

Pricing and Hedging Loss Derivatives in a Markovian Bottom-Up Model

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IMS Gothenburg 2010 Conference

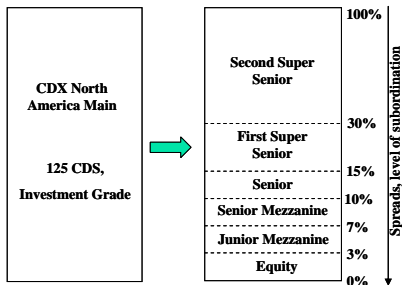
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Pricing and Hedging Portfolio Credit Derivatives in a Bottom-up Model
with Simultaneous Defaults

Risk management of portfolio credit derivatives



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

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

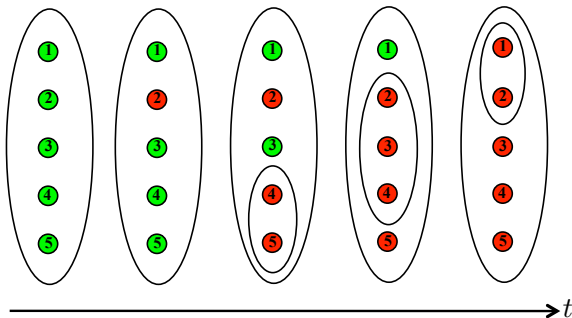
where R_i is the recovery rate and N_t^i is the default indicator of obligor i

Markovian portfolio credit risk model

Simultaneous default model

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

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



Markovian portfolio credit risk model

- $\{1, \dots, n\}$: credit references
- $\mathcal{Y} = \{\{1\}, \dots, \{n\}, I_1, \dots, I_m\}$: pre-specified groups
- $\lambda_Y(\cdot)$ intensity function associated with group Y
- N_t^i default indicator process of name $i = 1, \dots, n$
- $\mathbf{N}_t = (N_t^1, \dots, N_t^n)$ is a multivariate Markov chain in $\{0, 1\}^n$ such that for $\mathbf{k}, \mathbf{m} \in \{0, 1\}^n$:

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

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.

Markov copula condition

For any $i = 1, \dots, n$, N^i is a one dimensional Markov process:

$$\mathbb{E} \left[\Phi(N_T^i) \mid N_t^1, \dots, N_t^n \right] = \mathbb{E} \left[\Phi(N_T^i) \mid N_t^i \right]$$

Independent pricing and calibration of single-name products

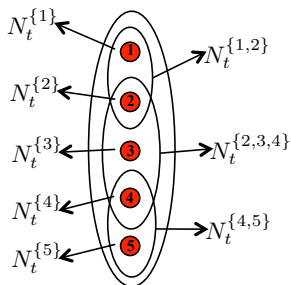
Hedging CDO tranches with single-name CDS

- Dynamics of CDO tranche prices and single-name CDS can be expressed in terms of some fundamental martingales
- Computation of min-variance hedging strategies
- Price of portfolio derivatives solves the Kolmogorov backward equations

Numerically intractable at least for large heterogeneous portfolios ($n > 20$)

Common Shocks Model Interpretation

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



$$\hat{N}_t^1 := \max \left\{ N_t^{\{1\}}, N_t^{\{1,2\}} \right\}$$

$$\hat{N}_t^2 := \max \left\{ N_t^{\{2\}}, N_t^{\{1,2\}}, N_t^{\{2,3,4\}} \right\}$$

$$\hat{N}_t^3 := \max \left\{ N_t^{\{3\}}, N_t^{\{2,3,4\}} \right\}$$

$$\hat{N}_t^4 := \max \left\{ N_t^{\{4\}}, N_t^{\{2,3,4\}}, N_t^{\{4,5\}} \right\}$$

$$\hat{N}_t^5 := \max \left\{ N_t^{\{5\}}, N_t^{\{4,5\}} \right\}$$

- N_t^Y , $Y \in \mathcal{Y}$ are **independent** $\{0, 1\}$ -point processes with intensity λ_Y :

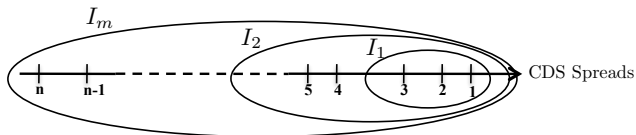
Common Shocks Model Interpretation

$$\left(\hat{N}_{t_1}^1, \dots, \hat{N}_{t_n}^n \right) \stackrel{d}{=} \left(N_{t_1}^1, \dots, N_{t_n}^n \right)$$

