# **Risk Management aspects of CDOs**

## CDOs after the crisis: Valuation and risk management reviewed 30 September 2008

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Presentation related to the paper Hedging default risks of CDOs in Markovian contagion models (2008) Available on <u>www.defaultrisk.com</u>

# **Preliminary or obituary?**

- On human grounds, shrinkage rather than enlargement of the job market
- Thanks to the crisis, our knowledge of the flaws of the various competing models has dramatically improved...
  - We know that we don't know and why
  - No new paradigm has yet emerged (if ever)
  - Paradoxically, academic research is making good progress
  - ... but at its own pace
- Model to be presented is low tech, unrealistic, nothing new
- But deserves to be known (this is pure speculation)
- Provides an academic view on practical issues

- Does not intend to give the insights of a trader or a risk manager

# Overview

- CDO Business context
  - Decline of the one factor Gaussian copula model for risk management purposes
  - Recent correlation crisis
  - Unsatisfactory credit deltas for CDO tranches
- Tree approach to hedging defaults
  - From theoretical ideas
  - To practical implementation of hedging strategies
  - Robustness of the approach?
- Mathematical framework
- Empirical results

- We are within a financial turmoil
  - Lots of restructuring and risk management of trading books
  - Collapse of highly leveraged products (CPDO)
  - February and March 2008 crisis on iTraxx and CDX markets
    - Surge in credit spreads
    - Extremely high correlations
    - ≻Trading of [60-100%] tranches
    - Emergence of recovery rate risk
  - What is really a default event ?
    - ≻How to cope with Fed or Treasury activism?
  - Questions about the pricing of bespoke tranches
     >Unreliability of projection techniques

#### Morgan Stanley

#### MORGAN STANLEY RESEARCH

March 10, 2008 Structured Credit Analytics

#### CDX and iTraxx – Correlation Analysis and Delta Neutral Return

#### **CDX Series 9**

| Index<br>5yr      | Tranche | Bid    |        | Mid 1 Week |             |        |       |        |       |        |        |            |           |           |        |                 |                                       |
|-------------------|---------|--------|--------|------------|-------------|--------|-------|--------|-------|--------|--------|------------|-----------|-----------|--------|-----------------|---------------------------------------|
| буг               |         | Did    |        |            | Mid 1 Month | Index  |       | Detach | Corr. | 1 Week | Skew Z | Corr. Skew | ITRAXX    | Value vs. |        | Neutral Returns |                                       |
|                   |         | Biu    | Ask    | Change     | Change      | Level  | Delta | Corr.1 | Skew  | Change | Score  | Rel. Value | Corr Skew | iTraxx    | 1 Week | 1 Month         | 1 Year                                |
|                   | Index   | 185.0  | 185.0  | 25.0       | 69.0        |        |       |        |       |        |        |            |           |           |        |                 |                                       |
| 5yr <sup>2</sup>  | 0-3%    | 67.5%  | 68.5%  | 10.4%      | 13.0%       | 185 bp | 2.7   | 38%    |       | -7.4%  | 1.19   | RICH       | (1.10)    | CHEAP     | -6.6%  | 0.6%            | 50.0%                                 |
| 5yr               | 3-7%    | 731    | 738    | 138        | 219         |        | 3.0   | 64%    | 25%   | 2.7%   | 1.07   | RICH       | 0.97      | FAIR      | -2.1%  | 3.2%            | 9.0%                                  |
| 5yr               | 7-10%   | 405    | 415    | 75         | 141         |        | 1.8   | 73%    | 9%    | 0.5%   | (0.31) | FAIR       | 0.21      | FAIR      | -1.2%  | 1.4%            | -4.7%                                 |
| 5yr               | 10-15%  | 207    | 214    | 12         | 73          |        | 1.3   | 37%    | 14%   | 2.1%   | 1.15   | RICH       | 0.96      | FAIR      | 0.6%   | 1.2%            | -1.8%                                 |
| 5yr               | 15-30%  | 126    | 130    | 13         | 62          |        | 1.0   | 198%   | 21%   | 1.2%   | (0.01) | FAIR       | 0.18      | FAIR      | 0.4%   | -0.3%           | -1.3%                                 |
| 5yr               | 30-100% | 78     | 80     | 10         | 45          |        | 0.7   |        |       |        |        |            |           |           | 0.3%   | -0.6%           | -1.6%                                 |
| 7yr               | Index   | 178.0  | 178.0  | 20.0       | 62.0        |        |       |        |       |        |        |            |           |           |        |                 |                                       |
| 7yr*              | 0-3%    | 71.0%  | 72.0%  | 8.0%       |             | 178 bp | 2.0   | 39%    |       | -4.2%  | 1.38   | RICH       |           |           | -5.0%  | 1.8%            | 42.9%                                 |
| 7yr               | 3-7%    | 785    | 795    | 130        | 193         |        | 3.2   | 62%    | 23%   | 1.6%   | 1.00   | FAIR       |           |           | -2.6%  | 5.7%            | 18.2%                                 |
| 7yr               | 7-10%   | 452    | 460    | 79         | 106         |        | 2.5   | 71%    | 8%    | 0.1%   | (0.66) | FAIR       |           |           | -1.8%  | 5.4%            | -2.8%                                 |
| 7yr               | 10-15%  | 256    | 265    | 21         | 73          |        | 1.6   | 33%    | 12%   | 1.6%   | (0.09) | FAIR       |           |           | 0.5%   | 3.0%            | -2.0%                                 |
| 7yr               | 15-30%  | 139    | 144    | 14         | 65          |        | 1.1   | 104%   | 21%   | 0.3%   | (0.73) | FAIR       |           |           | 0.3%   | -0.1%           | -1.7%                                 |
| 7yr               | 30-100% | 81     | 83     | 12         | 48          |        | 0.7   |        |       |        |        |            |           |           | 0.1%   | -0.9%           | -2.1%                                 |
| 10yr              | Index   | 174.0  | 174.0  | 18.0       | 57.0        |        |       |        |       |        |        |            |           |           |        |                 | i i i i i i i i i i i i i i i i i i i |
| 10yr <sup>2</sup> | 0-3%    | 74.0%  | 74.8%  | 7.9%       | 9.9%        | 174 bp | 1.5   | 39%    |       | -5.5%  | 1.33   | RICH       | 1.31      | RICH      | -5.2%  | 1.5%            | 31.6%                                 |
| 10yr              | 3-7%    | 910    | 920    | 110        | 204         |        | 3.4   | 57%    | 18%   | 3.3%   | 3.13   | RICH       | 3.16      | RICH      | -1.6%  | 7.5%            | 37.4%                                 |
| 10yr              | 7-10%   | 523    | 533    | 67         | 98          |        | 2.5   | 64%    | 7%    | 0.4%   | (0.64) | FAIR       | (0.64)    | FAIR      | -0.9%  | 8.5%            | 1.8%                                  |
| 10yr              | 10-15%  | 300    | 308    | 22         | 95          |        | 1.8   | 76%    | 12%   | 1.1%   | (0.60) | FAIR       | (0.60)    | FAIR      | 1.0%   | 2.6%            | 3.0%                                  |
| 10yr              | 15-30%  | 150    | 155    | 14         | 63          |        | 1.2   | 99%    | 23%   | 0.3%   | (1.99) | CHEAP      | (2.01)    | CHEAP     | 0.5%   | 0.0%            | -1.6%                                 |
| 10yr              | 30-100% | 79     | 82     | 10         | 45          |        | 0.7   |        |       |        |        |            |           |           | 0.2%   | -1.2%           | -2.9%                                 |
| HY                | Index   | 718.8  | 718.8  | 23.6       | 83.1        |        |       |        |       |        |        |            |           |           |        |                 |                                       |
| HY3               | 0-10%   | 92.1%  | 92.6%  | 1.6%       | 2.5%        | 719 bp | 0.7   | 44%    |       | -2.4%  | 1.05   | RICH       |           |           | -1.1%  | -0.4%           | 10.2%                                 |
| HY                | 10-15%  | 75.3%  | 75.8%  | 2.5%       | 6.1%        |        | 1.8   | 49%    | 5%    | 1.2%   | (0.12) | FAIR       |           |           | -1.4%  | -1.4%           | 8.0%                                  |
| HY                | 15-25%  | 1,290  | 1,305  | 81         | 222         |        | 2.3   | 72%    | 23%   | 1.4%   | 2.73   | RICH       |           |           | -1.0%  | -1.5%           | 1.6%                                  |
| HY                | 25-35%  | 680    | 695    | 48         | 145         |        | 1.5   | 92%    | 19%   | 0.2%   | (0.21) | FAIR       |           |           | -0.6%  | -1.6%           | -6.3%                                 |
| HY                | 35-100% | 202    | 208    | (2)        | 32          |        | 0.6   |        |       |        | . ,    |            |           |           | 0.5%   | 0.0%            | -2.1%                                 |
| LCDX              | Index   | 440.0  | 440.0  | 7.8        | 29.6        |        |       |        |       |        |        |            |           |           |        |                 |                                       |
| LCDX              | 0-5%    | 86.6%  | 87.5%  | 3.9%       | -0.2%       | 440 bp | 1.6   | 78%    |       | -6.2%  |        |            |           |           | -3.5%  | 3.0%            |                                       |
| LCDX3             | 5-8%    | 66.9%  | 67.6%  | 4.3%       | 1.0%        |        | 2.7   |        |       |        |        |            |           |           | -3.6%  | 3.6%            |                                       |
| LCDX              | 8-12%   | 1140.0 | 1152.0 | 30.0       | 35.0        |        | 3.5   |        |       |        |        |            |           |           | 0.0%   | 3.0%            |                                       |
| LCDX              | 12-15%  | 740.0  | 748.0  | -7.0       | 107.0       |        | 2.4   |        |       |        |        |            |           |           | 0.9%   | -1.4%           |                                       |
| LCDX              | 15-100% | 206.0  | 210.0  | -13.5      | 20.0        |        | 0.6   |        |       |        |        |            |           |           | 0.7%   | -0.8%           |                                       |

