



# *Comparative analysis of CDO pricing models*

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*Joint work with X. Burtschell & J. Gregory*

*A comparative analysis of CDO pricing models*

*Beyond the Gaussian copula: stochastic and local correlation*

*Available on [www.defaultrisk.com](http://www.defaultrisk.com)*



# *Comparative analysis of CDO pricing models*

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- 1 Factor based copulas
  - Collective & individual models of credit losses
  - Semi-explicit pricing
- 2 One factor Gaussian copula
  - Ordering of risks, Base correlation
  - correlation sensitivities
  - Stochastic recovery rates
- 3 Model dependence / choice of copula
  - Student  $t$ , double  $t$ , Clayton, Marshall-Olkin, Stochastic correlation
  - Distribution of conditional default probabilities
- 4 Beyond the Gaussian copula
  - Stochastic correlation and state dependent correlation
  - Marginal and local correlation



## *1 Factor based copulas*

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- CDO valuation, credit risk assessment
  - *Only need of loss distributions for different time horizons*
  - *Aggregate loss at time  $t$  on a given portfolio:  $L(t)$*
  - *Marginal loss distribution for time horizon  $t$*

$$l \rightarrow F_{L(t)}(l) = Q(L(t) \leq l)$$

- *VaR and quantile based risk measures for risk assessment*

$$\int_0^1 F_{L(t)}^{-1}(\alpha) v(\alpha) d\alpha$$

- *Pricing of CDOs<sup>0</sup> only involve options on aggregate loss*

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

- $K$  attachment – detachment points



## *1 Factor based copulas*

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- **Modelling approaches**

- *Direct modelling of  $L(t)$ : collective model*

- Dealing with heterogeneous portfolios
- non stationary, non Markovian
- Aggregation of portfolios, bespoke portfolios?
- Risk management of correlation risk?

- *Modelling of default indicators of names: individual model*

$$L(t) = \sum_{i=1}^n LGD_i 1_{\tau_i \leq t}$$

- *Numerical approaches*

- e.g. smoothing of base correlation of liquid tranches



## *1 Factor based copulas*

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- Individual model / factor based copulas
  - *Allows to deal with non homogeneous portfolios*
  - *Arbitrage free prices*
    - non standard attachment –detachment points
    - Non standard maturities
  - *Consistent pricing of bespoke, CDO<sup>2</sup>, zero-coupon CDOs*
  - *Computations*
    - Semi-explicit pricing, computation of Greeks, LHP
  - *But...*
    - Poor dynamics of aggregate losses (forward starting CDOs)
    - Risk management, credit deltas, theta effects
    - Calibration onto liquid tranches (matching the skew)



## *1 Factor based copulas*

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- Factor approaches to joint default times 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 (i-Traxx, CDX)*
- Need of tractable dependence between defaults:
  - *Parsimonious modelling*
  - *Semi-explicit computations for CDO tranches*
  - *Large portfolio approximations*



## *1 Factor based copulas*

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- Semi-explicit pricing for CDO tranches

- Laurent & Gregory [2003]

- *Default payments are based on the accumulated losses on the pool of credits:*

$$L(t) = \sum_{i=1}^n LGD_i 1_{\{\tau_i \leq t\}}, \quad LGD_i = N_i(1 - \delta_i)$$

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

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

- *This is equivalent to knowing the distribution of  $L(t)$*



## *1 Factor based copulas*

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- 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*
  - Similar approaches: recursion, inversion of Laplace transforms
- Only need of conditional default probabilities  $p_t^{i|V}$
- $p_t^{i|V}$  losses on a large homogeneous portfolio
  - *Approximation techniques for pricing CDOs*





# *Comparative analysis of CDO pricing models*

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- *2 One factor Gaussian copula*
  - Ordering of risks, Base correlation
  - correlation sensitivities
  - Stochastic recovery rates
- *3 Model dependence/Choice of copula*
  - Student  $t$ , double  $t$ , Clayton, Marshall-Olkin, Stochastic correlation
  - Distribution of conditional default probabilities
- *4 Beyond the Gaussian copula*
  - Stochastic correlation and state dependent correlation
  - Marginal and local correlation



