



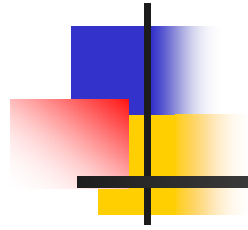
*A comparative analysis of
CDO pricing models*

*Credit Workshop
Isaac Newton Institute
26 February 2005*

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Joint work with X. Burtschell and J. Gregory



*A comparative analysis of
~~CDO~~ ad-hoc pricing models*

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Comparative analysis of CDO pricing models

■ Model dialectics

- *Benchmark model: one factor Gaussian model*
 - Used by all major investments banks to communicate quotes
 - On a large and liquid market market of synthetic CDO tranches
- *The oponents*
 - Within the « copula family »
 - Student t (O’Kane & Schloegl, Lindskog & McNeil), Clayton (Schönbucher), double t (Hull & White), multifactor Gaussian, Marshall-Olkin (Giesecke, Lindskog & McNeil), random factor loadings (Andersen & Sidenius)
 - Intensity models
 - Affine Jump Diffusion (Hutt), Gamma (Joshi), Stochastic Networks (Davis, Backhaus & Frey, Giesecke), Hawkes (Giesecke)
 - Structural models
- *United We Stand, Divided We Fall*



Comparative analysis of CDO pricing models

- **Model dialectics**

- *Try to look for a lowest common denominator*
- *« Homogeneity » assumption: default times are exchangeable*
 - Can be expressed through joint survival functions
 - Weaker form: the dependence structure is « exchangeable »
 - Expressed through the copula of default times
 - Weaker form: the dependence structure of default indicators is exchangeable
- *Leads to a one factor representation thanks to de Finetti theorem*
 - Usual de Finetti involves infinite sequences
 - Finite de Finetti, Jaynes (1986)
 - Factor representation involves signed measures



Comparative analysis of CDO pricing models

- Model dialectics
 - *Non homogeneous models*
 - *Usually involves a factor representation*
 - Loading factors will depend on names
- Factor approach holds for many intensity models
 - *Let us consider Joshi (25 February 2005)*
 - *Compensator follows a Gamma process $\Gamma(t)$*
 - *Conditional survival probabilities $\exp(-c_i\Gamma(t))$*
 - *$\Gamma(t)$ holds for the factor for time horizon t*
 - *Conditionally on the factor, default indicators are independent*



Comparative analysis of CDO pricing models

- Purpose of the presentation
 - *Assessment of CDO pricing models in a factor framework*
 - Comparisons of different models
 - Comparisons with market quotes
 - Based on tranche premiums
 - *Study the relevance of standard probabilistic tools such as tail dependence or non parametric measures of dependence (Kendall's tau)*
 - *Study how tranche premiums vary with parameters*
 - Stochastic orders theory
 - *Relate semi-analytical pricing approaches and large portfolio approximations (stochastic orders again)*



Comparative analysis of CDO pricing models

- $i = 1, \dots, n$ names.
- τ_1, \dots, τ_n default times.
- N_i nominal of credit i ,
- δ_i recovery rate
- $N_i(t) = 1_{\tau_i \leq t}$ default indicator
- $N_i(1 - \delta_i)$ loss given default



Comparative analysis of CDO pricing models

- Factor approaches to joint distributions:
 - *V: low dimensional factor*
 - *Conditionally on V, default times are independent.*
 - *Conditional default and survival probabilities:*

$$p_t^{i|V} = Q(\tau_i \leq t | V), \quad q_t^{i|V} = Q(\tau_i > t | V).$$

- Why factor models ?
 - *Tackle with large dimensions*
- Need tractable dependence between defaults:
 - *Parsimonious modelling*
 - *Semi-explicit computations for CDO tranches*



Comparative analysis of CDO pricing models

- Semi-explicit pricing for CDO tranches
 - *Default payments are based on the accumulated losses on the pool of credits:*

$$L(t) = \sum_{1 \leq i \leq n} N_i(1 - \delta_i)N_i(t)$$

- *Tranche premiums only involves call options on the accumulated losses*

$$E \left[(L(t) - K)^+ \right]$$



Comparative analysis of CDO pricing models

- Characteristic function: $\varphi_{L(t)}(u) = E \left[e^{iuL(t)} \right]$

- *By conditioning upon V and using conditional independence:*

$$\varphi_{L(t)}(u) = E \left[\prod_{1 \leq j \leq n} \left(1 - p_t^{j|V} + p_t^{j|V} \varphi_{1-\delta_j}(uN_j) \right) \right]$$

