

# Dynamic Hedging of Portfolio Credit Risk in a Markov Copula Model

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## Bielecki, Cousin, Crépey, Herbertsson (BCCH1)

Dynamic Modeling of Portfolio Credit Risk with Common-Shocks  
to appear in *Journal of Optimization Theory and Application*



## Bielecki, Cousin, Crépey, Herbertsson (BCCH2)

A Bottom-Up Dynamic Model of Portfolio Credit Risk.

Part I: Markov Copula Perspective

Part II: Common-Shock Interpretation, Calibration and Hedging Issues

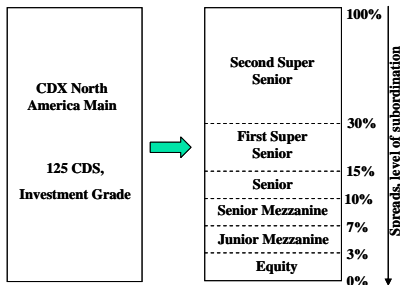


## Bielecki, Cousin, Crépey, Herbertsson (BCCH3)

A Bottom-Up Dynamic Model of Portfolio Credit Risk with Stochastic Intensities and Random Recoveries

# Introduction

## Main issue: hedging of portfolio credit derivatives



- Cash-flows driven by the realized path of the aggregate loss process

$$L_t = \frac{1}{n} \sum_{i=1}^n (1 - R_i) H_t^i$$

where  $R_i$  is the recovery rate and  $H_t^i$  is the default indicator of obligor  $i$

In this paper, we construct a **bottom-up dynamic** model consisting of

- $\mathbf{X} = (X^1, \dots, X^n)$  a vector of factor processes driving credit spreads
- $\mathbf{H} = (H^1, \dots, H^n)$  a vector of default processes ( $H_t^i = 1$  iif default of name  $i$  occurs before time  $t$ )
- $\mathcal{F}_t = \mathcal{F}_t^{\mathbf{X}, \mathbf{H}}$

and with the following **key features** :

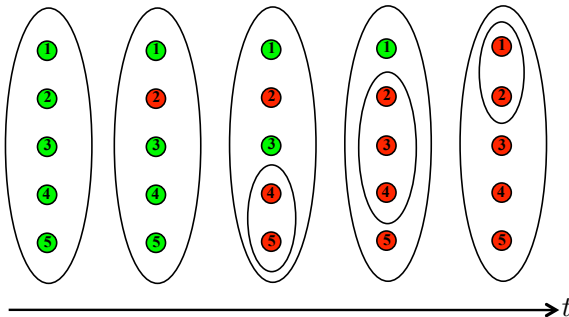
- **Markovian model**:  $(\mathbf{X}, \mathbf{H})$  is a Markov process with respect to  $\mathcal{F}$
- **Markov copula property**: Each pair  $(X^i, H^i)$  is a Markov process with respect to  $\mathcal{F}$
- **Tractable model**: Computation of CDS spreads, CDO tranche prices and hedging strategies can be achieved by fast numerical procedure

# Markov copula model of portfolio credit risk

## Simultaneous default model

- Defaults are the consequence of **triggering-events** affecting pre-specified groups of obligors

**Example:**  $n = 5$  and  $\mathcal{Y} = \{\{1\}, \{2\}, \{3\}, \{4\}, \{5\}, \{4, 5\}, \{2, 3, 4\}, \{1, 2\}\}$ .



# Markov copula model of portfolio credit risk

- $\{1, \dots, n\}$  set of credit references
- $\mathcal{Y} = \{\{1\}, \dots, \{n\}, I_1, \dots, I_m\}$  pre-specified groups of obligors
- $\lambda_Y = \lambda_Y(t)$  deterministic intensity function of the triggering-event associated with group  $Y \in \mathcal{Y}$
- $\mathbf{H}_t = (H_t^1, \dots, H_t^n)$  defined as an  **$n$ -dimensional Markov chain** in  $\{0, 1\}^n$  such that for  $\mathbf{k}, \mathbf{m} \in \{0, 1\}^n$ :

$$\mathbb{P}(\mathbf{H}_{t+dt} = \mathbf{m} \mid \mathbf{H}_t = \mathbf{k}) = \sum_{Y \in \mathcal{Y}} \lambda_Y(t) \mathbf{1}_{\{\mathbf{k}^Y = \mathbf{m}\}} dt$$

where  $\mathbf{k}^Y$  is obtained from  $\mathbf{k} = (k_1, \dots, k_n)$  by replacing the components  $k_j$ ,  $j \in Y$ , by number one. ex:  $(0, 1, 0, 0)^{\{1, 2, 4\}} = (1, 1, 0, 1)$

- $\mathcal{F}_t = \sigma(\mathbf{H}_u, u \leq t)$  natural filtration of  $\mathbf{H}$

