$.


## Multiple-decrement Case

## Some Important Quantities

- for each $t$, denote $\tau_{t}=\inf \left\{s \geq t: X_{s} \neq X_{t}\right\}$; and for $i=0, \ldots, m$, define $\tau_{t}^{i}=\tau_{t}$, if $X_{\tau_{t}}=i$ and $\infty$ otherwise.
- ${ }_{t} \bar{p}_{s}^{i} \triangleq P\left\{\tau_{s}>t \mid X_{s}=i\right\}$;
- ${ }_{t} \bar{q}_{s}^{i j} \triangleq P\left\{\tau_{s}^{j}=\tau_{s} \leq t \mid X_{s}=i\right\}, s \leq t, i, j \in\{0, \ldots, m\}$.
- Clearly, ${ }_{t} \bar{p}_{s}^{1}=1 ;{ }_{t} \bar{q}_{s}^{1 j}=0$, for all $j \neq 1$; and

$$
\begin{equation*}
{ }_{t} \bar{p}_{s}^{i}+\sum_{j \neq i} t \bar{q}_{s}^{i j}=1, \quad \forall i=0,1, \cdots, m, \quad 0 \leq s<t \tag{12}
\end{equation*}
$$

- "force of decrement of status i due to cause $j$ " as

$$
\begin{equation*}
\bar{\lambda}_{t}^{i j} \triangleq \lim _{h \rightarrow 0} \frac{t+h \bar{q}_{t}^{i j}}{h}, \quad i, j=0,1, \cdots m \tag{13}
\end{equation*}
$$

## Some Remarks

- If $m=1$, then the state process $X$ becomes the one as in the simple life model, and $\tau_{0}^{1}=T(x)$. In that case we should have

$$
{ }_{t} \bar{p}_{s}^{0}={ }_{t-s} p_{x+s}, \quad{ }_{t} q_{s}^{01}={ }_{t-s} q_{x+s}
$$

- Being a Markov chain, the process $X$ has its transition probability and the corresponding transition intensity

$$
{ }_{t} q_{s}^{i j}=P\left\{X_{t}=j \mid X_{s}=i\right\} ; \quad \lambda_{t}^{i j} \triangleq \lim _{h \downarrow 0} \frac{t+h q_{t}^{i j}}{h}, \quad i \neq j
$$

There are natural links between $p^{i j}$ 's and $\bar{p}^{i j}$ 's. For example:

- $\bar{\lambda}_{t}^{i j}=\lambda_{t}^{i j}$, for all $t \geq 0, i, j=0,1, \cdots, m$;

$$
\begin{aligned}
& -{ }_{t+h} \bar{p}_{t}^{i}=\exp \left\{-\int_{t}^{t+h} \sum_{j \neq i} \lambda_{s}^{i j} d s\right\} ;{ }_{t+h} p_{t}^{i j}=\int_{t}^{t+h}{ }_{\tau} \bar{p}_{t}^{i} \lambda_{\tau}^{i j} d \tau, \\
& \forall h>0, i, j=0, \cdots, m .
\end{aligned}
$$

## The Payment Process $A_{t}$ :

- Two types of payments will be considered: "life-annuity" and "life-insurance".
- Since the non-tradability of the asset $Z$ will not make significant difference in the optimization problem, we will not distinguish $Z$ from $S$.
- The cumulative payment process is defined by

$$
\begin{equation*}
A_{t}=\int_{0}^{t} \sum_{i} I_{u}^{i} a^{i}\left(u, S_{u}\right) d u+\sum_{i \neq j} a^{i j}\left(u, S_{u}\right) d N_{u}^{i j}, \quad t \geq 0 \tag{14}
\end{equation*}
$$

- an F-adapted, càdlàg, non-decreasing process in which
- $a^{i}(t, s)$ - rate of payments of annuity at state $i$, given $S_{t}=s$;
- $a^{i j}(t, s)$ - rate of payments of insurance when transit from state $i$ to $j$, given $S_{t}=s$.


## Dynamics of General Reserve

Dynamics of general reserve

$$
d \hat{W}_{t}^{\pi}=\left[r_{t} \hat{W}_{t}^{\pi}+\pi_{t}\left(\mu_{t}-r_{t}\right)\right] d t+\pi_{t} \sigma_{t} d B_{t}-d A_{t}
$$

where

- $d A_{t}=\sum_{i} I_{i}(t) a^{i}\left(t, S_{t}\right) d t+\sum_{i \neq j} a^{i j}\left(t, S_{t}\right) d N_{t}^{i j}$
- $I_{t}^{i}=\mathbf{1}_{\left\{X_{t}=i\right\}}, N_{t}^{i j} \triangleq \#\{$ jumps of $X$ from $i$ to $j$ during $[0, t]\}$


## Dynamics of General Reserve

## Dynamics of general reserve

$$
d \hat{W}_{t}^{\pi}=\left[r_{t} \hat{W}_{t}^{\pi}+\pi_{t}\left(\mu_{t}-r_{t}\right)\right] d t+\pi_{t} \sigma_{t} d B_{t}-d A_{t}
$$

where

- $d A_{t}=\sum_{i} I_{i}(t) a^{i}\left(t, S_{t}\right) d t+\sum_{i \neq j} a^{i j}\left(t, S_{t}\right) d N_{t}^{i j}$
- $I_{t}^{i}=\mathbf{1}_{\left\{X_{t}=i\right\}}, N_{t}^{i j} \triangleq \#\{$ jumps of $X$ from $i$ to $j$ during $[0, t]\}$


## Hamiltonian

$$
\left\{\begin{aligned}
\mathscr{H}^{k} \triangleq \frac{1}{2}\left|\sigma_{t} \pi\right|^{2} \psi+ & {\left[\left\langle\pi, \mu_{t}-r_{t} \mathbf{1}\right\rangle+r_{t} w-a^{k}(t, s)\right] \varphi } \\
& +\left\langle\pi, \sigma_{t} \sigma_{t}^{T} \operatorname{tr} D[s] p\right\rangle, \quad k=0,1, \cdots, m \\
H^{k}(t, w, s, \varphi, \psi, p) \triangleq & \sup _{\pi} \mathscr{H}^{k}(t, w, s, \varphi, \psi, p ; \pi)
\end{aligned}\right.
$$

## The HJB Equation

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

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

$$
\left\{\begin{array}{l}
U_{t}^{k}+F_{k}\left(t, w, s, D U^{k}, D^{2} U^{k}\right)+\left(\mathscr{H}_{k} U\right)=0  \tag{15}\\
U^{k}(T, w, s)=u(w), \quad k=0, \cdots, m
\end{array}\right.
$$

where

$$
\begin{aligned}
F_{k}(\cdots)= & \sup _{\pi \in \Pi}\left\{\pi\left(\mu_{t}-r_{t}\right) U_{w}^{k}+\frac{1}{2}\left|\sigma_{t} \pi\right|^{2} U_{w w}^{k}+\pi \sigma_{t}^{2} s U_{w s}^{k}\right\} \\
& +\mu_{t} s U_{s}^{k}+\frac{1}{2} \sigma_{t}^{2} s^{2} U_{s s}^{k}+\left(r_{t} w-a^{k}(t, s)\right) U_{w}^{k} \\
\left(\mathscr{H}_{k} U\right)= & \sum_{j \neq k} \lambda_{t}^{k j}\left(U^{j}\left(t, w-a^{k j}(t, s), s\right)-U^{k}(t, w, s)\right) .
\end{aligned}
$$

## Viscosity Solution for System of PDDEs

## Main Difficulties

- Definition of viscosity solution for the system of PDDE.
- Uniqueness
- Different from Ishii et al.'s results: Parabolic PDDE vs. Elliptic PDEs
- Different from Pardoux et al.'s results: Fully Nonlinear System vs. Semilinear System


## Viscosity Solution for System of PDDEs

## Main Difficulties

- Definition of viscosity solution for the system of PDDE.
- Uniqueness
- Different from Ishii et al.'s results: Parabolic PDDE vs. Elliptic PDEs
- Different from Pardoux et al.'s results: Fully Nonlinear System vs. Semilinear System


## Main idea:

- Taking the index vector of the value function as an additional "spatial" variable with values in a finite set: the system of PDDEs becomes a single PDDE!
- The abstract framework of viscosity solutions (e.g., Fleming \& Soner book) applies!


## Abstract Dynamic Programming Principle Revisited

## Recall Fleming-Soner (II.3)

- $\Sigma$ - a closed subset of a Banach space
- $\mathscr{C}$ - a collection of functions on $\Sigma$
- $\mathscr{T}_{t r}, 0 \leq t \leq r \leq T$ - a family of operators on $\mathscr{C}$, s.t.,
(i) $\mathscr{T}_{t t} \varphi=\varphi$;
(iia) $\mathscr{T}_{t r} \varphi \leq \mathscr{T}_{t s} \psi, \quad$ if $\varphi \leq\left(\mathscr{T}_{r s} \psi\right), \quad \forall 0 \leq t \leq r \leq s$;
(iib) $\mathscr{T}_{t r} \varphi \geq \mathscr{T}_{t s} \psi$, if $\varphi \geq\left(\mathscr{T}_{r s} \psi\right), \quad \forall 0 \leq t \leq r \leq s$.


## Note

- $r=s$ in $(\mathrm{ii}) \Longrightarrow$ monotonicity: $\mathscr{T}_{t r} \varphi \leq \mathscr{T}_{\operatorname{tr}} \psi$, if $\varphi \leq \psi$,
- (iia) $\oplus(\mathrm{iib}) \Longrightarrow$ semigroup property:

$$
\mathscr{T}_{t s} \varphi=\mathscr{T}_{t r}\left(\mathscr{T}_{r s} \varphi\right), t \leq r \leq s \leq T, \quad \text { if } \mathscr{T}_{t r} \varphi \in \mathscr{C}, \forall \varphi \in \mathscr{C} .
$$

Of course, the fact that $T_{t r} \varphi \in \mathscr{C}$ must be verified!

## Abstract Bellman (Dynamic Programming) Principle

- $\Sigma \subseteq \overline{\mathscr{O}}$, where $\mathscr{O}$ is an open set in $\mathbb{R}^{n}$, and $\mathscr{C}=\mathscr{M}(\Sigma)$,
- $T_{t, r ; u} \psi(x) \triangleq J(t, r ; u)=E_{t, x}\left\{\int_{t}^{r} L\left(s, X_{s}, u_{s}\right) d s+\psi\left(X_{r}\right)\right\}$.
- $\mathscr{T}_{t, r} \psi(x) \triangleq \inf _{u \in \mathscr{U}}^{a d}$ $T_{t, r ; u} \psi(x)$ (Thus, $T_{t, T} \psi(x)=V(t, x)!$ ).