- *Distribution of $L(t)$ can be obtained by FFT*
 - Or other inversion technique
- Only need of conditional (on factor) probabilities $p_t^{i|V}$



Comparative analysis of CDO pricing models

- One factor Gaussian copula:

- $V, \bar{V}_i, i = 1, \dots, n$ independent Gaussian,

$$V_i = \rho_i V + \sqrt{1 - \rho_i^2} \bar{V}_i$$

- *Default times:* $\tau_i = F_i^{-1}(\Phi(V_i))$
- F_i marginal distribution function of default times
- *Conditional default probabilities:*

$$p_t^{i|V} = \Phi \left(\frac{-\rho_i V + \Phi^{-1}(F_i(t))}{\sqrt{1 - \rho_i^2}} \right)$$



Comparative analysis of CDO pricing models

- Student t copula

- *Embrechts, Lindskog & McNeil, Greenberg et al, Mashal et al, O'Kane & Schloegl, Gilkes & Jobst*

$$\begin{cases} X_i = \rho V + \sqrt{1 - \rho^2} \bar{V}_i \\ V_i = \sqrt{W} \times X_i \\ \tau_i = F_i^{-1} (t_v (V_i)) \end{cases}$$

- V, \bar{V}_i independent Gaussian variables
- $\frac{v}{W}$ follows a χ_v^2 distribution
- Conditional default probabilities (two factor model)

$$P_t^{i|V,W} = \Phi \left(\frac{-\rho V + W^{-1/2} t_v^{-1} (F_i(t))}{\sqrt{1 - \rho^2}} \right)$$



Comparative analysis of CDO pricing models

- *Clayton copula*

- *Schönbucher & Schubert, Rogge & Schönbucher, Friend & Rogge, Madan et al*

$$V_i = \psi\left(-\frac{\ln U_i}{V}\right) \quad \tau_i = F_i^{-1}(V_i) \quad \psi(s) = (1+s)^{-1/\theta}$$

- *V: Gamma distribution with parameter θ*
- *U_1, \dots, U_n independent uniform variables*
- *Conditional default probabilities (one factor model)*

$$p_t^{i|V} = \exp\left(V\left(1 - F_i(t)^{-\theta}\right)\right)$$



Comparative analysis of CDO pricing models

- Double t model (Hull & White)

$$V_i = \rho_i \left(\frac{\nu - 2}{\nu} \right)^{1/2} V + \sqrt{1 - \rho_i^2} \left(\frac{\bar{\nu} - 2}{\bar{\nu}} \right)^{1/2} \bar{V}_i$$

- V, \bar{V}_i are independent Student t variables
 - with ν and $\bar{\nu}$ degrees of freedom

$$\tau_i = F_i^{-1} \left(H_i (V_i) \right)$$

- where H_i is the distribution function of V_i

$$P_t^{i|V} = t_{\bar{\nu}} \left(\left(\frac{\bar{\nu}}{\bar{\nu} - 2} \right)^{1/2} \frac{H_i^{-1} (F_i(t)) - \rho_i \left(\frac{\nu - 2}{\nu} \right)^{1/2} V}{\sqrt{1 - \rho_i^2}} \right)$$



Comparative analysis of CDO pricing models

- Shock models (multivariate exponential copulas)
 - *Duffie & Singleton, Giesecke, Elouerkhaoui, Lindskog & McNeil, Wong*
- Modelling of default dates: $V_i = \min(V, \bar{V}_i)$
 - V, \bar{V}_i exponential with parameters $\alpha, 1-\alpha$
 - Default dates $\tau_i = S_i^{-1}(\exp(-\min(V, \bar{V}_i)))$
 - S_i marginal survival function
 - Conditionally on V, τ_i are independent.
- Conditional default probabilities