# Markov copula model of portfolio credit risk

**Example:** Portfolio with  $n = 2$  names

$\mathcal{Y} = \{\{1\}, \{2\}, \{1, 2\}\}$ .  $\mathbf{H}_t = (H_t^1, H_t^2)$  is a bivariate continuous-time Markov chain with space set  $\{(0, 0), (1, 0), (0, 1), (1, 1)\}$  and generator matrix

$$\begin{array}{c} \begin{array}{cccc} & (0,0) & (1,0) & (0,1) & (1,1) \end{array} \\ \begin{array}{l} (0,0) \\ (1,0) \\ (0,1) \\ (1,1) \end{array} \left( \begin{array}{cccc} - & \lambda_{\{1\}} & \lambda_{\{2\}} & \lambda_{\{1,2\}} \\ 0 & - & 0 & \lambda_{\{2\}} + \lambda_{\{1,2\}} \\ 0 & 0 & - & \lambda_{\{1\}} + \lambda_{\{1,2\}} \\ 0 & 0 & 0 & 0 \end{array} \right) \end{array}$$

- Obligor 1 defaults with intensity  $\lambda_{\{1\}} + \lambda_{\{1,2\}}$  regardless of the state of the pool
- Obligor 2 defaults with intensity  $\lambda_{\{2\}} + \lambda_{\{1,2\}}$  regardless of the state of the pool

# Markov copula model of portfolio credit risk

## $n$ -dimensional case

Obligor  $i$  defaults with intensity

$$\lambda_i(t) = \lambda_{\{i\}}(t) + \sum_{k=1}^m \lambda_{I_k}(t) \mathbf{1}_{\{i \in I_k\}}$$

Then,

$$\mathbb{P}(H_{t+dt}^i - H_t^i = 1 \mid \mathcal{F}_t) = \mathbb{P}(H_{t+dt}^i - H_t^i = 1 \mid H_t^i) = (1 - H_t^i) \lambda_i(t) dt$$

- Each default indicator process  $H^i$ ,  $i = 1, \dots, n$  is Markov with respect to  $\mathcal{F}$ : **strong Markov copula property** (Bielecki, Vidozzi and Vidozzi 2008)
- On economic grounds, this means that there is **no contagion effect** : past defaults do not have any effect on intensities of surviving names



# Markov copula model of portfolio credit risk

The previous construction can be extended to the case of **stochastic intensity functions**:

$$\lambda_Y = \lambda_Y(t, X_t^Y), \quad Y \in \mathcal{Y}$$

where  $\mathbf{X} = (X_t^Y)_{Y \in \mathcal{Y}}$  is a strong solution to

$$dX_t^Y = b_Y(t, X_t^Y) dt + \sigma_Y(t, X_t^Y) dW_t^Y, \quad (1)$$

with

- $b_Y(t, x)$  and  $\sigma_Y(t, x)$  are suitable drift and diffusion functions
- $W^Y, Y \in \mathcal{Y}$  are independent standard brownian motions

# Markov copula model of portfolio credit risk

## Markov property of the model

Let  $\mathcal{F} = \mathcal{F}^{\mathbf{X}, \mathbf{H}}$  be the natural filtration of  $(\mathbf{X}, \mathbf{H})$ . The process  $(\mathbf{X}, \mathbf{H})$  is an  $\mathcal{F}$ -Markov process with generator  $\mathcal{A}$  given by

$$\begin{aligned}\mathcal{A}_t u(t, \mathbf{x}, \mathbf{k}) &= \sum_{Y \in \mathcal{Y}} \left( b_Y(t, x_Y) \partial_{x_Y} u(t, \mathbf{x}, \mathbf{k}) + \frac{1}{2} \sigma_Y^2(t, x_Y) \partial_{x_Y^2}^2 u(t, \mathbf{x}, \mathbf{k}) \right) \\ &\quad + \sum_{Y \in \mathcal{Y}} \lambda_Y(t, x_Y) \delta u^Y(t, \mathbf{x}, \mathbf{k})\end{aligned}$$

**Could we find  $\mathbf{X}$  and intensity functions  $\lambda_Y(\cdot, \cdot)$ ,  $Y \in \mathcal{Y}$  such that the Markov copula property holds ?**

- i.e., for every  $i = 1, \dots, n$ , the process  $(X^i, H^i)$  is an  $\mathcal{F}$ -Markov process ?