<sup>1</sup>Correlation of tranche with 0% attachment and the same detachment point as the benchmark tranche, implied from market prices of benchmark tranches

<sup>2</sup>Points upfront plus 500 bp running

<sup>3</sup>Points upfront plus 0 bp running

Source: Morgan Stanley

- Extremely high spreads for senior and super senior tranches
  - Issues with the right-end of the loss distribution
  - Prices in March were not consistent with a fixed recovery rate of 40%
    - See Burtschell et al. [2008], updated version
    - See anecdotal evidence from previous slide
  - In mid-September 2008, no implied correlation could be found for the [30-100%] tranche on the CDX IG even with a 0% recovery
    - Constraints related to the increase and concavity of expected losses on base tranches may not be fulfilled
    - Seemingly, inconsistencies between prices and arbitrage opportunities

#### • **Recovery rates**

#### - Market agreement of a fixed recovery rate of 40% is inadequate

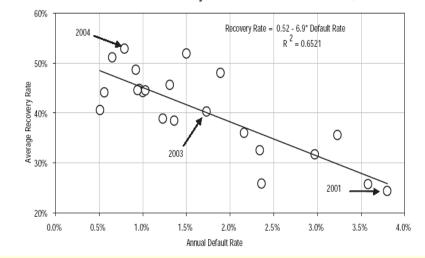


Exhibit 10 - Correlation between Recovery Rates and Annual Default Rates, 1983-2004

- Currently a major issue in the CDO market
   Krekel (2008), Amraoui and Hitier (2008)
- Use of state dependent stochastic recovery rates will dramatically change the credit deltas
- Management of recovery rate risk?

- Some basic issues are back:
  - What is really a default event?
    - ≻Restructuring, take over
    - ► Recent example of Wachovia
  - and can we define recovery rates properly and consistently across various products (CDS, CLNs, CDOs) and time horizons ?

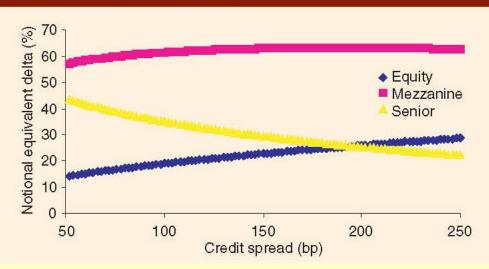
Recent example of Washington Mutual

Application of ISDA Protocol to Lehman Brothers

- Jarrow et al. (2008): "distressed debt prices and recovery rate estimation"
  - Raise serious doubts about recovery rate estimation
  - Question the notions of "economic" and "reported" default dates

- CDS hedge ratios are computed by bumping the marginal credit curves
  - In 1F Gaussian copula framework
  - Focus on credit spread risk
  - individual name effects
  - Bottom-up approach
  - Smooth effects
  - Pre-crisis...
- At first sight, poor theoretical properties
  - Does not lead to a replication of CDO tranche payoffs
  - Not a hedge against defaults...
  - Unclear issues with respect to the management of correlation risks
  - To be discussed further (break-even correlation)

#### 4. CDO tranche deltas

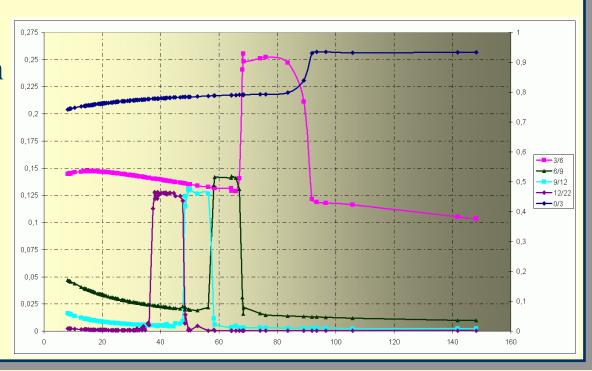


- Decline of the one factor Gaussian copula model
- Credit deltas in "high correlation states"
  - Close to comonotonic default dates (current market situation)
  - Deltas are equal to zero or one depending on the level of spreads

>Individual effects are too pronounced

➢Unrealistic i-gammas

≻Morgan & Mortensen



- The decline of the one factor Gaussian copula model + base correlation
  - This is rather a practical than a theoretical issue
- Negative tranche deltas frequently occur
  - Which is rather unlikely for out of the money call spreads
    - Though this could actually arise in an arbitrage-free model
    - Schloegl, Mortensen and Morgan (2008)
  - Especially with steep base correlations curves
    - In the base correlation approach, the deltas of base tranches are computed under different correlations
  - And with thin tranchelets
    - Often due to "numerical" and interpolation issues

- No clear agreement about the computation of credit deltas in the 1F Gaussian copula model
  - Sticky correlation, sticky delta?
  - Computation with respect to credit default swap index, individual CDS?
  - Volatility in the difference between CDS index spread and the average spread of the names in the index
- Weird effects when pricing and risk managing bespoke tranches
  - Price dispersion due to "projection" techniques
  - Negative deltas effects magnified
  - Sensitivity to names out of the considered basket

- Recent advances such as the notion of "break-even" correlation shed new light on the Gaussian copula model
- Think of a structural model with correlated Brownian motions:
  - Default leg of a CDS is like a barrier option
  - Defaults are predictable: no need to account for default risk
  - Correlation between credit spreads equals "correlation" between default events
  - Perfect replication of a CDO tranche can be achieved with the underlying CDS

#### > To cope with credit spread risk

- One factor Gaussian copula model close to structural model
  - Hull, Predescu and White (2005), Cousin and Laurent (2008)

- Credit deltas in Gaussian copula models may be viewed as approximations of replication deltas in the previous structural model
  - Provided that flat correlation is equal to the correlation of credit spreads
- Drawbacks in the previous approach:
  - Jumps in asset values

Associated with tail dependence in credit spreads
 And fat tails in loss distributions

- Creates incompleteness
- Spreads and stocks may move in the same way
  - > Due to state financial support
  - > Collapse of the standard structural model
  - ➤ And debt-equity arbitrage...