$$q_t^{i|V} = 1_{V > -\ln S_i(t)} S_i(t)^{1-\alpha}$$



Comparative analysis of CDO pricing models

- CDO margins (bps pa)
 - *With respect to correlation*
 - *Gaussian copula*
 - *Attachment points: 3%, 10%*
 - *100 names*
 - *Unit nominal*
 - *Credit spreads 100 bp*
 - *5 years maturity*

| | equity | mezzanine | senior |
|-------------|---------------|------------------|---------------|
| 0% | 5341 | 560 | 0.03 |
| 10% | 3779 | 632 | 4.6 |
| 30% | 2298 | 612 | 20 |
| 50% | 1491 | 539 | 36 |
| 70% | 937 | 443 | 52 |
| 100% | 167 | 167 | 91 |



Comparative analysis of CDO pricing models

- Equity tranche premiums are decreasing wrt ρ
 - *General result ?*
 - *Supermodular function f is such that:*

$$f : \mathbb{R}^n \rightarrow \mathbb{R} \quad \Delta_i^\varepsilon f(x) = f(x + \varepsilon e_i) - f(x)$$

$$\forall x \in \mathbb{R}^n, \forall \varepsilon, \delta > 0 \quad \Delta_i^\varepsilon \Delta_j^\delta f(x) \geq 0$$

- *Supermodular order*

$$X = (X_1, \dots, X_n) \quad Y = (Y_1, \dots, Y_n)$$

$$X \leq_{\text{sm}} Y \Leftrightarrow E[f(X)] \leq E[f(Y)], \quad \forall f \text{ supermodular}$$

Comparative analysis of CDO pricing models

- « Supermodular » order of Gaussian vectors

- *Let X and Y be Gaussian vectors with zero mean*

$$\Sigma^X = \begin{pmatrix} 1 & \sigma_{12}^X & & & \\ \sigma_{21}^X & 1 & & & \\ & & 1 & \sigma_{ij}^X & \\ & & & 1 & \\ \sigma_{ji}^X & & & & 1 \\ & & & & & 1 \end{pmatrix} \quad \Sigma^Y = \begin{pmatrix} 1 & \sigma_{12}^Y & & & \\ \sigma_{21}^Y & 1 & & & \\ & & 1 & \sigma_{ij}^Y & \\ & & & 1 & \\ \sigma_{ji}^Y & & & & 1 \\ & & & & & 1 \end{pmatrix}$$

$$\Sigma^X \leq \Sigma^Y \stackrel{\text{def}}{\Leftrightarrow} \sigma_{ij}^X \leq \sigma_{ij}^Y, \quad \forall i, j$$

- *Müller & Scarsini (2000), Müller (2001)*

$$\Sigma^X \leq \Sigma^Y \Leftrightarrow X \leq_{\text{sm}} Y$$



Comparative analysis of CDO pricing models

- « Stop-Loss » order
 - Accumulated losses: $L(t), L'(t)$

$$L(t) \leq_{sl} L'(t) \stackrel{\text{def}}{\Leftrightarrow} E \left[(L(t) - K)^+ \right] \leq E \left[(L'(t) - K)^+ \right], \forall K \geq 0$$

- Supermodular order of latent variables implies stop-loss order of accumulated losses
- Thus, equity tranche premium is always decreasing with correlation
- Guarantees uniqueness of « base correlation »
- Monotonicity properties extend to Student t , Marshall-Olkin copulas



Comparative analysis of CDO pricing models

- Second issue

- *Equity tranche premium decrease with correlation*
- *Does $\rho = 100\%$ correspond to some lower bound?*
- *$\rho = 100\%$ corresponds to « comonotonic » default dates:*

$$(\tau_1, \dots, \tau_n) \text{ comonotonic} \Leftrightarrow (\tau_1, \dots, \tau_n) \stackrel{d}{=} (F_1^{-1}(U), \dots, F_n^{-1}(U))$$

- where U is uniform

$$(\tau_1, \dots, \tau_n) \leq_{\text{sm}} (F_1^{-1}(U), \dots, F_n^{-1}(U))$$

- Tchen (1980)
- $\rho = 100\%$ is a model free lower bound for the equity tranche premium