## Possible solution : generalized CIR intensities

$$\lambda_Y(t, \mathbf{X}_t) = X_t^Y, Y \in \mathcal{Y}$$

where  $X_t^Y, Y \in \mathcal{Y}$  are generalized CIR processes such that

$$dX_t^Y = a \left( b_Y(t) - X_t^Y \right) dt + c \sqrt{X_t^Y} dW_t^Y$$

- $a$  and  $c$  are positive parameters common to all groups
- $b_Y(t)$  is specific to group  $Y$  and defined by a deterministic function of time
- $W^Y, Y \in \mathcal{Y}$  are independent brownian motions

# Markov copula model of portfolio credit risk

## Possible solution : generalized CIR intensities (cont.)

In this framework, the default intensity of name  $i$  is given by

$$X_t^i = X_t^{\{i\}} + \sum_{k=1}^m X_t^{I_k} \mathbf{1}_{\{i \in I_k\}}$$

Note that  $X^i$  is again a generalized CIR process such that

$$dX_t^i = a \left( b_i(t) - X_t^i \right) dt + c \sqrt{X_t^i} dW_t^i$$

where

- $b_i(t) = b_{\{i\}}(t) + \sum_{k=1}^m b_{I_k}(t) \mathbf{1}_{\{i \in I_k\}}$
- $W^i$  is a standard brownian motion

# Markov copula model of portfolio credit risk

## Markov copula property for extended CIR intensities (Bielecki et al. (2012a))

Let  $\mathcal{F} = \mathcal{F}^{\mathbf{X}, \mathbf{H}}$  be the natural filtration of  $(\mathbf{X}, \mathbf{H})$  where  $\mathbf{X} = (X^Y)_{Y \in \mathcal{Y}}$  are previously defined CIR processes. Then,

- $(\mathbf{X}, \mathbf{H})$  is an  $\mathcal{F}$ -Markov process
- For every  $i = 1, \dots, n$ , the process  $(X^i, H^i)$  is an  $\mathcal{F}$ -Markov process

Moreover, the  $\mathcal{F}_t$ -conditional **survival probability** of name  $i$  is given by

$$\mathbb{P}(\tau_i > T | \mathcal{F}_t) = (1 - H_t^i) \mathbb{E} \left\{ \exp \left( - \int_t^T X_u^i du \right) | X_t^i \right\}$$

- Analytical expression available in the case of piecewise-constant  $b^Y(\cdot)$  (See [BCCH3](#))
- Other stable-by-convolution processes can be considered (e.g., processes driving by Levy-subordinators)

# Common-Shock Model Interpretation

## Main result: equivalent $\mathcal{F}_t$ -related common-shocks model

- $\mathcal{Y}_t = \{Y \in \mathcal{Y}; Y \not\subseteq \text{supp}(\mathbf{H}_t)\}$  : set of pre-specified groups that contain at least one alive obligor
- For any pre-specified group  $Y \in \mathcal{Y}_t$ , we define

$$\tau_Y(t) = \inf \left\{ \theta \geq t \mid \int_t^\theta \lambda_Y(s, \mathbf{X}_s) ds > E_Y \right\}$$

where  $E_Y$ ,  $Y \in \mathcal{Y}_t$ , are **independent** and **exponentially distributed** random variables with parameter 1.

- In the  $\mathcal{F}_t$ -related common-shock model, the **individual default time** of a non-defaulted name  $i$  is defined by:

$$\hat{\tau}_i(t) = \min_{\{Y \in \mathcal{Y}_t; i \in Y\}} \tau_Y(t)$$

- $H_\theta^i(t) = \mathbf{1}_{\{\hat{\tau}_i(t) \leq \theta\}}$ : default indicator of name  $i$  at time  $\theta$  in the  $\mathcal{F}_t$ -related common-shock model

# Common-Shock Model Interpretation

## Main result

### Proposition

Let  $Z$  be a subset of  $\{1, \dots, n\}$ . For every  $\theta_1, \dots, \theta_n \geq t$ , one has on the event  $\{Z = \text{supp}^c(\mathbf{H}_t)\}$ :

$$\mathbb{P}(\tau_i > \theta_i, i \in \text{supp}^c(\mathbf{H}_t) \mid \mathcal{F}_t) = \mathbb{P}(\hat{\tau}_i(t) > \theta_i, i \in Z \mid \mathbf{X}_t)$$

Moreover, if

- $N_\theta = \sum_{i=1}^n H_\theta^i$  denotes the cumulative number of defaults at time  $\theta$  in the Markov model
- $N_\theta(t, Z) = n - |Z| + \sum_{i \in Z} H_\theta^i(t)$  denotes the cumulative number of defaults at time  $\theta$  in the  $\mathcal{F}_t$ -related common-shock model

then, for every  $\theta \geq t$ , one has on the event  $\{Z = \text{supp}^c(\mathbf{H}_t)\}$ :

$$\mathbb{P}(N_\theta = k \mid \mathcal{F}_t) = \mathbb{P}(N_\theta(t, Z) = k \mid \mathbf{X}_t)$$

for any  $k = n - |Z|, \dots, n$ .