## 2 One factor Gaussian copula

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- 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:

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



## 2 One factor Gaussian copula

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- Equity tranche premiums are decreasing wrt  $\rho$ 
  - *General result (use of stochastic orders theory)*
  - *Equity tranche premium is always decreasing with correlation parameter*
  - *Guarantees uniqueness of « base correlation »*
  - *Monotonicity properties extend to Student  $t$ , Clayton and Marshall-Olkin copulas*



## 2 One factor Gaussian copula

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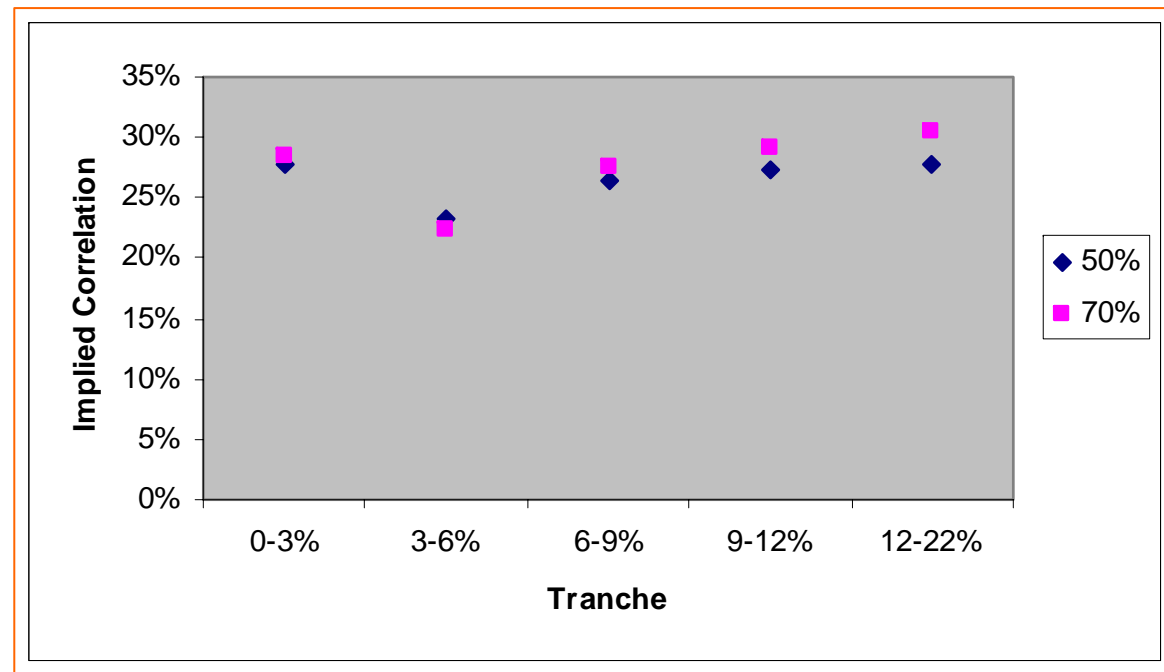
- $\rho = 100\%$ 
  - *Equity tranche premiums decrease with correlation*
  - *Does  $\rho = 100\%$  correspond to some lower bound?*
  - *$\rho = 100\%$  corresponds to « comonotonic » default dates:*
  - *$\rho = 100\%$  is a model free lower bound for the equity tranche premium*
- $\rho = 0\%$ 
  - *Does  $\rho = 0\%$  correspond 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 exist*





## 2 One factor Gaussian copula

- Correlation between default dates and recovery rates
  - *Correlation smile implied from the correlated recovery rates*
  - *Not as important as what is found in the market*





