



CDO modelling from a practitioner's point of view: What are the real problems?

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Bridging between academia and practice

- The speaker
- iTraxx, standard CDOs and conventions
- Gaussian copula model
 - CDO behaviour
 - Correlation smile
 - Compound \leftrightarrow base correlations
 - Some base correlation issues
- What is a good model?
 - Interpolation and extrapolation, non standard tranches/portfolios
 - Market information
 - Hedge ratios
- Implementation considerations
 - MC strategies, how to simulate
 - Risk numbers for all market data
 - Fast recursive techniques, conditional independence
 - Other model proposals
- What makes a good practitioner?
- Conclusion

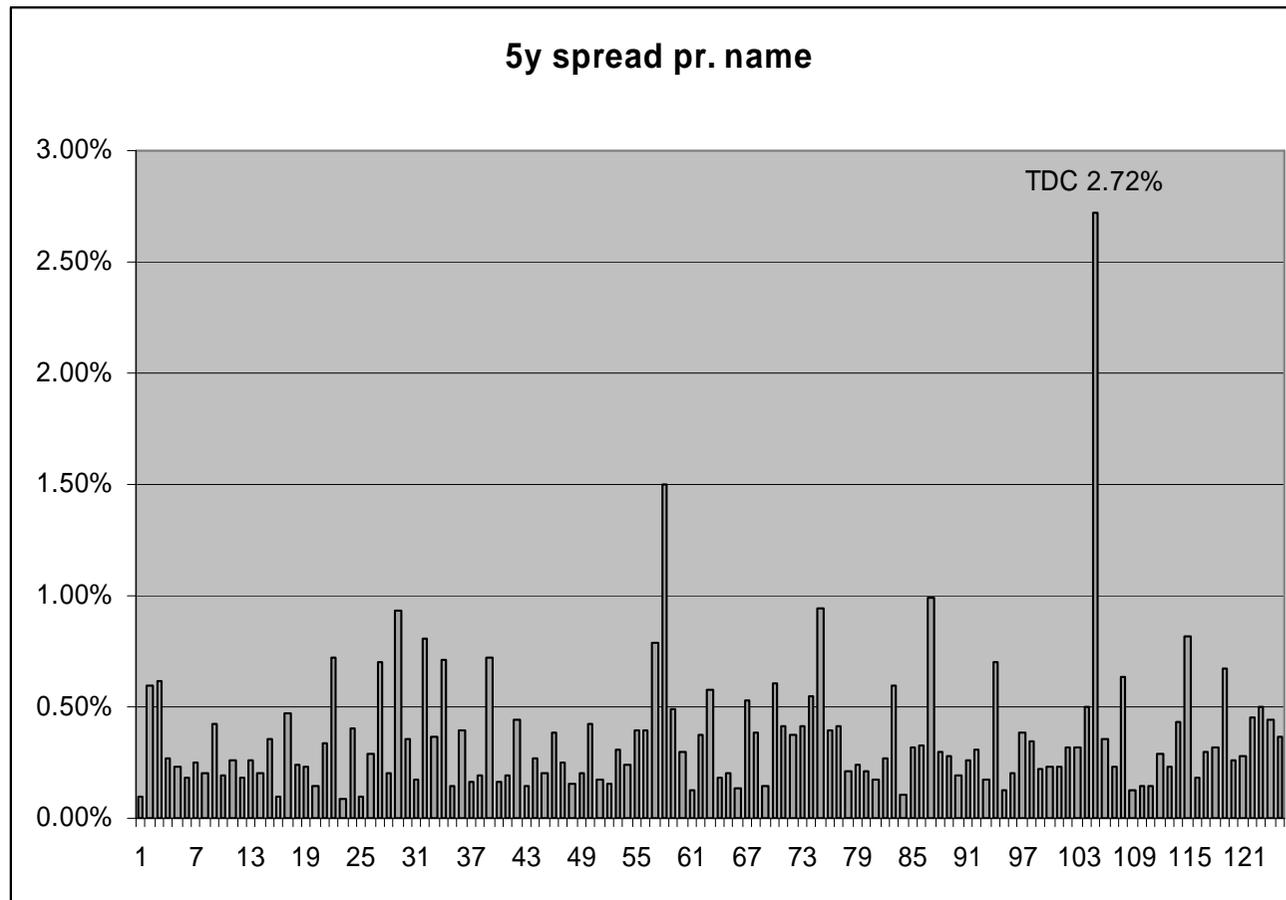
Jens Lund

- **Head of Product Development, Nordea Markets**
- **Background:**
 - Nov 1996: M.Sc. in statistics, University of Copenhagen
 - Feb 2000: Ph.D. in statistics, The Royal Veterinary and Agricultural University
 - Mar 2000 onwards: with Nordea, Product Development
 - Has done a lot of the credit modelling work in Nordea
- **Team:**
 - 5 members, various degrees of experience, mainly Ph.D. in natural science, looking for more people
 - 2 associated programmers helping with interface to trading system
 - Responsible for all derivatives modelling and calculations (NPV, risk,...)
 - Scripting language for description of all derivatives
 - Interest rates, credit derivatives, inflation, equity, ...