## Abstract Bellman (Dynamic Programming) Principle

- $\Sigma \subseteq \overline{\mathscr{O}}$, where $\mathscr{O}$ is an open set in $\mathbb{R}^{n}$, and $\mathscr{C}=\mathscr{M}(\Sigma)$,
- $T_{t, r ; u} \psi(x) \triangleq J(t, r ; u)=E_{t, x}\left\{\int_{t}^{r} L\left(s, X_{s}, u_{s}\right) d s+\psi\left(X_{r}\right)\right\}$.
- $\mathscr{T}_{t, r} \psi(x) \triangleq \inf _{u \in \mathscr{U}}^{a d}$ $T_{t, r ; u} \psi(x)$ (Thus, $T_{t, T} \psi(x)=V(t, x)!$ ).


## Note

Semigroup Property $=($ Abstract $)$ Bellman Principle(!)

## Abstract Bellman (Dynamic Programming) Principle

- $\Sigma \subseteq \overline{\mathscr{O}}$, where $\mathscr{O}$ is an open set in $\mathbb{R}^{n}$, and $\mathscr{C}=\mathscr{M}(\Sigma)$,
- $T_{t, r ; u} \psi(x) \triangleq J(t, r ; u)=E_{t, x}\left\{\int_{t}^{r} L\left(s, X_{s}, u_{s}\right) d s+\psi\left(X_{r}\right)\right\}$.
- $\mathscr{T}_{t, r} \psi(x) \triangleq \inf _{u \in \mathscr{U}_{a d}} T_{t, r ; u} \psi(x)$ (Thus, $\left.T_{t, T} \psi(x)=V(t, x)!\right)$.


## Note

## Semigroup Property $=($ Abstract $)$ Bellman Principle(!)

- Let $\left\{\mathscr{G}_{t}\right\}_{t \geq 0}$ be the "infinitesimal generator" of the semigroup $\mathscr{T}$, that is, for all $\varphi \in \mathscr{D}, y \in \Sigma$,

$$
\lim _{h \downarrow 0} \frac{1}{h}\left\{\left(\mathscr{T}_{t t+h} \varphi(t+h, \cdot)\right)(y)-\varphi(t, y)\right\}=\left[\frac{\partial}{\partial t}+\mathscr{G}_{t}\right] \varphi(t, y)
$$

- where $\mathscr{D} \subset C([0, T) \times \Sigma)$ is the set of "test functions" [i.e., $\forall \varphi \in \mathscr{D}, \frac{\partial}{\partial t} \varphi(t, y)$ and $\left(\mathscr{G}_{t} \varphi(t, \cdot)\right)(y)$ are continuous.]


## Abstract form of HJB Equation

Assume $V \in C^{1,2} \subset \mathscr{D}$. Then use the semigroup property one derives the HJB equation:

$$
\left\{\begin{align*}
0 & =\lim _{h \downarrow 0} \frac{1}{h}\left\{\left(\mathscr{T}_{t t+h} V(t+h, \cdot)\right)(y)-V(t, y)\right\}  \tag{16}\\
& =\left[\frac{\partial}{\partial t}+\mathscr{G}_{t}\right] V(t, y), \quad \forall y \in \Sigma \\
V & (T, y)=\psi(y)
\end{align*}\right.
$$

## Abstract form of HJB Equation

Assume $V \in C^{1,2} \subset \mathscr{D}$. Then use the semigroup property one derives the HJB equation:

$$
\left\{\begin{array}{l}
0=\lim _{h \downarrow 0} \frac{1}{h}\left\{\left(\mathscr{T}_{t t+h} V(t+h, \cdot)\right)(y)-V(t, y)\right\}  \tag{16}\\
\quad=\left[\frac{\partial}{\partial t}+\mathscr{G}_{t}\right] V(t, y), \quad \forall y \in \Sigma \\
V(T, y)=\psi(y)
\end{array}\right.
$$

## Theorem (Fleming-Soner, Theorem II.5.1)

If the value function of a control problem $V \in C[0, T] \times \Sigma)$, then $V$ is a viscosity solution to the (abstract) HJB equation (16).

## Abstract form of HJB Equation

Assume $V \in C^{1,2} \subset \mathscr{D}$. Then use the semigroup property one derives the HJB equation:

$$
\left\{\begin{align*}
0 & =\lim _{h \downarrow 0} \frac{1}{h}\left\{\left(\mathscr{T}_{t t+h} V(t+h, \cdot)\right)(y)-V(t, y)\right\} \\
& =\left[\frac{\partial}{\partial t}+\mathscr{G}_{t}\right] V(t, y), \quad \forall y \in \Sigma  \tag{16}\\
V & (T, y)=\psi(y)
\end{align*}\right.
$$

## Theorem (Fleming-Soner, Theorem II.5.1)

If the value function of a control problem $V \in C[0, T] \times \Sigma)$, then $V$ is a viscosity solution to the (abstract) HJB equation (16).

## Question:

What are $\mathscr{G}, \mathscr{D}, \ldots$, etc. in our case?

## Back to UVL Model

- $\Sigma=\{(w, s, k): w, s \in \mathbb{R}, k \in\{0,1, \ldots, m\}$,
- $\mathscr{C}=C(\Sigma)$.
- $\left(\mathscr{T}_{t r} \varphi\right)(w, s, k) \triangleq \sup _{\pi \in \mathscr{A}} E_{w, s, k}\left\{\varphi\left(\hat{W}_{r}^{\pi}, S_{r}, X_{r}\right)\right\}, \quad t \geq r$
- $\left(\mathscr{T}_{t T} u\right)(w, s, k)=U^{k}(t, w, s), \quad \forall(t, w, s)$ and $k$


## Back to UVL Model

- $\Sigma=\{(w, s, k): w, s \in \mathbb{R}, k \in\{0,1, \ldots, m\}$,
- $\mathscr{C}=C(\Sigma)$.
- $\left(\mathscr{T}_{t r} \varphi\right)(w, s, k) \triangleq \sup _{\pi \in \mathscr{A}} E_{w, s, k}\left\{\varphi\left(\hat{W}_{r}^{\pi}, S_{r}, X_{r}\right)\right\}, \quad t \geq r$
- $\left(\mathscr{T}_{t} T u\right)(w, s, k)=U^{k}(t, w, s), \quad \forall(t, w, s)$ and $k$


## Note

- It is easy to check that the family $\left\{\mathscr{T}_{t r}\right\}$ satisfies (i), (ii).
- Since $U^{k}(t, w, s)$ 's are all continuous, the function $(t, w, s, k) \mapsto U^{k}(t, w, s)$ (on $\Sigma$ ) should satisfy an abstract HJB equation!


## Back to UVL Model

- $\Sigma=\{(w, s, k): w, s \in \mathbb{R}, k \in\{0,1, \ldots, m\}$,
- $\mathscr{C}=C(\Sigma)$.
- $\left(\mathscr{T}_{t r} \varphi\right)(w, s, k) \triangleq \sup _{\pi \in \mathscr{A}} E_{w, s, k}\left\{\varphi\left(\hat{W}_{r}^{\pi}, S_{r}, X_{r}\right)\right\}, \quad t \geq r$
- $\left(\mathscr{T}_{t} T u\right)(w, s, k)=U^{k}(t, w, s), \quad \forall(t, w, s)$ and $k$


## Note

- It is easy to check that the family $\left\{\mathscr{T}_{t r}\right\}$ satisfies (i), (ii).
- Since $U^{k}(t, w, s)$ 's are all continuous, the function $(t, w, s, k) \mapsto U^{k}(t, w, s)$ (on $\Sigma$ ) should satisfy an abstract HJB equation!


## Problems:

- Identify the infinitesimal generator of the semigroup $\mathscr{T}$.
- Define the "viscosity solutions" to the corresponding abstract HJB equation (vs. the system of the HJB equations!)


## Abstract HJB Equation vs. System of PDDEs

Denote $U(t, w, s, k)=U^{k}(t, w, s)$, and recall the PDDEs (15):

$$
\begin{cases}\frac{\partial}{\partial t} U^{k}+F_{k}\left(t, w, s, D U^{k}, D^{2} U^{k}\right)+\left(\mathscr{H}_{k} U\right)(t, w, s)=0  \tag{17}\\ U^{k}(T, w, s)=u(w), & k=0, \cdots, m .\end{cases}
$$

## Abstract HJB Equation vs. System of PDDEs

Denote $U(t, w, s, k)=U^{k}(t, w, s)$, and recall the PDDEs (15):

$$
\begin{cases}\frac{\partial}{\partial t} U^{k}+F_{k}\left(t, w, s, D U^{k}, D^{2} U^{k}\right)+\left(\mathscr{H}_{k} U\right)(t, w, s)=0  \tag{17}\\ U^{k}(T, w, s)=u(w), & k=0, \cdots, m\end{cases}
$$

## Theorem

The viscosity solutions of the abstract HJB equation (16) with respect to the operator $\mathscr{T}$ and that of the system of PDDEs (17) are equivalent if and only if

$$
\begin{equation*}
\left(\mathscr{G}_{t} \varphi(t, \cdot)\right)(w, s, k)=\left[F_{k}\left(\cdot, \cdot, \cdot, D \varphi, D^{2} \varphi\right)+\left(\mathscr{H}_{k} \varphi\right)\right](t, w, s) . \tag{18}
\end{equation*}
$$

## Abstract HJB Equation vs. System of PDDEs

Denote $U(t, w, s, k)=U^{k}(t, w, s)$, and recall the PDDEs (15):

$$
\begin{cases}\frac{\partial}{\partial t} U^{k}+F_{k}\left(t, w, s, D U^{k}, D^{2} U^{k}\right)+\left(\mathscr{H}_{k} U\right)(t, w, s)=0  \tag{17}\\ U^{k}(T, w, s)=u(w), & k=0, \cdots, m\end{cases}
$$

## Theorem

The viscosity solutions of the abstract HJB equation (16) with respect to the operator $\mathscr{T}$ and that of the system of PDDEs (17) are equivalent if and only if

$$
\begin{equation*}
\left(\mathscr{G}_{t} \varphi(t, \cdot)\right)(w, s, k)=\left[F_{k}\left(\cdot, \cdot, \cdot, D \varphi, D^{2} \varphi\right)+\left(\mathscr{H}_{k} \varphi\right)\right](t, w, s) . \tag{18}
\end{equation*}
$$

## 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.


## 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.
- $T_{x_{1}}, T_{x_{2}}, \cdots, T_{x_{n}}$ - future life time random variables,
- $T_{m}=T_{x_{1}, \cdots, x_{n}} \triangleq \min \left\{T_{x_{1}}, \cdots, T_{x_{n}}\right\}$ - (Joint-life)
- $T_{M}=T_{\overline{x_{1}, \cdots, x_{n}}} \triangleq \max \left\{T_{x_{1}}, \cdots, T_{x_{n}}\right\}-$ (Last-survivor)


