# Multiple Defaults and Density Approach : Global and Default-free Information

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### Introduction

- The progressive enlargement of filtration plays an essential role in the credit risk modelling.
- Consider multiple default times τ = (τ<sub>1</sub>, · · · , τ<sub>n</sub>) on the market (Ω, A, P).
- There are default-free market information (*F<sub>t</sub>*)<sub>t≥0</sub> and default information D<sup>i</sup><sub>t</sub> = σ(τ<sub>i</sub> ∧ t), i ∈ Θ = {1, · · · , n}.
- ▶ The global market information  $\mathcal{G}_t = \mathcal{F}_t \vee \mathcal{D}_t^1 \vee \cdots \vee \mathcal{D}_t^n$ .
- The pricing and hedging problems are considered in *G<sub>t</sub>*, however it is technically difficult to work with *G<sub>t</sub>* in general since it is not generated by continuous processes.

## Single default: before-default and after-default

For a single defaul τ, the before-default pricing on {τ > t} (e.g. Bielecki-Jeanblanc -Rutkowski) is to establish a relationship between G<sub>t</sub> and F<sub>t</sub> by the key lemma of Dellacherie and Jeulin-Yor: for any A-measurable r.v. Y,

$$\mathbf{1}_{\{\tau > t\}} \mathbb{E}[\mathbf{Y}|\mathcal{G}_t] = \mathbf{1}_{\{\tau > t\}} \frac{\mathbb{E}[\mathbf{Y}\mathbf{1}_{\{\tau > t\}}|\mathcal{F}_t]}{\mathbb{P}(\tau > t|\mathcal{F}_t)} \quad a.s.$$

On the after-default set {τ ≤ t}, the default density approach (El Karoui-Jeanblanc-J.) is suitable: for any *F<sub>T</sub>* ⊗ *B*(ℝ<sub>+</sub>)-measurable *Y<sub>T</sub>*(*θ*),

$$\mathbf{1}_{\{\tau \leq t\}} E[\mathbf{Y}_{\mathcal{T}}(\tau) | \mathcal{G}_t] = \mathbf{1}_{\{\tau \leq t\}} \frac{E[\mathbf{Y}(\mathcal{T}, \theta) \alpha_{\mathcal{T}}(\theta) | \mathcal{F}_t]}{\alpha_t(\theta)} \Big|_{\theta=\tau} a.s.$$

where  $\alpha_t(\theta)$  is the conditional density of  $\tau$  given  $\mathcal{F}_t$  wrt the law of  $\tau$ .

# Generalization to multiple defaults

- The pricing problem in G<sub>t</sub> is decomposed into a before-default problem and an after-default one.
- Each new problem is considered wrt the default-free information  $\mathcal{F}_t$  and the impact of past default events can be examined.
- The  $\mathcal{F}_t$ -conditional law of default or its density will play an essential role.

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 Applications to pricing with multiple defaults and to counterparty risks.

### Decomposition on default scenarios

• Let  $I \subset \Theta = \{1, \cdots, n\}$  and  $\tau_I = (\tau_i)_{i \in I}$ .

Describe the default scenario by the event

$${\mathcal A}_t' := \Big( igcap_{i \in I} \{ au_i \leq t \} \Big) \cap \Big( igcap_{i \notin I} \{ au_i > t \} \Big)$$

at time t, I denotes the default set.

- All default scenarios:  $\Omega = \bigcup_{l \subset \Theta} A_t^l$ .
- Any *G<sub>t</sub>*-measurable random variable Y<sub>t</sub> can be written in the decomposed form

$$\mathbf{Y}_t = \sum_{I \subset \Theta} \mathbf{1}_{\mathcal{A}_t^I} \mathbf{Y}_t^I(\tau_I)$$

where  $Y'_t(\cdot)$  is  $\mathcal{F}_t \otimes \mathcal{B}(\mathbb{R}'_+)$ -measurable on  $\Omega \times \mathbb{R}'_+$ .

### Random measure of defaults

- We describe the  $\mathcal{F}_t$ -conditional law of  $\boldsymbol{\tau} = (\tau_1, \cdots, \tau_n)$ .
- Random measure μ<sup>τ</sup> on (Ω × ℝ<sup>n</sup><sub>+</sub>, F<sub>∞</sub> ⊗ B(ℝ<sup>n</sup><sub>+</sub>)): for any positive and F<sub>∞</sub> ⊗ B(ℝ<sup>n</sup><sub>+</sub>)-measurable function h<sub>∞</sub>(s), s = (s<sub>1</sub>, · · · , s<sub>n</sub>),

$$\mathbb{E}[h_{\infty}(\boldsymbol{ au})] = \int h_{\infty}(\boldsymbol{s}) \mu^{\boldsymbol{ au}}(\boldsymbol{d}\omega, \boldsymbol{d}\boldsymbol{s}).$$