iTraxx standard portfolio/CDS

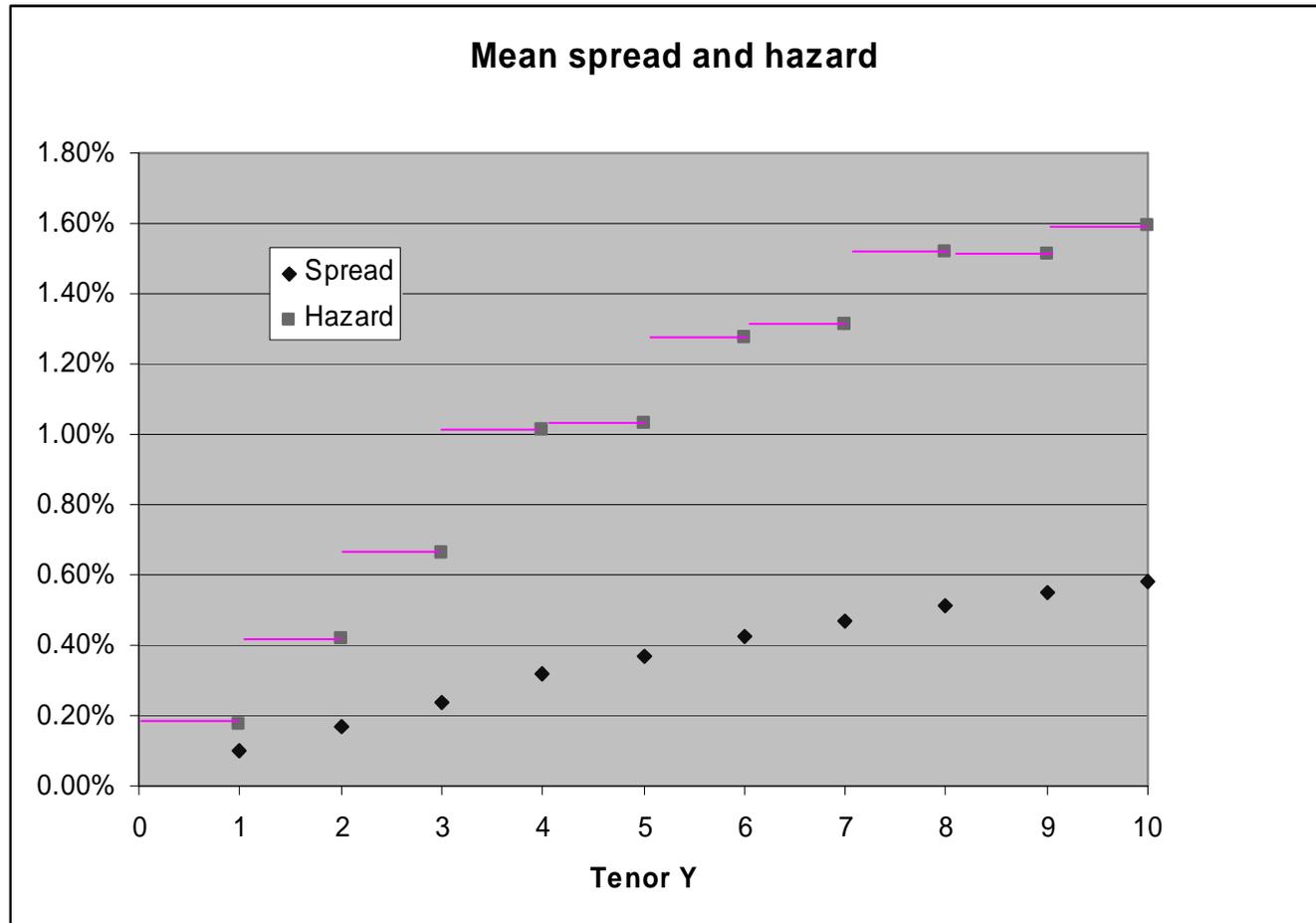
- **iTraxx Europe**
 - 125 liquid names
 - Underlying index CDSes for sectors
 - 5Y, 7Y & 10Y maturity
 - 5 standard CDO tranches, first to default baskets, options
 - US index CDX
- **3m, act/360, last 20 date roll, CDS pay accrued fee**
- **Index composition adjusted every 6m**
- **Index CDS trades at a fixed spread with accrued fee ---**
Traded with upfront premium (but quoted on spread)
 - Together with last 20 date roll this ensures liquidity and (minus counterparty risk) perfect netting of trades.