## 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.
- $T_{x_{1}}, T_{x_{2}}, \cdots, T_{x_{n}}$ - future life time random variables,
- $T_{m}=T_{x_{1}, \cdots, x_{n}} \triangleq \min \left\{T_{x_{1}}, \cdots, T_{x_{n}}\right\}-$ (Joint-life)
- $T_{M}=T_{\overline{x_{1}, \cdots, x_{n}}} \triangleq \max \left\{T_{x_{1}}, \cdots, T_{x_{n}}\right\}-$ (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 $F_{M}(t)=F_{T_{x_{1}}}(t) F_{T_{x_{2}}}(t) \ldots$

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

$$
\left\{\begin{array}{l}
\mu_{x_{1}}(t)=\lambda_{x_{1}}(t)+\mathbf{1}_{\left\{T_{x_{2}} \leq t\right\}} \gamma_{x_{1}}\left(t-T_{x_{2}}\right)  \tag{19}\\
\mu_{x_{2}}(t)=\lambda_{x_{2}}(t)+\mathbf{1}_{\left\{T_{x_{1}} \leq t\right\}} \gamma_{x_{2}}\left(t-T_{x_{1}}\right),
\end{array} \quad t \geq 0,\right.
$$

where $\lambda_{x_{i}}$ 's are the (marginal) force of mortality and

$$
\gamma_{x_{i}}(t)=\frac{n_{i}}{r_{i} e^{t}+1}, \quad i=1,2, \quad r_{1}, r_{2}, n_{1}, n_{2}>0
$$

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

$$
\left\{\begin{array}{l}
\mu_{x_{1}}(t)=\lambda_{x_{1}}(t)+\mathbf{1}_{\left\{T_{x_{2}} \leq t\right\}} \gamma_{x_{1}}\left(t-T_{x_{2}}\right)  \tag{19}\\
\mu_{x_{0}}(t)=\lambda_{x_{x_{2}}}(t)+\mathbf{1}_{\{\tau}(t\}, \gamma_{x_{0}}\left(t-T_{x_{1}}\right),
\end{array} \quad t \geq 0,\right.
$$

where $\lambda_{x_{i}}$ 's are the (marginal) force of mortality and

$$
\gamma_{x_{i}}(t)=\frac{n_{i}}{r_{i} e^{t}+1}, \quad i=1,2, \quad r_{1}, r_{2}, n_{1}, n_{2}>0
$$

## Note:

This essentially becomes a problem of "Counter-Party Risk", a well-know topic in "Contagion Models" of correlated default! Existing literature include

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


## Basic Setup

Let $\left(\Omega, \mathscr{F},\left\{\mathscr{F}_{t}\right\}, \mathbb{P}\right)$ be a given filtered probability space.

- $\mathbb{P}$ is risk neutral (in a default free bond market)
- $\exists$ a factor process $X=\left\{X_{t}: t \geq 0\right\}$
- There are $I$ firms, with default times $\tau^{i}, i=1, \cdots, I$


## Basic Setup

Let $\left(\Omega, \mathscr{F},\left\{\mathscr{F}_{t}\right\}, \mathbb{P}\right)$ be a given filtered probability space.

- $\mathbb{P}$ is risk neutral (in a default free bond market)
- $\exists$ a factor process $X=\left\{X_{t}: t \geq 0\right\}$
- There are $I$ firms, with default times $\tau^{i}, i=1, \cdots, l$

Denote

- $N_{t}^{i} \triangleq \mathbf{1}_{\left\{\tau^{i} \leq t\right\}}-$ default process with respect to $\tau^{i}$,
- $\mathscr{F}_{t} \triangleq \mathscr{F}_{t}^{X} \vee \mathscr{F}_{t}^{1} \vee \ldots \vee \mathscr{F}_{t}^{\prime}$, where $\mathscr{F}_{t}^{i}=\sigma\left\{N_{s}^{i}: 0 \leq s \leq t\right\}$, $\forall i$
- $\mathscr{H}_{t}^{i}=\mathscr{F}_{t}^{X} \vee \mathscr{F}_{t}^{1} \vee \ldots \vee \mathscr{F}_{t}^{i-1} \vee \mathscr{F}_{t}^{i+1} \vee \ldots \vee \mathscr{F}_{t}{ }^{\prime}$, $\Longrightarrow \mathscr{F}_{t}=\mathscr{H}_{t}^{i} \vee \mathscr{F}_{t}^{i}$.


## Basic Setup

## Define

- $S_{t}^{i}=\mathbb{P}\left\{\tau^{i}>t \mid \mathscr{H}_{t}^{i}\right\}>0\left(\Longrightarrow S^{i}\right.$ is an $\mathscr{H}^{i}$-supermg $)$
- $H_{t}^{i} \triangleq-\ln \left(S_{t}^{i}\right), t \geq 0$ - Hazard Process


## Basic Setup

Define

- $S_{t}^{i}=\mathbb{P}\left\{\tau^{i}>t \mid \mathscr{H}_{t}^{i}\right\}>0\left(\Longrightarrow S^{i}\right.$ is an $\mathscr{H}^{i}$-supermg $)$
- $H_{t}^{i} \triangleq-\ln \left(S_{t}^{i}\right), t \geq 0$ - Hazard Process

Note:

- $S_{t}^{i}>0$ implies that $\tau^{i}$ cannot be an $\mathscr{H}^{i}$-stopping time!


## Basic Setup

Define

- $S_{t}^{i}=\mathbb{P}\left\{\tau^{i}>t \mid \mathscr{H}_{t}^{i}\right\}>0\left(\Longrightarrow S^{i}\right.$ is an $\mathscr{H}^{i}$-supermg $)$
- $H_{t}^{i} \triangleq-\ln \left(S_{t}^{i}\right), t \geq 0$ - Hazard Process


## Note:

- $S_{t}^{i}>0$ implies that $\tau^{i}$ cannot be an $\mathscr{H}^{i}$-stopping time!
- If $\exists \lambda_{t}^{i} \in \mathscr{H}_{t}^{i}$, such that $H_{t}^{i}=\int_{0}^{t} \lambda_{s}^{i} d s, t \geq 0$, then

$$
\begin{equation*}
S_{t}^{i}=\mathbb{P}\left\{\tau^{i}>t \mid \mathscr{H}_{t}^{i}\right\}=\exp \left\{-\int_{0}^{t} \lambda_{s}^{i} d s\right\} \tag{20}
\end{equation*}
$$

- $\lambda^{i}$ is called the (conditional) intensity process of $\tau^{i}$, and it holds that $\lambda_{t}^{i}=-d S_{t}^{i} / S_{t}^{i}, t \geq 0$.


## A Useful Lemma

## Lemma

For any $\mathscr{F}$-measurable random variable $Z$ we have, for any $t \geq 0$,

$$
\begin{equation*}
\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{Z \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} Z \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} \tag{21}
\end{equation*}
$$

## A Useful Lemma

## Lemma

For any $\mathscr{F}$-measurable random variable $Z$ we have, for any $t \geq 0$,

$$
\begin{equation*}
\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{Z \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} Z \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} \tag{21}
\end{equation*}
$$

Idea: Define

$$
\mathscr{F}_{t}^{*} \triangleq\left\{A \in \mathscr{F} \mid \exists B \in \mathscr{H}_{t}^{i}, A \cap\left\{\tau^{i}>t\right\}=B \cap\left\{\tau^{i}>t\right\}\right\} .
$$

Then one can check that $\mathscr{F}_{t}=\mathscr{F}_{t}^{*}, t \geq 0$.

## A Useful Lemma

## Lemma

For any $\mathscr{F}$-measurable random variable $Z$ we have, for any $t \geq 0$,

$$
\begin{equation*}
\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{Z \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} Z \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} \tag{21}
\end{equation*}
$$

Idea: Define

$$
\mathscr{F}_{t}^{*} \triangleq\left\{A \in \mathscr{F} \mid \exists B \in \mathscr{H}_{t}^{i}, A \cap\left\{\tau^{i}>t\right\}=B \cap\left\{\tau^{i}>t\right\}\right\} .
$$

Then one can check that $\mathscr{F}_{t}=\mathscr{F}_{t}^{*}, t \geq 0$.
Applying "Monotone Class", one shows that, $\forall Z \in \mathscr{F}, \exists X \in \mathscr{H}_{t}^{i}$, s.t.

$$
\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} Z \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{Z \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} X
$$

Taking $\mathbb{E}\left\{\cdot \mid \mathscr{H}_{t}^{i}\right\}$ on both sides and solve for $X$.

## The Conditional Survival Probability

Note that $\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{F}_{t}\right\}$. Applying Lemma we have

$$
\begin{equation*}
\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left[\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} . \tag{22}
\end{equation*}
$$

## The Conditional Survival Probability

Note that $\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{F}_{t}\right\}$. Applying Lemma we have

$$
\begin{equation*}
\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left[\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} . \tag{22}
\end{equation*}
$$

Since

- $\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}=\mathbb{E}\left\{\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{H}_{T}^{i}\right\} \mid \mathscr{H}_{t}^{i}\right\}=$ $\mathbb{E}\left\{e^{-\int_{0}^{T} \lambda_{s}^{i} d s} \mid \mathscr{H}_{t}^{i}\right\}$.
- $\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}=e^{-\int_{0}^{t} \lambda_{s}^{i} d s}$


## The Conditional Survival Probability

Note that $\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{F}_{t}\right\}$. Applying Lemma we have

$$
\begin{equation*}
\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left[\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} . \tag{22}
\end{equation*}
$$

Since

- $\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}=\mathbb{E}\left\{\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{H}_{T}^{i}\right\} \mid \mathscr{H}_{t}^{i}\right\}=$

$$
\mathbb{E}\left\{e^{-\int_{0}^{T} \lambda_{s}^{i} d s} \mid \mathscr{H}_{t}^{i}\right\}
$$

- $\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}=e^{-\int_{0}^{t} \lambda_{s}^{i} d s}$

Consequently:

- $\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{e^{-\int_{t}^{T} \lambda_{s}^{i} d s} \mid \mathscr{H}_{t}^{i}\right\}$.


