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Some valuation models for CDOs and  
Basket Credit Derivatives<sup>1</sup>

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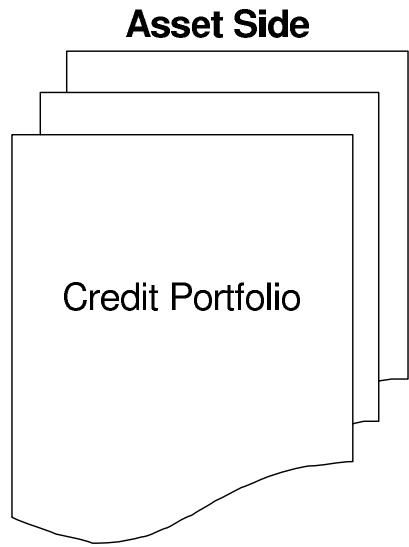
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<sup>1</sup>with Christian Bluhm, Credit Suisse, and Wolfgang Schmidt, HfB, Frankfurt,

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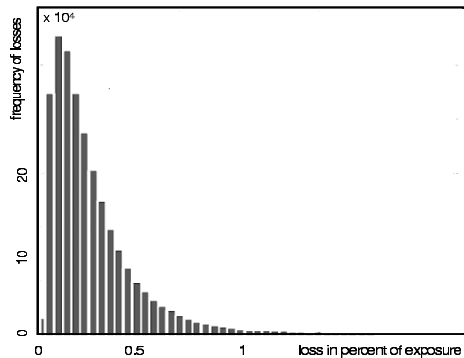
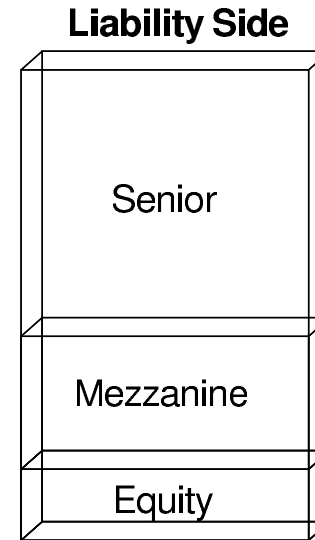
# Agenda

- Basic structure of CDOs
- Comonotonicity in Default Timing



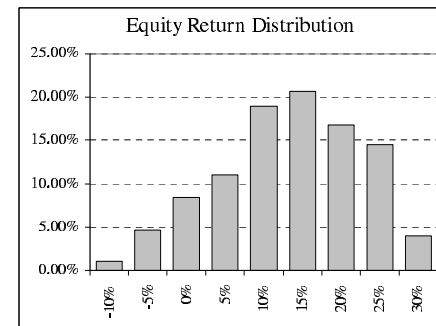
### Scenario Transformation

- draw an asset scenario at random
- apply the structural definitions of transaction
- obtain a scenario at the liability side

