# Comparison results for credit risk portfolios

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# Introduction

- In credit risk portfolio modelling, dependence among default events is a crucial assumption
- We will investigate tranches of Collateralized Debt Obligation (CDO)
- Which is the impact of dependence on
  - CDO tranche premiums ?
  - Risk measures on the aggregate loss associated with the reference portfolio ?



#### CDO tranches

Comparison results Application to several popular CDO pricing models Conclusion

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#### Comparison results

- Motivation
- De Finetti theorem and factor representation
- Stochastic orders
- Main results

### Application to several popular CDO pricing models

- Factor copula approaches
- Structural model
- Multivariate Poisson model



# CDO tranches

- Slice the credit portfolio into different risk levels or CDO tranches
- ex: CDO tranche on standardized Index such as CDX North America or Itraxx Europe



- [0, 3%] equity tranche is subordinated to [3, 6%] junior mezzanine tranche
- [3, 6%] junior mezzanine tranche is subordinated to [6, 9%] mezzanine tranche and so on,...



# CDO tranches

- Credit portfolio with *n* reference entities
- $\tau_1, \ldots, \tau_n$  default times
- $(D_1, \ldots, D_n) = (1_{\{\tau_1 \leq t\}}, \ldots, 1_{\{\tau_n \leq t\}})$  default indicators at time t
- $M_1, \ldots, M_n$  losses given default assumed to be independent of default times
- Aggregate loss:

$$L_t = \sum_{i=1}^n M_i \mathbb{1}_{\{\tau_i \le t\}}$$

• Dynamics of Losses  $\mathcal{L}_t^{[a,b]}$  affecting CDO tranche [a,b]:





### CDO tranches

•  $L_t^{[a,b]}$  has a call spread payoff with respect to the aggregate loss:



- Loss on CDO tranche [a, b]:  $L_t^{[a, b]} = (L_t a)^+ (L_t b)^+$
- Computation of CDO Tranche premiums only involves call options on the aggregate loss *L*<sub>t</sub>:

$$E\left[(L_t-a)^+
ight]-E\left[(L_t-b)^+
ight]$$

• for different time horizons t



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# Motivation

- Specify the dependence structure of default indicators  $D_1, \ldots, D_n$  which leads to:
  - an increase of the value of call options E [(L<sub>t</sub> a)<sup>+</sup>] for all strike level a > 0
  - an increase of convex risk measures on *L<sub>t</sub>* (TVaR, Wang risk measures)
- Comparison between homogeneous credit portfolios
  - $D_1, \ldots, D_n$  are assumed to be exchangeable Bernoulli random variables
  - De Finetti's theorem leads to a factor representation of  $D_1, \ldots, D_n$
- Application to several popular CDO pricing models



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### De Finetti theorem and factor representation

• Homogeneity assumption: default indicators  $D_1, \ldots, D_n$  forms an exchangeable Bernoulli random vector

#### Definition (Exchangeability)

A random vector  $(D_1, \ldots, D_n)$  is exchangeable if its distribution function is invariant for every permutations of its coordinates:  $\forall \sigma \in S_n$ 

$$(D_1,\ldots,D_n)\stackrel{d}{=}(D_{\sigma(1)},\ldots,D_{\sigma(n)})$$

Same marginals



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### De Finetti theorem and factor representation

- Assume that  $D_1, \ldots, D_n, \ldots$  is an exchangeable sequence of Bernoulli random variables
- Thanks to de Finetti's theorem, there exists a random factor  $\tilde{p}$  such that
- $D_1, \ldots, D_n$  are conditionally independent given  $\tilde{p}$
- Denote by  $F_{\tilde{p}}$  the distribution function of  $\tilde{p}$ , then:

$$P(D_1 = d_1, \ldots, D_n = d_n) = \int_0^1 p^{\sum_i d_i} (1-p)^{n-\sum_i d_i} F_{\bar{p}}(dp)$$

Finite exchangeability only leads to a sign measure Jaynes (1986) *p* is characterized by:

$$\frac{1}{n}\sum_{i=1}^n D_i \xrightarrow{\text{a.s.}} \tilde{p} \quad \text{as} \quad n \to \infty$$