# Common-Shock Model Interpretation

- But, for any time  $\theta \geq t$ ,  $H_\theta^i(t)$ ,  $i \in Z$ , are **conditionally independent Bernoulli's** given  $\left(H_\theta^{I_1}(t), \dots, H_\theta^{I_m}(t)\right)$
- **Fast convolution-recursion procedure** can be used to compute marginal loss distributions conditionally on any given set  $\{Z = \text{supp}^c(\mathbf{H}_t)\}$ .
- As far as standard **CDO tranches** are concerned, we will see that **pricing, calibration and computation of hedging strategies are numerically tractable**



# Hedging CDO tranches with single-name CDS-s

## Set of fundamental martingales for jump components of the Markov model

- $H_t^Z$  is the indicator process of simultaneous default of names in the set  $Z$ , for every subset  $Z$  of  $\{1, \dots, n\}$
- $Y_t = Y \cap \text{supp}^c(\mathbf{H}_{t-})$  stands for the set of survivors of set  $Y$  right before  $t$ , for every pre-specified group  $Y \in \mathcal{Y}$

## Set of fundamental martingales

The process  $M^Z$  defined by

$$dM_t^Z := dH_t^Z - \ell_Z(t, \mathbf{X}_t, \mathbf{H}_{t-})dt$$

is a martingale with respect to  $\mathcal{F}$ , where the intensity function  $\ell_Z(t, \mathbf{x}, \mathbf{k})$  is such that

$$\ell_Z(t, \mathbf{X}_t, \mathbf{H}_{t-}) = \sum_{Y \in \mathcal{Y}; Y_t = Z} \lambda_Y(t, \mathbf{X}_t)$$

# Hedging CDO tranches with single-name CDS-s

## Itô formula

Given a “regular enough” function  $u = u(t, \mathbf{x}, \mathbf{k})$ , one has, for  $t \in [0, T]$ ,

$$\begin{aligned} du(t, \mathbf{X}_t, \mathbf{H}_t) = & \left( \partial_t + \mathcal{A}_t \right) u(t, \mathbf{X}_t, \mathbf{H}_t) dt + \nabla u(t, \mathbf{X}_t, \mathbf{H}_t) \sigma(t, \mathbf{X}_t) d\mathbf{W}_t \\ & + \sum_{Z \in \mathcal{Z}_t} \delta u^Z(t, \mathbf{X}_t, \mathbf{H}_{t-}) dM_t^Z \end{aligned}$$

where

- $\sigma(t, \mathbf{x})$ : diagonal matrix with diagonal  $(\sigma_Y(t, x_Y))_{Y \in \mathcal{Y}}$
- $\nabla u(t, \mathbf{x}, \mathbf{k}) = (\partial_{x_1} u(t, \mathbf{x}, \mathbf{k}), \dots, \partial_{x_\nu} u(t, \mathbf{x}, \mathbf{k}))$
- $\delta u^Z(t, \mathbf{x}, \mathbf{k}) = u(t, \mathbf{x}, \mathbf{k}^Z) - u(t, \mathbf{x}, \mathbf{k})$
- $\mathcal{Z}_t = \{Y_t; Y \in \mathcal{Y}\} \setminus \emptyset$ : set of all non-empty sets of survivors of sets  $Y$  in  $\mathcal{Y}$  right before time  $t$

**Martingale dimension:**  $n + m + 2^n$

# Hedging CDO tranches with single-name CDS-s

## Price dynamics for single-name CDS-s (buy-protection position)

- $T$ : contract maturity
- $S_i$ :  $T$ -year contractual CDS-spread of obligor  $i$
- $t_1 < \dots < t_p = T$ : premium payment dates,  $h = t_j - t_{j-1}$  length between two premium payment dates (typically a quarter)
- $R_i$ : recovery rate of obligor  $i$

**Except for numerical results, we will assume zero interest rates**

# Hedging CDO tranches with single-name CDS-s

## Price dynamics for single-name CDS $i$

The price  $P^i$  and the cumulative value  $\hat{P}^i$  at time  $t \in [0, T]$  of a single-name CDS on obligor  $i$  are given by

$$\begin{aligned} P_t^i &= \mathbf{1}_{\{\tau_i > t\}} v_i(t, X_t^i) \\ d\hat{P}_t^i &= \mathbf{1}_{\{\tau_i > t\}} \partial_{x_i} v_i(t, X_t^i) \sigma_i(t, X_t^i) dW_t^i \\ &\quad + \sum_{Z \in \mathcal{Z}_t} \mathbf{1}_{\{i \in Z\}} \left( 1 - R_i - v_i(t, X_t^i) \right) dM_t^Z \end{aligned}$$

for a pre-default pricing function  $v_i(t, x_i)$  such that

$$\mathbf{1}_{\{\tau_i > t\}} v_i(t, X_t^i) = \mathbb{E}[(1 - R_i) \mathbf{1}_{\{t < \tau_i \leq T\}} - S_i h \sum_{t < t_j \leq T} \mathbf{1}_{\{\tau_i > t_j\}} | \mathcal{F}_t]$$