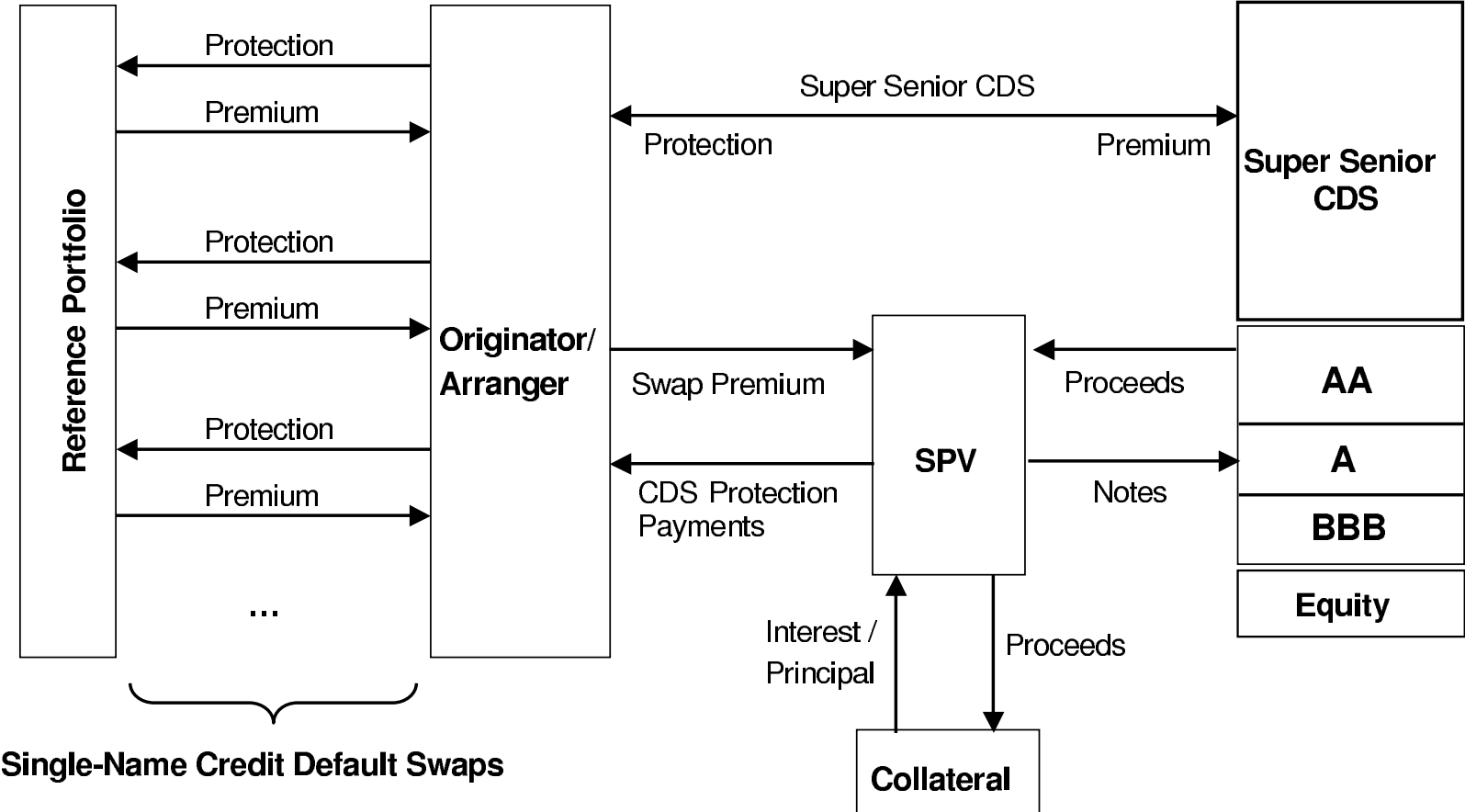


### Example:

obtain the equity return distribution of a CDO by transforming 100,000 asset scenarios and then aggregating accordingly



Synthetic transaction based on credit default swaps (CSO)



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# Comonotonicity in Default Quote Paths

Starting point: default times vector  $\boldsymbol{\tau} = (\tau_1, \dots, \tau_m)$  and default quote path  $(L_t)_{0 \leq t \leq T}$  with

$$L_t = \frac{1}{m} \sum_{k=1}^m \mathbf{1}_{\{\tau_k \leq t\}} \quad (1)$$

Intertemporal dependence of jumps  $L_t - L_s$  for times  $s < t$ , sampled at times

$$0 = t_0 < t_1 < \dots < t_{q-1} < t_q = T$$

is captured by some copula function

$$C : [0, 1]^m \rightarrow [0, 1], (u_1, \dots, u_m) \mapsto C(u_1, \dots, u_m)$$

(see SKLAR's theorem) such that

$$\mathbb{P}[L_{t_1} \leq x_1, \dots, L_{t_q} \leq x_q] = C(\mathbb{P}[L_{t_1} \leq x_1], \dots, \mathbb{P}[L_{t_q} \leq x_q]) \quad (2)$$

Note that always  $x_i = k_i/m$  with  $k_i \in \{1, \dots, m\}$  and that the copula  $C$  not unique!

