



Financial Risks

New Developments in Structured Products & Credit Derivatives

On the Pricing of CDOs

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Our Setup

- Reduced-form type of model .
- Use the **general quadratic setting** to obtain closed-form solutions
 - The class of GQTS include as special cases ATS and Gaussian-QTS.
 - It is as far as we can go in terms of exponential polynomials.
- Incorporate realistic features using **shot-noise processes**:
 - Clustering of defaults within firms
 - Correlation of defaults across firms
- Deal with **portfolio credit risk derivatives**



Previous Reduced-form Models

- **Closed-form solutions**
 - Deterministic intensity of default
 - Affine intensity models
 - Gaussian Quadratic models
- **Realistic features**
 - Hard to replicate.
 - It is hard to obtain reasonable correlation of defaults for reasonable intensity levels.
- **Other Problems**
 - Some affine intensity models do not guarantee positive intensity.

Our Model

The default time τ is a doubly stochastic random time with intensity

$$\mu_t = \eta_t + J_t$$

$$\eta_t = Z_t^\top Q(t) Z_t + g^\top(t) Z_t + f(t) \quad J_t = \sum_{\tilde{\tau}_i \leq t} Y_i h(t - \tilde{\tau}_i)$$



Predictable

component Finite state variable driven by Wiener Process.



