# Comparison results for exchangeable credit risk portfolios

## Areski COUSIN

ISFA, Université Lyon 1

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De Finetti theorem and factor representation Stochastic orders Main results

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- Factor copula approaches
- Multivariate Poisson model
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# CDO tranches

- Credit portfolio with *n* reference entities
- $\tau_1, \ldots, \tau_n$  default times
- $(D_{1,t},\ldots,D_{n,t})=(1_{\{ au_1\leq t\}},\ldots,1_{\{ au_n\leq t\}})$  default indicators at time t
- $M_1, \ldots, M_n$  losses given default assumed to be independent of default times
- Aggregate loss:

$$L_t = \sum_{i=1}^n M_i D_{i,t}$$

- Which is the impact of dependence on
  - CDO tranche premiums ?
  - Risk measures on the aggregate loss associated with the reference portfolio ?

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# De Finetti theorem and factor representation

 Homogeneity assumption: default indicators D<sub>1</sub>,..., D<sub>n</sub> form 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 unique random factor p
  such that
- $D_1,\ldots,D_n$  are conditionally independent given  $\widetilde{p}$
- Denote by  $F_{\tilde{p}}$  the distribution function of  $\tilde{p}$ , then:

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

p̃ is characterized by:

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

p
 is exactly the loss of the infinitely granular portfolio (Basel 2 terminology)

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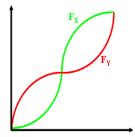
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## Stochastic orders

- The convex order compares the dispersion level of two random variables
- Convex order:  $X \leq_{cx} Y$  if  $E[f(X)] \leq E[f(Y)]$  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 functions 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

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

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

## ㅣ Müller(1997)

Stop-loss order for portfolios of dependent risks

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

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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<sub>1</sub>,..., D<sub>n</sub> be exchangeable Bernoulli random variables associated with the mixing probability p̃
- Let D<sup>\*</sup><sub>1</sub>,..., D<sup>\*</sup><sub>n</sub> exchangeable Bernoulli random variables associated with the mixing probability p<sup>\*</sup>

#### Theorem

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

- 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

# Main results

- Let D<sub>1</sub>,..., D<sub>n</sub>,... be exchangeable Bernoulli random variables associated with the mixing probability p̃
- Let D<sup>\*</sup><sub>1</sub>,..., D<sup>\*</sup><sub>n</sub>,... be exchangeable Bernoulli random variables associated with the mixing probability p<sup>\*</sup>

Theorem (reverse implication)

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

Factor copula approaches Multivariate Poisson model Structural model

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# Ordering of CDO tranche premiums

- Analysis of the dependence structure in several popular CDO pricing models
- An increase of the dependence parameter leads to:
  - a decrease of [0%, b] equity tranche premiums (which guaranties the uniqueness of the market base correlation)
  - an increase of [a, 100%] senior tranche 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  $V_i$
- $D_i = 1_{\{\tau_i \leq t\}}, i = 1 \dots n$  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  $D_i = 1_{\{\tau_i \leq t\}}$ ,  $i = 1 \dots n$  and  $D_i^* = 1_{\{\tau_i^* \leq t\}}$ ,  $i = 1 \dots n$  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)
  - ullet V is a positive random variable with Laplace transform  $arphi^{-1}$
  - $U_1,\ldots,U_n$  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 arphi-archimedean copula

• 
$$P(V_1 \leq v_1, \ldots, V_n \leq v_n) = \varphi^{-1}(\varphi(v_1) + \ldots + \varphi(v_n))$$

• 
$$\tau_i = G^{-1}(V_i)$$

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

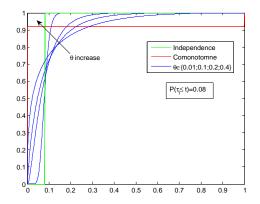
Copula	Generator $arphi$	Parameter
Clayton	$t^{- heta}-1$	$ heta \geq 0$
Gumbel	$(-\ln(t))^{ heta}$	$ heta \geq 1$
Franck	$-\ln\left[(1-e^{- heta t})/(1-e^{- heta}) ight]$	$ heta\in I\!\!R^*$