# Hedging CDO tranches with single-name CDS-s

## Price dynamics for CDO tranche $[a, b]$ (buy-protection position)

- $T$ : contract maturity
- $a$ : attachment point,  $b$ : detachment point,  $0 \leq a < b \leq 1$
- $S^{a,b}$ :  $T$ -year contractual spread of CDO tranche  $[a, b]$
- $t_1 < \dots < t_p = T$ : premium payment dates,  $h = t_j - t_{j-1}$  length between two premium payment dates (typically a quarter)
- CDO tranche cash-flows are driven by the **tranche loss process**

$$L_t^{a,b} = L_{a,b}(\mathbf{H}_t) = (L_t - a)^+ - (L_t - b)^+$$

where

$$L_t = L_t(\mathbf{H}_t) = \frac{1}{n} \sum_{i=1}^n (1 - R_i) H_t^i$$

is the credit loss process for the underlying portfolio

# Hedging CDO tranches with single-name CDS-s

## Price dynamics for CDO tranche $[a, b]$

The price  $\Pi$  and the cumulative value  $\hat{\Pi}$  at time  $t \in [0, T]$  of a CDO-tranche  $[a, b]$  are given by

$$\Pi_t = u(t, \mathbf{X}_t, \mathbf{H}_t)$$

$$\begin{aligned} d\hat{\Pi}_t &= \nabla u(t, \mathbf{X}_t, \mathbf{H}_t) \sigma(t, \mathbf{X}_t) d\mathbf{W}_t \\ &\quad + \sum_{Z \in \mathcal{Z}_t} \left( L_{a,b}(\mathbf{H}_{t-}^Z) - L_{a,b}(\mathbf{H}_{t-}) + \delta u^Z(t, \mathbf{X}_t, \mathbf{H}_{t-}) \right) dM_t^Z \end{aligned}$$

for a pricing function  $u(t, \mathbf{x}, \mathbf{k})$  such that

$$u(t, \mathbf{X}_t, \mathbf{H}_t) = \mathbb{E} \left[ L_T^{a,b} - L_t^{a,b} - S^{a,b} h \sum_{t < t_j \leq T} \left( b - a - L_{t_j}^{a,b} \right) \middle| \mathcal{F}_t \right]$$

**The pricing function  $u(t, \mathbf{x}, \mathbf{k})$  solves a very large system of Kolmogorov pde. Thanks to the common-shock interpretation, it can be computed by fast recursion procedures.**

# Hedging CDO tranches with single-name CDS-s

**Hedging portfolio:** first  $d$  single-name CDS-s and the savings account

The vector of cumulative values  $\hat{\mathbf{P}} = (\hat{P}^1, \dots, \hat{P}^d)^\top$  associated with the first  $d$  CDS-s has the following dynamics:

$$d\hat{\mathbf{P}}_t = \nabla \mathbf{v}(t, \mathbf{X}_t, \mathbf{H}_t) \sigma(t, \mathbf{X}_t) d\mathbf{W}_t + \sum_{Z \in \mathcal{Z}_t} \Delta \mathbf{v}^Z(t, \mathbf{X}_t, \mathbf{H}_{t-}) dM_t^Z$$

where

- $\nabla \mathbf{v}$  is a  $d \times \nu$ -matrix such that  $\nabla \mathbf{v}(t, \mathbf{x}, \mathbf{k})_i^j = \mathbf{1}_{\{k_j=0\}} \partial_{x_j} v_i(t, x_i)$ , for every  $1 \leq i \leq d$  and  $1 \leq j \leq \nu$

- $\Delta \mathbf{v}^Z(t, \mathbf{x}, \mathbf{k})$  is a  $d$ -dimensional column vector equal to

$$(\mathbf{1}_{\{1 \in Z, k_1=0\}} ((1 - R_1) - v_1(t, x_1)), \dots, \mathbf{1}_{\{d \in Z, k_d=0\}} ((1 - R_d) - v_d(t, x_d)))^\top$$