## The Conditional Survival Probability

Note that $\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{F}_{t}\right\}$. Applying Lemma we have

$$
\begin{equation*}
\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \frac{\mathbb{E}\left[\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}} . \tag{22}
\end{equation*}
$$

Since

- $\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>T\right\}} \mid \mathscr{H}_{t}^{i}\right\}=\mathbb{E}\left\{\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{H}_{T}^{i}\right\} \mid \mathscr{H}_{t}^{i}\right\}=$

$$
\mathbb{E}\left\{e^{-\int_{0}^{T} \lambda_{s}^{i} d s} \mid \mathscr{H}_{t}^{i}\right\}
$$

- $\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\}=e^{-\int_{0}^{t} \lambda_{s}^{i} d s}$

Consequently:

- $\mathbb{P}\left\{\tau^{i}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}\left\{e^{-\int_{t}^{T} \lambda_{s}^{i} d s} \mid \mathscr{H}_{t}^{i}\right\}$.
- $M_{t}^{i} \triangleq N_{t}^{i}-H_{t \wedge \tau^{i}}^{i}=\mathbf{1}_{\left\{\tau^{i} \leq t\right\}}-\int_{0}^{t} \mathbf{1}_{\left\{\tau^{i}>s\right\}} \lambda_{s}^{i} d s, i=1, \ldots, l$, are $\left\{\mathscr{F}_{t}\right\}$-martingales.


## Standing Assumptions

(H1) $\lambda_{t}^{i}$ satisfy the following condition:

$$
\mathbb{E}\left\{\exp \left(2 \int_{0}^{t} \sum_{i=1}^{1} \lambda_{s}^{i} d s\right)\right\}<\infty, \quad \forall t<\infty .
$$

(H2) For each $i, \mathbb{P}\left\{\tau^{i}>0\right\}=1$. Furthermore, there are no simultaneous defaults among the $I$ firms. In other words, it holds that $\mathbb{P}\left\{\tau^{i} \neq \tau^{j}\right\}=1$, whenever $i \neq j$.

## Standing Assumptions

(H1) $\lambda_{t}^{i}$ satisfy the following condition:

$$
\mathbb{E}\left\{\exp \left(2 \int_{0}^{t} \sum_{i=1}^{l} \lambda_{s}^{i} d s\right)\right\}<\infty, \quad \forall t<\infty
$$

(H2) For each $i, \mathbb{P}\left\{\tau^{i}>0\right\}=1$. Furthermore, there are no simultaneous defaults among the $I$ firms. In other words, it holds that $\mathbb{P}\left\{\tau^{i} \neq \tau^{j}\right\}=1$, whenever $i \neq j$.

## Main Task

Find effective, tractable way to calculate the joint distribution (survival probability):

$$
\mathbb{P}\left\{\tau^{1} \leq t_{1}, \cdots, \tau^{\prime} \leq t_{l}\right\}, \quad \text { and } / \text { or } \quad \mathbb{P}\left\{\tau^{1}>t_{1}, \cdots, \tau^{\prime}>t_{l}\right\},
$$

given the conditional intensities.

## Representation of Joint Survival Probability

Define, for $i=1, \ldots, I, \Gamma_{t}^{i} \triangleq \exp \left\{\int_{0}^{t} \lambda_{s}^{i} d s\right\}$, and

$$
\begin{equation*}
Z_{t}^{i} \triangleq \mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \exp \left\{\int_{0}^{t} \lambda_{s}^{i} d s\right\} \tag{23}
\end{equation*}
$$

## Representation of Joint Survival Probability

Define, for $i=1, \ldots, I, \Gamma_{t}^{i} \triangleq \exp \left\{\int_{0}^{t} \lambda_{s}^{i} d s\right\}$, and

$$
\begin{equation*}
Z_{t}^{i} \triangleq \mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \exp \left\{\int_{0}^{t} \lambda_{s}^{i} d s\right\} \tag{23}
\end{equation*}
$$

Then

- $Z_{t}^{i} \geq 0$; and $Z_{0}^{i}=1, \forall i$.
- $Z^{i}$ 's are $\left\{\mathscr{F}_{t}\right\}$-adapted, and $\mathbb{E}\left\{Z_{t}^{i}\right\}=1$.


## Representation of Joint Survival Probability

Define, for $i=1, \ldots, I, \Gamma_{t}^{i} \triangleq \exp \left\{\int_{0}^{t} \lambda_{s}^{i} d s\right\}$, and

$$
\begin{equation*}
Z_{t}^{i} \triangleq \mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i}=\mathbf{1}_{\left\{\tau^{i}>t\right\}} \exp \left\{\int_{0}^{t} \lambda_{s}^{i} d s\right\} \tag{23}
\end{equation*}
$$

Then

- $Z_{t}^{i} \geq 0$; and $Z_{0}^{i}=1, \forall i$.
- $Z^{i}$ 's are $\left\{\mathscr{F}_{t}\right\}$-adapted, and $\mathbb{E}\left\{Z_{t}^{i}\right\}=1$.


## Proposition

Assume (H1) and (H2). Then, for $k=1, \ldots, l$, the processes

$$
\begin{equation*}
\prod_{i=1}^{k} Z_{t}^{i} \triangleq \prod_{i=1}^{k} \mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i}, \quad t \geq 0 \tag{24}
\end{equation*}
$$

are all $\left\{\mathscr{F}_{t}\right\}$-martingales.

## Representation of Joint Survival Probability

[Sketch of the proof.] (i) $Z_{t}^{i}$ 's are martingales.

## Representation of Joint Survival Probability

[Sketch of the proof.] (i) $Z_{t}^{i}$ 's are martingales.

$$
\begin{aligned}
\mathbb{E}\left\{Z_{t}^{i} \mid \mathscr{F}_{s}\right\} & =\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\}=\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\} \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\left.\mathbf{1}_{\left\{\tau^{i}>t\right\}}\right|_{t} ^{i} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mid \mathscr{H}_{s}^{i}\right\}} \quad \text { (Lemma) } \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\left.\mathbf{1}_{\left\{\tau^{i}>t\right\}}\right|_{t} ^{i} \mid \mathscr{H}_{s}^{i}\right\}}{\left(\Gamma_{t}^{i}\right)^{-1}}=Z_{s}^{i} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\} \\
& =Z_{s}^{i} \mathbb{E}\left\{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\}=Z_{s}^{i} .
\end{aligned}
$$

## Representation of Joint Survival Probability

[Sketch of the proof.] (i) $Z_{t}^{i \text { 's }}$ are martingales.

$$
\begin{aligned}
\mathbb{E}\left\{Z_{t}^{i} \mid \mathscr{F}_{s}\right\} & =\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\}=\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\} \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\left.\mathbf{1}_{\left\{\tau^{i}>t\right\}}\right|_{t} ^{i} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mid \mathscr{H}_{s}^{i}\right\}} \quad \text { (Lemma) } \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\left.\mathbf{1}_{\left\{\tau^{i}>t\right\}}\right|_{t} ^{i} \mid \mathscr{H}_{s}^{i}\right\}}{\left(\Gamma_{t}^{i}\right)^{-1}}=Z_{s}^{i} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\} \\
& =Z_{s}^{i} \mathbb{E}\left\{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\}=Z_{s}^{i}
\end{aligned}
$$

(ii) If $\tilde{Z}_{t}^{k} \triangleq \prod_{i=1}^{k} Z_{t}^{i}$ is an mg , then so is $\prod_{i=1}^{k+1} Z_{t}^{i}=\tilde{Z}_{t}^{k} Z_{t}^{k+1}$.

## Representation of Joint Survival Probability

[Sketch of the proof.] (i) $Z_{t}^{i \text { 's }}$ are martingales.

$$
\begin{aligned}
\mathbb{E}\left\{Z_{t}^{i} \mid \mathscr{F}_{s}\right\} & =\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\}=\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\} \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\left.\mathbf{1}_{\left\{\tau^{i}>t\right\}}\right|_{t} ^{i} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mid \mathscr{H}_{s}^{i}\right\}} \quad \text { (Lemma) } \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\left.\mathbf{1}_{\left\{\tau^{i}>t\right\}}\right|_{t} ^{i} \mid \mathscr{H}_{s}^{i}\right\}}{\left(\Gamma_{t}^{i}\right)^{-1}}=Z_{s}^{i} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\} \\
& =Z_{s}^{i} \mathbb{E}\left\{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\}=Z_{s}^{i} .
\end{aligned}
$$

(ii) If $\tilde{Z}_{t}^{k} \triangleq \prod_{i=1}^{k} Z_{t}^{i}$ is an mg , then so is $\prod_{i=1}^{k+1} Z_{t}^{i}=\tilde{Z}_{t}^{k} Z_{t}^{k+1}$.

$$
\tilde{Z}_{t}^{k} Z_{t}^{k+1}=\int_{0^{+}}^{t} \tilde{Z}_{s-}^{k} d Z_{s}^{k+1}+\int_{0^{+}}^{t} Z_{s-}^{k+1} d \tilde{Z}_{s}^{k}+\left[\tilde{Z}^{k}, Z^{k+1}\right]_{t}
$$

## Representation of Joint Survival Probability

[Sketch of the proof.] (i) $Z_{t}^{i}$ 's are martingales.

$$
\begin{aligned}
\mathbb{E}\left\{Z_{t}^{i} \mid \mathscr{F}_{s}\right\} & =\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\}=\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{F}_{s}\right\} \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{H}_{t}^{i}\right\}}{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>s\right\}} \mid \mathscr{H}_{s}^{i}\right\}} \quad \text { (Lemma) } \\
& =\mathbf{1}_{\left\{\tau^{i}>s\right\}} \frac{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\}}{\left(\Gamma_{t}^{i}\right)^{-1}}=Z_{s}^{i} \mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\} \\
& =Z_{s}^{i} \mathbb{E}\left\{\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{i}>t\right\}} \mid \mathscr{H}_{t}^{i}\right\} \Gamma_{t}^{i} \mid \mathscr{H}_{s}^{i}\right\}=Z_{s}^{i}
\end{aligned}
$$

(ii) If $\tilde{Z}_{t}^{k} \triangleq \prod_{i=1}^{k} Z_{t}^{i}$ is an mg , then so is $\prod_{i=1}^{k+1} Z_{t}^{i}=\tilde{Z}_{t}^{k} Z_{t}^{k+1}$.

$$
\tilde{Z}_{t}^{k} Z_{t}^{k+1}=\int_{0^{+}}^{t} \tilde{Z}_{s-}^{k} d Z_{s}^{k+1}+\int_{0^{+}}^{t} Z_{s-}^{k+1} d \tilde{Z}_{s}^{k}+\left[\tilde{Z}^{k}, Z^{k+1}\right]_{t}
$$

Since both $\tilde{Z}^{k}$ and $Z^{k+1}$ are FV and quadratic pure jump,
$\left[\tilde{Z}^{k}, Z^{k+1}\right]_{t}=\tilde{Z}_{0}^{k} Z_{0}^{k+1}+\sum_{0<s \leq t} \Delta \tilde{Z}_{s}^{k} \Delta Z_{s}^{k+1}=\tilde{Z}_{0}^{k} Z_{0}^{k+1}$.