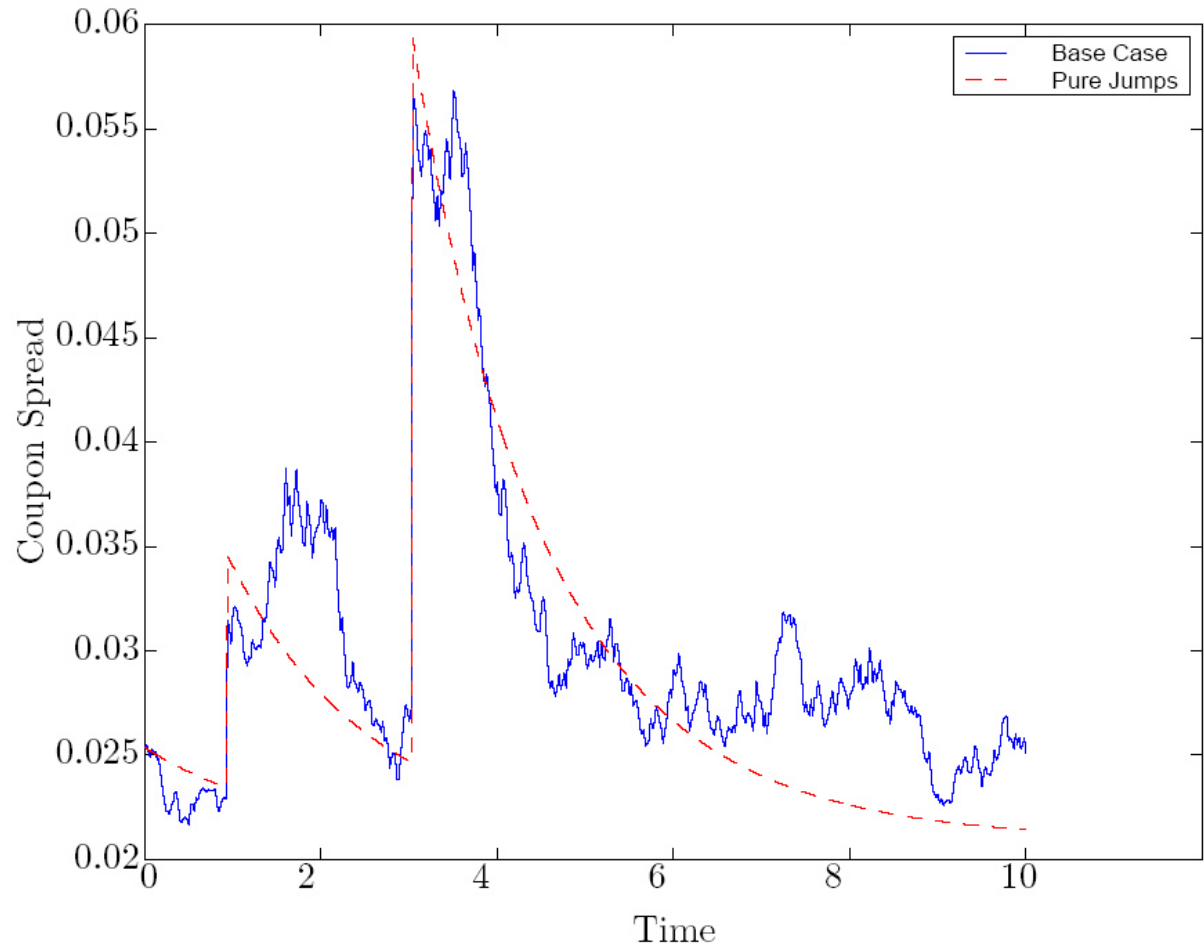
Unpredictable

component $\tilde{\tau}_1, \tilde{\tau}_2, \dots$
Jump process where
are the jump times of a
standard Poisson process.
The $f h(x) = e^{-bx}$ be, for
instance,

Our intensity

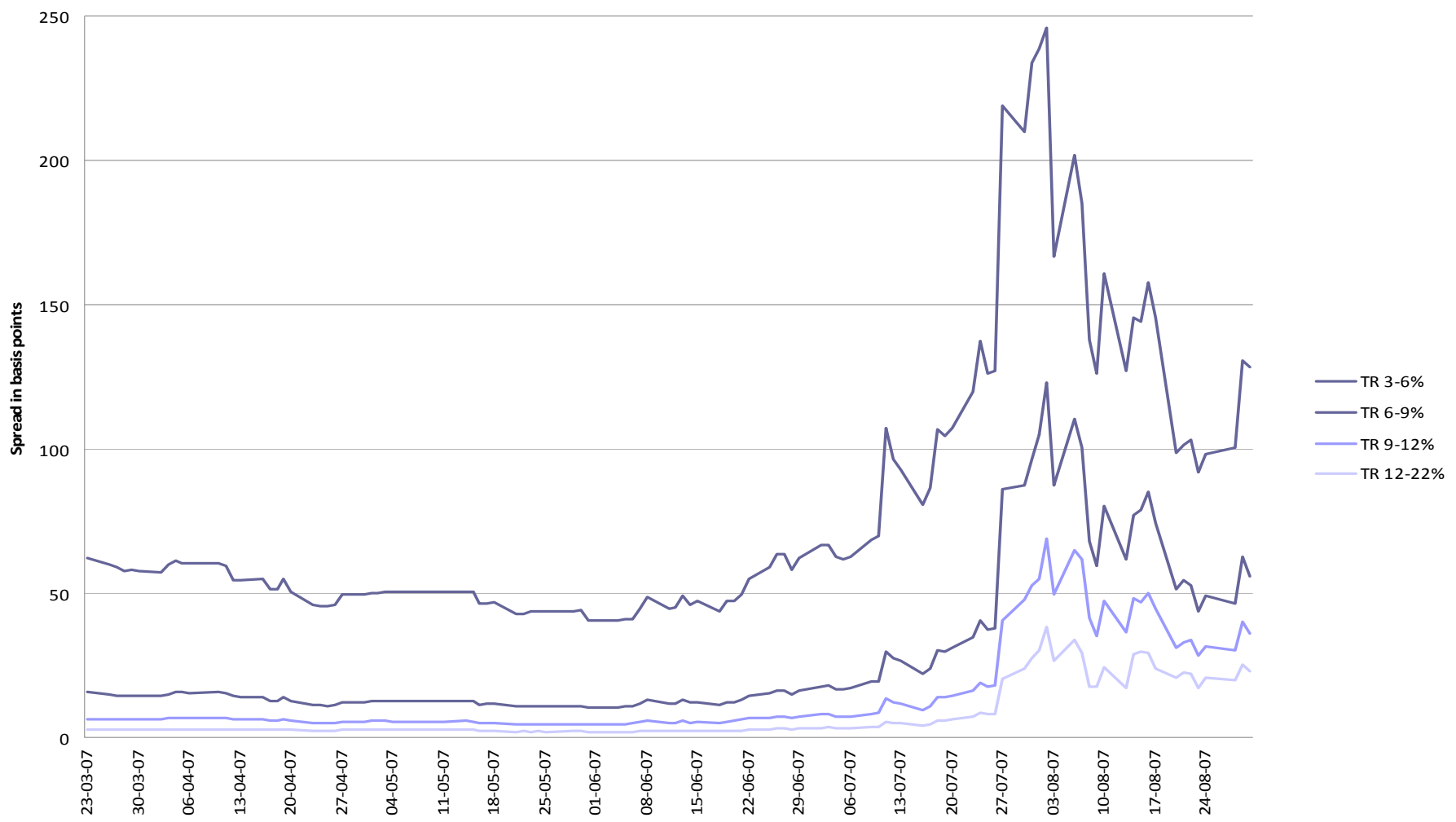


How realistic is such a process ?