Comparative analysis of CDO pricing models

- **Third issue**

- *Does $\rho = 0\%$ corresponds to the higher bound on the equity tranche premium?*
- *$\rho = 0\%$ corresponds to the independence case between default dates*
- *The answer is no, negative dependence can occur*
- *Base correlation does not always exists*
 - Even in Gaussian copula models
- *Factor models are usually associated with positive dependence*
 - i.e. independent default dates are smaller with respect to supermodular order



Comparative analysis of CDO pricing models

- Calibration issues
 - *One parameter copulas*
 - *Well suited for homogeneous portfolios*
 - *Dependence is « monotonic » in the parameter*
- Calibration procedure
 - *Fit Clayton, Student t , double t , Marshall Olkin parameters onto CDO equity tranches*
 - Computed under one factor Gaussian model
 - Or given market quotes on equity tranches
 - *Reprice mezzanine and senior CDO tranches*
 - Given the previous parameters



Comparative analysis of CDO pricing models

- CDO margins (bps pa)
 - *With respect to correlation*
 - *Gaussian copula*
 - *Attachment points: 3%, 10%*
 - *100 names*
 - *Unit nominal*
 - *Credit spreads 100 bp*
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Comparative analysis of CDO pricing models

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| 70% | 937 | 443 | 52 |
| 100% | 167 | 167 | 91 |



Comparative analysis of CDO pricing models

| | | | | | | |
|--------------------|----|------|------|------|------|----------|
| ρ | 0% | 10% | 30% | 50% | 70% | 100% |
| θ | 0 | 0.05 | 0.18 | 0.36 | 0.66 | ∞ |
| ρ_6 | 0% | 0% | 14% | 39% | 63% | 100% |
| ρ_{12} | 0% | 0% | 22% | 45% | 67% | 100% |
| $\rho_{t(4)-t(4)}$ | 0% | 12% | 34% | 55% | 73% | 100% |
| $\rho_{t(5)-t(4)}$ | 0% | 13% | 36% | 56% | 74% | 100% |
| $\rho_{t(4)-t(5)}$ | 0% | 12% | 34% | 54% | 73% | 100% |
| $\rho_{t(3)-t(4)}$ | 0% | 10% | 32% | 53% | 75% | 100% |
| $\rho_{t(4)-t(3)}$ | 0% | 11% | 33% | 54% | 73% | 100% |
| α | 0 | 28% | 53% | 69% | 80% | 100% |

Table 5: correspondence between parameters



Comparative analysis of CDO pricing models

| ρ | 0% | 10% | 30% | 50% | 70% | 100% |
|--------------|-----|-----|-----|-----|-----|------|
| Gaussian | 560 | 633 | 612 | 539 | 443 | 167 |
| Clayton | 560 | 637 | 628 | 560 | 464 | 167 |
| Student (6) | 676 | 676 | 637 | 550 | 447 | 167 |
| Student (12) | 647 | 647 | 621 | 543 | 445 | 167 |
| $t(4)-t(4)$ | 560 | 527 | 435 | 369 | 313 | 167 |
| $t(5)-t(4)$ | 560 | 545 | 454 | 385 | 323 | 167 |
| $t(4)-t(5)$ | 560 | 538 | 451 | 385 | 326 | 167 |
| $t(3)-t(4)$ | 560 | 495 | 397 | 339 | 316 | 167 |
| $t(4)-t(3)$ | 560 | 508 | 406 | 342 | 291 | 167 |
| MO | 560 | 284 | 144 | 125 | 134 | 167 |

Table 6: mezzanine tranche (bps pa)



Comparative analysis of CDO pricing models

| ρ | 0% | 10% | 30% | 50% | 70% | 100% |
|--------------|------|-----|-----|-----|-----|------|
| Gaussian | 0.03 | 4.6 | 20 | 36 | 52 | 91 |
| Clayton | 0.03 | 4.0 | 18 | 33 | 50 | 91 |
| Student (6) | 7.7 | 7.7 | 17 | 34 | 51 | 91 |
| Student (12) | 2.9 | 2.9 | 19 | 35 | 52 | 91 |
| $t(4)-t(4)$ | 0.03 | 11 | 30 | 45 | 60 | 91 |
| $t(5)-t(4)$ | 0.03 | 10 | 29 | 45 | 59 | 91 |
| $t(4)-t(5)$ | 0.03 | 10 | 29 | 44 | 59 | 91 |
| $t(3)-t(4)$ | 0.03 | 12 | 32 | 47 | 71 | 91 |
| $t(4)-t(3)$ | 0.03 | 12 | 32 | 47 | 61 | 91 |
| MO | 0.03 | 25 | 49 | 62 | 73 | 91 |