#### 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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## Multivariate Poisson model

## 

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

- $ar{N_t^i}$  Poisson with parameter  $ar{\lambda}$ : 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

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

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

## Multivariate Poisson model

- Dependence structure of  $( au_1, \dots, au_n)$  is the Marshall-Olkin copula
- $\tau_i \sim Exp(\bar{\lambda} + p\lambda)$
- D<sub>i</sub> = 1<sub>{τi</sub>≤t}</sub>, i = 1...n are conditionally independent given N<sub>t</sub>

• 
$$\frac{1}{n}\sum_{i=1}^{n}D_i \xrightarrow{a.s} E[D_i \mid N_t] = P(\tau_i \leq t \mid N_t)$$

• Conditional default probability:

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

## 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  $p$ 

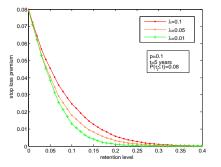
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## Multivariate Poisson model

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:

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



- Computation of E[(L<sub>t</sub> a)<sup>+</sup>]:
  - 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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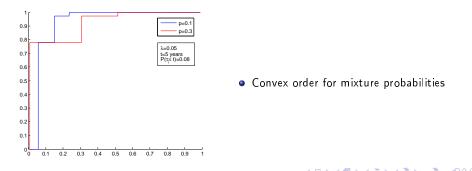
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## Multivariate Poisson model

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:

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



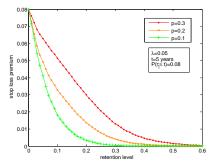
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## Multivariate Poisson model

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:

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



- 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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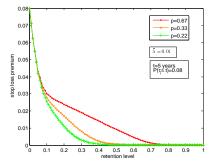
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## Multivariate Poisson model

Theorem  $(\bar{\lambda} = \bar{\lambda}^*)$ 

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

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



- 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

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

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

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

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## Structural model

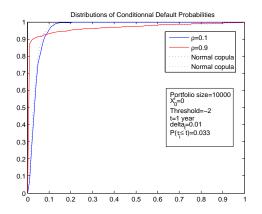
#### Theorem

For any fixed time horizon T, denote by  $D_i = 1_{\{\tau_i \leq T\}}$ ,  $i = 1 \dots n$  and  $D_i^* = 1_{\{\tau_i^* \leq T\}}$ ,  $i = 1 \dots n$  the default indicators corresponding to (resp.)  $\rho$  and  $\rho^*$ , then:  $\rho < \rho^* \Rightarrow (D_1, \dots, D_n) <_{sm} (D_1^*, \dots, D_n^*)$ 

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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: *p* <<sub>cx</sub> *p*<sup>\*</sup>

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# Conclusion

- When considering an exchangeable vector of default indicators, the conditional independence assumption is not restrictive thanks to de Finetti's theorem
- The mixing 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 mixing probabilities
- We show that the mixing 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 class of CDO pricing models

# Exchangeability: how realistic is a homogeneous assumption?

- Homogeneity of default marginals is an issue when considering the pricing and the hedging of CDO tranches
  - ex: Sudden surge of GMAC spreads in the CDX index in May, 2005
  - This event dramatically impacts the equity tranche compared to others tranches
- But composition of standard indices are updated every semester, resulting in an increase of portfolio homogeneity
- It may be reasonable to split a credit portfolio in several homogeneous sub-portfolios (by economic sectors for example)
  - Then, for each sub-portfolio, we can find a specific factor and apply the previous comparison analysis
  - The initial credit portfolio can thus be associated with a vector of factors (one by sector)
  - Is it possible to relate comparison between global aggregate losses to comparison between vectors of random factors?

## Are comparisons in a static framework restrictive?

- Are comparisons among aggregate losses at fixed horizons too restrictive?
- Computation of CDO tranche premiums only requires marginal loss distributions at several horizons
  - Comparison among aggregate losses at different dates is sufficient
- However, comparison of more complex products such as options on tranche or forward started CDOs are not possible in this framework
- Building a framework in which one can compare directly aggregate loss processes is a subject of future research