- Complete markets
  - As many risks as hedging instruments
- Perfect replication of payoffs by dynamically trading a small number of « underlying assets »
  - Local volatility type framework
  - Obviously, a stylized view on risk management: model risk
- That is further investigated in the presentation
  - Dynamic trading of CDS to replicate CDO tranche payoff
  - Based on Laurent, Cousin and Fermanian (2008)

- What are we trying to achieve?
- Show that under some (stringent) assumptions the market for CDO tranches is complete
  - CDO tranches can be perfectly replicated by dynamically trading CDS
  - > Exhibit the building of the unique risk-neutral measure
- Display the analogue of the local volatility model of Dupire or Derman & Kani for credit portfolio derivatives

One to one correspondence between CDO tranche quotes and model dynamics (continuous time Markov chain for losses)

• Show the practical implementation of the model with market data

> Deltas correspond to "sticky implied tree"

- Main theoretical features of the complete market model
  - No simultaneous defaults
    - Unlike multivariate Poisson models
  - Credit spreads are driven by defaults
  - Jumps in credit spreads at default times
    - Contagion model (Jarrow & Yu)
      - Enron failure was informative
      - Not consistent with the "conditional independence" assumption
    - Credit spreads are deterministic between two defaults
  - Bottom-up approach
    - > Aggregate loss intensity is derived from individual loss intensities
  - Correlation dynamics is also driven by defaults
    - > Defaults lead to an increase in dependence

- Changes in the dependence structure between default times
  - In the Gaussian copula world, change in the correlation parameters in the copula
  - The present value of the default leg of an equity tranche decreases when correlation increases
- Dependence parameters and credit spreads may be highly correlated

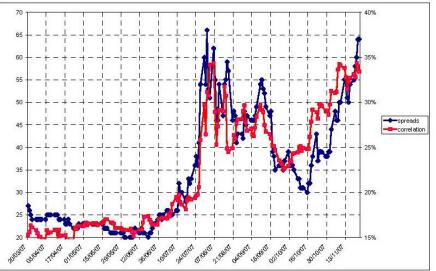
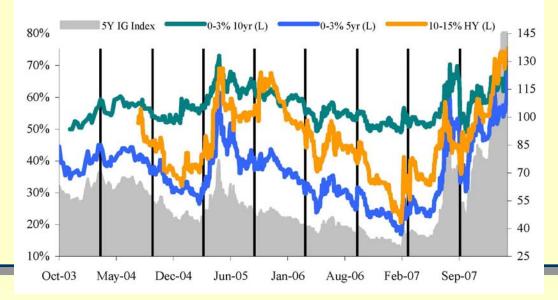
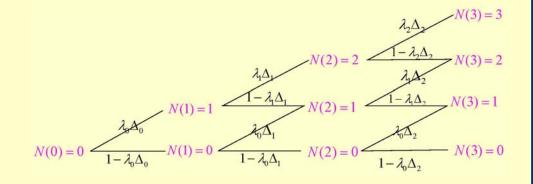


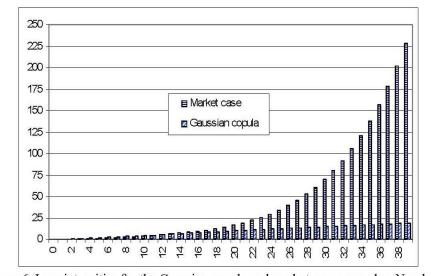
Figure 9. Credit spreads on the five years iTraxx index (Series 7) in bps on the left axis. Implied correlation on the equity tranche on the right axis

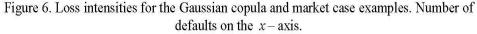
- Without additional assumptions the model is intractable
  - Homogeneous portfolio
    - ≻ Only need of the CDS index
    - ≻ No individual name effect, no i-Gamma
    - Top-down approach
      - Only need of the aggregate loss dynamics
- Parallel shifts in credit spreads
  - On March 10, 2008, the 5Y CDX IG index spread quoted at 194 bp pa
  - starting from 30 bp pa on February 2007



- Without additional assumptions the model is intractable
  - Markovian dynamics
    - Pricing and hedging CDO tranches within a binomial tree
  - Perfect calibration the loss dynamics from CDO tranche quotes
    - Thanks to forward induction in the tree
    - On the right, loss intensities wrt number of defaults







- We will start with two names only
- Firstly in a static framework
  - Look for a First to Default Swap
  - Discuss historical and risk-neutral probabilities
- Further extending the model to a dynamic framework
  - Computation of prices and hedging strategies along the tree
  - Pricing and hedging of tranchelets
- Multiname case: homogeneous Markovian model
  - Computation of risk-neutral tree for the loss
  - Computation of dynamic deltas
- Technical details can be found in the paper:
  - "hedging default risks of CDOs in Markovian contagion models"

- Some notations :
  - $-\tau_1, \tau_2$  default times of counterparties 1 and 2,
  - $\mathcal{H}_t$  available information at time *t*,
  - -P historical probability,

-  $\alpha_1^P, \alpha_2^P$  : (historical) default intensities:  $P[\tau_i \in [t, t + dt[|H_t] = \alpha_i^P dt, i = 1, 2]$ 

• Assumption of « local » independence between default events – Probability of 1 and 2 defaulting altogether:  $P[\tau_1 \in [t, t + dt[, \tau_2 \in [t, t + dt[|H_t]] = \alpha_1^P dt \times \alpha_2^P dt \text{ in } (dt)^2]$ 

- Local independence: simultaneous joint defaults can be neglected

- Building up a tree:
  - Four possible states: (*D*,*D*), (*D*,*ND*), (*ND*,*D*), (*ND*,*ND*)
  - Under no simultaneous defaults assumption  $p_{(D,D)}=0$
  - Only three possible states: (*D*,*ND*), (*ND*,*D*), (*ND*,*ND*)
  - Identifying (historical) tree probabilities:

$$\alpha_{1}^{P}dt \quad (D, ND)$$

$$\alpha_{2}^{P}dt \quad (ND, D)$$

$$1 - (\alpha_{1}^{P} + \alpha_{2}^{P})dt \quad (ND, ND)$$

$$0 \Rightarrow n = -n + n = -n$$

л

$$\begin{cases} p_{(D,D)} = 0 \Rightarrow p_{(D,ND)} = p_{(D,D)} + p_{(D,ND)} = p_{(D,.)} = \alpha_1^P dt \\ p_{(D,D)} = 0 \Rightarrow p_{(ND,D)} = p_{(D,D)} + p_{(ND,D)} = p_{(.,D)} = \alpha_2^P dt \\ p_{(ND,ND)} = 1 - p_{(D,.)} - p_{(.,D)} \end{cases}$$

- Stylized cash flows of short term digital CDS on counterparty 1:
  - $\alpha_1^Q dt$  CDS 1 premium

$$\alpha_{1}^{P}dt \quad 1 - \alpha_{1}^{Q}dt \quad (D, ND)$$

$$\alpha_{2}^{P}dt \quad -\alpha_{1}^{Q}dt \quad (ND, D)$$

$$1 - (\alpha_{1}^{P} + \alpha_{2}^{P})dt \quad -\alpha_{1}^{Q}dt \quad (ND, ND)$$

• Stylized cash flows of short term digital CDS on counterparty 2:

 $\alpha_{1}^{P}dt -\alpha_{2}^{Q}dt \quad (D,ND)$   $0 \qquad \alpha_{2}^{P}dt \qquad 1-\alpha_{2}^{Q}dt \quad (ND,D)$   $1-(\alpha_{1}^{P}+\alpha_{2}^{P})dt \qquad -\alpha_{2}^{Q}dt \quad (ND,ND)$ 

• Cash flows of short term digital first to default swap with premium  $\alpha_F^Q dt$ :