### 3 Model dependence / choice of copula

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- Stochastic correlation copula
  - $V, \bar{V}_i, i = 1, \dots, n$  independent Gaussian variables
  - $B_i = 1$  correlation  $\rho$ ,  $B_i = 0$  correlation  $\beta$

$$V_i = B_i \left( \rho V + \sqrt{1 - \rho^2} \bar{V}_i \right) + (1 - B_i) \left( \beta V + \sqrt{1 - \beta^2} \bar{V}_i \right)$$

$$\tau_i = F_i^{-1}(\Phi(V_i))$$

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





### 3 Model dependence / choice of copula

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- Student  $t$  copula

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

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

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



### 3 Model dependence / choice of copula

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- *Clayton copula*

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

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

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



### 3 Model dependence / choice of copula

- 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  $\nu$  and  $\bar{\nu}$  degrees of freedom

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

- where  $H_i$  is the distribution function of  $V_i$

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



### 3 Model dependence / choice of copula

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- Shock models (multivariate exponential copulas)
  - *Marshall-Olkin copulas*
- 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  $V$ ,  $\tau_i$  are independent.*
- Conditional default probabilities

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



### *3 Model dependence / choice of copula*

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- Calibration procedure
  - *One parameter copulas*
  - *Fit Clayton, Student t, double t, Marshall Olkin parameters onto CDO equity tranches*
    - Computed under one factor Gaussian model
  - *Reprice mezzanine and senior CDO tranches*
    - Given the fitted parameter
    - Look for departures from the Gaussian copula
    - Look for ability to explain the correlation skew



### 3 Model dependence / choice of copula

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



### 3 Model dependence / choice of copula

$\rho$	0%	10%	30%	50%	70%	100%
Gaussian	560	633	612	539	443	167
Clayton	560	637	628	560	464	167
Student (6)			637	550	447	167
Student (12)			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)



### 3 Model dependence / choice of copula

$\rho$	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)			17	34	51	91
Student (12)			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)

Gaussian, Clayton and Student  $t$  CDO premiums are close



### 3 Model dependence / choice of copula

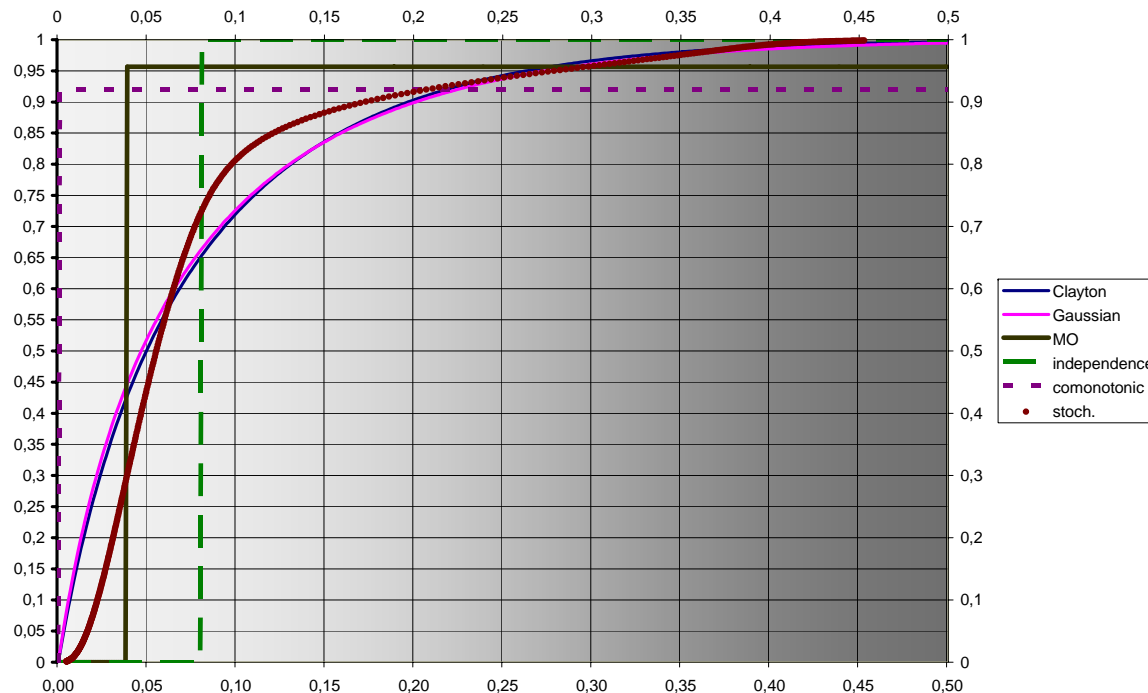
- Why do Clayton and Gaussian copulas provide same premiums?