iTraxx, distribution of 5y spreads

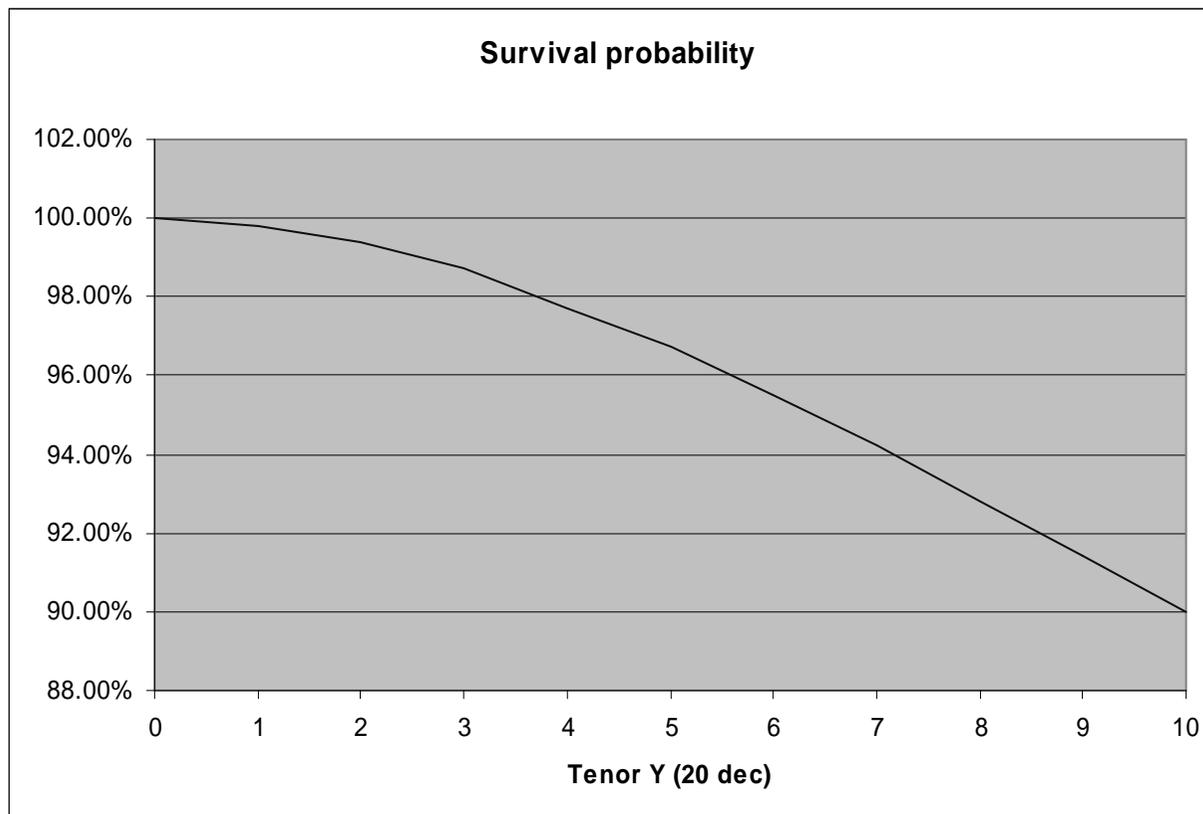


TDC 2 Nov 2005: 293bp -> 268bp -> 290bp, Mid March at approx 270bp

iTraxx average spreads, 5y mean = 37bp

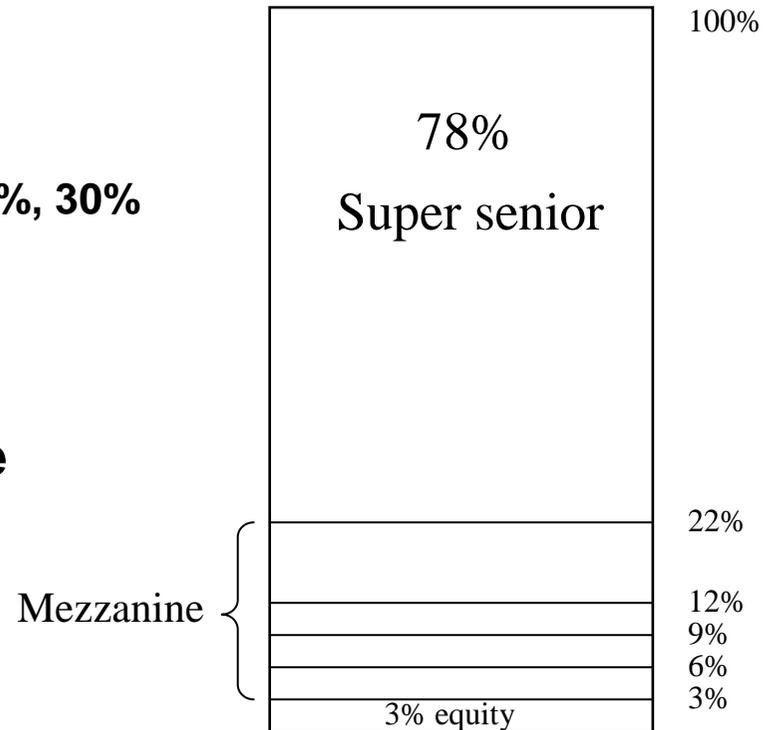


Average market implied survival probability



Standardized CDO tranches

- **iTraxx Europe 3%, 6%, 9%, 12%, 22%**
 - US index CDX has points 3%, 7%, 10%, 15%, 30%
- **Has done a lot to provide liquidity in structured credit**
- **Reliable pricing information available**
- **Quotation:**
 - bp running fee
 - Equity tranche:
 - 500bp running, quoted on upfront payment!**
 - Due to timing risk of events**



Reference Gaussian copula model

- **N credit names, $i = 1, \dots, N$**
- **Default times:** $T_i \sim F_i(t) = 1 - \exp\left(-\int_0^t \lambda_i(u) du\right)$
- **λ_i curves bootstrapped from CDS quotes**
- **T_i correlated through the copula:**

$$F_i(T_i) = \Phi(X_i) \text{ with } X = (X_1, \dots, X_N)^t \sim N(0, \Sigma)$$

Σ correlation matrix, variance 1, constant correlation ρ

Could take

$$X_i = \sqrt{\rho}M + \sqrt{1-\rho}Z_i \quad \Sigma = \begin{pmatrix} 1 & & & \\ & \ddots & \rho & \\ & \rho & \ddots & \\ & & & 1 \end{pmatrix}$$