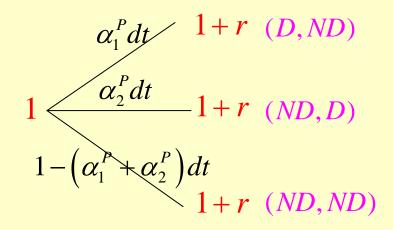
$$\alpha_{1}^{P}dt = 1 - \alpha_{F}^{Q}dt \quad (D, ND)$$

$$\alpha_{2}^{P}dt = 1 - \alpha_{F}^{Q}dt \quad (ND, D)$$

$$1 - (\alpha_{1}^{P} + \alpha_{2}^{P})dt - \alpha_{F}^{Q}dt \quad (ND, ND)$$

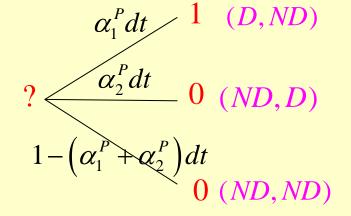
- Cash flows of holding CDS 1 + CDS 2:  $\alpha_1^P dt \quad 1 - (\alpha_1^Q + \alpha_2^Q) dt \quad (D, ND)$   $0 \quad \alpha_2^P dt \quad 1 - (\alpha_1^Q + \alpha_2^Q) dt \quad (ND, D)$   $1 - (\alpha_1^P + \alpha_2^P) dt \quad (-(\alpha_1^Q + \alpha_2^Q) dt \quad (ND, ND))$
- Perfect hedge of first to default swap by holding 1 CDS 1 + 1 CDS 2
  - Delta with respect to CDS 1 = 1, delta with respect to CDS 2 = 1

- Absence of arbitrage opportunities imply:
  - $\alpha_F^Q = \alpha_1^Q + \alpha_2^Q$
- Arbitrage free first to default swap premium
  - Does not depend on historical probabilities  $\alpha_1^P, \alpha_2^P$
- Three possible states: (*D*,*ND*), (*ND*,*D*), (*ND*,*ND*)
- Three tradable assets: CDS1, CDS2, risk-free asset



• For simplicity, let us assume r = 0

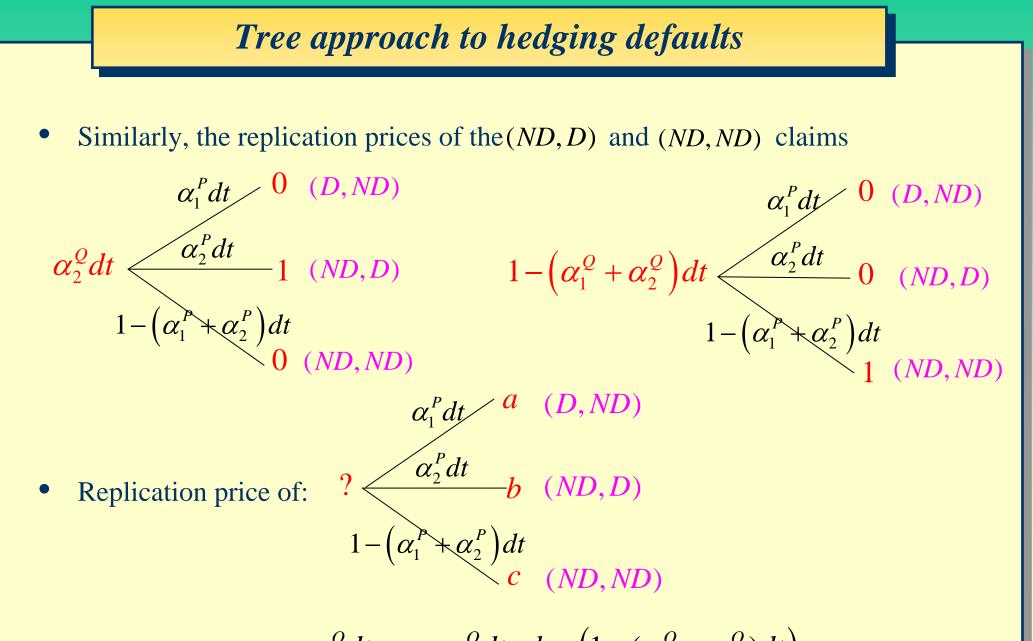
- Three state contingent claims
  - Example: claim contingent on state (*D*, *ND*)
  - Can be replicated by holding
  - 1 CDS 1 +  $\alpha_1^Q dt$  risk-free asset



- Replication price = 
$$\alpha_1^Q dt$$

 $\alpha_1^P dt$ 

 $\alpha_2^P dt$  $\alpha_1^Q dt$ 



• Replication price =  $\alpha_1^Q dt \times a + \alpha_2^Q dt \times b + (1 - (\alpha_1^Q + \alpha_2^Q) dt)c$ 

- Replication price obtained by computing the expected payoff
  - Along a risk-neutral tree

$$\alpha_1^{\mathcal{Q}}dt \times a + \alpha_2^{\mathcal{Q}}dt \times b + \left(1 - (\alpha_1^{\mathcal{Q}} + \alpha_2^{\mathcal{Q}})dt\right)c \xrightarrow{\alpha_2^{\mathcal{Q}}dt} b (ND, D)$$

$$1 - \left(\alpha_1^{\mathcal{Q}} + \alpha_2^{\mathcal{Q}}\right)dt \xrightarrow{C} (ND, ND)$$

- Risk-neutral probabilities
  - Used for computing replication prices
  - Uniquely determined from short term CDS premiums
  - No need of historical default probabilities

- Computation of deltas
  - Delta with respect to CDS 1:  $\delta_1$
  - Delta with respect to CDS 2:  $\delta_2$
  - Delta with respect to risk-free asset: p

 $\succ p$  also equal to up-front premium

$$\begin{cases} a = p + \delta_1 \times \overbrace{\left(1 - \alpha_1^{\mathcal{Q}} dt\right)}^{\text{payoff CDS 1}} + \delta_2 \times \overbrace{\left(-\alpha_2^{\mathcal{Q}} dt\right)}^{\text{payoff CDS 2}} \\ b = p + \delta_1 \times \left(-\alpha_1^{\mathcal{Q}} dt\right) + \delta_2 \times \left(1 - \alpha_2^{\mathcal{Q}} dt\right) \\ c = p + \delta_1 \times \underbrace{\left(-\alpha_1^{\mathcal{Q}} dt\right)}_{\text{payoff CDS 1}} + \delta_2 \times \underbrace{\left(-\alpha_2^{\mathcal{Q}} dt\right)}_{\text{payoff CDS 2}} \end{cases}$$

- As for the replication price, deltas only depend upon CDS premiums

Dynamic case:

 $\lambda_{2}^{\varrho}dt \quad (D,D)$   $\alpha_{1}^{\varrho}dt \quad (D,ND) \quad 1-\lambda_{2}^{\varrho}dt \quad (D,ND)$   $\alpha_{2}^{\varrho}dt \quad (ND,D) \quad \kappa_{1}^{\varrho}dt \quad (D,D)$   $1-(\alpha_{1}^{\varrho}+\alpha_{2}^{\varrho})dt \quad (ND,ND) \quad \pi_{1}^{\varrho}dt \quad (ND,D)$   $\pi_{2}^{\varrho}dt \quad (ND,D)$   $1-(\pi_{1}^{\varrho}+\pi_{2}^{\varrho})dt \quad (ND,ND)$ emium after default of name 1  $\lambda_2^{Q} dt$  CDS 2 premium after default of name 1

- $-\kappa_1^{\varrho} dt$  CDS 1 premium after default of name 2
- $\pi_1^Q dt$  CDS 1 premium if no name defaults at period 1
- $\pi_2^Q dt$  CDS 2 premium if no name defaults at period 1
- Change in CDS premiums due to contagion effects
  - Usually,  $\pi_1^Q < \alpha_1^Q < \kappa_1^Q$  and  $\pi_2^Q < \alpha_2^Q < \lambda_2^Q$