iTraxx Europe Series 7 T=5year



Advantages of our Framework

- The intensity is positive in a natural way
- We get (quasi) closed form expressions to **all** key ingredients.
- The processes η_t and J_t are **independent**.
- **Trivial to extend** to multiple firms



For each name k we have:

$$\lambda_t^k = \underbrace{\mu_t^k}_{\text{Firm-specific term}} + \underbrace{\epsilon^k \mu_t^c}_{\text{Systematic term that depends upon the sensitivity of each name to the overall economy.}}$$

Firm-specific term

Systematic term that depends upon the sensitivity of each name to the overall economy.

Each term is of quadratic + shot noise form



- **Jump part:** Clustering/Contagion
- **Quadratic part:** Business Cycle

A concrete instance

$$\mu_t^k = \eta_t^k = (Z_t^k)^2 \quad \left| \quad \begin{aligned} dZ_t^i &= [\beta_i(t) - \alpha_i Z_t^i] dt + \sigma_i dW_t^i, \quad i = 1, 2 \\ dr_t &= \alpha_r [\beta_r - r_t] dt + \sigma_r \sqrt{r_t} dW_t^r \end{aligned} \right.$$

$$\mu^c = J^c + \delta r \quad \left| \quad \begin{aligned} J_t^c &= \sum_{\tilde{\tau}_i < t} Y_i h(t - \tilde{\tau}_i) \quad Y_i \sim \chi^2(2) \quad h(t) = e^{-bt}, \quad b > 0 \\ \tilde{\tau}_i &\text{ are jumps of a Poisson process with intensity } l^c \end{aligned} \right.$$