## Representation of Joint Survival Probability

Define

$$
\begin{equation*}
\left.\frac{d \mathbb{P}^{i}}{d \mathbb{P}^{1}}\right|_{\mathscr{F}_{T}} \triangleq Z_{T}^{i} ;\left.\quad \frac{d \mathbb{P}^{1, \cdots, k}}{d \mathbb{P}^{P}}\right|_{\mathscr{F}_{T}} \triangleq \tilde{Z}_{T}^{k}=\prod_{i=1}^{k} Z_{T}^{i} \tag{25}
\end{equation*}
$$

and $\mathbb{E}^{1, \cdots, k}\{X\} \triangleq \mathbb{E}^{\mathbb{P}^{1, \cdots, k}}\{X\}=\mathbb{E}\left\{Z_{T}^{1} Z_{T}^{2} \ldots Z_{T}^{k} X\right\}$.

## Representation of Joint Survival Probability

Define

$$
\begin{equation*}
\left.\frac{d \mathbb{P}^{i}}{d \mathbb{P}^{\prime}}\right|_{\mathscr{F}_{T}} \triangleq Z_{T}^{i} ;\left.\quad \frac{d \mathbb{P}^{1, \cdots, k}}{d \mathbb{P}}\right|_{\mathscr{F}_{T}} \triangleq \tilde{Z}_{T}^{k}=\prod_{i=1}^{k} Z_{T}^{i} . \tag{25}
\end{equation*}
$$

and $\mathbb{E}^{1, \cdots, k}\{X\} \triangleq \mathbb{E}^{\mathbb{P}^{1, \cdots, k}}\{X\}=\mathbb{E}\left\{Z_{T}^{1} Z_{T}^{2} \ldots Z_{T}^{k} X\right\}$.
Then,for each $k$ and $A \in \mathscr{F}_{t}$, it holds that

$$
\begin{aligned}
\mathbb{E}\left\{\mathbf{1}_{A} \tilde{Z}_{t}^{k} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}\right\} & =\mathbb{E}\left\{\mathbf{1}_{A} \mathbb{E}\left\{\tilde{Z}_{T}^{k} \mid \mathscr{F}_{t}\right\} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}\right\} \\
& =\mathbb{E}^{1, \cdots, k}\left\{\mathbf{1}_{A} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}\right\} \\
& =\mathbb{E}^{1, \cdots, k}\left\{\mathbf{1}_{A} X\right\}=\mathbb{E}\left\{\mathbf{1}_{A} \mathbb{E}\left\{\tilde{Z}_{T}^{k} X \mid \mathscr{F}_{t}\right\}\right\} .
\end{aligned}
$$

## Representation of Joint Survival Probability

Define

$$
\begin{equation*}
\left.\left.\frac{d \mathbb{P}^{i}}{d \mathbb{P}_{\mathscr{F}_{T}}}\right|_{\mathscr{F}_{T}} \triangleq \quad \frac{d \mathbb{P}^{1, \cdots, k}}{d \mathbb{P}^{i}}\right|_{\mathscr{F}_{T}} \triangleq \tilde{Z}_{T}^{k}=\prod_{i=1}^{k} Z_{T}^{i} \tag{25}
\end{equation*}
$$

and $\mathbb{E}^{1, \cdots, k}\{X\} \triangleq \mathbb{E}^{\mathbb{P}^{1, \cdots, k}}\{X\}=\mathbb{E}\left\{Z_{T}^{1} Z_{T}^{2} \ldots Z_{T}^{k} X\right\}$.
Then, for each $k$ and $A \in \mathscr{F}_{t}$, it holds that

$$
\begin{aligned}
\mathbb{E}\left\{\mathbf{1}_{A} \tilde{Z}_{t}^{k} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}\right\} & =\mathbb{E}\left\{\mathbf{1}_{A} \mathbb{E}\left\{\tilde{Z}_{T}^{k} \mid \mathscr{F}_{t}\right\} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}\right\} \\
& =\mathbb{E}^{1, \cdots, k}\left\{\mathbf{1}_{A} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}\right\} \\
& =\mathbb{E}^{1, \cdots, k}\left\{\mathbf{1}_{A} X\right\}=\mathbb{E}\left\{\mathbf{1}_{A} \mathbb{E}\left\{\tilde{Z}_{T}^{k} X \mid \mathscr{F}_{t}\right\}\right\} .
\end{aligned}
$$

This leads to

$$
\begin{equation*}
\mathbb{E}\left\{Z_{T}^{1} Z_{T}^{2} \ldots Z_{T}^{k} X \mid \mathscr{F}_{t}\right\}=Z_{t}^{1} Z_{t}^{2} \ldots Z_{t}^{k} \mathbb{E}^{1, \cdots, k}\left\{X \mid \mathscr{F}_{t}\right\}, \quad \mathbb{P}-\text { a.s. } \tag{26}
\end{equation*}
$$

## Representation of Joint Survival Probability

Assume $I=2$, and $t_{1} \leq t_{2}$. Apply (26) we get

$$
\begin{aligned}
\mathbb{P}\left\{\tau^{1}>t_{1}, \tau^{2}>t_{2}\right\} & \left.=\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{1}>t_{1}\right\}} \mathbb{E}\left\{Z_{t_{2}}^{2}\left(\Gamma_{t_{2}}^{2}\right)^{-1}\right\} \mid \mathscr{F}_{t_{1}}\right\}\right\} \\
& \left.=\mathbb{E}\left\{\mathbf{1}_{\left\{\tau^{1}>t_{1}\right\}} Z_{t_{1}}^{2} \mathbb{E}^{\mathbb{P}^{2}}\left\{\left(\Gamma_{t_{2}}^{2}\right)^{-1}\right\} \mid \mathscr{F}_{t_{1}}\right\}\right\} \\
& \left.=\mathbb{E}\left\{Z_{t_{1}}^{1} Z_{t_{1}}^{2} \mathbb{E}^{\mathbb{P}^{2}}\left\{\left(\Gamma_{t_{1}}^{1}\right)^{-1}\left(\Gamma_{t_{2}}^{2}\right)^{-1}\right\} \mid \mathscr{F}_{t_{1}}\right\}\right\} \\
& \left.=\mathbb{E} \mathbb{E}^{1,2}\left\{\mathbb{E}^{\mathbb{P}^{2}}\left\{\left(\Gamma_{t_{1}}^{1}\right)^{-1}\left(\Gamma_{t_{2}}^{2}\right)^{-1}\right\} \mid \mathscr{F}_{t_{1}}\right\}\right\} .
\end{aligned}
$$

In particular, if $t_{1}=t_{2}=t$, then we have

$$
\mathbb{P}\left\{\tau^{1}>t, \tau^{2}>t\right\}=\mathbb{E}^{1,2}\left\{\exp \left\{-\int_{0}^{t}\left(\lambda_{s}^{1}+\lambda_{s}^{2}\right) d s\right\}\right\} .
$$

## Representation of Joint Survival Probability

## Theorem

Assume (H1) and (H2). Then,
(i) For any $0 \leq t_{1} \leq t_{2} \leq \ldots \leq t_{l}<\infty$, it holds that

$$
\begin{aligned}
& \mathbb{P}\left\{\tau^{1}>t_{1}, \tau^{2}>t_{2}, \ldots, \tau^{\prime}>t_{l}\right\} \\
= & \mathbb{E}^{1, \cdots, l}\left\{\cdots\left\{\mathbb{E}^{\mathbb{P}^{\prime}}\left\{\prod_{i=1}^{\prime}\left(\Gamma_{t_{i}}^{i}\right)^{-1}\right\} \mid \mathscr{F}_{t_{l-1}}\right\} \cdots \mid \mathscr{F}_{t_{1}}\right\} ;
\end{aligned}
$$

## Representation of Joint Survival Probability

## Theorem

Assume (H1) and (H2). Then,
(i) For any $0 \leq t_{1} \leq t_{2} \leq \ldots \leq t_{l}<\infty$, it holds that

$$
\begin{aligned}
& \mathbb{P}\left\{\tau^{1}>t_{1}, \tau^{2}>t_{2}, \ldots, \tau^{\prime}>t_{l}\right\} \\
= & \mathbb{E}^{1, \cdots, l}\left\{\cdots\left\{\mathbb{E}^{\mathbb{P}^{\prime}}\left\{\prod_{i=1}^{\prime}\left(\Gamma_{t_{i}}^{i}\right)^{-1}\right\} \mid \mathscr{F}_{t_{l-1}}\right\} \cdots \mid \mathscr{F}_{t_{1}}\right\} ;
\end{aligned}
$$

(ii) Denote $\tau^{*}=\min \left\{\tau^{1}, \cdots, \tau^{\prime}\right\}$, then for any $0 \leq t \leq T$ a) $\mathbb{P}\left\{\tau^{*}>t\right\}=\mathbb{E}^{1, \cdots, I}\left\{e^{-\int_{0}^{t} \sum_{i=1}^{l} \lambda_{s}^{i} d s}\right\}$;

## Representation of Joint Survival Probability

## Theorem

Assume (H1) and (H2). Then,
(i) For any $0 \leq t_{1} \leq t_{2} \leq \ldots \leq t_{l}<\infty$, it holds that

$$
\begin{aligned}
& \mathbb{P}\left\{\tau^{1}>t_{1}, \tau^{2}>t_{2}, \ldots, \tau^{\prime}>t_{l}\right\} \\
= & \mathbb{E}^{1, \cdots, I}\left\{\cdots\left\{\mathbb{E}^{\mathbb{P}^{\prime}}\left\{\prod_{i=1}^{l}\left(\Gamma_{t_{i}}^{i}\right)^{-1}\right\} \mid \mathscr{F}_{t_{l-1}}\right\} \cdots \mid \mathscr{F}_{t_{1}}\right\} ;
\end{aligned}
$$

(ii) Denote $\tau^{*}=\min \left\{\tau^{1}, \cdots, \tau^{\prime}\right\}$, then for any $0 \leq t \leq T$
a) $\mathbb{P}\left\{\tau^{*}>t\right\}=\mathbb{E}^{1, \cdots, I}\left\{e^{-\int_{0}^{t} \sum_{i=1}^{l} \lambda_{s}^{i} d s}\right\}$;
b) $\mathbb{P}\left\{\tau^{*}>T \mid \mathscr{F}_{t}\right\}=\prod_{i=1}^{l} \mathbf{1}_{\left\{\tau^{i}>t\right\}} \mathbb{E}^{1, \cdots, I}\left\{e^{-\int_{t}^{T} \sum_{i=1}^{l} \lambda_{s}^{i} d s} \mid \mathscr{F}_{t}\right\}$.