- Loss distributions depend on the **distribution** of conditional default probabilities

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

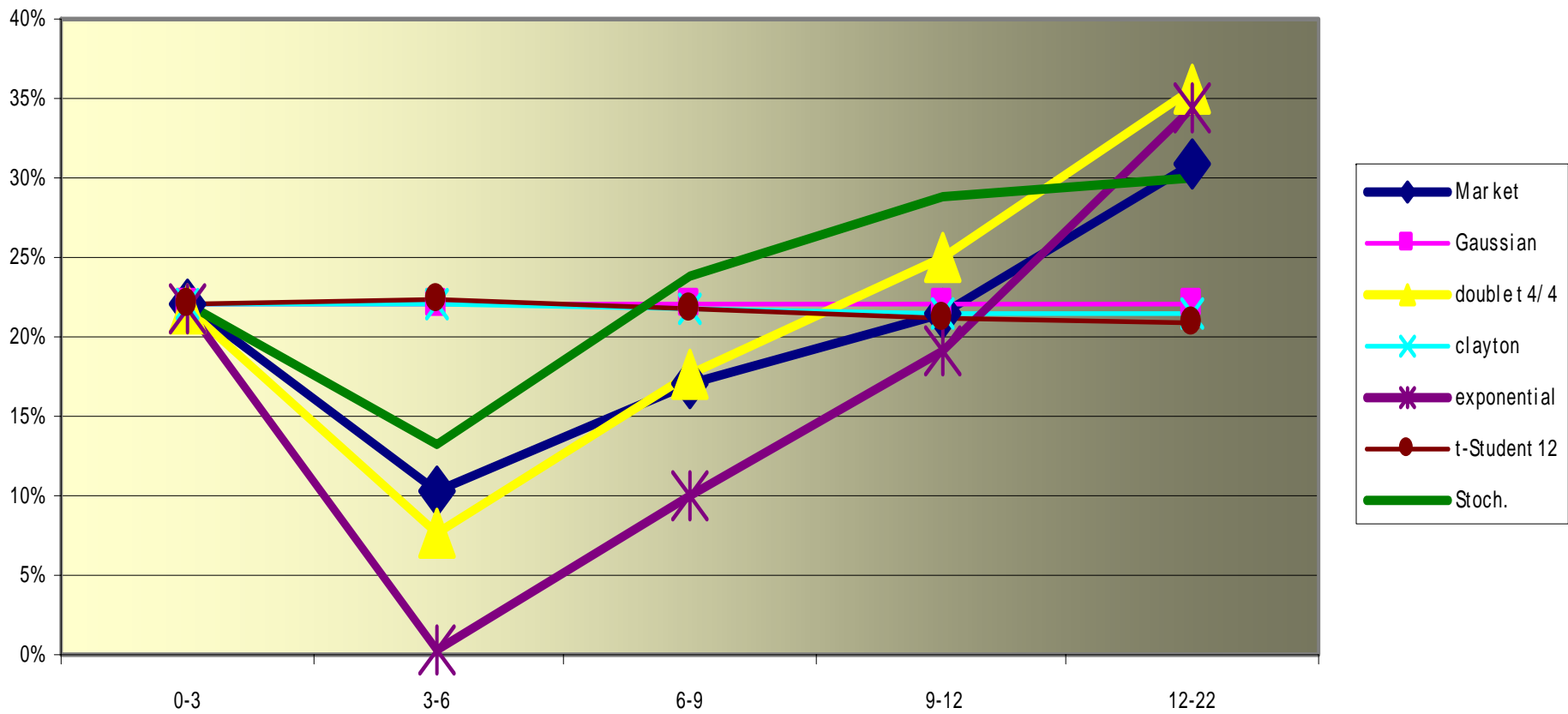
$$p_t^{i|V} = \Phi\left(\frac{-\rho V + \Phi^{-1}\left(F_i(t)\right)}{\sqrt{1 - \rho^2}}\right)$$