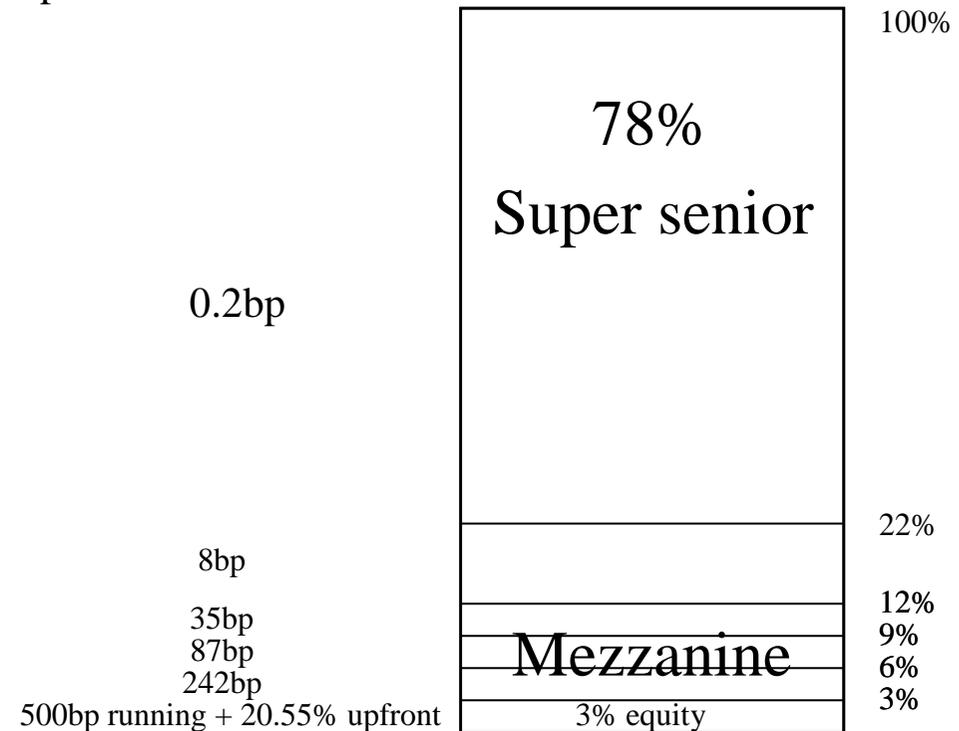
- **In model: correlation independent of product to be priced**

CDO behaviour

- **Structure:**

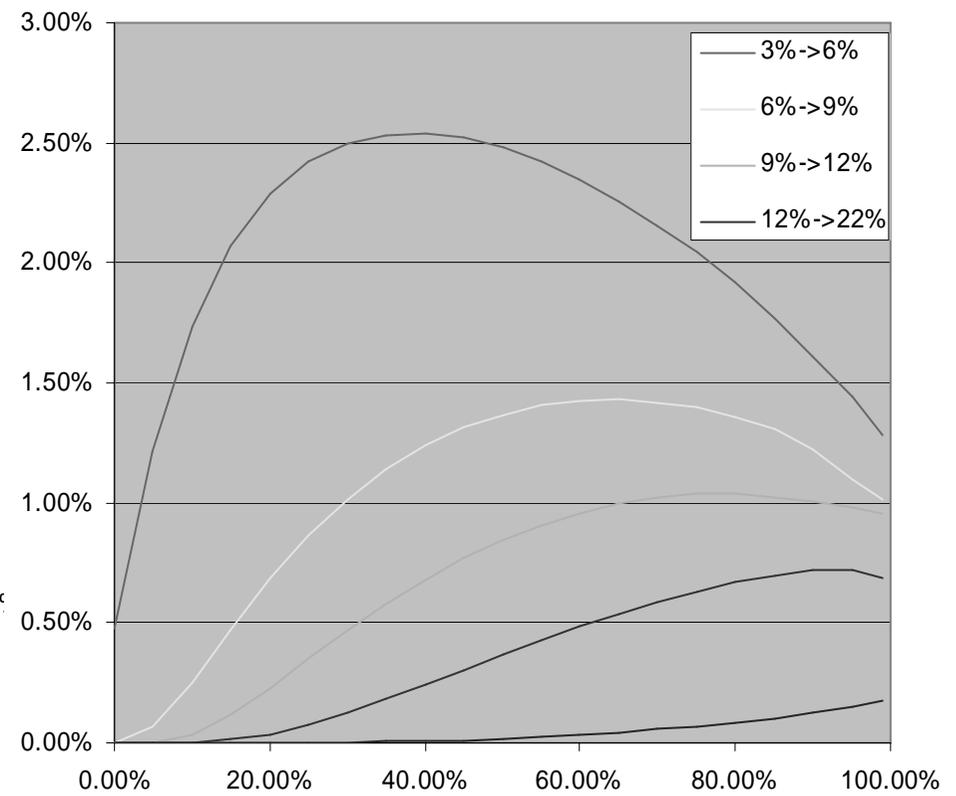
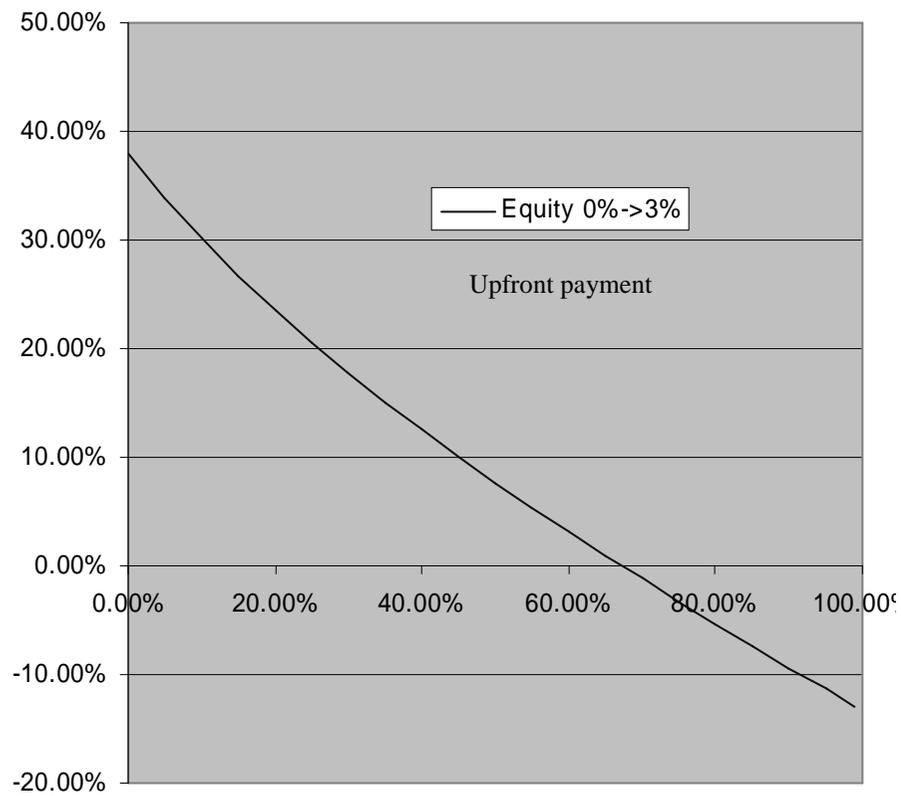
- 125 name, iTraxx
- RR almost all 40%
- Avg CDS = 37bp
- Corr = 25%
- Start 11-oct-2005
- 5y structure, ends 20-dec-2010
- Premium: 3m, act/360
- Valuation 10-oct-2005