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**Definition 1** A default quote path  $(L_t)_{0 \leq t \leq T}$  is called comonotonic if

$$\mathbb{P}[L_{t_1} \leq x_1, \dots, L_{t_q} \leq x_q] = \tilde{C}(\mathbb{P}[L_{t_1} \leq x_1], \dots, \mathbb{P}[L_{t_q} \leq x_q])$$

where  $C = \tilde{C}$  is the comonotonic copula defined by

$$\tilde{C}(u_1, \dots, u_m) = \min\{u_1, \dots, u_m\} \quad (u_1, \dots, u_m \in [0, 1]).$$

Note that  $\tilde{C}$  sometimes is called the upper Frechet copula. It is the copula inducing the strongest possible multivariate dependency

Next, we will construct the comonotone version of any given default quote path:

Define the following probabilities w.r.t. any given portfolio model,

$$p_{t,k} = \mathbb{P}[L_t = k/m], \tag{3}$$

where  $m$  denotes the number of names in the portfolio and  $L_t$  is from (1)

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Given these probabilities for a discrete time grid  $t = t_1, \dots, t_q$ ,

$$(p_{t,k})_{t=t_1, \dots, t_q; k=0,1,2, \dots, m} \in \mathbb{R}^{q \times (m+1)},$$

we write  $[z] = \max\{k \in \mathbb{N}_0 : k \leq z\}$  for the greatest nonnegative integer below  $z \geq 0$  and define for any time  $t_j > 0$  a distribution function

$$G_{t_j}(x) = \sum_{k=0}^{[mx]} p_{t_j, k} \quad (x \in [0, 1]) \quad (4)$$

$G_{t_j}$  is a step function due to the finite granularity  $m$  of the portfolio

We now compare the original default quote path

$$\mathbf{L} = (L_{t_1}, \dots, L_{t_q}) = \left( \frac{1}{m} \sum_{k=1}^m \mathbf{1}_{\{\tau_k \leq t_1\}}, \dots, \frac{1}{m} \sum_{k=1}^m \mathbf{1}_{\{\tau_k \leq t_q\}} \right),$$

determined by the chosen portfolio model in line with (1) with the default quote path

$$\tilde{\mathbf{L}} = (\tilde{L}_{t_1}, \dots, \tilde{L}_{t_q}) = \left( G_{t_1}^{-1}(Z), \dots, G_{t_q}^{-1}(Z) \right) \quad (Z \sim U[0, 1])$$