•  $\tilde{p}$  is exactly the loss of the infinitely granular portfolio (Bâle 2 terminology)



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### Stochastic orders

- The convex order compares the dispersion level of two random variables
- Convex order: X ≤<sub>cx</sub> Y if E[f(X)] ≤ E[f(Y)] for all convex functions f
- Stop-loss order:  $X \leq_{sl} Y$  if  $E[(X K)^+] \leq E[(Y K)^+]$  for all  $K \in \mathbb{R}$

• 
$$X \leq_{sl} Y$$
 and  $E[X] = E[Y] \Leftrightarrow X \leq_{cx} Y$ 

•  $X \leq_{cx} Y$  if E[X] = E[Y] and  $F_X$ , the distribution function of X and  $F_Y$ , the distribution function of Y are such that:





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# Supermodular order

- The supermodular order captures the dependence level among coordinates of a random vector
- $(X_1, \ldots, X_n) \leq_{sm} (Y_1, \ldots, Y_n)$  if  $E[f(X_1, \ldots, X_n)] \leq E[f(Y_1, \ldots, Y_n)]$  for all supermodular function f

#### Definition (Supermodular function)

A function  $f : \mathbb{R}^n \to \mathbb{R}$  is supermodular if for all  $x \in \mathbb{R}^n$ ,  $1 \le i < j \le n$  and  $\varepsilon, \delta > 0$  holds

$$f(x_1,\ldots,x_i+\varepsilon,\ldots,x_j+\delta,\ldots,x_n) - f(x_1,\ldots,x_i+\varepsilon,\ldots,x_j,\ldots,x_n)$$

 $\geq f(x_1,\ldots,x_i,\ldots,x_j+\delta,\ldots,x_n) - f(x_1,\ldots,x_i,\ldots,x_j,\ldots,x_n)$ 

• Consequences of new defaults are always worse when other defaults have already occurred



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### Review of literature

Müller(1997)

Stop-loss order for portfolios of dependent risks

$$(D_1,\ldots,D_n)\leq_{sm}(D_1^*,\ldots,D_n^*)\Rightarrow\sum_{i=1}^nM_iD_i\leq_{sl}\sum_{i=1}^nM_iD_i^*$$



### Bäuerle and Müller(2005)

Stochastic orders ans risk measures: Consistency and bounds

$$X \leq_{sl} Y \Rightarrow \rho(X) \leq \rho(Y)$$

for all law-invariant, convex risk measures  $\rho$ 



### Lefèvre and Utev(1996)

Comparing sums of exchangeable Bernoulli random variables

$$\tilde{p} \leq_{cx} \tilde{p}^* \Rightarrow \sum_{i=1}^n D_i \leq_{sl} \sum_{i=1}^n D_i^*$$



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### Main results

- Let us compare two credit portfolios with aggregate loss  $L_t = \sum_{i=1}^n M_i D_i$ and  $L_t^* = \sum_{i=1}^n M_i D_i^*$
- Let  $D_1, \ldots, D_n$  be exchangeable Bernoulli random variables associated with the mixture probability  $\tilde{p}$
- Let D<sub>1</sub><sup>\*</sup>,..., D<sub>n</sub><sup>\*</sup> exchangeable Bernoulli random variables associated with the mixture probability p̃<sup>\*</sup>

#### Theorem

$$\tilde{\rho} \leq_{cx} \tilde{\rho}^* \Rightarrow (D_1, \dots, D_n) \leq_{sm} (D_1^*, \dots, D_n^*)$$

• In particular, if  $\tilde{p} \leq_{cx} \tilde{p}^*$ , then:

- $E[(L_t a)^+] \le E[(L_t^* a)^+]$  for all a > 0.
- $ho(L_t) \leq 
  ho(L_t^*)$  for all convex risk measures ho



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### Main results

- Let  $D_1, \ldots, D_n, \ldots$  be exchangeable Bernoulli random variables associated with the mixture probability  $\tilde{p}$
- Let D<sup>\*</sup><sub>1</sub>,..., D<sup>\*</sup><sub>n</sub>,... be exchangeable Bernoulli random variables associated with the mixture probability p<sup>\*</sup>