The restriction μ<sup>τ</sup><sub>t</sub> of μ<sup>τ</sup> on F<sub>t</sub> ⊗ B(ℝ<sup>n</sup><sub>+</sub>) represents the conditional law of τ on F<sub>t</sub>: for any positive and F<sub>t</sub> ⊗ B(ℝ<sup>n</sup><sub>+</sub>)-measurable function h<sub>t</sub>(s),

$$\mathbb{E}[h_t(\boldsymbol{\tau})] = \int h_t(\boldsymbol{s}) \mu^{\boldsymbol{\tau}}(d\omega, d\boldsymbol{s}) = \int h_t(\boldsymbol{s}) \mu_t^{\boldsymbol{\tau}}(d\omega, d\boldsymbol{s}).$$

# General pricing formula

#### Theorem

Let  $Y_T(\tau)$  be the payoff function where  $Y_T(\mathbf{s})$  is positive and  $\mathcal{F}_T \otimes \mathcal{B}(\mathbb{R}^n_+)$ - measurable. Then for any t < T,

$$\mathbb{E}[\mathbf{Y}_{T}(\boldsymbol{\tau})|\mathcal{G}_{t}] = \sum_{I \subset \Theta} \mathbf{1}_{\mathcal{A}_{t}^{\prime}} \frac{\int_{]t, \infty['^{c}} \mathbb{E}[\mathbf{Y}_{T}(\boldsymbol{s}_{I})\mu_{T}^{\boldsymbol{\tau}}|\mathcal{F}_{t}](\boldsymbol{d}\omega, \boldsymbol{d}\boldsymbol{s})}{\int_{]t, \infty['^{c}} \mu_{t}^{\boldsymbol{\tau}}(\boldsymbol{d}\omega, \boldsymbol{d}\boldsymbol{s})} \bigg|_{\boldsymbol{s}_{I}=\tau_{I}}$$
(1)

- On each default scenario, (1) is interpreted as a Randn-Nikodym derivative on *F<sub>t</sub>* ⊗ *B*(ℝ<sup>*l*</sup><sub>+</sub>) and depends on the past default events *τ<sub>l</sub>*.
- ▶ We observe a jump of the price at each default time.
- Choose models of μ<sup>τ</sup> for default correlations and explicit pricing results.

# Default density

#### **Hypothesis**

We say that  $\tau = (\tau_1, \cdots, \tau_n)$  satisfies the density hypothesis if the measure  $\mu^{\tau}$  is absolutely continuous wrt  $\mathbb{P} \otimes \nu^{\tau}$  where  $\nu^{\tau}$  is the law of  $\tau$ .

- ▶  $\nu^{\tau}$  is the marginal measure  $\nu^{\tau}$  of  $\mu^{\tau}$  on  $\mathcal{B}(\mathbb{R}^{n}_{+})$ , i.e.,  $\nu^{\tau}(U) = \mu^{\tau}(\Omega \times U), \forall U \in \mathcal{B}(\mathbb{R}^{n}_{+}).$
- ► Denote by  $\alpha_t(\cdot)$  the density of  $\mu^{\tau}$  wrt  $\mathbb{P} \otimes \nu^{\tau}$  on  $(\Omega \times \mathbb{R}^n_+, \mathcal{F}_t \otimes \mathcal{B}(\mathbb{R}^n_+)),$

$$\mu_t^{\boldsymbol{\tau}}(\boldsymbol{d}\omega,\boldsymbol{d}\boldsymbol{s}) = \alpha_t(\boldsymbol{s}) \mathbb{P}(\boldsymbol{d}\omega) \otimes \nu^{\boldsymbol{\tau}}(\boldsymbol{d}\boldsymbol{s}).$$

For any positive Borel function on  $\mathbb{R}^n_+$ ,

$$\mathbb{E}[f(\boldsymbol{\tau})|\mathcal{F}_t] = \int_{\mathbb{R}^n_+} f(\boldsymbol{s}) \alpha_t(\boldsymbol{s}) \nu^{\boldsymbol{\tau}}(d\boldsymbol{s}).$$