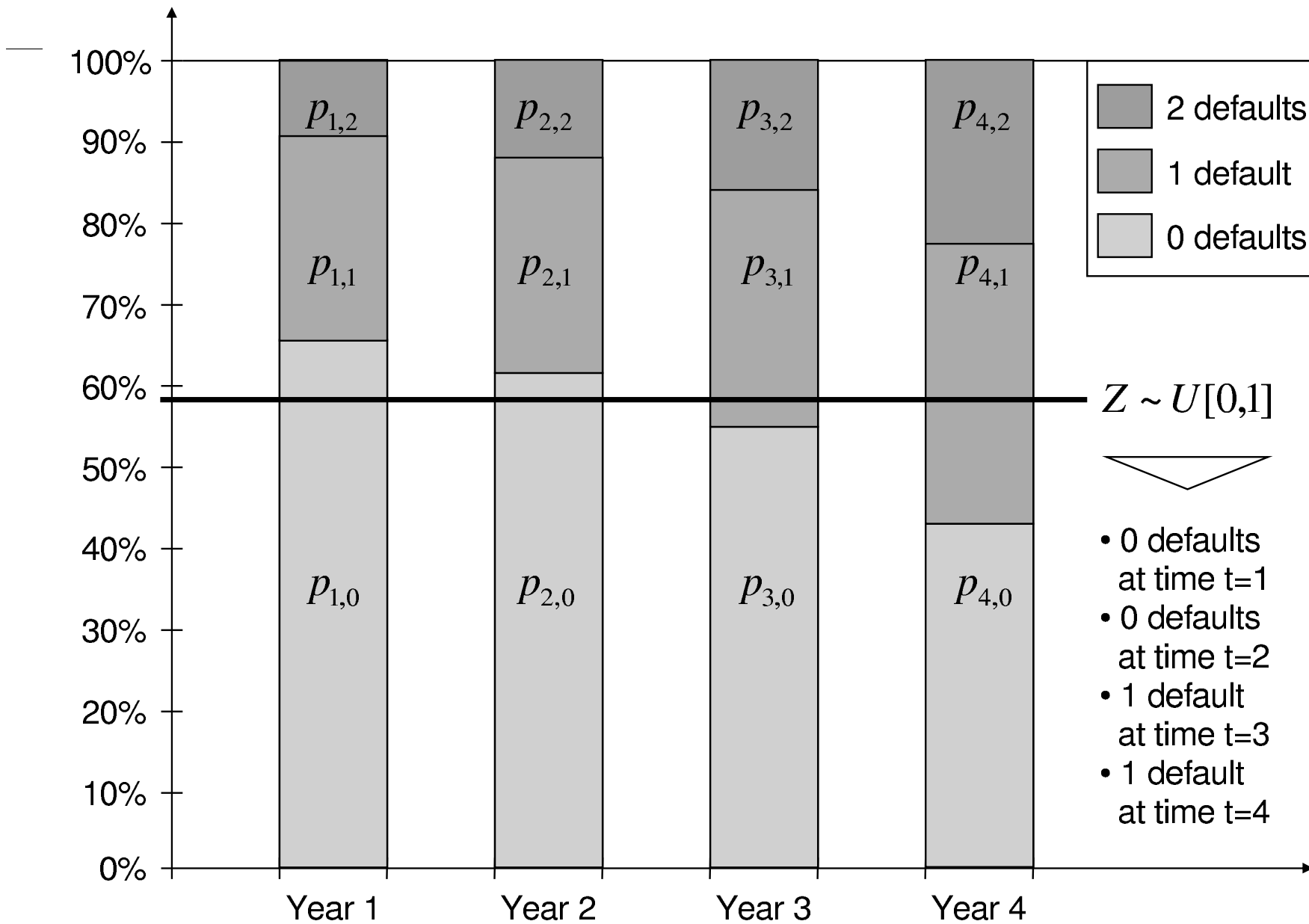

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**Proposition 1** *The marginal distributions of  $\mathbf{L}$  and  $\tilde{\mathbf{L}}$  coincide and  $\tilde{\mathbf{L}}$  is comonotonic*

From Proposition 1 we can derive a simple simulation scheme for  $\tilde{\mathbf{L}}$ :

- Simulate  $Z \sim U[0, 1]$
- Compare  $Z$  with the ‘partition of unity’ applied in (4)



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# Comonotonic Default Timing

Denote by  $\tau_{n:m}$  the time until the  $n$ -th default out of  $m$  obligors w.r.t.  $L$ ,

$$\tau_{n:m} = \inf\{t \geq 0 : L_t \geq n/m\}$$

and write  $\tilde{\tau}_{n:m}$  for the time until the  $n$ -th default w.r.t.  $\tilde{L}$ ,

$$\tilde{\tau}_{n:m} = \inf\{t \geq 0 : G_t^{-1}(Z) \geq n/m\} \quad (Z \sim U[0, 1])$$