Spreads with corr = 25%



Correlation dependence

Fair spreads as function of correlation



CDO behaviour depends on

- **Number of names**
- **Spreads of the underlying names**
- **Tranching:**
 - **Size of tranche**
 - **Smaller tranches are more leverage/exposed to changes**
 - **Order of tranche**
- **Correlation**
- **Recovery rate**

Prices in the market have a correlation smile

- In practice:

Correlation depends on product, 10-oct-2005, 5Y iTraxx Europe

- ◆ Tranche
- ◆ Maturity
- ◆ Fair coupons

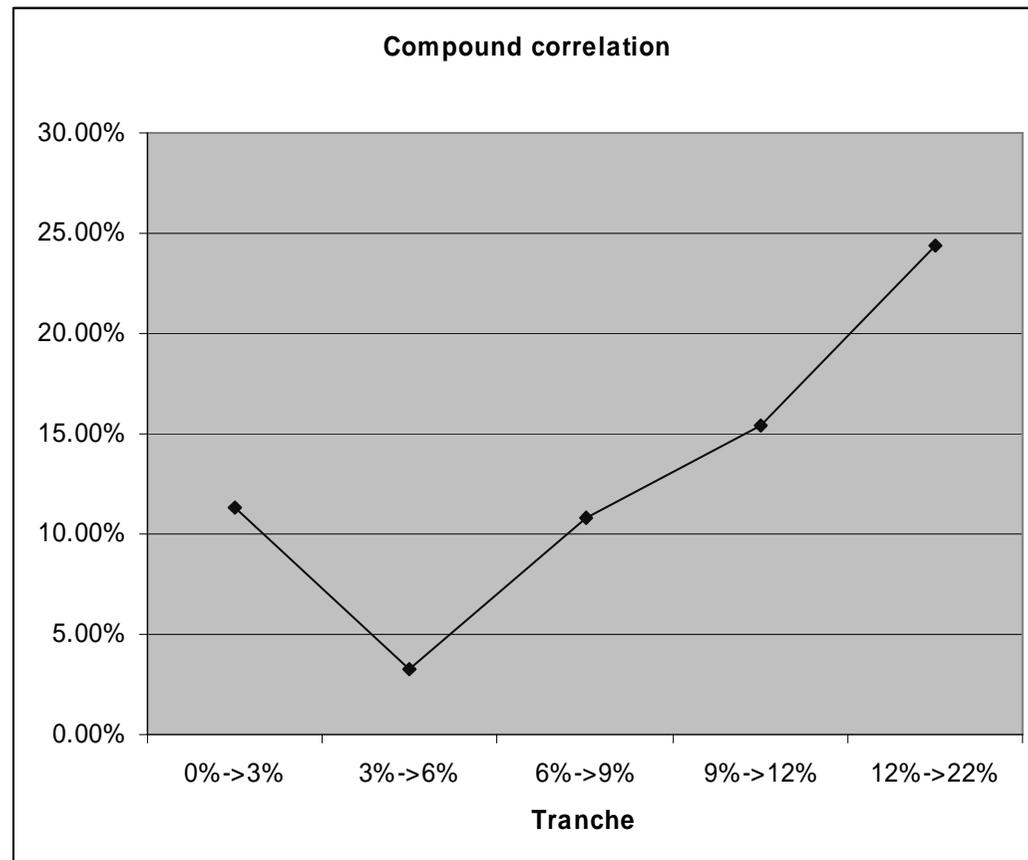
Equity upfront: 29.2%

3-6%: 0.97%

6-9%: 0.28%

9-12% 0.13%

12-22%: 0.07%

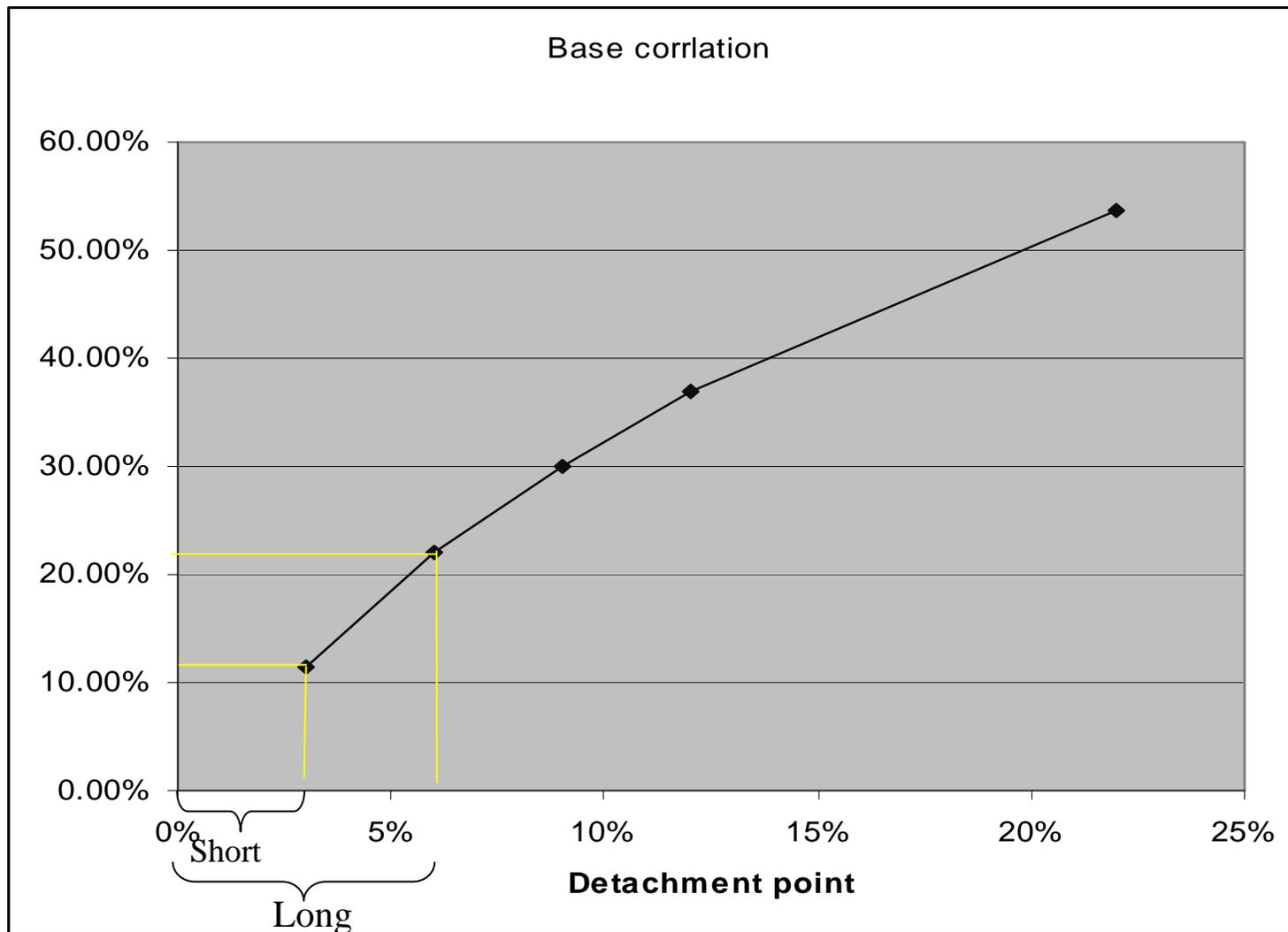


Compound correlations

- **The correlation on the individual tranches**
- **Mezzanine tranches have low correlation sensitivity and even non-unique or non-existent correlation for given spreads!**
- **No way to extend to, say, 2%-5% tranche or bespoke tranches**

- **What alternatives exists?**

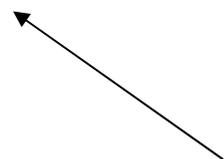
Base correlations



Steep base correlation curves

Base smile	Corr	Fair coupons
3%	5%	33.94%
6%	30%	-0.65%
9%	45%	-0.13%
12%	50%	0.41%
22%	55%	0.31%

Negative spreads!!



- **Base correlations depends on previous points**
- **Somewhat contradicting the whole idea of base tranches!**

Are base correlations a real solution?

- **No, it is merely a convenient way of describing prices on CDO tranches**
- **An intermediate step towards better models that exhibit a smile**
- **No general extension to other products**
- **No smile dynamics**
- **Interpolation issues**
- **Correlation smile modelling, versus**
- **Models with a smile and correlation dynamics**
- **Base correlation is NOT a model!!**
- **Nevertheless: they are used a lot!**

Why have models? How to use them?