- Distribution of conditional default probabilities are close for Gaussian and Clayton



# 3 Model dependence / choice of copula

## implied compound correlation





## 4 Beyond the Gaussian copula

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- Stochastic correlation

- Latent variables  $V_i = \tilde{\rho}_i V + \sqrt{1 - \tilde{\rho}_i^2} \bar{V}_i, \quad i = 1, \dots, n$

$$\tilde{\rho}_i = (1 - B_s)(1 - B_i)\rho + B_s$$

$\tilde{\rho}_i$ , stochastic correlation,

$Q(B_s = 1) = q_s$ , systemic state,

$Q(B_i = 1) = q$ , idiosyncratic state

- Conditional default probabilities

$$p_t^{V, B_s=0} = (1 - q)\Phi\left(\frac{\Phi^{-1}(F(t)) - \rho V}{\sqrt{1 - \rho^2}}\right) + qF(t), \quad F(t) \text{ default probability}$$

$$p_t^{V, B_s=1} = \mathbf{1}_{V \leq \Phi^{-1}(F(t))}, \quad \text{comonotonic}$$



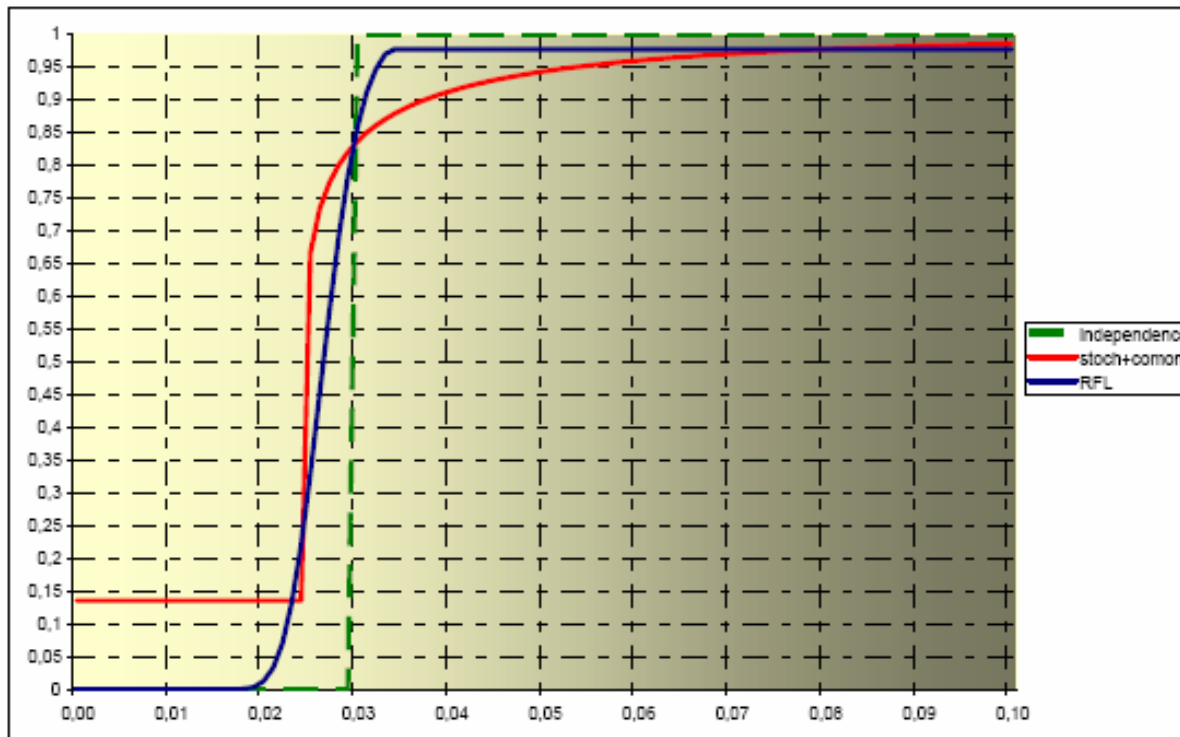
## 4 Beyond the Gaussian copula

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- Stochastic correlation  $\tilde{\rho}_i = (1 - B_s)(1 - B_i)\rho + B_s$ 
  - *Semi-analytical techniques for pricing CDOs available*