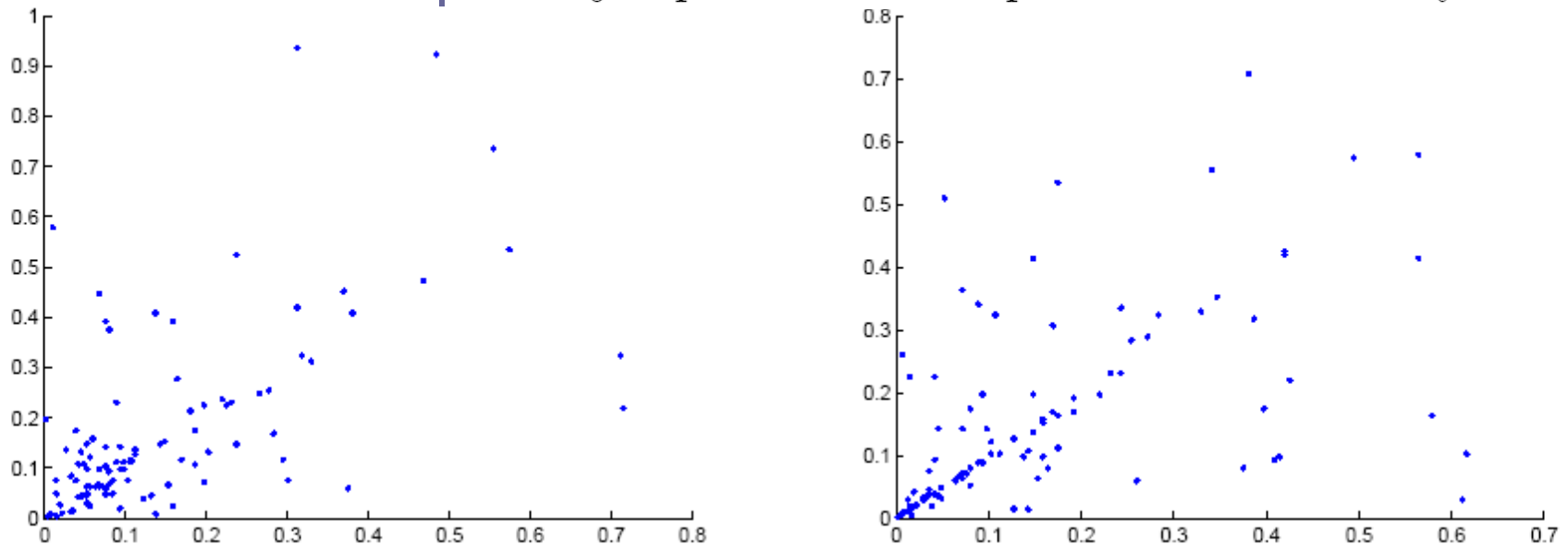


Figure 3: Simulated defaults of two companies according to the concrete model. Parameters are for $i = 1, 2$: $\beta_i = 1, \alpha_i = 0.5, \sigma_i = 0.2, l^c = 2, b = 0.5$. $Y_i \sim \chi^2(2)$ and $r = 0$. The left picture is for $\epsilon_i = 0.1$, the right $\epsilon_i = 0.5$.

Default Correlation

$$\mu_t^k = \eta_t^k = (Z_t^k)^2 \quad \left| \quad \begin{aligned} dZ_t^i &= [\beta_i(t) - \alpha_i Z_t^i] dt + \sigma_i dW_t^i, \quad i = 1, 2 \\ dr_t &= \alpha_r [\beta_r - r_t] dt + \sigma_r \sqrt{r_t} dW_t^r \end{aligned} \right.$$

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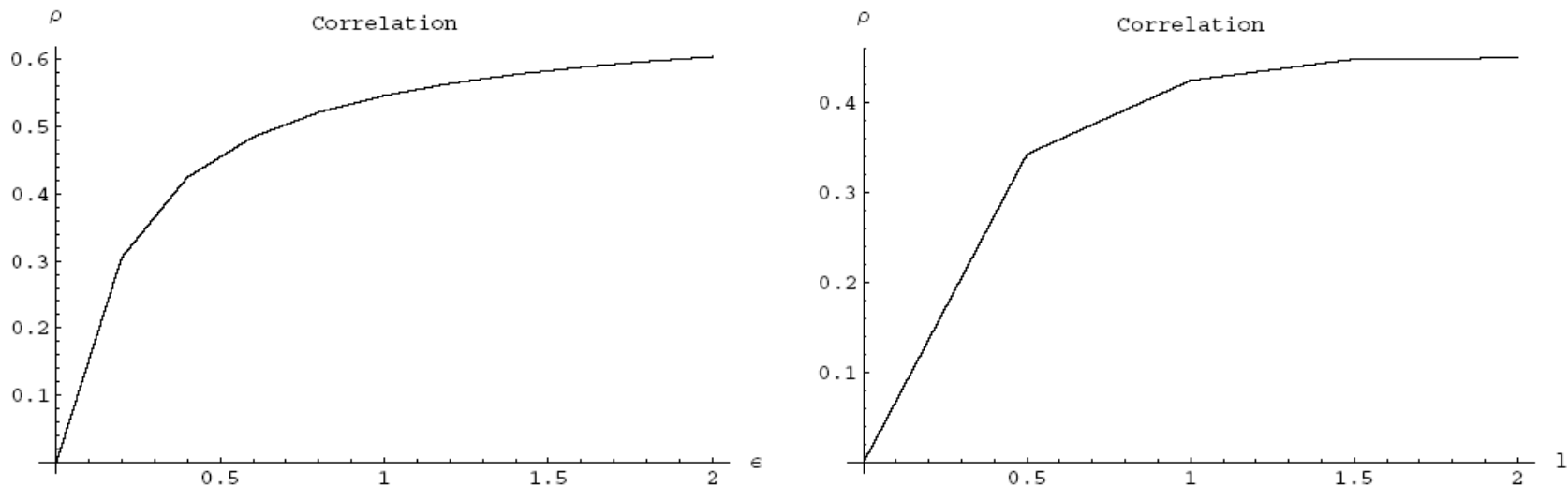


Figure 4: Model parameters: $\alpha = 0.5$, $\beta = 0.1\alpha$, $\sigma = 0.1$, $b = 0.5$. The Graph shows the correlation for varying $\epsilon = \epsilon_1 = \epsilon_2$ (left, $l^c = 1$) and $l = l^c$ (right, $\epsilon_1 = \epsilon_2 = 0.4$).