Table 7: senior tranche (bps pa)



Comparative analysis of CDO pricing models

| ρ | 0% | 10% | 30% | 50% | 70% | 100% |
|--------------|----|------|------|------|------|------|
| Gaussian | 0% | 0% | 0% | 0% | 0% | 100% |
| Clayton | 0% | 0% | 2% | 15% | 35% | 100% |
| Student (6) | 3% | 3% | 4% | 6% | 13% | 100% |
| Student (12) | 0% | 0% | 0% | 1% | 4% | 100% |
| $t(4)-t(4)$ | 0% | 0% | 1% | 10% | 48% | 100% |
| $t(5)-t(4)$ | 0% | 0% | 0% | 0% | 0% | 100% |
| $t(4)-t(5)$ | 0% | 100% | 100% | 100% | 100% | 100% |
| $t(3)-t(4)$ | 0% | 100% | 100% | 100% | 100% | 100% |
| $t(4)-t(3)$ | 0% | 0% | 0% | 0% | 0% | 100% |
| MO | 0% | 28% | 53% | 69% | 80% | 100% |

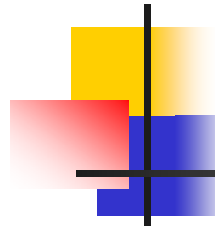
Table 8: coefficient of lower tail dependence (%)



Comparative analysis of CDO pricing models

| ρ | 0% | 10% | 30% | 50% | 70% | 100% |
|--------------|----|-----|-----|-----|-----|------|
| Gaussian | 0% | 1% | 6% | 16% | 33% | 100% |
| Clayton | 0% | 3% | 8% | 15% | 25% | 100% |
| Student (6) | 0% | 0% | 1% | 10% | 26% | 100% |
| Student (12) | 0% | 0% | 3% | 13% | 30% | 100% |
| MO | 0% | 16% | 36% | 53% | 67% | 100% |

Table 9: Kendall's τ (%)



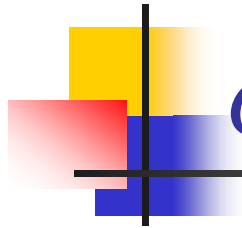
Comparative analysis of CDO pricing models

| Tranches | Market | Gaussian | $t(4)-t(4)$ | Clayton | MO | Student (12) | Student (6) |
|----------|--------|----------|-------------|---------|-----|--------------|-------------|
| 0%-3% | 916 | 916 | 916 | 916 | 916 | 916 | 849 |
| 3%-6% | 101 | 163 | 82 | 163 | 15 | 164 | 186 |
| 6%-9% | 33 | 48 | 34 | 47 | 12 | 47 | 61 |
| 9%-12% | 16 | 17 | 22 | 16 | 12 | 15 | 21 |
| 12%-22% | 9 | 3 | 13 | 2 | 12 | 2 | 3 |

Table 15: CDO tranche premiums I-TRAXX (bps pa)