  - *Large portfolio approximation can be derived*
  - *Allows for Monte Carlo*
  - $\nearrow \rho, \searrow q_s, \searrow q$  leads to increase senior tranche premiums
- State dependent correlation  $V_i = m_i(V)V + \sigma_i(V)\bar{V}_i, \quad i = 1, \dots, n$ 
  - *Local correlation*  $V_i = -\rho(V)V + \sqrt{1 - \rho^2(V)}\bar{V}_i$ 
    - Turc et al
  - *Random factor loadings*  $V_i = m + (l1_{V < e} + h1_{V \geq e})V + v\bar{V}_i$ 
    - Andersen & Sidenius

## 4 Beyond the Gaussian copula

- Distribution functions of conditional default probabilities
  - *stochastic correlation vs RFL*



- *With respect to level of aggregate losses*
- *Also correspond to loss distributions on large portfolios*



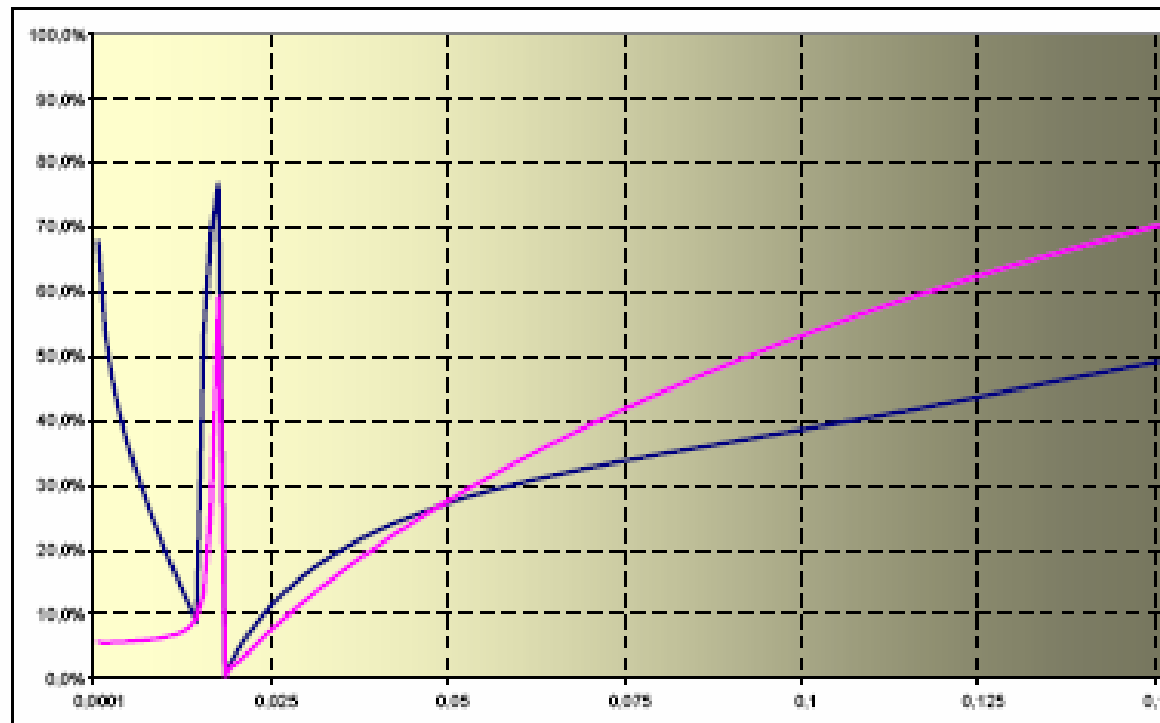
## 4 Beyond the Gaussian copula

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- Marginal compound correlation
  - *Compound correlation of a  $[\alpha, \alpha]$  tranche*
    - Digital call on aggregate loss
  - *obtained from conditional default probability distribution*
  - *Need to solve a second order equation*
  - *zero, one or two marginal compound correlations*