We now come to a very useful result

**Proposition 2** *The distributions of  $\tau_{n:m}$  and  $\tilde{\tau}_{n:m}$  agree,*

$$\mathbb{P}[\tau_{n:m} \leq t] = \mathbb{P}[\tilde{\tau}_{n:m} \leq t] \quad \text{for all } t \geq 0.$$

*Default timing in the comonotonic approach therefore coincides with default timing according to the original model in line with (1)*

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*Proof.* We start with the comonotonic path  $\tilde{\mathbf{L}}$ . Define numbers

$$q_{t,n:m} = \sum_{k=0}^{n-1} p_{t,k} \quad (t \geq 0; n = 1, \dots, m)$$

Note that  $t \mapsto q_{t,n:m}$  is a strictly decreasing function

Observing  $n$  out of  $m$  defaults at time  $t$  in a comonotonic simulation is equivalent to

$$q_{t,n:m} \leq Z < q_{t,(n+1):m} \tag{5}$$

Therefore, the  $n$ -th default time agrees with the first time  $t$  s.t.  $Z \geq q_{t,n:m}$  for  $Z \sim U[0, 1]$ ,

$$\tilde{\tau}_{n:m} = \inf\{t \geq 0 : Z \geq q_{t,n:m}\}$$

We turn our attention to the original  $\mathbf{L}$

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The statement  $\tau_{n:m} \leq t$  means that at least  $n$  out of  $m$  defaults occurred before or at time  $t$

The likelihood for  $n$  or more defaults in the time interval  $[0, t]$  is given by

$$\begin{aligned} \mathbb{P}[\#\text{defaults} \in \{n, n+1, \dots, m\} \text{ until time } t] &= \sum_{i=n}^m \mathbb{P}[L_t = i/m] \\ &= \sum_{i=n}^m p_{t,i} = 1 - q_{t,n:m} \end{aligned}$$

Combining the pieces we have yield the conclusion

$$\mathbb{P}[\tau_{n:m} \leq t] = 1 - q_{t,n:m} = \mathbb{P}[Z \geq q_{t,n:m}] = \mathbb{P}[\tilde{\tau}_{n:m} \leq t].$$

Therefore, the  $n$ -th default times exactly match in distribution  $\square$

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# Implications of Comonotonicity

We illustrate implications and benefits by means of a one-factor model,

$$p_{t,k} = \mathbb{P}[L_t = k/m] = \binom{m}{k} \int_{-\infty}^{\infty} p(t, y)^k (1 - p(t, y))^{m-k} dN(y), \quad (6)$$

$$p(t, Y) = N \left[ \frac{N^{-1}(p_t) - \sqrt{\varrho} Y}{\sqrt{1 - \varrho}} \right] \quad (7)$$

where  $Y \sim N(0, 1)$ ,  $(p_t)_{t \geq 0}$  is a uniform credit curve, and  $\varrho$  denotes the uniform asset correlation between (uniformly factored) latent variables

$$r_i = \sqrt{\varrho} Y + \sqrt{1 - \varrho} \varepsilon_i \quad (i = 1, \dots, m) \quad (8)$$

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triggering default indicators leading to default quotes

$$L_t = \frac{1}{m} \sum_{i=1}^m \mathbf{1}_{\{r_i < N^{-1}[p_t]\}} \quad (Y, \varepsilon_1, \dots, \varepsilon_m \sim N(0, 1) \text{ iid}) \quad (9)$$

Comparing (1) and (9) suggests to define the default time of any obligor  $i$  by

$$\tau_i = F^{-1}(N[r_i]) \quad (i = 1, \dots, m) \quad (10)$$

where  $r_i$  is defined in (8) and  $F(t) = \mathbb{P}[\tau_i \leq t] = p_t$  (distribution function of  $\tau_i$  on  $[0, \infty)$ )

In the original model we have to simulate  $(m + 1)$  variables  $Y, \varepsilon_1, \dots, \varepsilon_m \sim N(0, 1)$

In the comonotonic approach we only have to simulate one variable  $Z \sim U[0, 1]$

Both methods give the same  $n$ -th to default times.