Key Ingredients

We consider the following risk-neutral expectations the key ingredients in credit risk reduced-form models

$$S_{\eta}^k(\theta, t, T) := \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T \theta \eta_s^k ds} \mid \mathbf{F}_t^W \right]$$

$$\Gamma_{\eta}^k(\theta, t, T) := \mathbb{E}^{\mathbb{Q}} \left[\theta \eta_T^k e^{-\int_t^T \theta \eta_s^k ds} \mid \mathbf{F}_t^W \right]$$

$$\bar{S}_{\eta}^k(\theta, t, T) := \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T r_s + \theta \eta_s^k ds} \mid \mathbf{F}_t^W \right]$$

$$S_J^k(\theta, t, T) := \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T \theta J_s^k ds} \mid \mathbf{F}_t^J \right]$$

$$\Gamma_J^k(\theta, t, T) := \mathbb{E}^{\mathbb{Q}} \left[\theta J_T^k e^{-\int_t^T \theta J_s^k ds} \mid \mathbf{F}_t^J \right]$$

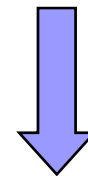
$$\bar{\Gamma}^k(\theta, t, T) := \mathbb{E}^{\mathbb{Q}} \left[\theta \eta_T^k e^{-\int_t^T r_s + \theta \eta_s^k ds} \mid \mathbf{F}_t^W \right]$$

All can be obtained either in **explicit form** or up to the **solution of a ODE system**



$$\begin{aligned} S^k(\theta, t, T) &:= \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T \theta \mu_s^k ds} \mid \mathbf{F}_t^W \right] = S_{\eta}^k(\theta, t, T) \cdot S_J^k(\theta, t, T) \\ \bar{S}^k(\theta, t, T) &:= \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_t^T r_s + \theta \mu_s^k ds} \mid \mathbf{F}_t^W \right] = \bar{S}_{\eta}^k(\theta, t, T) \cdot S_J^k(\theta, t, T) \\ \Gamma^k(\theta, t, T) &:= \mathbb{E}^{\mathbb{Q}} \left[\theta \mu^k e^{-\int_t^T \theta \mu_s^k ds} \mid \mathbf{F}_t^W \right] \\ &= \Gamma_{\eta}^k(\theta, t, T) S_J^k(\theta, t, T) + \Gamma_J^k(\theta, t, T) S_{\eta}^k(\theta, t, T) \\ \bar{\Gamma}^k(\theta, t, T) &= \mathbb{E}^{\mathbb{Q}} \left[\theta \mu^k e^{-\int_t^T r_s + \theta \mu_s^k ds} \mid \mathbf{F}_t^W \right] \\ &= \bar{\Gamma}_{\eta}^k(\theta, t, T) S_J^k(\theta, t, T) + \Gamma_J^k(\theta, t, T) \bar{S}_{\eta}^k(\theta, t, T). \end{aligned}$$

**What can we obtain in
(quasi) closed form using
these key ingredients?**



A lot...

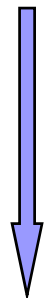
For each name

- Survival (and default) probabilities: $\mathbb{Q}_S^k(t, t_n)$
- Prices of defaultable discount bonds: $\bar{p}_0^k(t, t_n)$
- Prices of digitals : $e^{*k}(t, t_{n-1}, t_n)$
(price of a payoff of 1 if default occurs in $(t_{n-1}, t_n]$)