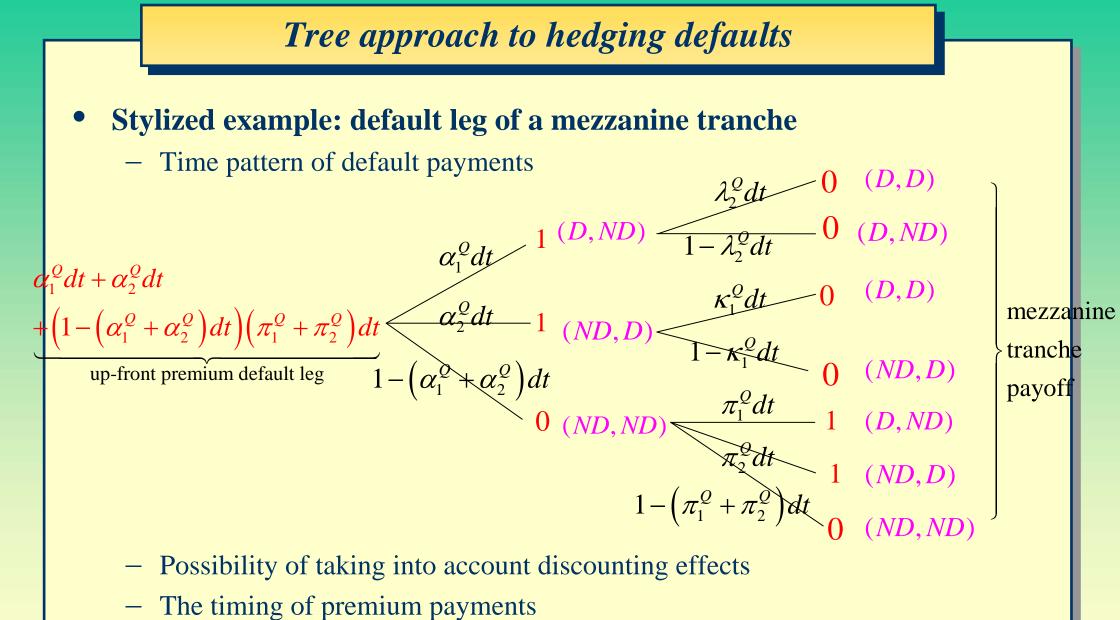
- Computation of prices and hedging strategies by backward induction
  - use of the dynamic risk-neutral tree
  - Start from period 2, compute price at period 1 for the three possible nodes
  - + hedge ratios in short term CDS 1,2 at period 1
  - Compute price and hedge ratio in short term CDS 1,2 at time 0
- Example: term structure of credit spreads
  - computation of CDS 1 premium, maturity = 2
  - $p_1 dt$  will denote the periodic premium
  - Cash-flow along the nodes of the tree

Computations CDS on name 1, maturity =  $2 \frac{\lambda_2^Q dt}{\lambda_2^Q dt}$ (D,D) $\begin{array}{c} \alpha_{1}^{Q}dt & 1 - p_{1}dt \ (D,ND) & 1 - \lambda_{2}^{Q}dt & 0 \ (D,ND) \\ 0 & \alpha_{2}^{Q}dt & -p_{1}dt \ (ND,D) & \kappa_{1}^{Q}dt & 1 - p_{1}dt \ (D,D) \\ 1 - (\alpha_{1}^{Q} + \alpha_{2}^{Q})dt & -p_{1}dt \ (ND,ND) & \pi_{1}^{Q}dt & 1 - p_{1}dt \ (ND,D) \\ -p_{1}dt \ (ND,ND) & \pi_{2}^{Q}dt & -p_{1}dt \ (D,ND) \\ 1 - (\pi_{1}^{Q} + \pi_{2}^{Q})dt & -p_{1}dt \ (ND,D) \\ 1 - (\pi_{1}^{Q} + \pi_{2}^{Q})dt & -p_{1}dt \ (ND,ND) \end{array}$ 

• Premium of CDS on name 1, maturity = 2, time = 0,  $p_1 dt$  solves for:  $0 = (1 - p_1)\alpha_1^Q + (-p_1 + (1 - p_1)\kappa_1^Q - p_1(1 - \kappa_1^Q))\alpha_2^Q + (-p_1 + (1 - p_1)\pi_1^Q - p_1\pi_2^Q - p_1(1 - \pi_1^Q - \pi_2^Q))(1 - \alpha_1^Q - \alpha_2^Q)$ 

- Stylized example: default leg of a senior tranche
  - Zero-recovery, maturity 2
  - Aggregate loss at time 2 can be equal to 0,1,2
    - > Equity type tranche contingent on no defaults
    - > Mezzanine type tranche : one default
    - Senior type tranche : two defaults

| $\lambda_{2}^{\varrho} dt = \begin{pmatrix} D, ND \end{pmatrix} = \begin{pmatrix} \lambda_{2}^{\varrho} dt & 1 & (D, D) \\ \hline 1 - \lambda_{2}^{\varrho} dt & 0 & (D, ND) \end{pmatrix}$                     |                                |
|---|--------------------------------|
| $\underbrace{\alpha_1^{\varrho}dt \times \kappa_2^{\varrho}dt + \alpha_2^{\varrho}dt \times \kappa_1^{\varrho}dt}_{(ND,D)} \underbrace{\kappa_1^{\varrho}dt}_{(D,D)} \underbrace{\kappa_1^{\varrho}dt}_{(D,D)}$ | senior                         |
| up-front premium default leg<br>$1 - (\alpha_1^Q + \alpha_2^Q) dt$ $1 - \kappa_1^Q dt$ $0$ (ND, D)  | <pre>&gt; tranche payoff</pre> |
| $(ND,ND) = \frac{\pi_1^Q dt}{(ND,ND)} = 0  (D,ND)$  | puyon                          |
| $\pi_2^{\Phi} dt = 0$ (ND, D)   |                                |
| $1 - \left(\pi_1^Q + \pi_2^Q\right) dt \qquad (ND, ND)$   | J                              |

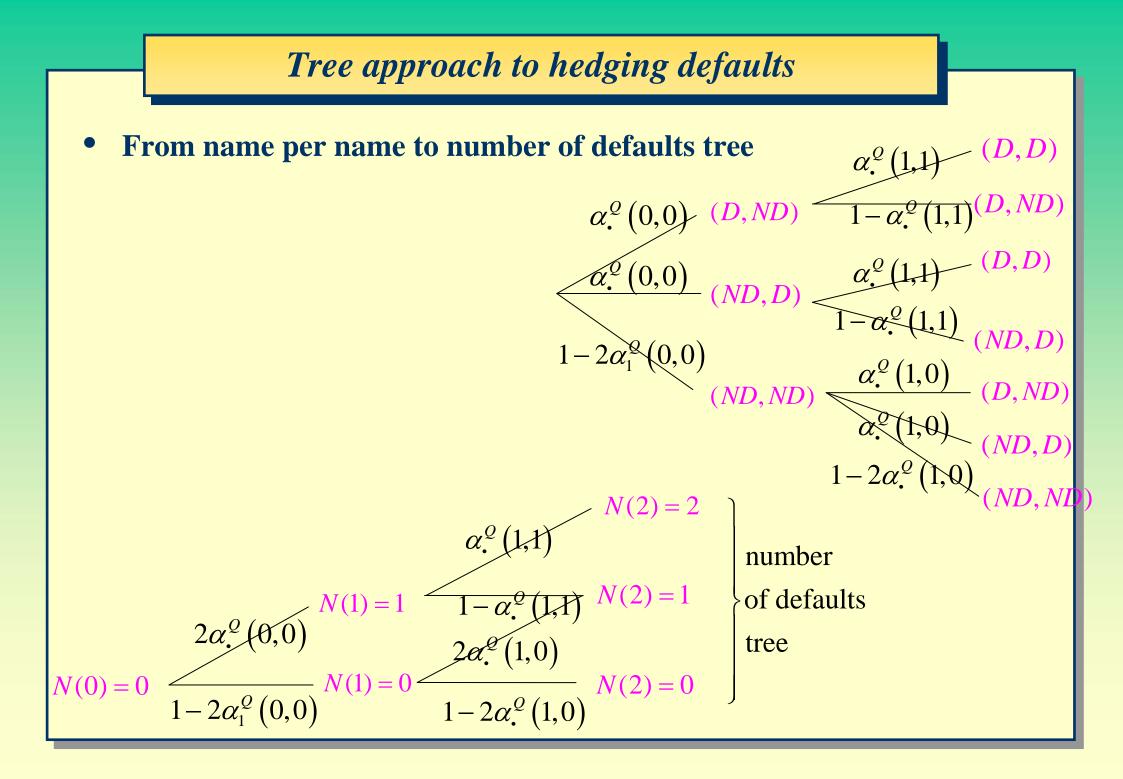


- The timing of premium payments
- Computation of dynamic deltas with respect to short or actual CDS on names 1,2

