

# Dependence Properties of Dynamic Credit Risk Models

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## Outline of Presentation

- Reduced-form portfolio credit risk model with default feedback (contagion)
- Concept of association and its properties
- Association of default intensities and implications for default times
- Properties of associated default times
- Association of accumulated hazard processes
- Applications to credit swaps

## Reduced-Form Portfolio Credit Risk Model

On a filtered probability space  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$  consider for every obligor  $i \in \{1, \dots, d\}$

- an adapted, increasing, right-continuous, accumulated hazard process  $\Lambda_i = \{\Lambda_i(t)\}_{t \geq 0}$  with  $\Lambda_i(0) = 0$
- a standard exponentially distributed threshold  $E_i$ ,
- the default time  $\tau_i = \inf\{t \geq 0 \mid \Lambda_i(t) \geq E_i\}$  with

$$\mathbb{P}(\tau_i > t \mid \Lambda_i(t)) \stackrel{\text{a.s.}}{=} e^{-\Lambda_i(t)}, \quad t \geq 0,$$

- the default indicator process  $Y_i(t) = 1_{[E_i, \infty)}(\Lambda_i(t))$ ,
- possibly a default intensity process  $\lambda_i$  satisfying

$$\Lambda_i(t) = \int_0^t \lambda_i(s) ds, \quad t \geq 0.$$

## Model 1: Default Feedback by Default Intensities

- $\Psi = \{\Psi_t\}_{t \geq 0}$  an  $\mathbb{R}^m$ -valued environment process (contains relevant economic information like interest rates, stock price indices, economic indices, etc.)
- Thresholds  $E = (E_1, \dots, E_d)$  independent of  $\Psi$
- Default intensity  $\lambda_i(t, \Psi_t, Y_t)$  of obligor  $i \in \{1, \dots, d\}$

**Aim:** Investigate and control how dependence through environment process  $\Psi$  and previous defaults, given by the default indicator process  $Y_t = (Y_1(t), \dots, Y_d(t))$ , transfers to dependence of default times  $\tau_1, \dots, \tau_d$ .

## Definition of Association

- An  $\mathbb{R}^d$ -valued random vector  $X = (X_1, \dots, X_d)$  and its distribution  $\mathcal{L}(X)$  are called **associated**, if

$$\text{Cov}(f(X), g(X)) \geq 0$$

for all measurable, componentwise increasing functions  $f, g : \mathbb{R}^d \rightarrow \mathbb{R}$  for which  $f(X)$ ,  $g(X)$  and the product  $f(X)g(X)$  are integrable.

- An  $\mathbb{R}^d$ -valued process  $\{X_t\}_{t \geq 0}$  is called **associated** if for all  $k \in \mathbb{N}$  and times  $0 \leq t_1 < \dots < t_k$  the  $\mathbb{R}^{dk}$ -valued vector  $(X(t_1), \dots, X(t_k))$  is associated.

## Properties of Association\*

- If  $X_1, \dots, X_d$  are independent, then the random vector  $X = (X_1, \dots, X_d)$  is associated.
- If  $X = (X_1, \dots, X_d)$  and  $Y = (Y_1, \dots, Y_k)$  are associated random vectors, which are independent, then  $(X_1, \dots, X_d, Y_1, \dots, Y_k)$  is associated.
- If  $X = (X_1, \dots, X_d)$  is associated, then the vector  $(f_1(X), \dots, f_k(X))$  is associated for every  $k \in \mathbb{N}$  and every choice of measurable increasing (or decreasing) functions  $f_1, \dots, f_k : \mathbb{R}^d \rightarrow \mathbb{R}$ .
- If  $\{X_n\}_{n \in \mathbb{N}}$  is a sequence of associated  $\mathbb{R}^d$ -valued random vectors and  $X_n \xrightarrow{d} X$ , then  $X$  is associated.

\*see Esary, Proschan, Walkup (1967), Ann. Math. Statist. 38

## Association is a Copula Property

Let  $X = (X_1, \dots, X_d)$  be an  $\mathbb{R}^d$ -valued random vector with marginal distributions  $F_1, \dots, F_d$ . Define the copula  $C_X : [0, 1]^d \rightarrow [0, 1]$  of  $X$  as distribution function of  $(F_1(X_1), \dots, F_d(X_d))$ .