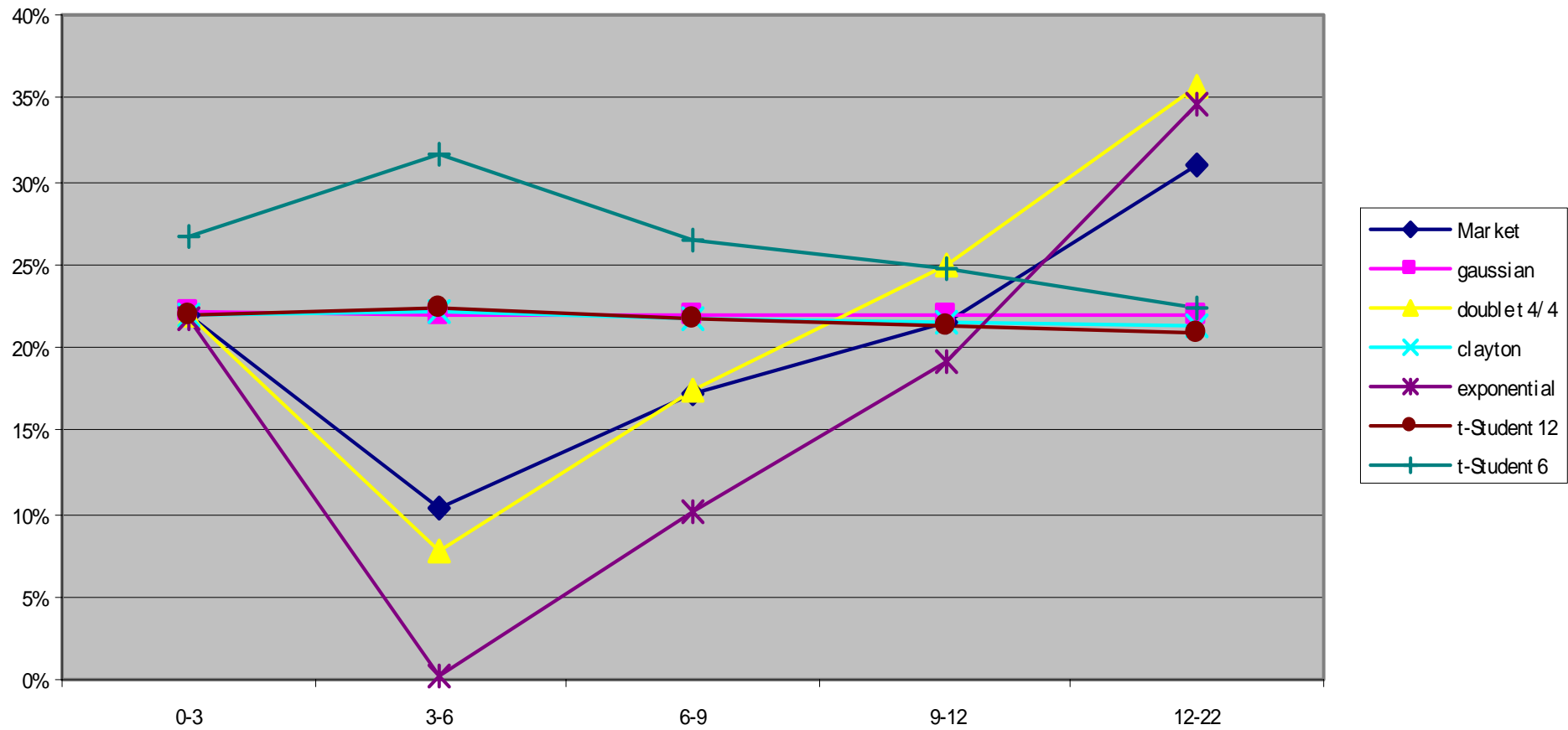
| Tranches | Market | Gaussian | $t(4)-t(4)$ | Clayton | MO | Student (12) | Student (6) |
|----------|--------|----------|-------------|---------|-----|--------------|-------------|
| 0%-3% | 916 | 916 | 916 | 916 | 916 | 916 | 849 |
| 0%-6% | 466 | 503 | 456 | 504 | 417 | 504 | 490 |
| 0%-9% | 311 | 339 | 305 | 339 | 273 | 340 | 336 |
| 0%-12% | 233 | 253 | 230 | 253 | 203 | 254 | 253 |
| 0%-22% | 128 | 135 | 128 | 135 | 114 | 135 | 135 |

Table 16: CDO tranche premiums I-TRAXX (bps pa)