- In theory, one could also derive dynamic hedging strategies for standardized CDO tranches
  - Numerical issues: large dimensional, non recombining trees
  - Homogeneous Markovian assumption is very convenient
    - CDS premiums at a given time *t* only depend upon the current number of defaults N(t)
  - CDS premium at time 0 (no defaults)  $\alpha_1^Q dt = \alpha_2^Q dt = \alpha_2^Q (t = 0, N(0) = 0)$
  - CDS premium at time 1 (one default)  $\lambda_2^Q dt = \kappa_1^Q dt = \alpha_{\cdot}^Q (t = 1, N(t) = 1)$
  - CDS premium at time 1 (no defaults)  $\pi_1^Q dt = \pi_2^Q dt = \alpha_{\cdot}^Q (t = 1, N(t) = 0)$

## Tree approach to hedging defaults

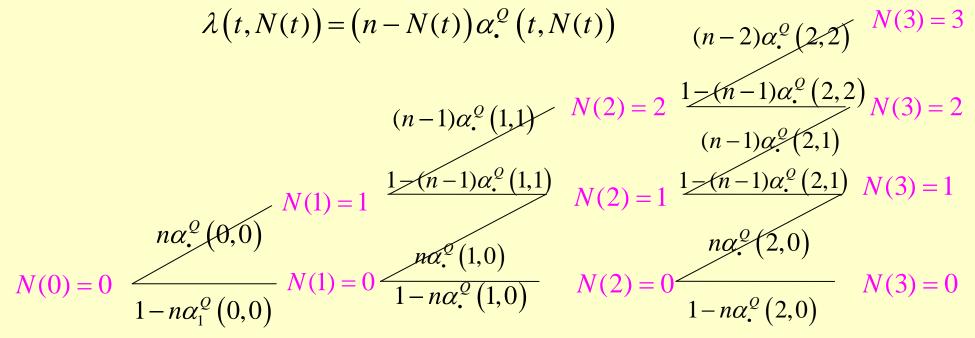
- Tree in the homogeneous case
  - If we have N(1) = 1, one default at t=1  $a \cdot (0,0) + (D, ND) = 1 - a \cdot (1,1) + (D,D)$   $a \cdot (1,1) + (D,D)$   $a \cdot (1,1) + (D,D)$   $a \cdot (1,1) + (D,D)$   $1 - a \cdot (1,1) + (D,D)$   $1 - 2a \cdot (1,0) + (D,D)$  $1 - 2a \cdot (1,0)$
  - The probability to have N(2) = 1, one default at t=2...
  - Is  $1 \alpha_{\cdot}^{Q}(1,1)$  and does not depend on the defaulted name at t=1
  - N(t) is a **Markov process**
  - Dynamics of the number of defaults can be expressed through a binomial tree



#### Tree approach to hedging defaults

#### • Easy extension to *n* names

- Predefault name intensity at time t for N(t) defaults:  $\alpha^Q_{\cdot}(t, N(t))$
- Number of defaults intensity : sum of surviving name intensities:



- $\alpha_{\cdot}^{\mathcal{Q}}(0,0), \alpha_{\cdot}^{\mathcal{Q}}(1,0), \alpha_{\cdot}^{\mathcal{Q}}(1,1), \alpha_{\cdot}^{\mathcal{Q}}(2,0), \alpha_{\cdot}^{\mathcal{Q}}(2,1), \dots \text{ can be easily calibrated}$
- on marginal distributions of N(t) by forward induction.

- *n* obligors
- Default times:  $\tau_1, \ldots, \tau_n$ 
  - $(\Omega, A, P)$  Probability space
- Default indicator processes:  $N_i(t) = 1_{\{\tau_i \le t\}}, i = 1, ..., n$
- $H_{i,t} = \sigma(N_i(s), s \le t), i = 1, ..., n; H_t = \bigvee_{i=1}^n H_{i,t}$ 
  - Natural filtration of default times
  - Ordered default times:  $\tau^1, \ldots, \tau^n$
  - No simultaneous defaults:  $\tau^1 < \ldots < \tau^n, P-a.s.$

• 
$$\alpha_1^P, \dots, \alpha_n^P$$
  $(P, H_t)$  intensities  
-  $t \to N_i(t) - \int_0^t \alpha_i^P(s) ds$   $(P, H_t)$  martingales

- Instantaneous digital CDS
  - Traded at t
  - Stylized cash-flow at *t*+d*t*:

 $dN_i(t) - \alpha_i(t)dt$ premium default payment payment

- Default free interest rate: *r*
- Payoffs of self-financed strategies:

$$\underbrace{V_{0}}_{initial} e^{rT} + \sum_{i=1}^{n} \int_{0}^{T} \underbrace{\delta_{i}(s)}_{holdings} e^{r(T-s)} \left( dN_{i}(s) - \alpha_{i}(s) ds \right)$$

-  $\delta_1(\bullet), \dots, \delta_n(\bullet)$   $H_t$  - predictable processes

- Absence of arbitrage opportunities:  $\left\{ \begin{array}{l} \alpha_{i}(t) > 0 \\ \sum_{CDS} \end{array} \right\} = \left\{ \begin{array}{l} \alpha_{i}^{P}(t) > 0 \\ \sum_{(P-H_{t})} \end{array} \right\}$
- As a consequence:  $\exists ! Q \sim P$ , - such that  $\alpha_1, \dots, \alpha_n$  are the  $(Q, H_t)$  intensities of default times
- $M: H_T$  measurable, Q –integrable payoff
- Integral representation theorem of point processes (Brémaud)

$$M = E^{Q}[M] + \sum_{i=1}^{n} \int_{0}^{T} \underbrace{\theta_{i}(s)}_{\substack{H_{s}-\\ \text{predictable}}} \left( \underbrace{\frac{dN_{i}(s) - \alpha_{i}(s)ds}_{\text{CDS}}}_{\text{cash-flow}} \right)$$

- Integral representation theorem implies completeness of the credit market
  - Perfect replication of claims which depend only upon the default history
    - With CDS on underlying names and default-free assetCDO tranches
  - -Q: unique martingale measure
  - Replication price of *M* at time *t*:  $V_t = E^Q \left[ M e^{-r(T-t)} | H_t \right]$
  - Note that the holdings of CDS only depend upon default history

Credit spread risk is not taken into account

• Need of additional assumptions to effectively compute dynamic hedging strategies:

 $\begin{cases} \alpha_i(t) = \alpha(t, N(t)), \ i = 1, ..., n \\ N(t) = \sum_{i=1}^n N_i(t), \text{ number of defaults at time } t \end{cases}$ 

- CDS spreads only depend upon the current credit status
   ➤Markov property
- CDS spreads only depend on the number of defaults
   ➢Mean-field
- All names have the same short-term credit spread
   Homogeneity

- $N(t) = \sum_{i=1}^{n} 1_{\{\tau_i \le t\}}$  number of default process
- is a continuous time *Q* Markov chain
  - $-\lambda(t,0)$   $\lambda(t,0)$  0 0 0 0 0  $0 \qquad -\lambda(t,1) \quad \lambda(t,1) \quad 0$ - Pure death process 0 () - Generator of the Chain  $\Lambda(t) =$ 0 0 0 0  $-\lambda(t,n-1)$   $\lambda(t,n-1)$ 0 0 0 0 0 0 0
    - λ(t, N(t)) is the intensity of the pure jump process N(t)
       ≻ is also the aggregate loss intensity

$$\lambda(t, N(t)) = (\underline{n - N(t)}) \times \alpha(t, N(t))$$

number of non-defaulted names individual pre-default intensity

- Replication price for a CDO tranche  $V_t = V_{CDO}(t, N(t))$
- Only depends on the number of defaults
  - And of the individual characteristics of the tranche

Seniority, maturity, features of premium payments

- Thanks to the "homogeneity" between names:
  - All hedge ratios with respect to individual CDS are equal
  - Only hedge with the CDS index + risk-free asset
- Replicating hedge ratio:

$$\delta(t, N(t)) = \frac{V_{CDO}(t, N(t) + 1) - V_{CDO}(t, N(t))}{V_{CDS \text{ Index}}(t, N(t) + 1) - V_{CDS \text{ Index}}(t, N(t))}$$

#### • Calibration of loss intensities

- From marginal distributions of aggregate losses
- Or onto CDO tranche quotes
- Use of forward Kolmogorov equations
  - ➢ For the Markov chain
- Easy to solve for a pure death process
- Loss intensities with respect to the number of defaults
  - For simplicity, assumption of time homogeneous intensities
  - Increase in intensities: contagion effects
  - Compare flat and steep base correlation structures

| 3% | 6% 9%  | / <u> </u> | 6 <mark>22</mark> % |
|----|--------|------------|---------------------|
| 3% | 28% 36 | % 42%      | 6 58%               |

Table 8. Base correlations with respect to attachment points.