**Lemma:**  $X$  is associated  $\iff C_X$  is associated.

**Proof:** “ $\implies$ ” By property of association.

“ $\impliedby$ ” Use the lower quantile functions

$$F_i^{\leftarrow}(t) := \inf\{x \in \mathbb{R} \mid F_i(x) \geq t\}, \quad t \in [0, 1],$$

for  $i \in \{1, \dots, d\}$  to see that

$$(X_1, \dots, X_d) \stackrel{\text{a.s.}}{=} (F_1^{\leftarrow}(F_1(X_1)), \dots, F_d^{\leftarrow}(F_d(X_d))).$$



## Application to Defaultable Zero-Coupon Bonds

Let  $R_t$  denote the integrated stochastic interest intensity, i.e.,  $e^{-R_t}$  is the factor for discounting from  $t$  to 0. Let  $\Lambda_t$  denote the accumulated hazard for default up to  $t$ .

**Lemma:** For a defaultable payment of 1 at time  $t$ , assume that  $(R_t, \Lambda_t)$  is associated under an equivalent pricing measure  $\mathbb{P}$ . Then for the price at time 0:

$$\mathbb{E}[e^{-R_t} \mathbf{1}_{\{\tau > t\}}] \geq \mathbb{E}[e^{-R_t}] \mathbb{P}(\tau > t).$$

**Proof:** The vector  $(e^{-R_t}, e^{-\Lambda_t})$  is associated. Since  $\{\tau > t\} = \{\Lambda_t < E\}$  and  $\mathbb{P}(\Lambda_t < E | \Lambda_t, R_t) = e^{-\Lambda_t}$ , the definition of association implies

$$\mathbb{E}[e^{-R_t} \mathbf{1}_{\{\tau > t\}}] = \mathbb{E}[e^{-R_t} e^{-\Lambda_t}] \geq \mathbb{E}[e^{-R_t}] \mathbb{E}[e^{-\Lambda_t}].$$

## Monotone Mixtures and Association

**Definition:**  $X = (X_1, \dots, X_d)$  is called a **monotone mixture** of  $\Theta = (\Theta_1, \dots, \Theta_k)$  if for every measurable, bounded and componentwise increasing  $f : \mathbb{R}^d \rightarrow \mathbb{R}$  there exists a componentwise increasing  $h : \mathbb{R}^k \rightarrow \mathbb{R}$  such that

$$h(\Theta) \stackrel{\text{a.s.}}{=} \mathbb{E}[f(X) | \Theta].$$

**Lemma:**\* If the conditional distribution  $\mathcal{L}(X|\Theta)$  is associated,  $\Theta$  is associated and  $X$  is a monotone mixture of  $\Theta$ , then the vector  $(X, \Theta)$  is associated.

\*see K. Jogdeo (1978), Ann. Statist. 6, 232–234.

## Conditional Increasing in Sequence and Association

**Definition:**  $X = (X_1, \dots, X_d)$  is called **conditional increasing in sequence (CIS)** if for every  $k \in \{2, \dots, d\}$  and bounded increasing  $f : \mathbb{R} \rightarrow \mathbb{R}$

$$(x_1, \dots, x_{k-1}) \mapsto \mathbb{E}[f(X_k) \mid X_1 = x_1, \dots, X_{k-1} = x_{k-1}]$$

is increasing in every  $x_1, \dots, x_{k-1}$ .

**Lemma:**\* If  $X$  is conditional increasing in sequence, then  $X$  is associated.

\*cf. A. Müller & D. Stoyan, *Comparison Methods for Stochastic Models and Risks*, Wiley (2002), Theorem 3.10.11.

## Implication of Associated (Integrated) Intensities

Write  $\lambda_t = (\lambda_1(t), \dots, \lambda_d(t))$  for the joint  $\mathbb{R}^d$ -valued intensity process and  $\Lambda_t = (\Lambda_1(t), \dots, \Lambda_d(t))$  for the integrated version (accumulated hazard) at time  $t \geq 0$ .