## Counter-Party Risk Models

## Two firm case:

$$
\left\{\begin{array}{l}
\lambda_{t}^{A}=a_{0}(t)+\mathbf{1}_{\left\{\tau^{B} \leq t\right\}} a_{1}\left(t-\tau^{B}\right),  \tag{27}\\
\lambda_{t}^{B}=b_{0}(t)+\mathbf{1}_{\left\{\tau^{A} \leq t\right\}} b_{1}\left(t-\tau^{A}\right),
\end{array}\right.
$$

where $a_{0}, a_{1}, b_{0}$, and $b_{1}$ are deterministic functions.

## Counter-Party Risk Models

## Two firm case:

$$
\left\{\begin{array}{l}
\lambda_{t}^{A}=a_{0}(t)+\mathbf{1}_{\left\{\tau^{B} \leq t\right\}} a_{1}\left(t-\tau^{B}\right),  \tag{27}\\
\lambda_{t}^{B}=b_{0}(t)+\mathbf{1}_{\left\{\tau^{A} \leq t\right\}} b_{1}\left(t-\tau^{A}\right),
\end{array}\right.
$$

where $a_{0}, a_{1}, b_{0}$, and $b_{1}$ are deterministic functions.
Jarrow-Yu (2004) — a $a_{1}$, $b_{1}$ constants.

## Counter-Party Risk Models

## Two firm case:

$$
\left\{\begin{array}{l}
\lambda_{t}^{A}=a_{0}(t)+\mathbf{1}_{\left\{\tau^{B} \leq t\right\}} a_{1}\left(t-\tau^{B}\right),  \tag{27}\\
\lambda_{t}^{B}=b_{0}(t)+\mathbf{1}_{\left\{\tau^{A} \leq t\right\}} b_{1}\left(t-\tau^{A}\right),
\end{array}\right.
$$

where $a_{0}, a_{1}, b_{0}$, and $b_{1}$ are deterministic functions.

## Jarrow-Yu (2004) — $a_{1}$, $b_{1}$ constants.

(H3) (i) $a_{0}$ and $b_{0}$ are positive functions;
(ii) $a_{1}$ and $b_{1}$ are either positive and decreasing or negative and increasing, such that

$$
\begin{equation*}
\lim _{t \rightarrow \infty} a_{1}(t)=0 \quad \lim _{t \rightarrow \infty} b_{1}(t)=0 ; \tag{28}
\end{equation*}
$$

and such that both $\lambda_{t}^{A}$ and $\lambda_{t}^{B}$ are positive functions.

## Counter-Party Risk Models

## Proposition

Assume ( H 1$)-(\mathrm{H} 3)$. Then the joint survival probability $\mathbb{P}\left\{\tau^{A}>t_{1}, \tau^{B}>t_{2}\right\}$ is given by

$$
\begin{aligned}
& \mathbb{P}\left\{\tau^{A}>t_{1}, \tau^{B}>t_{2}\right\} \\
& = \begin{cases}c\left(t_{1}, t_{2}\right)\left(\int_{t_{1}}^{t_{2}} a_{0}(x) e^{-\int_{x}^{t_{2}} b_{1}(s-x) d s-\int_{t_{1}}^{x} a_{0}(s) d s} d x\right. \\
& \left.+\int_{t_{2}}^{\infty^{2}} a_{0}(x) e^{-\int_{t_{1}}^{x} a_{0}(s) d s} d x\right) \\
c\left(t_{1}, t_{2}\right)\left(\int_{1}^{t_{1}} b_{0}(x) e^{-\int_{x}^{t_{1}} a_{1}(s-x) d s-\int_{t_{2}}^{x} b_{0}(s) d s} d x\right. \\
& \left.+\int_{t_{1}}^{\infty_{2}} b_{0}(x) e^{-\int_{t_{2}}^{x} b_{0}(s) d s} d x\right)\end{cases} \\
&
\end{aligned}
$$

where $c\left(t_{1}, t_{2}\right)=\exp \left\{-\int_{0}^{t_{1}} a_{0}(s) d s-\int_{0}^{t_{2}} b_{0}(s) d s\right\}$.

## Counter-Party Risk Models

Main Observation: $\lambda_{s}^{A}=a_{0}(s), \lambda_{s}^{B}=b_{0}(s), \mathbb{P}^{A, B}$-a.s.

$$
\Longrightarrow \quad 1-F_{\tau^{A}}^{B}(x)=\mathbb{P}^{B}\left(\tau^{A}>x\right)=\mathbb{P}^{A, B}\left(\left(\Gamma_{x}^{A}\right)^{-1}\right)=e^{-\int_{0}^{x} a_{0}(s) d s}
$$

## Counter-Party Risk Models

Main Observation: $\lambda_{s}^{A}=a_{0}(s), \lambda_{s}^{B}=b_{0}(s), \mathbb{P}^{A, B}$-a.s.

$$
\Longrightarrow \quad 1-F_{\tau^{A}}^{B}(x)=\mathbb{P}^{B}\left(\tau^{A}>x\right)=\mathbb{P}^{A, B}\left(\left(\Gamma_{x}^{A}\right)^{-1}\right)=e^{-\int_{0}^{x} a_{0}(s) d s} .
$$

Applying the change of measure, we have

$$
\begin{aligned}
& \left.\mathbb{P}\left\{\tau^{A}>t_{1}, \tau^{B}>t_{2}\right\}=\mathbb{E}\left[\mathbf{1}_{\left\{\tau^{A}>t_{1}\right\}} \mathbf{1}_{\left\{\tau^{B}>t_{2}\right\}} \Gamma_{t_{2}}^{B}\left(\Gamma_{t_{2}}^{B}\right)^{-1}\right\}\right] \\
= & \mathbb{E}^{B}\left[\mathbf{1}_{\left\{\tau^{A}>t_{1}\right\}} \exp \left(-\int_{0}^{t_{2}}\left(b_{0}(s)+\mathbf{1}_{\left\{\tau^{A} \leq s\right\}} b_{1}\left(s-\tau^{A}\right)\right) d s\right)\right] \\
= & c\left(t_{2}\right)\left\{\int_{t_{1}}^{t_{2}} e^{-\int_{x}^{t_{2}} b_{1}(s-x) d s} F_{\tau^{A}}^{B}(d x)+\int_{t_{2}}^{\infty} F_{\tau^{A}}^{B}(d x)\right\} \\
= & c\left(t_{2}\right)\left\{\int_{t_{1}}^{t_{2}} e^{-\int_{x}^{t_{2}} b_{1}(s-x) d s} f_{\tau^{A}}(x) d x+\int_{t_{2}}^{\infty} f_{\tau^{A}}(x) d x\right\} \\
= & \operatorname{RHS}\left(t_{1} \leq t_{2}\right)
\end{aligned}
$$

## Multiple Firm Case

Assume that $I>2$, and that the default intensities are given by

$$
\begin{equation*}
\lambda_{t}^{i}=a_{0}^{i}(t)+\sum_{\substack{j=1 \\ j \neq i}} \mathbf{1}_{\{\tau j \leq t\}} a_{j-1}^{i}\left(t-\tau^{j}\right), \quad i=1, \cdots, l, \tag{29}
\end{equation*}
$$

where $a_{j}^{i}$ 's are deterministic functions satisfying (H3).

## Multiple Firm Case

Assume that $I>2$, and that the default intensities are given by

$$
\begin{equation*}
\lambda_{t}^{i}=a_{0}^{i}(t)+\sum_{\substack{j=1 \\ j \neq i}} \mathbf{1}_{\{\tau j \leq t\}} a_{j-1}^{i}\left(t-\tau^{j}\right), \quad i=1, \cdots, l, \tag{29}
\end{equation*}
$$

where $a_{j}^{i}$ 's are deterministic functions satisfying (H3).

- For $1 \leq m \leq I$, denote $f_{m}\left(t_{1}, t_{2}, \cdots, t_{m}\right)$ to be the joint density function of the default times $\tau^{1}, \tau^{2}, \cdots, \tau^{m}$.
- For example, $f_{1}\left(t_{1}\right)=f_{\tau^{1}}\left(t_{1}\right)=a_{1,0}\left(t_{1}\right) e^{-\int_{0}^{t_{1}} a_{1,0}(s) d s}$.


## Proposition

$$
\text { For } \begin{aligned}
0 & =t_{0}<t_{1}<t_{2}<\ldots<t_{m+1} \\
& f_{m+1}\left(t_{1}, t_{2}, \cdots, t_{m+1}\right) \\
= & \left\{\sum_{j=0}^{m} a_{j}^{m+1}\left(t_{m+1}-t_{j}\right)\right\} e^{-\sum_{j} \int_{t_{j}}^{t_{m+1}} a_{j}^{m+1}\left(s-t_{j}\right) d s} f_{m}\left(t_{1}, \cdots, t_{m}\right)
\end{aligned}
$$

## Multiple Firm Case (General)

- Let $\mathscr{P}(I)$ be all the permutations $p=p(1, \cdots, I)$, then $|\mathscr{P}(I)|=I!$.


## Multiple Firm Case (General)

- Let $\mathscr{P}(I)$ be all the permutations $p=p(1, \cdots, I)$, then $|\mathscr{P}(I)|=I!$.
- $\forall p \in \mathscr{P}(I)$, permute $\left(t_{1}, \cdots, t_{l}\right)$ to $\left(t_{1}^{(p)}, \cdots, t_{l}^{(p)}\right)$, and

$$
\mathscr{D}^{(p)} \triangleq\left\{\left(t_{1}, \cdots, t_{l}\right) \in \mathbb{R}_{+}^{\prime}: t_{1}^{(p)}<\cdots<t_{l}^{(p)}\right\}
$$

## Multiple Firm Case (General)

- Let $\mathscr{P}(I)$ be all the permutations $p=p(1, \cdots, I)$, then $|\mathscr{P}(I)|=I!$.
- $\forall p \in \mathscr{P}(I)$, permute $\left(t_{1}, \cdots, t_{l}\right)$ to $\left(t_{1}^{(p)}, \cdots, t_{l}^{(p)}\right)$, and

$$
\mathscr{D}^{(p)} \triangleq\left\{\left(t_{1}, \cdots, t_{l}\right) \in \mathbb{R}_{+}^{\prime}: t_{1}^{(p)}<\cdots<t_{l}^{(p)}\right\}
$$

- $\mathbb{R}_{+}^{\prime}=\bigcup_{i \in \mathscr{P}(I)} \mathscr{D}^{(p)} ; \mathscr{D}^{(p)} \cap \mathscr{D}^{(p)}=\emptyset$.