Number of names: 125 Default-free rate: 4% 5Y credit spreads: 20 bps Recovery rate: 40%

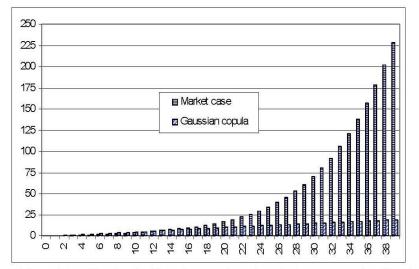


Figure 6. Loss intensities for the Gaussian copula and market case examples. Number of defaults on the x-axis.

• Dynamics of the credit default swap index in the Markov chain

| Nb Defaults - |      | We   | eks  |      |
|---------------|------|------|------|------|
| ND Delauits - | 0    | 14   | 56   | 84   |
| 0             | 20   | 19   | 17   | 16   |
| 1             | 0    | 31   | 23   | 20   |
| 2             | 0    | 95   | 57   | 43   |
| 3             | 0    | 269  | 150  | 98   |
| 4             | 0    | 592  | 361  | 228  |
| 5             | 0    | 1022 | 723  | 490  |
| 6             | 0 14 |      | 1193 | 905  |
| 7             | 0    | 1870 | 1680 | 1420 |
| 8             | 0    | 2243 | 2126 | 1945 |
| 9             | 0    | 2623 | 2534 | 2423 |
| 10            | 0    | 3035 | 2939 | 2859 |

Table 9. Dynamics of credit default swap index spread  $s_{is}(i,k)$  in basis points per annum.

- The first default leads to a jump from 19 bps to 31 bps
- The second default is associated with a jump from 31 bps to 95 bps
- Explosive behavior associated with upward base correlation curve

- What about the credit deltas?
  - In a homogeneous framework, deltas with respect to CDS are all the same
  - Perfect dynamic replication of a CDO tranche with a credit default swap index and the default-free asset
  - Credit delta with respect to the credit default swap index
  - = change in PV of the tranche / change in PV of the CDS index

| Nb Defaults | OutStanding | Weeks |       |       |       |
|-------------|-------------|-------|-------|-------|-------|
| ND Deraults | Nominal     | 0     | 14    | 56    | 84    |
| 0           | 3.00%       | 0.541 | 0.617 | 0.823 | 0.910 |
| 1           | 2.52%       | 0     | 0.279 | 0.510 | 0.690 |
| 2           | 2.04%       | 0     | 0.072 | 0.166 | 0.304 |
| 3           | 1.56%       | 0     | 0.016 | 0.034 | 0.072 |
| 4           | 1.08%       | 0     | 0.004 | 0.006 | 0.012 |
| 5           | 0.60%       | 0     | 0.002 | 0.002 | 0.002 |
| 6           | 0.12%       | 0     | 0.001 | 0.000 | 0.000 |
| 7           | 0.00%       | 0     | 0     | 0     | 0     |

Table 11. Delta of the default leg of the [0,3%] equity tranche with respect to the credit default swap index ( $\delta_d(i,k)$ ).

# • Dynamics of credit deltas:

| Nb Defaults | OutStanding | Weeks |       |       |       |
|-------------|-------------|-------|-------|-------|-------|
| ND Defaults | Nominal     | 0     | 14    | 56    | 84    |
| 0           | 3.00%       | 0.541 | 0.617 | 0.823 | 0.910 |
| 1           | 2.52%       | 0     | 0.279 | 0.510 | 0.690 |
| 2           | 2.04%       | 0     | 0.072 | 0.166 | 0.304 |
| 3           | 1.56%       | 0     | 0.016 | 0.034 | 0.072 |
| 4           | 1.08%       | 0     | 0.004 | 0.006 | 0.012 |
| 5           | 0.60%       | 0     | 0.002 | 0.002 | 0.002 |
| 6           | 0.12%       | 0     | 0.001 | 0.000 | 0.000 |
| 7           | 0.00%       | 0     | 0     | 0     | 0     |

Table 11. Delta of the default leg of the [0,3%] equity tranche with respect to the credit default swap index ( $\delta_d(i,k)$ ).

- Deltas are between 0 and 1
- Gradually decrease with the number of defaults
   Concave payoff, negative gammas
- When the number of defaults is > 6, the tranche is exhausted
- Credit deltas increase with time

≻Consistent with a decrease in time value

- Market and theoretical deltas at inception
  - Market deltas computed under the Gaussian copula model
    - Base correlation is unchanged when shifting spreads
    - ➤ "Sticky strike" rule
    - > Standard way of computing CDS index hedges in trading desks

|               | [0-3%] | [3-6%] | [6-9%] | [9-12%] | [12-22%] |
|---------------|--------|--------|--------|---------|----------|
| market deltas | 27     | 4.5    | 1.25   | 0.6     | 0.25     |
| model deltas  | 21.5   | 4.63   | 1.63   | 0.9     | NA       |

Smaller equity tranche deltas for in the Markov chain model
 How can we explain this?

- Smaller equity tranche deltas in the Markov chain model
  - Default is associated with an increase in dependence

➤Contagion effects

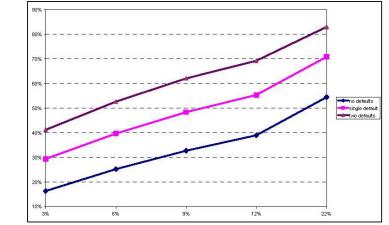


Figure 8. Dynamics of the base correlation curve with respect to the number of defaults. Detachment points on the x-axis. Base correlations on the y-axis.

- Increasing correlation leads to a decrease in the PV of the equity tranche
  - Sticky implied tree deltas
- Recent market shifts go in favour of the contagion model

- The current crisis is associated with joint upward shifts in credit spreads
  - Systemic risk
- And an increase in base correlations

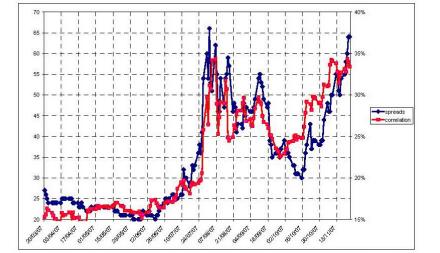


Figure 9. Credit spreads on the five years iTraxx index (Series 7) in bps on the left axis. Implied correlation on the equity tranche on the right axis

- Sticky implied tree deltas are well suited in regimes of fear
  - Derman: "regimes of volatility" (1999)

- We have experienced three defaults on the CDX NA IG so far
- Good material to look further for contagion effects
  - Financials are subject to systemic and thus contagion risk
  - However, the model does not account for feedback effects to central banks looking for stabilizing the financial system
  - The model translates a base correlation skew into contagion effects
  - Does not provide more than market expectations in CDO tranches
  - For example, large spreads in senior and super senior tranches are a sign of fear of a global market turndown

- Comparing with results provided by:
  - Arnsdorf and Halperin "BSLP: Markovian Bivariate Spread-Loss Model for Portfolio Credit Derivatives" Working Paper, JP Morgan (2007), Figure 7

|               | [0-3%] | [3-6 %] | [6-9%] | [9-12%] | [12-22%] |
|---------------|--------|---------|--------|---------|----------|
| market deltas | 26.5   | 4.5     | 1.25   | 0.65    | 0.25     |
| model deltas  | 21.9   | 4.81    | 1.64   | 0.79    | 0.38     |

- Computed in March 2007 on the iTraxx tranches
- Two dimensional Markov chain, shift in credit spreads

|               | [0-3%] | [3-6%] | [6-9%] | [9-12%] | [12-22%] |
|---------------|--------|--------|--------|---------|----------|
| market deltas | 27     | 4.5    | 1.25   | 0.6     | 0.25     |
| model deltas  | 21.5   | 4.63   | 1.63   | 0.9     | 0.6      |