- **We do see prices on the standard tranches in the market, so why have a model at all?**
- **Interpolation**
 - Non standard tranches, e.g. 2%-4%
- **Extrapolation**
 - Attachment/detachment points below 3%
 - Bespoke portfolios
 - Other products: CDS -> CDO, CDO -> CDO², etc.
- **Usually: map expected losses to find corr for other tranches**
- **Risk numbers**

Delta hedges

- **Delta risk: how much does the NPV change when the underlying credit spreads widen by 1bp?**
- **CDO tranches typical traded with initial credit hedge, i.e. only correlation risk left!**
- **Conveniently quoted as amount of underlying index CDS to buy in order to hedge credit risk, i.e. $\text{deltaCDO}/\text{deltaCDS}$**
- **Split out on individual names or just consider index?**
- **Base correlation: find by long/short strategy in the same way as NPV!**

Deltas in different models

- **Deltas differ between models:**

Tranche	0%-3%	3%-6%	6%-9%	9%-12%	12%-22%
Compound corr	22.1	9.1	2.7	1.2	0.6
Base corr	22.1	6.1	2.0	0.9	0.5
RFL	25.9	7.5	0.4	0.1	0.1

- **Agreement on delta amounts requires model agreement**
- **Non-unique deltas when spread & correlation is connected, i.e. in models with smile dynamics**

Last 20 schedule and date roll convention

- End date will be 20th of Mar, Jun, Sep or Dec.
- If we have passed any of these dates we roll to the next date, so e.g. end date 21st Jun will roll to 20st Sep, etc.
- First period will be long if we would otherwise get less than 1m to first date in scheudle!
- Stub/long period in the beginning.
- Intermediate points are rolled Following.
- Usually in the credit market start and end dates can fall on non-business days.
- Always start protection the day after the trade day, even if a non business day.

Risk ladders

- **CDS curve most often bootstrapped from yearly quotes**

- Risk on the yearly quotes, 1Y, 2Y, 3Y, 4Y, 5Y, 6Y, etc.

- **However: trades end every quarter**

- **Risk might move around when crossing 20 Mar, Jun, Sep, Dec**

- **Example:**

Date	4Y	4.25Y	4.50Y	4.75Y	5Y
QTR before roll			200		
Y before roll	100				100
QTR after roll		200			
Y after roll	150				50

- **Get risk on a quarterly ladder, even though the curve is still bootstrapped from yearly quotes. Be aware how your risk changes on rolls.**

Implementation strategies

- **Key: efficiency, flexibility and fast + accurate risk!**
- **Copula type models:**
 - Monte Carlo
 - Recursive/FFT techniques

Implementation of Gaussian copula by MC

- **Monte Carlo simulation of $X \sim N(0, \Sigma)$**
 - Simulate $Y \sim N(0, I)$
 - Find A such that $AA' = \Sigma$
 - $X = AY$
- **How to find A ?**
 - Cholesky decomposition
 - Eigenvalue decomposition: $A = P \text{sqrt}(\lambda)$
 - The latter is better, in particular with Sobol sequences
- **Simulation:**
 - Simulate default time $T_i = F_i^{-1}(\Phi(X_i))$ for all names, and price.
 - Do it, say, 10000 times.
 - Can price any derivative, simple.

Risk numbers in MC pricing

- “Naive” risk numbers: $\frac{\partial V}{\partial \lambda_i} = \frac{V(\lambda_i + \varepsilon) - V(\lambda_i)}{\varepsilon}$
- For credit risk we can exchange differentiation and integration:

$$V = E[g(\tau)] = \int g(\tau) f(\tau | \lambda_1, \dots, \lambda_N) d\tau$$

$$\begin{aligned} \frac{\partial}{\partial \varepsilon} V(\lambda_1, \dots, \lambda_i + \varepsilon s, \dots, \lambda_N) |_{\varepsilon=0} &= \int g(\tau) \frac{\partial}{\partial \varepsilon} f(\tau | \lambda_1, \dots, \lambda_i + \varepsilon s, \dots, \lambda_N) |_{\varepsilon=0} d\tau \\ &= \int g(\tau) \frac{\partial}{\partial \varepsilon} \log f(\tau | \lambda_1, \dots, \lambda_i + \varepsilon s, \dots, \lambda_N) |_{\varepsilon=0} f(\tau) d\tau \\ &= E[g(\tau) \frac{\partial}{\partial \varepsilon} \log f(\tau | \lambda_1, \dots, \lambda_i + \varepsilon s, \dots, \lambda_N) |_{\varepsilon=0}] \end{aligned}$$