## Multiple Firm Case (General)

- Let $\mathscr{P}(I)$ be all the permutations $p=p(1, \cdots, I)$, then $|\mathscr{P}(I)|=I!$.
- $\forall p \in \mathscr{P}(I)$, permute $\left(t_{1}, \cdots, t_{l}\right)$ to $\left(t_{1}^{(p)}, \cdots, t_{l}^{(p)}\right)$, and

$$
\mathscr{D}^{(p)} \triangleq\left\{\left(t_{1}, \cdots, t_{l}\right) \in \mathbb{R}_{+}^{\prime}: t_{1}^{(p)}<\cdots<t_{l}^{(p)}\right\} .
$$

- $\mathbb{R}_{+}^{\prime}=\bigcup_{i \in \mathscr{P}(I)} \mathscr{D}^{(p)} ; \mathscr{D}^{(p)} \cap \mathscr{D}^{(p)}=\emptyset$.
- $\forall p \in \mathscr{P}(I)$, define $\left(\tau_{1}^{(p)}, \cdots, \tau_{l}^{(p)}\right)$ accordingly, and

$$
\lambda_{t}^{i,(p)}=a_{0}^{i,(p)}(t)+\sum_{\substack{j=1 \\ j \neq i}} \mathbf{1}_{\left\{\tau_{j}^{(p)} \leq t\right\}} b_{j-1}^{i}\left(t-\tau_{j}^{(p)}\right),
$$

where $b_{j, 0}(t)=a_{j(p), 0}(t), j=1, \cdots, l, j^{(p)}$ is the image position of $j$ after the permutation $p \in \mathscr{P}(I)$, and $b_{j}^{i}$ are appropriately defined functions from $a_{j}^{i}$ 's.

## Multiple Firm Case (General)

$\forall p \in \mathscr{P}(I)$ apply the Proposition on the region $D^{(i)}$, with $\left(\lambda_{1}, \cdots, \lambda_{I}\right)$ being replaced by $\left(\lambda_{1}^{(p)}, \cdots \lambda_{I}^{(p)}\right)$, to obtain the joint density function on $D^{(p)}$, denoted by $f_{l}{ }^{(p)}$. We can then define

$$
g_{l}\left(t_{1}, \cdots, t_{l}\right)=f_{l}^{(p)}\left(t_{1}^{(p)}, \cdots, t_{l}^{(p)}\right),\left(t_{1}, \cdots, t_{l}\right) \in D^{(p)}
$$

## Multiple Firm Case (General)

$\forall p \in \mathscr{P}(I)$ apply the Proposition on the region $D^{(i)}$, with $\left(\lambda_{1}, \cdots, \lambda_{I}\right)$ being replaced by $\left(\lambda_{1}^{(p)}, \cdots \lambda_{I}^{(p)}\right)$, to obtain the joint density function on $D^{(p)}$, denoted by $f_{l}^{(p)}$. We can then define

$$
g_{l}\left(t_{1}, \cdots, t_{l}\right)=f_{l}^{(p)}\left(t_{1}^{(p)}, \cdots, t_{l}^{(p)}\right),\left(t_{1}, \cdots, t_{l}\right) \in D^{(p)}
$$

## Theorem

Assume (H1)-(H3). The joint distribution of $\tau_{1}, \tau_{2}, \cdots, \tau_{I}$ can be expressed as
$\mathbb{P}\left\{\tau^{1} \leq t_{1}, \cdots, \tau^{\prime} \leq t_{l}\right\}=\int_{0}^{t_{1}} \cdots \int_{0}^{t_{l}} g_{l}\left(u_{1}, \cdots, u_{l}\right) d u_{1} d u_{2} \cdots d u_{l}$.
where $g$,'s are defined above.

## Joint-life vs. Last-survivor

Let $T_{X_{1}}, T_{x_{2}}, \cdots, T_{X_{n}}$ be $n$ future life time random variables, then their and are given by, respectively:

$$
\begin{aligned}
T_{m}= & T_{x_{1}, \cdots, x_{n}} \triangleq \\
& \min \left\{T_{x_{1}}, T_{x_{2}}, \cdots, T_{x_{n}}\right\} \\
& \quad-(\text { Joint-life }=\text { first default }) \\
T_{M}= & T_{\overline{x_{1}, \cdots, x_{n}}} \triangleq \max \left\{T_{x_{1}}, T_{x_{2}}, \cdots, T_{x_{n}}\right\} \\
& \quad-(\text { Last-survivor }=\text { last default })
\end{aligned}
$$

## Joint-life vs. Last-survivor

Let $T_{X_{1}}, T_{x_{2}}, \cdots, T_{X_{n}}$ be $n$ future life time random variables, then their and are given by, respectively:

$$
\begin{aligned}
T_{m}= & T_{x_{1}, \cdots, x_{n}} \triangleq \\
& \min \left\{T_{x_{1}}, T_{x_{2}}, \cdots, T_{x_{n}}\right\} \\
& \quad-(\text { Joint-life }=\text { first default }) \\
T_{M}= & T_{\overline{x_{1}, \cdots, x_{n}}} \triangleq \max \left\{T_{x_{1}}, T_{x_{2}}, \cdots, T_{x_{n}}\right\} \\
& \quad-(\text { Last-survivor }=\text { last default })
\end{aligned}
$$

If $n=2$, one has

- $T_{M}+T_{m}=T_{x_{1}}+T_{x_{2}}, T_{M} T_{m}=T_{x_{1}} T_{x_{2}}$.
- $\left\{T_{x_{1}} \leq t\right\} \cap\left\{T_{x_{2}} \leq t\right\}=\left\{T_{M} \leq t\right\}$, $\left\{T_{x_{1}} \leq t\right\} \cup\left\{T_{x_{2}} \leq t\right\}=\left\{T_{m} \leq t\right\}$,
- $F_{M}(t)+F_{m}(t)=F_{T_{x_{1}}}(t)+F_{T_{x_{2}}}(t), t \geq 0$ where $F_{T}$ is the distribution function of $T$.


## First Default in Multi-firm Case

Assume for $i=1, \cdots, I$,

$$
\lambda_{t}^{i}=a_{0}^{i}(t)+\sum_{k \neq i} a_{k}^{i}(t) \mathbf{1}_{\left\{\tau^{k} \leq t\right\}}=a_{0}^{i}(t)+\sum_{k \neq i} a_{k}^{i}(t) N_{s}^{i},
$$

Assume for $i=1, \cdots, l$,

$$
\lambda_{t}^{i}=a_{0}^{i}(t)+\sum_{k \neq i} a_{k}^{i}(t) \mathbf{1}_{\left\{\tau^{k} \leq t\right\}}=a_{0}^{i}(t)+\sum_{k \neq i} a_{k}^{i}(t) N_{s}^{i},
$$

Then

$$
\begin{aligned}
\mathbb{P}\left\{\tau_{m}>t\right\} & =\mathbb{P}\left\{\tau^{1}>t, \tau^{2}>t, \cdots, \tau^{l}>t\right\} \\
& =\mathbb{E}^{1,2, \cdots, I}\left\{e^{-\int_{0}^{t}\left(\lambda_{s}^{1}+\lambda_{s}^{2}+\ldots+\lambda_{s}^{\prime}\right) d s}\right\} \\
& =\mathbb{E}^{1,2, \cdots, I}\left\{e^{-\int_{0}^{t}\left[a_{0}^{1}(s)+a_{0}^{2}(s)+\ldots+a_{0}^{l}(s)\right] d s}\right\} .
\end{aligned}
$$

If all $a_{0}^{i}$ 's are deterministic, then

$$
\mathbb{P}\left\{\tau_{m}>t\right\}=\exp \left\{-\int_{0}^{t}\left[a_{0}^{1}(s)+a_{0}^{2}(s)+\ldots+a_{0}^{\prime}(s)\right] d s\right\}
$$

Similarly one can obtain the conditional survival probability of $\tau_{m}$ :

$$
\begin{aligned}
& \mathbb{P}\left\{\tau_{m}>T \mid \mathscr{F}_{t}\right\}=\mathbb{P}\left\{\tau^{1}>T, \tau^{2}>T, \cdots, \tau^{\prime}>T \mid \mathscr{F}_{t}\right\} \\
= & \prod_{i=1}^{\prime} \mathbf{1}_{\left\{\tau_{t}^{\prime}>t\right\}} \mathbb{E}^{1,2, \cdots, I}\left\{\exp \left\{-\int_{t}^{T}\left[\sum_{i=1}^{\prime} \lambda_{s}^{i}\right] d s\right\} \mid \mathscr{F}_{t}\right\} \\
= & \mathbf{1}_{\left\{\tau_{m}>t\right\}} \mathbb{E}^{1,2, \cdots, l}\left\{\exp \left\{-\int_{t}^{T}\left[\sum_{i=1}^{\prime} a_{0}^{i}(s)\right] d s\right\} \mid \mathscr{F}_{t}\right\} .
\end{aligned}
$$

If $a_{0}^{i}$ 's are all deterministic, then

$$
\mathbb{P}\left\{\tau_{m}>T \mid \mathscr{F}_{t}\right\}=\mathbf{1}_{\left\{\tau_{m}>t\right\}} \exp \left\{-\int_{t}^{T} \sum a_{0}^{i}(s) d s\right\}
$$

## Flight to Quality

- The term "flight to quality" refers to the phenomenon that investors move their capital away from riskier investments to the safest possible investment vehicles, e.g., treasury bonds.


## Flight to Quality

- The term "flight to quality" refers to the phenomenon that investors move their capital away from riskier investments to the safest possible investment vehicles, e.g., treasury bonds.
- One firm model (Collins-Dufresne et al. $(03,04)$ ):

$$
\begin{equation*}
r_{t}=r_{0}+J \mathbf{1}_{\{\tau \leq t\}} \geq 0, \quad t \geq 0 \tag{30}
\end{equation*}
$$

## Flight to Quality

- The term "flight to quality" refers to the phenomenon that investors move their capital away from riskier investments to the safest possible investment vehicles, e.g., treasury bonds.
- One firm model (Collins-Dufresne et al. $(03,04)$ ):

$$
\begin{equation*}
r_{t}=r_{0}+J \mathbf{1}_{\{\tau \leq t\}} \geq 0, \quad t \geq 0 \tag{30}
\end{equation*}
$$

- We will consider multi-firm model:

$$
\begin{equation*}
r_{t}=r_{0}\left(X_{t}\right)+J \mathbf{1}_{\left\{\tau_{M} \leq t\right\}}, \quad t \geq 0 \tag{31}
\end{equation*}
$$

where $\tau_{M} \triangleq \max \left\{\tau^{1}, \cdots, \tau^{\prime}\right\}$ is the last-to-default time, $X$ is a factor process.

- Main purpose: pricing defaultable zero-coupon bonds.