Theorem (reverse implication)

$$(D_1,\ldots,D_n)\leq_{sm} (D_1^*,\ldots,D_n^*), \forall n\in\mathbb{N}\Rightarrow \tilde{p}\leq_{cx} \tilde{p}^*.$$



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# Ordering of CDO tranche premiums



### Burtschell, Gregory, and Laurent(2008)

A comparative analysis of CDO pricing models

- Analysis of the dependence structure within some factor copula models such as:
  - Gaussian, Student t, Double t, Clayton, Marshall-Olkin copula
- An increase of the dependence parameter leads to:
  - a decrease of [0%, b] equity tranches premiums (which guaranties the uniqueness of the market base correlation)
  - an increase of [a, 100%] senior tranches premiums



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### Additive factor copula approaches

• The dependence structure of default times is described by some latent variables  $V_1, \ldots, V_n$  such that:

• 
$$V_i = \rho V + \sqrt{1 - \rho^2} \bar{V}_i, \ i = 1 \dots n$$

•  $V, \bar{V}_i, i = 1 \dots n$  independent

• 
$$\tau_i = G^{-1}(H_{\rho}(V_i)), \ i = 1 \dots n$$

- G: distribution function of  $\tau_i$
- $H_{\rho}$ : distribution function of  $V_i$
- $D_i = \mathbb{1}_{\{\tau_i \leq t\}}, i = 1 \dots n$  are conditionally independent given V
- $\frac{1}{n}\sum_{i=1}^{n}D_i \xrightarrow{a.s} E[D_i \mid V] = P(\tau_i \leq t \mid V) = \tilde{p}$



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### Additive factor copula approaches

#### Theorem

For any fixed time horizon t, denote by  $D_i = 1_{\{\tau_i \leq t\}}$ , i = 1...n and  $D_i^* = 1_{\{\tau_i^* \leq t\}}$ , i = 1...n the default indicators corresponding to (resp.)  $\rho$  and  $\rho^*$ , then:

$$\rho \leq \rho^* \Rightarrow \tilde{p} \leq_{cx} \tilde{p}^* \Rightarrow (D_1, \dots, D_n) \leq_{sm} (D_1^*, \dots, D_n^*)$$

- This framework includes popular factor copula models:
  - One factor Gaussian copula the industry standard for the pricing of CDO tranches
  - Double t: Hull and White(2004)
  - NIG, double NIG: Guegan and Houdain(2005), Kalemanova, Schmid and Werner(2007)
  - Double Variance Gamma: Moosbrucker(2006)



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# Archimedean copula

- Schönbucher and Schubert(2001), Gregory and Laurent(2003), Madan *et al.*(2004), Friend and Rogge(2005)
  - V is a positive random variable with Laplace transform  $\varphi^{-1}$
  - $U_1, \ldots, U_n$  are independent Uniform random variables independent of V
  - $V_i = \varphi^{-1}\left(-\frac{\ln U_i}{V}\right), i = 1...n$  (Marshall and Olkin (1988))
    - $(V_1,\ldots,V_n)$  follows a  $\varphi$ -archimedean copula
    - $P(V_1 \le v_1, ..., V_n \le v_n) = \varphi^{-1}(\varphi(v_1) + ... + \varphi(v_n))$
  - $\tau_i = G^{-1}(V_i)$ 
    - G: distribution function of  $\tau_i$
  - $D_i = 1_{\{\tau_i \leq t\}}, i = 1 \dots n$  independent knowing V
  - $\frac{1}{n}\sum_{i=1}^{n}D_{i} \xrightarrow{a.s} E[D_{i} \mid V] = P(\tau_{i} \leq t \mid V)$



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### Archimedean copula

• Conditional default probability:  $\tilde{p} = \exp \{-\varphi(G(t)V)\}$ 

Copula	Generator $arphi$	Parameter
Clayton	$t^{- heta}-1$	$ heta \geq 0$
Gumbel	$(-\ln(t))^{ heta}$	$ heta \geq 1$
Franck	$-\ln\left[(1-e^{- heta t})/(1-e^{- heta}) ight]$	$ heta\in I\!\!R^*$