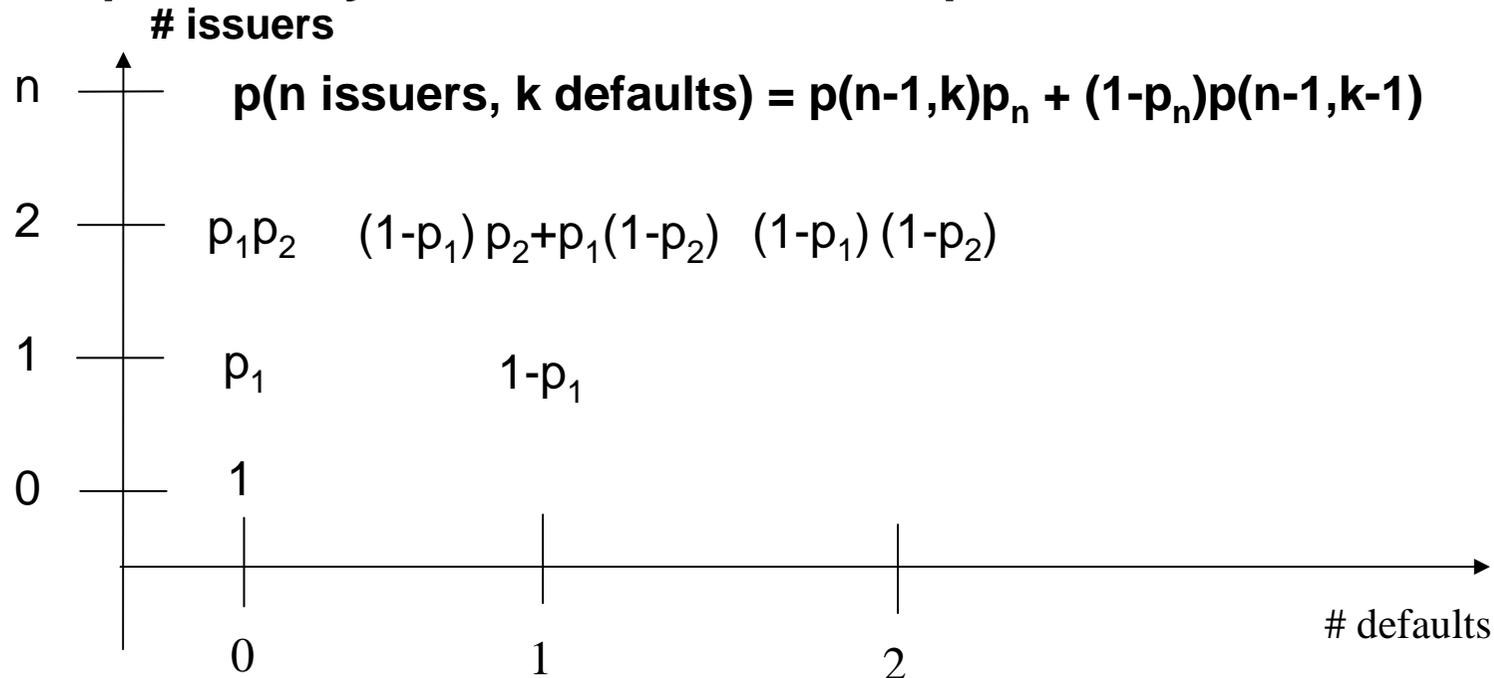
- Calculate derivative and price in same simulation, but for a different payout function
- Speed up of factor $5 \times 125 = 625$. Also improves stability!

Recursive/FFT implementation

- Write $X \sim N(0, \Sigma)$ as $X_i = aM + \sqrt{1-a^2}Z_i$ $a = \sqrt{\rho}$
- Only viable for derivatives that depend on the number of defaults or the cumulative loss (perhaps discretised if RR or notionals are not equal)
- Conditional on M:
 - X_i 's are independent \Rightarrow
 For given horizon T, the default indicators $1_{(X_i \leq C_i(T))}$ are independent
 - Calculate distribution of number of defaults recursively in $N = \text{\#names}$
 - Binomial expression
- Find loss distribution by integrating over M
- Fast and no MC noise

Recursive build-up of loss distribution

- Conditional on M , given a time horizon t : independence and p_i is the probability name i has survived up to time t .



- Next: integrate over M

What is the survival probability?

- Let $X_i = aM + \sqrt{1-a^2}Z_i$, with $X_i \sim H_i$
- The model matches quantiles: $F_i(T_i) = H_i(X_i)$
- This means the conditional survival probabilities are:

$$\begin{aligned}
 P(T_i \geq t \mid M) &= P(F_i^{-1}(H_i(X_i)) \geq t \mid M) \\
 &= P(X_i \geq H_i^{-1}(F_i(t)) \mid M) \\
 &= P\left(Z_i \geq \frac{H_i^{-1}(F_i(t)) - aM}{\sqrt{1-a^2}} \mid M\right)
 \end{aligned}$$

The search for better copulas has started...

- **“Better” means**

- describing the observed prices in the market for iTraxx
- produces a correlation smile
- has a reasonable low number of parameters
one can have a view on and interpret
- has a plausible dynamics for the correlation smile
- constant parameters can be used on a range of
 - tranches / products
 - maturities
 - (portfolios)
- Efficient pricing and risk numbers

- **Often start from Gaussian model described as a 1 factor model**

- Computational efficiency!

$$X_i = aM + \sqrt{1-a^2}Z_i$$

What makes a good practitioner?

- **Has a lot of common sense**
 - Understands the difference between up and down, elbow and head...
- **Understands products and markets**
- **IT knowledge**
 - Excel, Visual Basic, system functionality, ..., C++
 - Make things operational, streamline repetitive processes, can implement the math so it works, etc.
- **Mathematical skills**
 - Required to develop models and make efficient implementation
 - Numerical analysis, analysis, algebra, stochastic models, etc.

Conclusion

- **Models will have to be developed further**
 - Smile description and dynamics
 - Delta amounts and other relevant risk numbers
 - Bespoke tranches
 - Computational efficiency
 - Will have to go through a couple of iterations
- **Market and products are changing over time**
- **Practical challenges**
 - Manage market data & information
 - Provide smooth infrastructure for all the numbers/trades/etc.