- Note that our results, related to default deltas, are quite similar

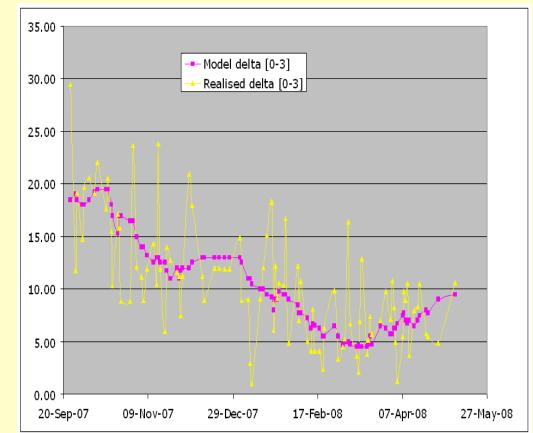
Equity tranche deltas are smaller in contagion models than Gaussian copula credit deltas

- Cont and Kan: "Dynamic hedging of portfolio credit derivatives" (2008)
- Spread deltas
  - Gaussian copula model
  - Local intensity corresponds to our contagion model
  - BSLP corresponds to Arnsdorf and Halperin (2007)
  - GPL: generalized Poisson loss model of Brigo *et al.* (2006)
- This shows some kind of robustness
- Picture becomes more complicated when considering other hedging criteria...

| Tranche  | Gauss | Local | BSLP  | GPL   |
|----------|-------|-------|-------|-------|
| 0 - 3    | 24.48 | 24.52 | 24.79 | 24.48 |
| 3 - 6    | 5.54  | 5.45  | 5.30  | 5.54  |
| 6 - 9    | 1.79  | 1.80  | 1.80  | 1.79  |
| 9 - 12   | 0.87  | 0.85  | 0.88  | 0.87  |
| 12 - 22  | 0.35  | 0.35  | 0.32  | 0.35  |
| 22 - 100 | 0.08  | 0.08  | 0.09  | 0.08  |

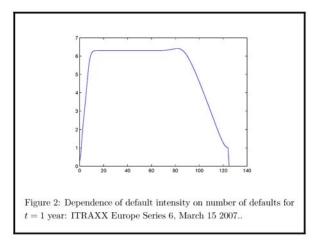
#### <u>Spread deltas</u> computed for 5Y Europe iTraxx on 20 September 2006

- Back-test study on iTraxx Series 8 equity tranche
- Comparison of realized spread deltas on the equity tranche and model (implied tree) deltas
- Good hedging performance compared with the Gaussian copula model
  - During the credit crisis
  - Discrepancy with results of Cont and Kan (2008)?

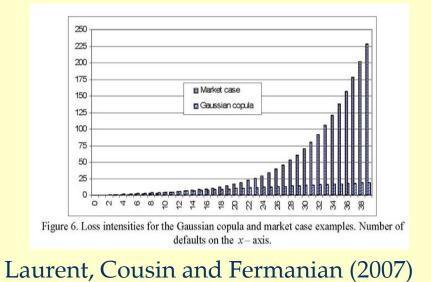


Source: S. Amraoui BNP Paribas

- Cont and Kan (2008) show rather poor performance of "jump to default" deltas
  - Even in the recent crisis period
- However, unsurprisingly, the credit deltas ("jump to default") seem to be rather sensitive to the calibration of contagion parameters on quoted CDO tranches
- Right pictures represent aggregate loss intensities
  - Huge contagion effects for the first six defaults in Cont *et al.* (2008)
  - Much smaller contagion effects for the first defaults in Laurent *et al.* (2007)



#### Cont, Minca and Savescu (2008)



• Frey and Backhaus: "Dynamic hedging of synthetic CDO tranches with spread risk and default contagion" (2007)

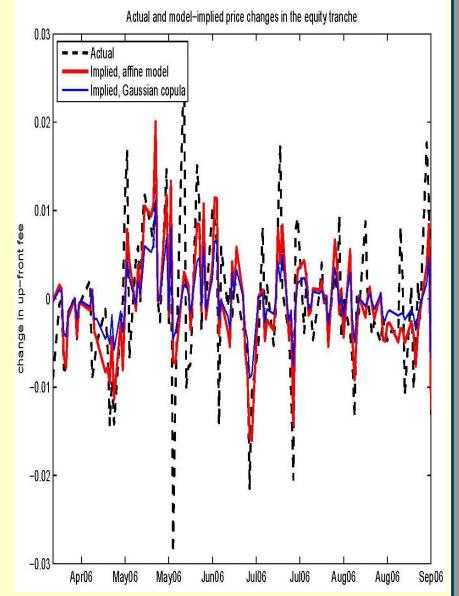
| Tranche                       | [0,3]           | [3,6]           | [6,9]           | [9,12]           | [12,22]          |
|-------------------------------|-----------------|-----------------|-----------------|------------------|------------------|
| Spread<br>Tranche Correlation | 26 %<br>17.30 % | 84 bp<br>3.22 % | 24 bp<br>9.93 % | 14 bp<br>15.81 % | 11 bp<br>27.46 % |
| Gauss Cop. $\Delta$           | 0.61            | 0.23            | 0.06            | 0.03             | 0.07             |

#### VOD: Value on default

|          | VOD in the Markov model | VOD in the Copula model |
|----------|-------------------------|-------------------------|
| [0, 3]   | 0.344                   | 1.002                   |
| [3, 6]   | 0.138                   | 0.171                   |
| [6, 9]   | 0.058                   | 0.023                   |
| [9, 12]  | 0.039                   | 0.008                   |
| [12, 22] | 0.107                   | 0.010                   |

Much smaller deltas in the contagion model than in Gaussian copula model

- Laurent: "A note on the risk management of CDO" (2007)
  - provides a theoretical framework for hedging credit spread risk only while default risk is diversified at the portfolio level
  - no default contagion, correlation between defaults are related to "correlation" between credit spreads
- Feldhütter: "An empirical investigation of an intensity-based model for pricing CDO tranches" (2008)
  - comparison of hedging performance of a Duffie and Garleanu (2001) reduced-form model and one factor Gaussian copula
  - Use of information at time *t*+1 to compute hedge ratios at time *t*
  - Higher deltas for the equity tranche in the affine model compared with the 1F Gaussian copula (market deltas)



 Consistent results with the affine model of Eckner (2007) based on December 2005 CDX data

| Tranches               | [0-3%] | [3-7%] | [7-10%] | [10-15%] | [15-30%] |
|------------------------|--------|--------|---------|----------|----------|
| market deltas          | 18.5   | 5.5    | 1.5     | 0.8      | 0.4      |
| AJD deltas             | 21.7   | 6.0    | 1.1     | 0.4      | 0.1      |
| contagion model deltas | 17.9   | 6.3    | 2.5     | 1.3      | 0.8      |

- Market deltas, "intensity" model credit deltas in Eckner (2007) and contagion model deltas
- Goes into the opposite direction when comparing with the contagion model
- Note that Feldhütter (2008) and Eckner (2007) are pre-crisis
- And are according to a "sticky delta rule" (Derman) which is reflects irrational exuberance or greed
  - And might be appropriate for the pre-crisis period

# Conclusion

- Main theoretical features of the complete market model
  - No simultaneous defaults
    - Unlike multivariate Poisson models
  - Credit spreads are driven by defaults
    - Contagion model
      - Jumps in credit spreads at default times
    - Credit spreads are deterministic between two defaults
  - Bottom-up approach
    - Aggregate loss intensity is derived from individual loss intensities
  - Correlation dynamics is also driven by defaults
     Defaults lead to an increase in dependence

# Conclusion

- What did we learn from the previous approaches?
  - Thanks to stringent assumptions:
    - credit spreads driven by defaults
    - homogeneity
    - Markov property
  - It is possible to compute a dynamic hedging strategy
    - Based on the CDS index
  - That fully replicates the CDO tranche payoffs
    - Model matches market quotes of liquid tranches
    - Very simple implementation
    - Credit deltas are easy to understand
  - Improve the computation of default hedges
    - Since it takes into account credit contagion
    - Provide some meaningful results in the current credit crisis

#### Additional selected references

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