## 4 Beyond the Gaussian copula

- Marginal compound correlations:
  - *With respect to attachment – detachment point*



- *Stochastic correlation vs RFL*
- *zero marginal compound correlation at the expected loss*



## 4 *Beyond the Gaussian copula*

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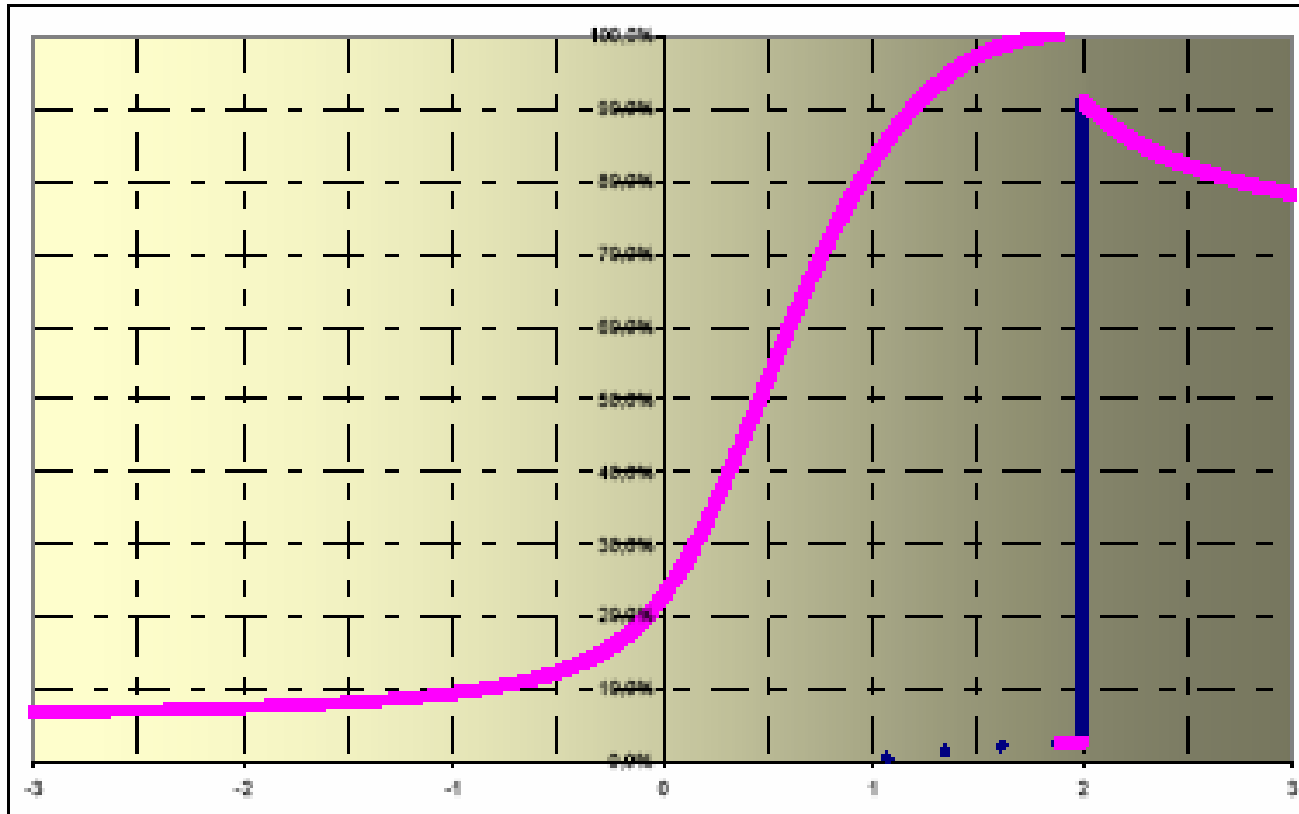
- Local correlation