**Lemma:** (B. & S.)

- If  $\{\lambda_t\}_{t \geq 0}$  is associated and càdlàg, then  $\{\Lambda_t\}_{t \geq 0}$  is associated.
- If  $\{\Lambda_t\}_{t \geq 0}$  is associated, componentwise right-continuous and  $\Lambda_i(t) \nearrow \infty$  a.s. as  $t \rightarrow \infty$ , and the thresholds  $(E_1, \dots, E_d)$  are associated and independent of  $\{\Lambda_t\}_{t \geq 0}$ , then the default times  $(\tau_1, \dots, \tau_d)$  are associated.

## Association in Model 1 with Default Intensities

Theorem: (B. & S.) If

- environment process  $\Psi$  is associated,
- $\lambda_i(t, \Psi_t, Y_t)$  is increasing in 2<sup>nd</sup> and 3<sup>rd</sup> argument,
- $\int_0^\infty \lambda_i(t, \Psi_t, y) dt \stackrel{\text{a.s.}}{=} \infty$  for every  $y \in \{0, 1\}^d$ ,  $y_i = 0$ ,
- technical conditions (suitable meas. & continuity),

then the accumulated hazard process

$$\Lambda_t = \left( \int_0^t \lambda_i(s, \Psi_s, Y_s) ds \right)_{i=1, \dots, d}, \quad t \geq 0,$$

is associated, and the default times  $\tau = (\tau_1, \dots, \tau_d)$  are associated, too.

## Association and Positive Supermodular Dependence

**Definition:**  $f : \mathbb{R}^d \rightarrow \mathbb{R}$  is called **supermodular** if

$$f(x) + f(y) \leq f(x \vee y) + f(x \wedge y), \quad \forall x, y \in \mathbb{R}^d.$$

**Definition:** Let  $X = (X_1, \dots, X_d)$  be a random vector and  $X^\perp = (X_1^\perp, \dots, X_d^\perp)$  a copy with independent components. Then  $X$  and its distribution  $\mathcal{L}(X)$  are called **positive supermodular dependent** (PSD) if

$$\mathbb{E}[f(X^\perp)] \leq \mathbb{E}[f(X)]$$

for all measurable, supermodular  $f : \mathbb{R}^d \rightarrow \mathbb{R}$  for which the expectations exist.

**Lemma:**  $X$  is associated  $\implies X$  is PSD

## Implications of Associated Default Times

If  $(\tau_1, \dots, \tau_d)$  are associated, then, for all non-void  $I \subset \{1, \dots, d\}$  and  $\{t_i\}_{i \in I} \subset [0, \infty)$ ,

$$\mathbb{P}(\tau_i^\perp > t_i \text{ for all } i \in I) \leq \mathbb{P}(\tau_i > t_i \text{ for all } i \in I),$$

$$\mathbb{P}(\tau_i^\perp \leq t_i \text{ for all } i \in I) \leq \mathbb{P}(\tau_i \leq t_i \text{ for all } i \in I),$$

because the indicator functions are supermodular.

**Definition:** A r.v.  $X$  is smaller in **usual stochastic order** than  $Y$ , if  $\mathbb{P}(X > t) \leq \mathbb{P}(Y > t)$  for all  $t \in \mathbb{R}$ .

**Consequence:** With  $\leq_{\text{st}}$  for usual stochastic order,

$$\min_{i \in I} \tau_i^\perp \leq_{\text{st}} \min_{i \in I} \tau_i \quad \text{and} \quad \max_{i \in I} \tau_i \leq_{\text{st}} \max_{i \in I} \tau_i^\perp.$$

## Implications of Associated Default Times (Cont.)

For default times  $\tau_1, \dots, \tau_d$  let  $\tau_{1:d} \leq \dots \leq \tau_{d:d}$  denote the order statistics.

**Lemma:**  $(\tau_1, \dots, \tau_d)$  is associated  
 $\implies (\tau_{1:d}, \dots, \tau_{d:d})$  is associated.

**Proof:** Since for  $k \in \{1, \dots, d\}$

$$\tau_{k:d} = \min_{\substack{I \subset \{1, \dots, d\} \\ |I|=k}} \max_{i \in I} \tau_i,$$

every  $\tau_{k:d}$  is an increasing function of  $(\tau_1, \dots, \tau_d)$ .