# Pricing with density

- For the k<sup>th</sup>-to-default swap, consider the ordered set of defaults τ<sub>(1)</sub> < · · · < τ<sub>(n)</sub>.
- ► The key term for the pricing is the indicator default process  $\mathbf{1}_{\{\tau_{(k)} > T\}}$  with respect to the market filtration  $\mathcal{G}_t$ :

$$\mathbb{E}[\mathbf{1}_{\{\tau_{(k)} > T\}} | \mathcal{G}_t] = \sum_{|I| < k} \mathbf{1}_{\mathcal{A}_t^I} \sum_{J \supset I, |J| < k} \frac{\int_{]T, \infty[J^c} \int_{]t, T]^{J \setminus I}} \alpha_t(\mathbf{s}) \nu^{\tau}(d\mathbf{s})}{\int_{]t, \infty[J^c} \alpha_t(\mathbf{s}) \nu^{\tau}(d\mathbf{s})} \Big|_{\mathbf{s}_I = \tau_I}$$

For CDO, the cumulative loss  $L_t = \sum_{i=1}^n \mathbf{1}_{\tau_i \leq t}$ , pricing via the *k*tD

$$(L_T-a)_+=\sum_{k\geq a}\min(k-a,1)\mathbf{1}_{\{\tau_{(k)}\leq T\}}.$$

Both ktD and CDO depend on the ordered successive defaults.

# Several remarks

- The density approach can also be applied to ordered defaults, making a link with the top-down models for the loss process L.
- The joint density characterize the correlation structure of defaults in a dynamic manner.
- Part of the joint density can be deduced from the individual default intensity processes. To obtain the whole term structure, we need more information.

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# Application: a contagion risk model

- The density approach and the decomposition methodology can be applied to multiple credit problems in a general way.
- Consider a portfolio of assets whose value process S is subjected to contagion default risks: an N-dimensional G-adapted process S<sub>t</sub> = ∑<sub>I⊂Θ</sub> 1<sub>A'<sub>t</sub></sub>S'<sub>t</sub>(τ<sub>I</sub>).
- The asset value is affected by the counterparty defaults and has a regime switching at each default

$$dS_t^{\prime}(s_l) = S_t^{\prime}(s_l) * (\mu_t^{\prime}(s_l)dt + \Sigma_t^{\prime}(s_l)dW_t), \quad t > s_{\vee t}$$

and there is jump given default

$$S_{\mathbf{S}_{\vee I}}^{I}(\mathbf{s}_{I}) = S_{\mathbf{S}_{\vee I}}^{J}(\mathbf{s}_{J}) * (\mathbf{1} - \gamma_{\mathbf{S}_{\vee I}}^{J,k}(\mathbf{s}_{J})),$$

where  $k = \min\{i \in I | s_i = s_{\vee I}\}$ ,  $J = I \setminus \{k\}$  and  $\gamma^{J,k}(\cdot)$  is  $\mathcal{P}_{\mathbb{F}} \otimes \mathcal{B}(\mathbb{R}^J_+)$ -measurable

### Utility maximization of wealth

- ► Investment strategy is characterized by a G-predictable process  $\pi$  which satisfies  $\pi_t = \sum_l \mathbf{1}_{A'_l} \pi_t^l(\tau_l)$
- ► The wealth process *X* has the decomposed form  $X_t = \sum \mathbf{1}_{A_t^l} X_t^l(\tau_l)$  such that

$$dX_t^{I}(s_l) = X_t^{I}(s_l)\pi_t^{I}(s_l) \cdot \left(\mu_t^{I}(s_l)dt + \Sigma_t^{I}(s_l)dW_t\right), \quad t > s_{\vee I}$$

and

$$X_{s_{\vee I}}^{I}(s_{I}) = X_{s_{\vee I}}^{J}(s_{J})(1 - \pi_{s_{\vee I}}^{J}(s_{J}) \cdot \gamma_{s_{\vee I}}^{J,k})$$

Optimal investment E[U(X<sub>T</sub>)] for admissible trading strategies π or equivalently (π<sup>I</sup>(·))<sub>I⊂Θ</sub>. The admissible set A<sup>I</sup>(s<sub>I</sub>) such that ∫<sub>0</sub><sup>T</sup> |π<sup>I</sup><sub>t</sub>(s<sub>I</sub>)σ<sup>I</sup><sub>t</sub>(s<sub>I</sub>)|<sup>2</sup>dt < ∞ and π<sup>I</sup><sub>t</sub>(s<sub>I</sub>) · γ<sup>I,i</sup><sub>t</sub> < 1 for any *i* ∉ *I* and any *t* ∈]s<sub>∨I</sub>, *T*].