Common Shocks Model Interpretation

Calibration of individual intensities on single-name CDS

- Individual shocks + Common shocks: $\mathcal{Y} = \{\{1\}, \dots, \{n\}, I_1, \dots, I_m\}$
- Names are ordered with respect to riskiness



- Price of CDS i can be expressed as a function of $\mathbb{E} [N_t^i]$, $t = 0, \dots, T$

$$\mathbb{E} [N_t^i] = 1 - \exp \left(- \int_0^t \eta_i(u) dt \right)$$

where

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

- η_i , $i = 1, \dots, n$ calibrated on individual CDS spreads by a bootstrap procedure

Common Shocks Model Interpretation

Calibration of common-shocks intensities on CDO tranches

- Pricing of CDO tranches only involves marginal loss distributions
- Thanks to the common-shock model interpretation:

$$L_t = \frac{1}{n} \sum_{i=1}^n (1 - R_i) N_t^i \stackrel{d}{=} \frac{1}{n} \sum_{i=1}^n (1 - R_i) \hat{N}_t^i$$

- Conditionally on $(N_t^{I_1}, \dots, N_t^{I_m})$, $\hat{N}^1, \dots, \hat{N}^n$ are independent Bernoulli random variables with parameters

$$p_t^i = \begin{cases} 1 & i \in \cup_{k=1}^m \{I_k ; \hat{N}_t^{I_k} = 1\} \\ 1 - \exp\left(-\int_0^t \lambda_{\{i\}}(u) du\right) & \text{else} \end{cases}$$

where

$$\lambda_{\{i\}}(u) = \eta_i(u) - \sum_{k=1}^m \lambda_{I_k}(u) \mathbf{1}_{\{i \in I_k\}} \geq 0$$

Fast convolution-recursion procedure for computing loss distributions

Data set: 5-years CDX North-America index on 20 December 2007

- Quoted spreads (at different pillars) of the 125 index constituents
- Quoted spreads of standard CDO tranches

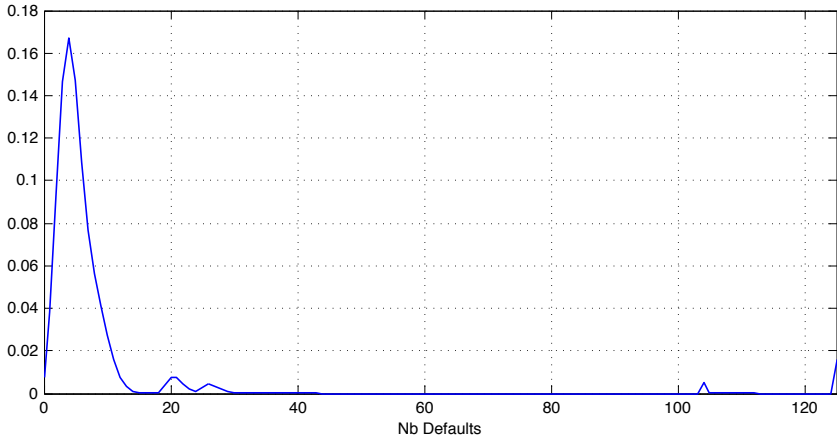
Model specification:

- 5 groups $I_1 \subset \dots \subset I_5$ such that $I_1 = \{1, \dots, 6\}$, $I_2 = \{1, \dots, 18\}$, $I_3 = \{1, \dots, 25\}$, $I_4 = \{1, \dots, 100\}$, $I_5 = \{1, \dots, 125\}$
- Piecewise constant intensities $\lambda_{\{1\}}, \dots, \lambda_{\{125\}}$, $\lambda_{I_1}, \dots, \lambda_{I_5}$ with grid points corresponding to CDS pillars
- Recovery rate: 40%
- Interest rate: 3%





Calibration results:

	0%-3%	3%-7%	7%-10%	10%-15%	15%-30%	index
Market quotes	54	265	125	62	43	81
Model outputs	48	254	124	61	41	78

5-years calibrated loss distribution:



- Hedging CDO tranches with individual CDS
- Computation of min-variance hedging strategies
- Comparison with Gaussian copula spread-sensitivity deltas

-  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