# Hedging CDO tranches with single-name CDS-s

**Tracking error:** Process  $(e_t)$  such that  $e_0 = 0$  and for  $t \in [0, T]$ :

$$\begin{aligned} de_t &= d\widehat{\Pi}_t - \zeta_t d\widehat{\mathbf{P}}_t \\ &= \left( \nabla u(t, \mathbf{X}_t, \mathbf{H}_t) - \zeta_t \nabla \mathbf{v}(t, \mathbf{X}_t, \mathbf{H}_t) \right) \sigma(t, \mathbf{X}_t) d\mathbf{W}_t \\ &\quad + \sum_{Z \in \mathcal{Z}_t} \left( \Delta u^Z(t, \mathbf{X}_t, \mathbf{H}_{t-}) - \zeta_t \Delta \mathbf{v}^Z(t, \mathbf{X}_t, \mathbf{H}_{t-}) \right) dM_t^Z \end{aligned}$$

where

- $\zeta_t = (\zeta_t^1, \dots, \zeta_t^d)$  gives the positions held at time  $t$  in CDS  $1, \dots, d$
- $\nabla u(t, \mathbf{x}, \mathbf{k}) = (\partial_{x_1} u(t, \mathbf{x}, \mathbf{k}), \dots, \partial_{x_\nu} u(t, \mathbf{x}, \mathbf{k}))$
- $\Delta u^Z(t, \mathbf{x}, \mathbf{k}) = \delta^Z u(t, \mathbf{x}, \mathbf{k}) + L_{a,b}(\mathbf{k}^Z) - L_{a,b}(\mathbf{k})$



# Hedging CDO tranches with single-name CDS-s

## Min-variance hedging strategies

The min-variance hedging strategy  $\zeta$  for the CDO-tranche  $[a, b]$  is

$$\zeta_t = \frac{d\langle \hat{\Pi}, \hat{\mathbf{P}} \rangle_t}{dt} \left( \frac{d\langle \hat{\mathbf{P}} \rangle_t}{dt} \right)^{-1} = \zeta(t, \mathbf{X}_t, \mathbf{H}_{t-})$$

where  $\zeta = (u, \mathbf{v})(\mathbf{v}, \mathbf{v})^{-1}$ , with

$$\begin{aligned}(u, \mathbf{v}) &= (\nabla u) \sigma^2 (\nabla \mathbf{v})^T + \sum_{Y \in \mathcal{Y}} \lambda_Y \Delta u^Y (\Delta \mathbf{v}^Y)^T \\ (\mathbf{v}, \mathbf{v}) &= (\nabla \mathbf{v}) \sigma^2 (\nabla \mathbf{v})^T + \sum_{Y \in \mathcal{Y}} \lambda_Y \Delta \mathbf{v}^Y (\Delta \mathbf{v}^Y)^T\end{aligned}$$

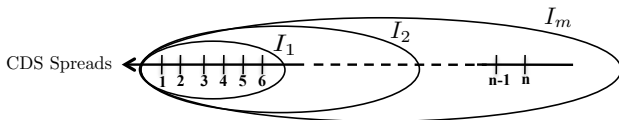
# Calibration on CDX index

**Data set:** 5-year CDX North-America IG index on 20 December 2007

- Quoted spreads at different pillars of the  $n = 125$  index constituents
- Quoted spreads of standard tranches  $[0,3]$ ,  $[3,7]$ ,  $[7,10]$ ,  $[10,15]$ ,  $[15,30]$

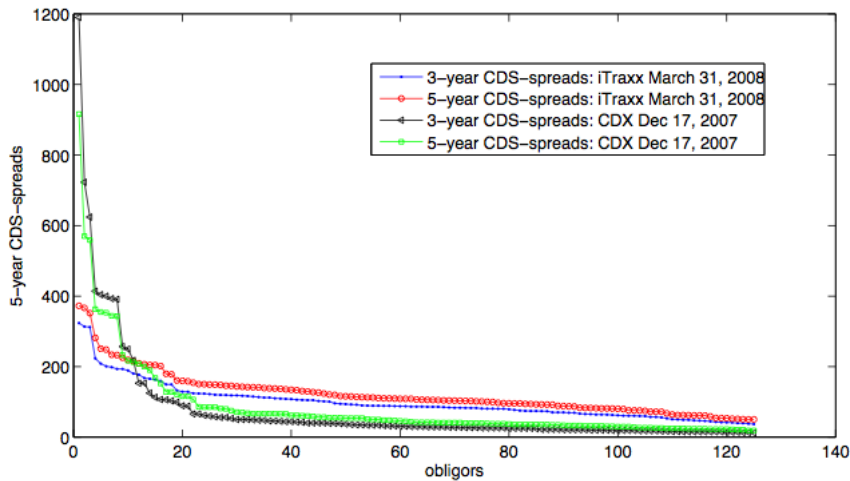
**Model specification:**

- Names are labelled with respect to decreasing level of spreads



- $m = 5$  groups  $I_1 \subset \dots \subset I_5$  such that  $I_1 = \{1, \dots, 6\}$ ,  $I_2 = \{1, \dots, 19\}$ ,  $I_3 = \{1, \dots, 25\}$ ,  $I_4 = \{1, \dots, 61\}$ ,  $I_5 = \{1, \dots, 125\}$
- Piecewise-constant intensities  $\lambda_{\{1\}}, \dots, \lambda_{\{125\}}$ ,  $\lambda_{I_1}, \dots, \lambda_{I_5}$  with grid points corresponding to CDS pillars
- Homogeneous and constant recovery rates: 40%
- Constant short-term interest rate: 3%