# Optimization decomposition

- The global problem can be treated by the general theory of optimization.
- Motivations for adopting the decomposition methodology :
  - financially, to analyze the impact of past defaults on the optimal trading strategy
  - mathematically, to avoid the difficulty related to jump processes
- By decomposition of the terminal wealth

$$\mathbb{E}[U(X_T)] = \sum_{I \subseteq \Theta} \mathbb{E}[\mathbf{1}_{A_T'} U(X_T'(\tau_I))] = \sum_{I \subseteq \Theta} \mathbb{E}\big[\mathbb{E}[\mathbf{1}_{A_T'} U(X_T'(\tau_I))|\mathcal{F}_T]\big]$$
$$= \mathbb{E}\Big[\sum_{I \subseteq \Theta} \int_{[0,T]' \times ]T, \infty['^c} U(X_T'(s_I)) \alpha_T(\mathbf{s}) d\mathbf{s}\Big].$$

The idea is to consider a family of optimization problems at each default scenario. We shall treat the optimization problem in a backward and recursive way. Let

$$J_{\Theta}(\boldsymbol{x}, \boldsymbol{s}, \pi^{\Theta}) := \mathbb{E}[U(X_{T}^{\Theta}(\boldsymbol{s}))\alpha_{T}(\boldsymbol{s}) \,|\, \mathcal{F}_{\boldsymbol{s}_{\vee\Theta}}]_{X_{\boldsymbol{s}_{\vee\Theta}}^{\Theta}(\boldsymbol{s})=\boldsymbol{x}}$$

and

$$V_{\Theta}(x, \mathbf{s}) = \operatorname*{esssup}_{\pi^{\Theta} \in \mathcal{A}^{\Theta}(\mathbf{s})} J_{\Theta}(x, \mathbf{s}, \pi^{\Theta}).$$

We define recursively for  $I \subset \Theta$ ,

$$\begin{aligned} J_{l}(\boldsymbol{x},\boldsymbol{s}_{l},\boldsymbol{\pi}^{l}) &:= \mathbb{E} \bigg[ U(\boldsymbol{X}_{T}^{l}(\boldsymbol{s}_{l})) \int_{]T,+\infty[^{l^{c}}} \alpha_{T}(\boldsymbol{s}) d\boldsymbol{s}_{l^{c}} \\ &+ \sum_{i \in l^{c}} \int_{]\boldsymbol{s}_{\vee l},T]} V_{l \cup \{i\}} \big( \boldsymbol{X}_{\boldsymbol{s}_{i}}^{l \cup \{i\}}(\boldsymbol{s}_{l \cup \{i\}}), \boldsymbol{s}_{l \cup \{i\}} \big) d\boldsymbol{s}_{i} \, \bigg| \, \mathcal{F}_{\boldsymbol{s}_{\vee l}} \bigg]_{\boldsymbol{X}_{\boldsymbol{s}_{\vee l}}^{l}(\boldsymbol{s}_{l})=\boldsymbol{x}} \end{aligned}$$

and correspondingly

$$V_{I}(x, s_{I}) := \underset{\pi^{I} \in \mathcal{A}^{I}(s_{I})}{essup} J_{I}(x, s_{I}, \pi^{I}).$$

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- ▶ We obtain a family of optimization problems  $(V_l(x, s_l))_{l \subset \Theta}$ .
- ► The whole system need to be dealt with in a recursive manner backwardly, each problem concerning the filtration (*F<sub>t</sub>*)<sub>t≥0</sub> and the time interval [s<sub>∨l</sub>, *T*].
- At each step, the problem V₁ involves the resolution of other ones V₁∪{i}.
- By resolving recursively the problems, we obtain a family of optimal strategies (*π*<sup>*l*</sup>(·))<sub>*l*⊂Θ</sub>, which shall give the optimal strategy of the initial optimization problem.

### Theorem

$$\sup_{\pi \in \mathcal{A}} \mathbb{E}[U(X_{T})]_{X_{0}=x} = V_{\emptyset}(x). \tag{2}$$

### Conclusions

- The default density approach provides a suitable framework for financial problems concerning multiple defaults.
- By the two applications to pricing and to optimal investment, we show that the initial problem in global information filtration can be decomposed to problems in default-free information filtration (often a Brownian filtration), with which we are more familiar.
- The decomposition method also allows to analyze the contagion default impact in an explicit manner.

### **Related works**

- El Karoui, N., Jeanblanc, M. and Jiao, Y. (2009), "What happens after a default: the conditional density approach", *Stochastic Processes and their Applications*, 120(7), 1011-1032.
- El Karoui, N., Jeanblanc, M. and Jiao, Y. (2010), "Modelling successive default events", working paper.
- Jiao, Y. (2010), "Multiple defaults and contagion risks with global and default-free information", working paper.
- Jiao, Y. and Pham, H. (2009), "Optimal investment with counterparty risk: a default-density model approach", to appear in *Finance and Stochastics*.

Thanks for your attention !