## Model 2: Accumulated Hazard Processes

For every obligor  $i \in \{1, \dots, d\}$  and time  $t \geq 0$  put

$$\Lambda_i(t) = \Psi_i(t) + \sum_{j \in \{1, \dots, d\} \setminus \{i\}} \mathbf{1}_{\{\tau_j \leq t\}} \Gamma_{i,j}(t - \tau_j).$$

**Theorem:** (B. & S.) Assume that

- $\Psi_i$  and  $\Gamma_{i,j}$  are positive processes with increasing paths,
- $\{\Gamma_t\}_{t \geq 0}$  with  $\Gamma_t = (\Gamma_{1,2}(t), \dots, \Gamma_{d,d-1}(t))$ , thresholds  $(E_1, \dots, E_d)$ , and the environment process  $\{\Psi_t\}_{t \geq 0}$  are associated and independent,
- technical conditions (suitable continuity,  $\Lambda_i(t) \rightarrow \infty$ ).

Then the accumulated hazard processes  $\{\Lambda_t\}_{t \geq 0}$  as well as the default times  $\tau_1, \dots, \tau_d$  are associated.

## Application to Credit Swap Contracts

- Reference party  $R$  issues a bond with maturity  $T^*$ .
- Party  $A$  buys the bond and pays swap rate  $c$  continuously to  $B$  until swap maturity  $T \leq T^*$  or  $\tau_A$  ( $A$  pays even if  $B$  or  $R$  have already defaulted).
- Party  $B$  pays 1 € at time  $T$  to  $A$  if  $\tau_R \leq T$  and  $\tau_B > T$ .

With pricing measure  $\mathbb{P}$  and spot rate process  $\{r_t\}_{t \in [0, T]}$ , the fair swap rate at time 0 is

$$c = \frac{\mathbb{E}\left[\exp\left(-\int_0^T r_s ds\right) \mathbf{1}_{\{\tau_B > T, \tau_R \leq T\}}\right]}{\mathbb{E}\left[\int_0^T \exp\left(-\int_0^t r_s ds\right) \mathbf{1}_{\{\tau_A > t\}} dt\right]}.$$

## Application to Credit Swap Contracts (Cont.)

**Lemma:** Assume that  $(\tau_B, \tau_R)$  is associated and independent of the spot rate process. Then  $c \leq c^\perp$ , where  $c^\perp$  denotes the fair swap rate when  $\tau_B^\perp$  and  $\tau_R^\perp$  are independent with  $\tau_B \stackrel{d}{=} \tau_B^\perp$  and  $\tau_R \stackrel{d}{=} \tau_R^\perp$ .

**Proof:** Note that

$$\mathbf{1}_{\{\tau_B > T, \tau_R \leq T\}} = \mathbf{1}_{\{\tau_R \leq T\}} - \mathbf{1}_{\{\tau_B \leq T, \tau_R \leq T\}}.$$

Since association of  $(\tau_B, \tau_R)$  implies positive supermodular dependence,

$$\mathbb{P}(\tau_B^\perp \leq T, \tau_R^\perp \leq T) \leq \mathbb{P}(\tau_B \leq T, \tau_R \leq T),$$

which yields the statement.

## Generalization to $k^{\text{th}}$ -to-Default Swaps

Suppose  $A$  buys a collateralized debt obligation which defaults if the  $k^{\text{th}}$  default happens in a portfolio of  $d$  obligors with default times  $\tau_1, \dots, \tau_d$ . Fair swap rate:

$$c_k = \frac{\mathbb{E} \left[ \exp \left( - \int_0^T r_s ds \right) \mathbf{1}_{\{\tau_B > T, \tau_{k:d} \leq T\}} \right]}{\mathbb{E} \left[ \int_0^T \exp \left( - \int_0^t r_s ds \right) \mathbf{1}_{\{\tau_A > t\}} dt \right]}.$$

**Lemma:** Assume that  $(\tau_B, \tau_1, \dots, \tau_d)$  is associated and independent of the spot rate process. Then  $c_k \leq c_k^\perp$ , where  $c_k^\perp$  denotes the fair swap rate when  $\tau_B^\perp, \tau_1^\perp, \dots, \tau_d^\perp$  are independent.

**Proof:** Note that  $(\tau_B, \tau_{k:d})$  is an increasing function of  $(\tau_B, \tau_1, \dots, \tau_d)$ , hence associated. Use previous lemma.