For any two names

- Default correlation: $\rho^{i,j}(t, T)$

For portfolios

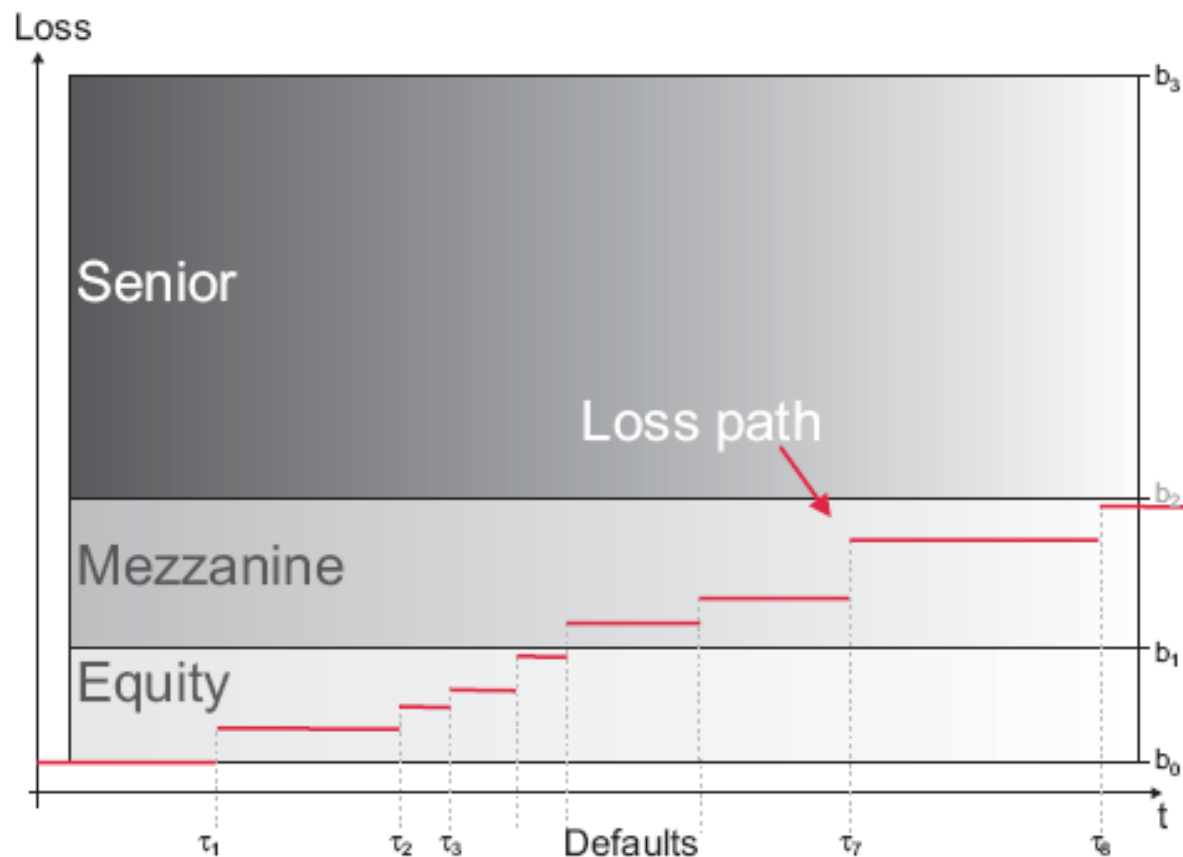


- First-to-default spreads: $s^{FD}(t)$
- Total loss distributions under $\mathbb{Q}(L_T \leq x | \mathbf{G}_t)$ $\mathbb{Q}^T(L_T \leq x | \mathbf{G}_t)$
- Par spreads for CDO tranches on credit indices: $S(t, b_1, b_2)$

CDOs

The loss process:

$$L(t) := \sum_{\tau^k \leq t} q^k M^k$$



Tranche Losses:

$$L^i(t) = \begin{cases} 0 & \text{if } L(t) < b^{i-1} \\ L(t) - b^{i-1} & \text{if } b^{i-1} \leq L(t) < b^i \\ b^i - b^{i-1} & \text{if } L(t) \geq b^i \end{cases}$$



Credit Indices

- The iTraxx is effectively a **portfolio of 125 single CDS**.
- To guarantee liquidity, the portfolio is reorganized (the so-called series) semiannually.
- The aim of this procedure is to guarantee that the underlying portfolio stays in a certain class of credit worthiness.

Mathematical Setting:

- Assume the notional is 1
- The credit index is on K names, each represented by a CDS with spread
- All names are in the same credit class so it reasonable to assume some **homogeneity** in terms of losses (given default), sensitivity to common factors and tenor structure

$$q^k = q \quad \epsilon^k = \epsilon \quad t_j = j\Delta, 1 \leq j \leq N^*$$

- Each name have equal weight ($M^k = M$)
- Typically, the recovery in traded indices is set to zero, but we stay a bit more general at this point.