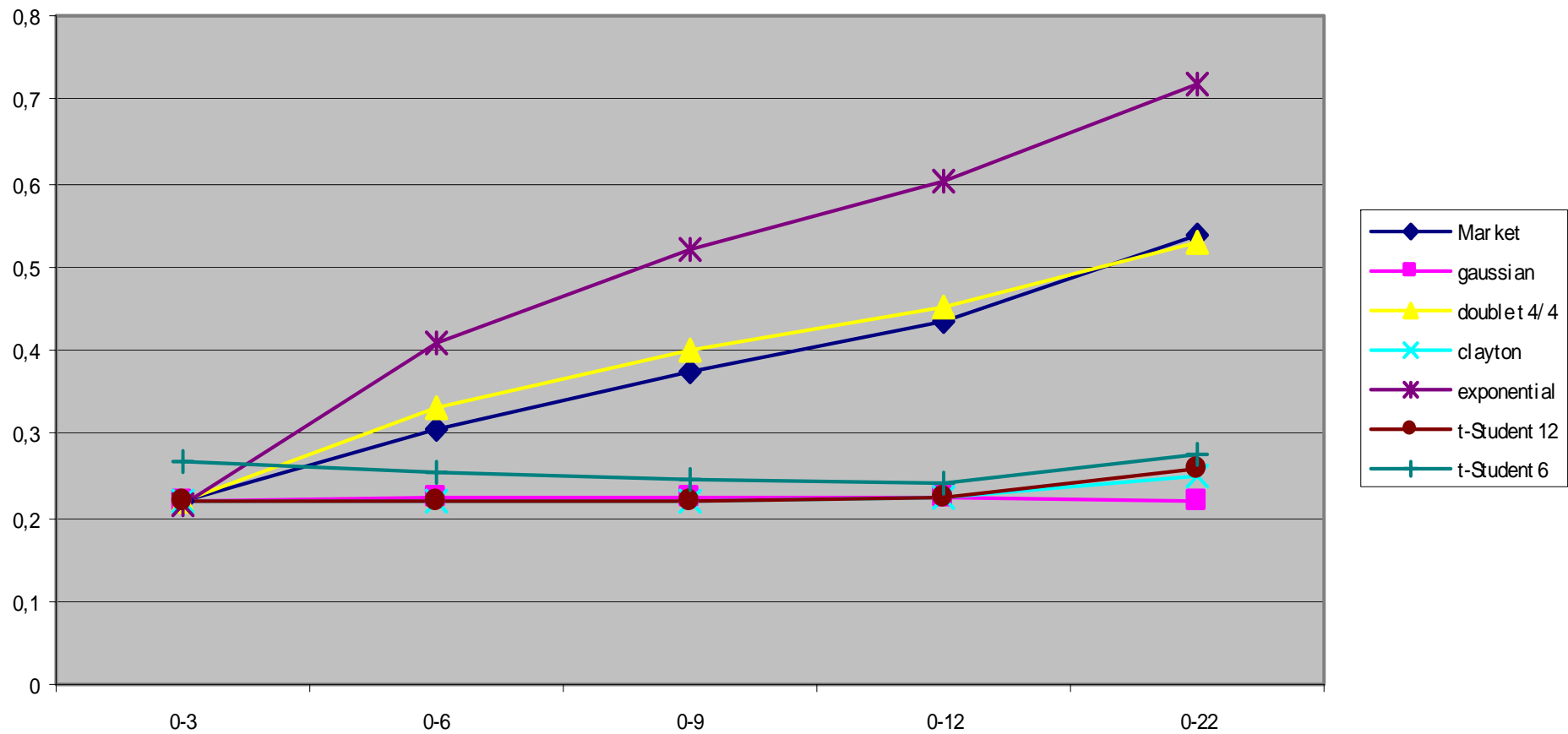
Comparative analysis of CDO pricing models

Implied Compound Correlation



Comparative analysis of CDO pricing models

Implied Base Correlation





Model risk: choice of copula

- Related results:
 - *Student vs Gaussian*
 - Frey & McNeil, Mashal et al
 - Calibration on asset correlation
 - Distance between Gaussian and Student is bigger for low correlation levels
 - And extremes of the loss distribution
 - Joint default probabilities increase as number of degrees of freedom decreases
 - *Calibration onto joint default probabilities*
 - or default correlation, or aggregate loss variance
 - O'Kane & Schloegl, Schonbucher
 - *Gaussian, Clayton and Student t are all very similar*



Conclusion

- Factor models of default times:
 - *Simple computation of CDO's*
- Gaussian, Clayton and Student t copulas provide very similar patterns
- Shock models (Marshall-Olkin) quite different
- Double t provides intermediate results