- *obtained from conditional default probability distribution*
- *Fixed point algorithm*
- *Local correlation at step one: rescaled marginal compound correlation*
- *Same issues of uniqueness and existence as marginal compound correlation*



## 4 Beyond the Gaussian copula

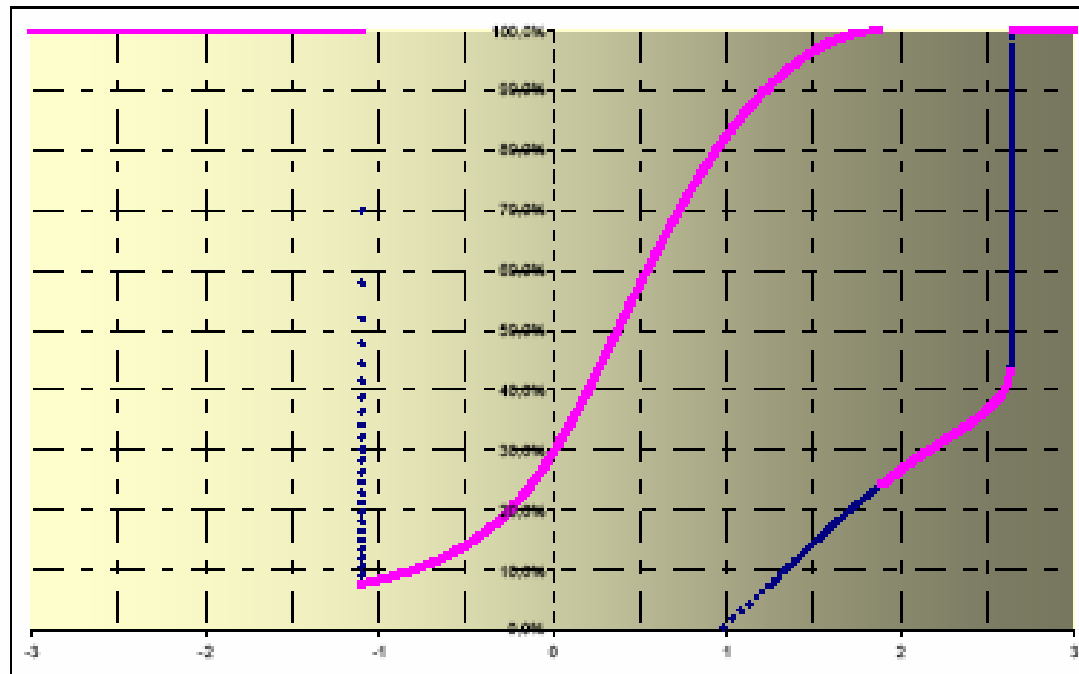
- Local correlation associated with RFL (as a function of the factor)



- Jump at threshold 2, low correlation level 5%, high correlation level 85%
- Possibly two local correlations

## 4 Beyond the Gaussian copula

- Local correlation associated with stochastic correlation model
  - *With respect to factor  $V$*



- *Correlations of 1 for high-low values of  $V$  (comonotonic state)*
- *Possibly two local correlations leading to the same prices*
- *As for RFL, rather irregular pattern*



## Conclusion

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- Analysis of dependence through factor models
  - *Usefulness of stochastic orders*
  - *Correlation sensitivities, base correlations*
- Matching the correlation skew
  - *Conditional default probability distributions are the drivers*
- Beyond the Gaussian copula
  - *Stochastic, local & marginal compound correlation*
- Further work
  - *Matching term structure of correlation skews*
  - *Integrating factor copulas and intensity approaches*