# Calibration on CDX index



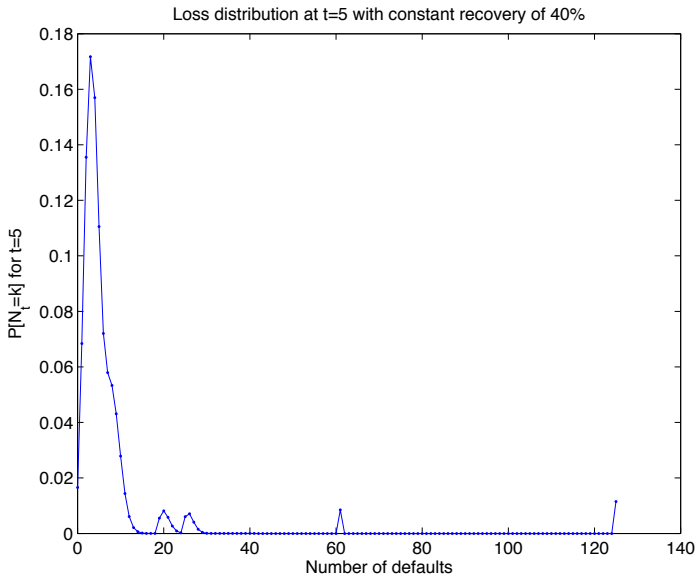
## Calibration results:

Tranche	[0,3]	[3,7]	[7,10]	[10,15]	[15,30]
Model spread in bps	48.0701	254.0000	124.0000	61.0000	38.9390
Market spread in bps	48.0700	254.0000	124.0000	61.0000	41.0000
Abs. Err. in bps	0.0001	0.0000	0.0000	0.0000	2.0610
% Rel. Err.	0.0001	0.0000	0.0000	0.0000	5.0269

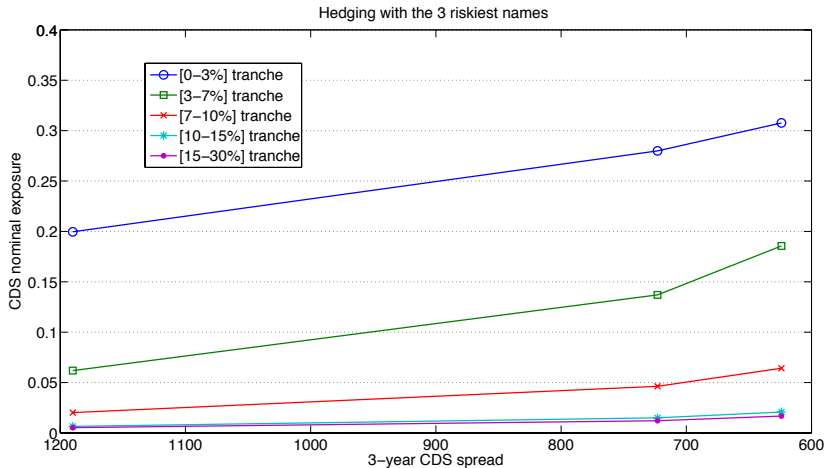
- Names in the set  $I_5 \setminus I_4$  are excluded from the calibration constraints (they can only default within the Armageddon shock  $I_5$ )

# Calibration on CDX index

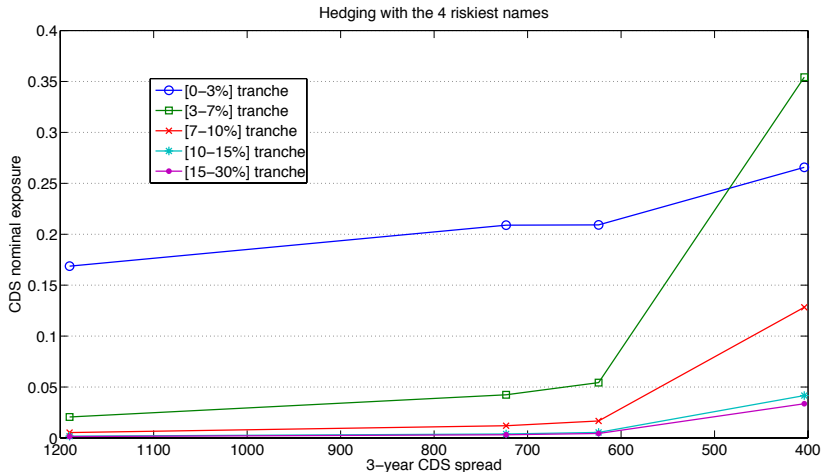
## Implied 5-year loss distribution:



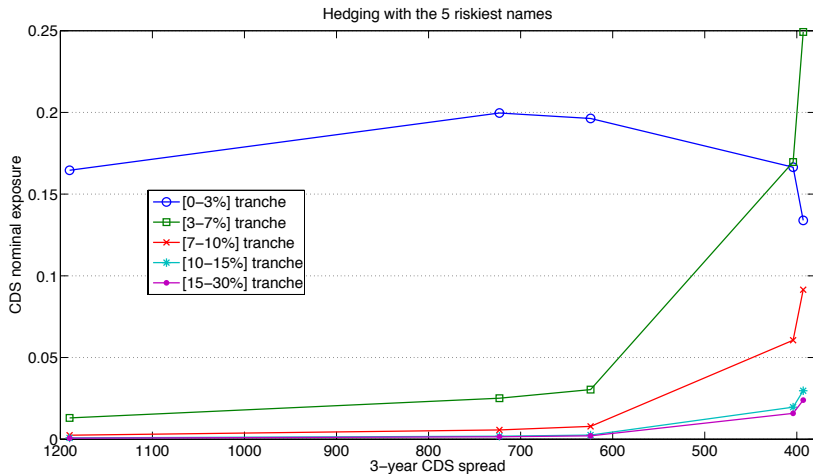
# Min-variance hedging strategies



# Min-variance hedging strategies

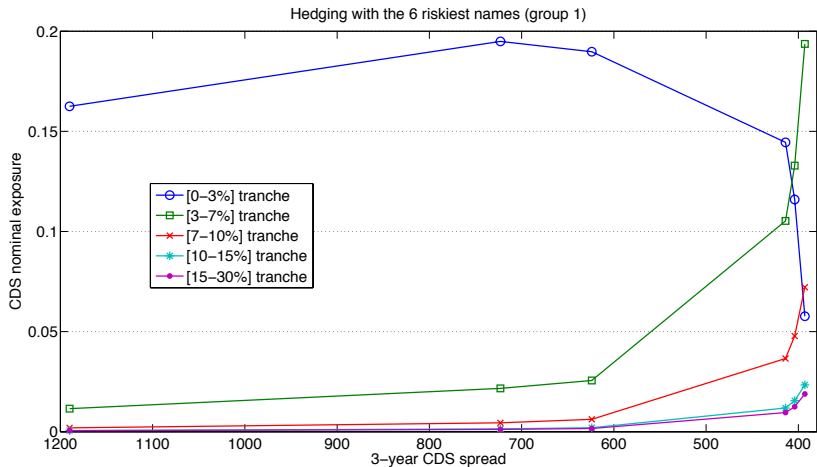


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In this paper, we construct a **dynamic bottom-up model** of portfolio credit risk:

- **Markov-copula construction** of default times: two-steps calibration procedure of model parameters
- **Common-shocks representation** of default times conditionally on any given state of the Markov model: fast numerical computation of conditional loss distributions
- The model allows us to **hedge CDO tranches using single-name CDS-s** in a theoretically sound and practical convenient way

Thank you for your attention!

-  BIELECKI, T.R., VIDOZZI, A. AND VIDOZZI, L.: A Markov Copulae Approach to Pricing and Hedging of Credit Index Derivatives and Ratings Triggered Step-Up Bonds, *J. of Credit Risk*, 2008.
-  BRIGO, D., PALLAVICINI, A., TORRENTIAL, R.: Calibration of CDO Tranches with the Dynamical Generalized-Poisson Loss Model. *Working Paper*, 2006.
-  ELOUERKHAOU, Y.: Pricing and Hedging in a Dynamic Credit Model. *International Journal of Theoretical and Applied Finance*, Vol. 10, Issue 4, 703–731, 2007.
-  LINDSKOG, F. AND MCNEIL, A. J.: Common Poisson Shock Models: Applications to Insurance and Credit Risk Modelling. *ASTIN Bulletin*, 33(2), 209-238, 2003

## Comment on Markov copula property

- The Markov copula property satisfied by the model is known as the *strong Markov copula property*. This property prohibits default contagion between individual credit names.
- A weaker form of the Markov copula property, where for every  $i = 1, \dots, n$ , the process  $(X^i, H^i)$  is an  $\mathcal{F}^i$ -Markov but not-necessarily  $\mathcal{F}$ -Markov, has also been studied. Such weak Markov copula property allows for default contagion between individual credit names.