Index Spread:

The payment stream of the credit index is as follows.

$$S \Delta \frac{K - N_{t_n}}{K}$$

Fixed leg

Defaulting leg

$$\sum_{\tau^k \in (t_{n-1}, t_n]} (1 - q) = (1 - q)(N_{t_n} - N_{t_{n-1}})$$

- The spread of a credit index is an average of single CDS spreads **only until** the first default occurs.

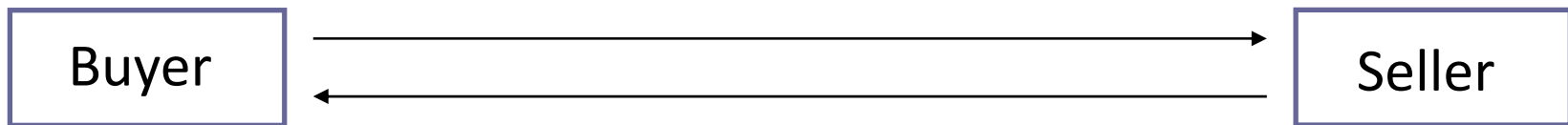
- The exact expression of the index spread is given by

$$S_t = \bar{q} K \frac{\sum_{n=2}^{N^*} \mathbf{1}_{\{t_n \geq t\}} \sum_{k=1}^K e^{*k}(t, t_{n-1}, t_n)}{\Delta \sum_{t_n \geq t}^{t_{N^*-1}} \sum_{k=1}^K \bar{p}_0^k(t, t_n)}$$

Tranches on Credit Indices

- We consider the overall nominal to be 1.
- A **tranche** refers to an interval $(b_1; b_2]$ $[0; 1]$.
- Investing (selling protection) in a tranche is again done by a swap where the following payments are exchanged:

$$S \left(\mathbf{1}_{\{L_{t_n} \leq b_1\}} + \frac{(b_2 - L_{t_n})^+}{b_2 - b_1} \mathbf{1}_{\{L_{t_n} > b_1\}} \right)$$



$$\left((b_2 - L_{t_{n-1}})^+ - (b_2 - L_{t_n})^+ \right) \mathbf{1}_{\{L_{t_n} > b_1\}}$$

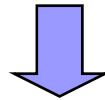
- The par spread of tranche (b1; b2] is given by

$$S(t, b_1, b_2) =$$

$$= \frac{\int_{b_1}^{b_2} \left(p(t, t_1) \mathbf{1}_{\{L_t < y\}} - C(t, t_{N^*}, y) + \sum_{n=2}^{N^*-1} C(t, t_n, y) \cdot \left(1 - \frac{p(t, t_{n+1})}{p(t, t_n)} \right) \right) dy}{\sum_{n=1}^{N^*-1} \frac{1}{b_2 - b_1} \int_{b_1}^{b_2} C(t, t_n, y) dy}$$

where

$$C(t, T, y) = p(t, T) \mathbb{Q}^T (L_T < y | \mathbf{G}_t)$$



Thus also in (quasi) closed-form !



Conclusion

- We propose to use a **class of quadratic shot-noise reduced form models** when dealing with credit risk.
- We show that this class of models:
 - is flexible enough to capture realistic **correlation and clustering** of defaults
 - it is rather **parsimonious** as all (default) correlations result from a common term in the default intensities of the names
 - allow to derive in (quasi) **close form solutions** to all interesting variables of credit risk. Even in the case of portfolio credit products
 - **consistently** prices single-name and portfolio credit products.



References

- **This paper:**

- Gaspar, R.M. and T. Schmidt (2008) On the Pricing of CDOs, in *Credit Derivatives* (Ed. P.U. Ali and G. Gregouriou), McGraw-Hill.

- **Related papers:**

- Gaspar, R.M. and T. Schmidt (2007), Term Structure Models with shot-noise effects, *available at SSRN and DefaultRisk.com*
- Gaspar, R.M. (2004), "General Quadratic Term Structures of Bond, Futures and Forward Prices", SSE/EFI working papers Series in Economics and Finance, n.559.

Further Research

- Calibrate a concrete instance of our class of models to market data!

Other Suggestions Welcome...