Comonotonicity, however, works only for a homogenous basket!!

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The comonotonic approach does not capture the fact that in general defaults follow a multinomial pattern.

Again we illustrate this by means of the one-factor model with default quotes  $L_t$  from (9)

Given a time grid  $t_1, \dots, t_q$ , the corresponding default quote *jumps* are given by

$$S_j = L_{t_j} - L_{t_{j-1}} \quad (j = 1, \dots, q; L_0 = 0). \quad (11)$$

Additionally we define a ‘residual jump’

$$S_\infty = 1 - L_{t_q}$$

for the time period  $[t_q, \infty)$

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Based on Equation (7) we define functions

$$g_1(y) = p(t_1, y), \quad g_j(y) = p(t_j, y) - p(t_{j-1}, y), \quad g_\infty(y) = 1 - p(t_q, y), \quad (12)$$

where  $y$  denotes a realization of the factor  $Y$

Obligor  $k \in \{1, \dots, m\}$  admits  $q+1$  possible time intervals capturing her/his default time,

$$[0, t_1), \quad [t_1, t_2), \quad \dots, \quad [t_{q-1}, t_q), \quad \text{or the residual period} \quad [t_q, \infty)$$

Denoting the default time of obligor  $k$  by  $\tau_k$  we see that

$$\mathbb{P}[t_{j-1} \leq \tau_k < t_j \mid Y = y] = g_j(y) \quad \text{and} \quad \mathbb{P}[t_q \leq \tau_k < \infty \mid Y = y] = g_\infty(y) \quad (13)$$

Note that by construction we have

$$\sum_{j=1}^q g_j(y) + g_\infty(y) = 1 \quad (\forall y \in \mathbb{R}) \quad (14)$$

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**Proposition 3** *The distribution of the jump vector  $m \times (S_1, \dots, S_q, S_\infty)$  conditional on  $Y = y$  follows a multinomial distribution  $M[m; g_1(y), \dots, g_q(y), g_\infty(y)]$*

*Proof.* Given  $Y = y$ , the vector  $(\mathbf{1}_{\{\tau_k \in [0, t_1]\}}, \dots, \mathbf{1}_{\{\tau_k \in [t_{q-1}, t_q]\}}, \mathbf{1}_{\{\tau_k \in [t_q, \infty)\}})$  is multivariate Bernoulli with occurrence probabilities  $g_1(y), \dots, g_q(y), g_\infty(y)$ ; see also (14)

Due to the independence of default times conditional on  $Y$ ,

$$\sum_{k=1}^m (\mathbf{1}_{\{\tau_k \in [0, t_1]\}}, \dots, \mathbf{1}_{\{\tau_k \in [t_{q-1}, t_q]\}}, \mathbf{1}_{\{\tau_k \in [t_q, \infty)\}})$$

given  $Y = y$  is a convolution of (independent) multivariate Bernoulli distributions and therefore must follow a multinomial distribution  $M[m; g_1(y), \dots, g_q(y), g_\infty(y)]$

See Equation (8) regarding conditional independence  $\square$

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## Conclusions:

- First analysis of the dependency of defaults in different time buckets
- Comonotonicity provides a powerful tool for efficient and variance-reducing simulation of  $n$ -th-to-default times in baskets and CDOs
- In copula models the dependency of default events in different time slices are close to the comonotonic copula, especially in a one-factor normal copula.
- However, one has to keep in mind that comonotonicity (as a side effect) leads to a lack of precision if the underlying assets are too far from being ‘exchangeable’
- Additionally we pointed out that the multinomial distribution governs the time dependency (in a one-factor) copula model)
- Implication for using asset correlation for the copula of default time models (which is currently an industry standard) requires further investigation, cf Fingers (2000).

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# Literature Remarks

## References

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