#### Theorem

$$\theta \leq \theta^* \Rightarrow \tilde{p} \leq_{cx} \tilde{p}^* \Rightarrow (D_1, \dots, D_n) \leq_{sm} (D_1^*, \dots, D_n^*)$$



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### Archimedean copula



- Clayton copula
- Mixture distributions are ordered with respect to the convex oder



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### Structural model

### Hull, Predescu and White(2005)

- Consider *n* firms
- Let  $V_{i,t}$ ,  $i = 1 \dots n$  be their asset dynamics

$$V_{i,t} = \rho V_t + \sqrt{1 - \rho^2} \overline{V}_{i,t}, \quad i = 1 \dots n$$

- V,  $\overline{V}_i$ ,  $i = 1 \dots n$  are independent standard Wiener processes
- Default times as first passage times:

 $\tau_i = \inf\{t \in \mathbf{R}^+ | V_{i,t} \le f(t)\}, \ i = 1 \dots n, \ f : \mathbf{R} \to \mathbf{R} \text{ continuous}$ 

•  $D_i = 1_{\{\tau_i \leq T\}}$ ,  $i = 1 \dots n$  are conditionally independent given  $\sigma(V_t, t \in [0, T])$ 



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### Structural model

#### Theorem

For any fixed time horizon *T*, denote by  $D_i = 1_{\{\tau_i \leq \tau\}}$ ,  $i = 1 \dots n$  and  $D_i^* = 1_{\{\tau_i^* \leq \tau\}}$ ,  $i = 1 \dots n$  the default indicators corresponding to (resp.)  $\rho$  and  $\rho^*$ , then:

 $\rho \leq \rho^* \Rightarrow (D_1, \ldots, D_n) \leq_{sm} (D_1^*, \ldots, D_n^*)$ 



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### Structural model



•  $\frac{1}{n}\sum_{i=1}^{n}D_{i} \xrightarrow{a.s} \tilde{p}$ 

• 
$$\frac{1}{n} \sum_{i=1}^{n} D_i^* \xrightarrow{a.s} \tilde{p}^*$$

• Empirically, mixture probabilities are ordered with respect to the convex order:  $\tilde{p} \leq_{cx} \tilde{p}^*$ 



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### Multivariate Poisson model

### Duffie(1998), Lindskog and McNeil(2003), Elouerkhaoui(2006)

- $\bar{N}_t^i$  Poisson with parameter  $\bar{\lambda}$ : idiosyncratic risk
- $N_t$  Poisson with parameter  $\lambda$ : systematic risk
- $(B_i^i)_{i,j}$  Bernoulli random variable with parameter p
- All sources of risk are independent

• 
$$N_t^i = \bar{N}_t^i + \sum_{j=1}^{N_t} B_j^i, \ i = 1 \dots n$$

• 
$$\tau_i = \inf\{t > 0 | N_t^i > 0\}, \ i = 1 \dots n$$



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### Multivariate Poisson model

- Dependence structure of  $(\tau_1, \ldots, \tau_n)$  is the Marshall-Olkin copula
- $\tau_i \sim Exp(\bar{\lambda} + p\lambda)$
- $D_i = \mathbb{1}_{\{\tau_i \leq t\}}, i = 1 \dots n$  are conditionally independent given  $N_t$
- $\frac{1}{n}\sum_{i=1}^{n}D_{i} \xrightarrow{a.s} E[D_{i} \mid N_{t}] = P(\tau_{i} \leq t \mid N_{t})$
- Conditional default probability:

$$\tilde{p} = 1 - (1 - p)^{N_t} \exp(-\bar{\lambda}t)$$



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### Multivariate Poisson model

- Comparison of two multivariate Poisson models with parameter sets  $(\bar{\lambda},\lambda,p)$  and  $(\bar{\lambda}^*,\lambda^*,p^*)$
- Supermodular order comparison requires equality of marginals:  $\bar{\lambda} + p\lambda = \bar{\lambda}^* + p^*\lambda^*$
- 3 comparison directions:

• 
$$p = p^*$$
:  $\overline{\lambda}$  v.s  $\lambda$   
•  $\lambda = \lambda^*$ :  $\overline{\lambda}$  v.s  $p$   
•  $\overline{\lambda} = \overline{\lambda}^*$ :  $\lambda$  v.s  $p$ 



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### Multivariate Poisson model

#### Theorem $(p = p^*)$

Let parameter sets  $(\bar{\lambda}, \lambda, p)$  and  $(\bar{\lambda}^*, \lambda^*, p^*)$  be such that  $\bar{\lambda} + p\lambda = \bar{\lambda}^* + p\lambda^*$ , then:

$$\lambda \leq \lambda^*, \ ar{\lambda} \geq ar{\lambda}^* \Rightarrow \widetilde{p} \leq_{cx} \widetilde{p}^* \Rightarrow (D_1, \dots, D_n) \leq_{sm} (D_1^*, \dots, D_n^*)$$



- Computation of  $E[(L_t a)^+]$ :
  - 30 names
  - $M_i = 1, i = 1 \dots n$
- When  $\lambda$  increases, the aggregate loss increases with respect to stop-loss order



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### Multivariate Poisson model

#### Theorem $(\lambda = \lambda^*)$

Let parameter sets  $(\bar{\lambda}, \lambda, p)$  and  $(\bar{\lambda}^*, \lambda^*, p^*)$  be such that  $\bar{\lambda} + p\lambda = \bar{\lambda}^* + p^*\lambda$ , then:

$$p \leq p^*, \ ar{\lambda} \geq ar{\lambda}^* \Rightarrow ar{p} \leq_{\sf cx} ar{p}^* \Rightarrow (D_1, \dots, D_n) \leq_{\sf sm} (D_1^*, \dots, D_n^*)$$



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### Multivariate Poisson model

#### Theorem $(\lambda = \lambda^*)$

Let parameter sets  $(\bar{\lambda}, \lambda, p)$  and  $(\bar{\lambda}^*, \lambda^*, p^*)$  be such that  $\bar{\lambda} + p\lambda = \bar{\lambda}^* + p^*\lambda$ , then:

$$p \leq p^*, \ ar{\lambda} \geq ar{\lambda}^* \Rightarrow ar{p} \leq_{\sf cx} ar{p}^* \Rightarrow (D_1, \dots, D_n) \leq_{\sf sm} (D_1^*, \dots, D_n^*)$$



- Computation of  $E[(L_t K)^+]$ :
  - 30 names
  - $M_i = 1, i = 1 \dots n$
- When *p* increases, the aggregate loss increases with respect to stop-loss order



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### Multivariate Poisson model

### Theorem $(ar{\lambda}=ar{\lambda}^*)$

Let parameter sets  $(\bar{\lambda}, \lambda, p)$  and  $(\bar{\lambda}^*, \lambda^*, p^*)$  be such that  $p\lambda = p^*\lambda^*$ , then:

$$p \leq p^*, \ \lambda \geq \lambda^* \Rightarrow \tilde{p} \leq_{\mathsf{cx}} \tilde{p}^* \Rightarrow (D_1, \dots, D_n) \leq_{\mathsf{sm}} (D_1^*, \dots, D_n^*)$$



- Computation of  $E[(L_t K)^+]$ :
  - 30 names
  - $M_i = 1, i = 1 \dots n$
- When p increases, the aggregate loss increases with respect to stop-loss order



# Conclusion

- When considering an exchangeable vector of default indicators, the conditional independence assumption is not restrictive thanks to de Finetti's theorem
- The mixture probability (the factor) can be viewed as the loss of an infinitely granular portfolio
- We completely characterize the supermodular order between exchangeable default indicator vectors in term of the convex ordering of corresponding mixture probabilities
- We show that the mixture probability is the key input to study the impact of dependence on CDO tranche premiums
- Comparison analysis can be performed with the same method within a large number of popular CDO pricing models