## Pricing of UVL Insurance Involving Married Couples

- Let $T_{x_{1}}$ and $T_{x_{2}}$ be two future life time r.v.'s. Denote $N_{t}^{i}=\mathbf{1}_{\left\{T_{x_{i}} \leq t\right\}}, i=1,2$, and

$$
\mathscr{F}_{t}=\mathscr{F}_{t}^{X} \vee \mathscr{F}_{t}^{1} \vee \mathscr{F}_{t}^{2}, \quad t \geq 0
$$

where $\mathscr{F}_{t}^{i}=\sigma\left\{N_{s}^{i}, 0 \leq s \leq t\right\}, t \geq 0, i=1,2$, and $X$ is a factor process, assumed to be a diffusion process

## Pricing of UVL Insurance Involving Married Couples

- Let $T_{x_{1}}$ and $T_{x_{2}}$ be two future life time r.v.'s. Denote $N_{t}^{i}=\mathbf{1}_{\left\{T_{x_{i}} \leq t\right\}}, i=1,2$, and

$$
\mathscr{F}_{t}=\mathscr{F}_{t}^{X} \vee \mathscr{F}_{t}^{1} \vee \mathscr{F}_{t}^{2}, \quad t \geq 0
$$

where $\mathscr{F}_{t}^{i}=\sigma\left\{N_{s}^{i}, 0 \leq s \leq t\right\}, t \geq 0, i=1,2$, and $X$ is a factor process, assumed to be a diffusion process

- Death benefit is a lump-sum (e.g., \$1) payable at a terminal time $T$, contingent on the survivorship of a married couple.
- Let $T_{x_{1}}$ and $T_{x_{2}}$ be two future life time r.v.'s. Denote $N_{t}^{i}=\mathbf{1}_{\left\{T_{x_{i}} \leq t\right\}}, i=1,2$, and

$$
\mathscr{F}_{t}=\mathscr{F}_{t}^{X} \vee \mathscr{F}_{t}^{1} \vee \mathscr{F}_{t}^{2}, \quad t \geq 0
$$

where $\mathscr{F}_{t}^{i}=\sigma\left\{N_{s}^{i}, 0 \leq s \leq t\right\}, t \geq 0, i=1,2$, and $X$ is a factor process, assumed to be a diffusion process

- Death benefit is a lump-sum (e.g., \$1) payable at a terminal time $T$, contingent on the survivorship of a married couple.
- Let $K_{t}$ be a generic status process, e.g., $K$ could be one of the following:

$$
J L I_{t}=\mathbf{1}_{\left\{T_{x_{1} x_{2}} \leq t\right\}}, \quad S L I_{t}=\mathbf{1}_{\left\{T_{\overline{x_{1} x_{2}}} \leq t\right\}}, \quad t \geq 0,
$$

- Let $T_{x_{1}}$ and $T_{x_{2}}$ be two future life time r.v.'s. Denote $N_{t}^{i}=\mathbf{1}_{\left\{T_{x_{i}} \leq t\right\}}, i=1,2$, and

$$
\mathscr{F}_{t}=\mathscr{F}_{t}^{X} \vee \mathscr{F}_{t}^{1} \vee \mathscr{F}_{t}^{2}, \quad t \geq 0
$$

where $\mathscr{F}_{t}^{i}=\sigma\left\{N_{s}^{i}, 0 \leq s \leq t\right\}, t \geq 0, i=1,2$, and $X$ is a factor process, assumed to be a diffusion process

- Death benefit is a lump-sum (e.g., \$1) payable at a terminal time $T$, contingent on the survivorship of a married couple.
- Let $K_{t}$ be a generic status process, e.g., $K$ could be one of the following:

$$
J L I_{t}=\mathbf{1}_{\left\{T_{x_{1} x_{2}} \leq t\right\}}, \quad S L I_{t}=\mathbf{1}_{\left\{T_{\overline{x_{1} x_{2}}} \leq t\right\}}, \quad t \geq 0,
$$

## Bereaved Partner Case (M.-Yun '10)

Assume that the individual $T_{x_{i}}$ 's follow the Gompertz's law (1825): $\lambda_{x_{1}}(t)=h_{1} e^{g_{1}\left(x_{1}+t\right)}, \lambda_{x_{2}}(t)=h_{2} e^{g_{2}\left(x_{2}+t\right)}, h_{i}>0, g_{i}>0$. Then

$$
\begin{aligned}
& \mathbb{P}\left\{T_{x_{1}}>t_{1}, T_{x_{2}}>t_{2}\right\} \\
= & \left\{\begin{array}{cc}
\frac{c\left(t_{1}, t_{2}\right)}{\left(r_{2}+1\right)^{n_{2}}} \sum_{k=0}^{n_{2}}\binom{n_{2}}{k} \frac{h_{1}}{g_{1}} r_{2}^{n_{2}-k} B^{1}\left(\tilde{\mathbb{D}}_{k}^{1}\left(t_{2}\right)-\tilde{\mathbb{D}}_{k}^{1}\left(t_{1}\right)\right) \\
+c\left(t_{2}, t_{2}\right) & t_{1} \leq t_{2} ; \\
\left.\frac{c\left(t_{1}, t_{2}\right)}{\left(r_{1}+1\right)^{n_{1}}} \sum_{k=0}^{n_{1}}\binom{n_{1}}{k} \frac{h_{2}}{g_{2}} r_{1}^{n_{1}-k} B^{2}\left(\tilde{\mathbb{D}}_{k}^{2}\left(t_{1}\right)\right)-\tilde{\mathbb{D}}_{k}^{2}\left(t_{2}\right)\right) \\
+c\left(t_{1}, t_{1}\right) & t_{1}>t_{2},
\end{array}\right.
\end{aligned}
$$

where

- $\Delta_{k}^{i}(t)=\int_{0}^{t} y^{\frac{k}{g_{i}}} e^{-\frac{h_{i}}{g_{i}} y} d y, \tilde{\mathbb{D}}_{k}^{i}(t)=\mathbb{D}_{k}^{i}\left(\frac{\lambda_{x_{i}}(t)}{h_{i}}\right), i=1,2$,
- $B^{1}=e^{-k\left(t_{2}+x_{1}\right)+\frac{h_{1}}{g_{1}} e^{g_{1}\left(x_{1}+t_{1}\right)}}, B^{2}=e^{-k\left(t_{1}+x_{2}\right)+\frac{h_{2}}{g_{2}} e^{g_{2}\left(x_{2}+t_{2}\right)}}$,
- $c\left(t_{1}, t_{2}\right)=\exp \left\{-\frac{h_{1}}{g_{1}}\left[e^{g_{1}\left(x_{1}+t_{1}\right)}-e^{g_{1} x_{1}}\right]-\frac{h_{2}}{g_{2}}\left[e^{g_{2}\left(x_{2}+t_{2}\right)}-e^{g_{2} x_{2}}\right]\right\}$.


## Back to UVL Insurance Pricing

- 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:

$$
J L I_{t}=\mathbf{1}_{\left\{T_{x_{1} x_{2}} \leq t\right\}}, \quad S L I_{t}=\mathbf{1}_{\left\{T_{\bar{x}_{1} x_{2}} \leq t\right\}}, \quad t \geq 0,
$$

[Then the pdf of $K_{T}$ could be computable!]

## Back to UVL Insurance Pricing

- 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:

$$
J L I_{t}=\mathbf{1}_{\left\{T_{x_{1} x_{2}} \leq t\right\}}, \quad S L I_{t}=\mathbf{1}_{\left\{T_{\bar{x}_{1} x_{2}} \leq t\right\}}, \quad t \geq 0,
$$

[Then the pdf of $K_{T}$ could be computable!]

- Let $u$ be an exponential utility function:

$$
\begin{equation*}
u(w)=-\frac{1}{\alpha} e^{-\alpha w}, \quad w \in \mathbb{R} \tag{32}
\end{equation*}
$$

- Define $J(t, w ; \pi) \triangleq \mathbb{E}_{t, w}\left\{u\left(W_{T}^{\pi}-K_{T}\right)\right\}$, where $W$ is the wealth process with investment portfolio $\pi$.


## Back to UVL Insurance Pricing

- 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:

$$
J L I_{t}=\mathbf{1}_{\left\{T_{x_{1} x_{2}} \leq t\right\}}, \quad S L I_{t}=\mathbf{1}_{\left\{T_{\overline{x_{1} x_{2}}} \leq t\right\}}, \quad t \geq 0,
$$

[Then the pdf of $K_{T}$ could be computable!]

- Let $u$ be an exponential utility function:

$$
\begin{equation*}
u(w)=-\frac{1}{\alpha} e^{-\alpha w}, \quad w \in \mathbb{R} \tag{32}
\end{equation*}
$$

- Define $J(t, w ; \pi) \triangleq \mathbb{E}_{t, w}\left\{u\left(W_{T}^{\pi}-K_{T}\right)\right\}$, where $W$ is the wealth process with investment portfolio $\pi$.
- If $K_{T} \equiv 0$, then denote $J^{0}(t, w ; \pi) \triangleq \mathbb{E}_{t, w}\left\{u\left(W_{T}^{\pi}\right)\right\}, \pi \in \mathscr{A}$.
- $U(t, w) \triangleq \sup _{\pi \in \mathscr{A}} J(t, w ; \pi), V(t, w) \triangleq \sup _{\pi \in \mathscr{A}} J^{0}(t, w ; \pi)$.


## 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 $\Phi$ ?

## 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 $\Phi$ ?

## Theorem (M.-Yun '10)

- $\Phi(t, w)=\mathbb{E}_{t, w}\left\{e^{\alpha K_{T}}\right\}$.
$\left[\right.$ Note that $\left.J(t, w ; \pi)=J^{0}(t, w ; \pi) \mathbb{E}_{t, w}\left\{e^{\alpha K_{T}}\right\}!\right]$


## 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 $\Phi$ ?

## Theorem (M.-Yun '10)

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

$$
p_{t}^{*}=\frac{1}{\alpha} e^{-r(T-t)} \log \Phi(t, w)=\frac{1}{\alpha} e^{-r(T-t)} \log \mathbb{E}_{t, w}\left[e^{\alpha K_{T}}\right]
$$

R- Ma, J. and Yu, Y., (2006), Principle of Equivalent Utility and Universal Variable Life Insurance, Scand. Actuarial J., 6, pp. 311-337.
R Ma, J., Yu, Y. (2007), Indifference Pricing of Universal Variable Life Insurance, pp. 107-121. World Sci. Publ., Hackensack, NJ. Control Theory and Related Topics.
雷 Ma, J., Yun, Y. (2010) Dependent Default Probability in Intensity-Based Cox Models, preprint.

围 Young, V. R. and Zariphopoulou, T. (2002) Pricing Insurance Risks Using the Principle of Equivalent Utility, Scand. Actuarial J., 4, 246-279.

