

	Apt	Ppt	Size	Spd	Cash flow ①	Cash flow ②
ET	0	3%	3m	$Y_{ET}$	$3m \cdot Y_{ET}$	$2m \cdot Y_{ET}$
MT	3%	7%	4m	$Y_{MT}$	$4m \cdot Y_{MT}$	$4m \cdot Y_{MT}$
ST	7%	100%	93m	$Y_{ST}$	$93m \cdot Y_{ST}$	$93m \cdot Y_{ST}$

①  $PL = 0\%$

②  $PL = 1\%$

③  $PL = 4\%$

Synthetic CDO

ET	0
MT	$3m \cdot Y_{MT}$
ST	$93m \cdot Y_{ST}$



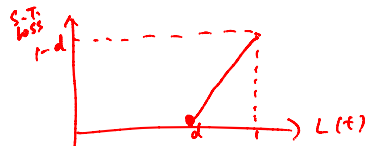
Tranche Payoff  $f^{\pi}$

$$L(t; d, u) = \max(\min(L(t), u) - d, 0)$$

S.T.  $u = 1$

$$L(t; d, 1) = \max(L(t) - d, 0)$$

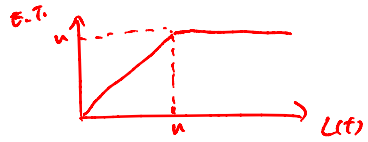
Call  $\max(S_t - K, 0)$



E.T.  $d=0$

$$L(t; 0, u) = \max\{\min(L(t), u) - d, 0\}$$

$$= \min\{L(t), u\}$$



M.T. ①  $L(t) < d < u$

$$L(t; d, u) = \max\{\min(L, u) - d, 0\}$$

$$= \max\{L - d, 0\}$$

$$= 0$$

②  $d < L(t) < u$

$$L(t; d, u) = \max\{\min(L, u) - d, 0\}$$

$$= \max\{L - d, 0\}$$

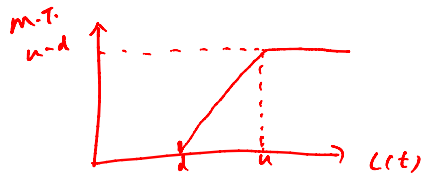
$$= L - d$$

③  $L(t) > u > d$

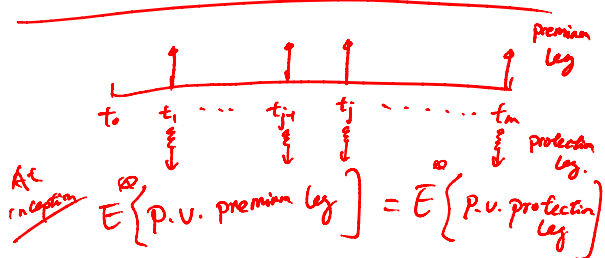
$$L(t; d, u) = \max\{\min(L, u) - d, 0\}$$

$$= \max\{u - d, 0\}$$

$$= u - d$$



CDO pricing



① premium leg

def:  $S$  spot  
 $\Delta$  coupon freq.

$$S \cdot \Delta \cdot \sum_{j=1}^m (u-d) - L(t_j; d, u) Z(t_0, t_j)$$

② protection leg.

$$\sum_{j=1}^m \left[ L(t_j; d, u) - L(t_{j-1}; d, u) \right] Z(t_0, t_j)$$

$$S = f(L(t_j; d, u); u, d, z)$$

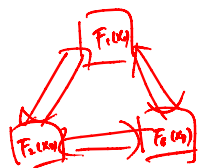
$$E[L(t_j; d, u)]$$

$$L(t_j; d, u) = \max(\min(L(t), u) - d, 0)$$

Loss Dist<sup>n</sup> of portfolio  
 $L(t)$

$$L(t) \sim f(z_1, z_2, \dots, z_n, EAD_i, LGD_i, etc)$$

target  
 $\rightarrow$  Dist<sup>n</sup>:  $f(z_1, z_2, \dots, z_n)$



$$F(x_1, x_2, x_3)$$

$$F(x_1, \dots, x_n)$$

$$= C(F_1(x_1), \dots, F_n(x_n))$$

$$u_i = F_i(x_i)$$

$$= C(u_1, \dots, u_n)$$

$$Y = F(X) \quad \text{any r.v. } X$$

$$Y \sim \text{uniform}(0, 1) \quad \text{hold all dis.}$$

$$C(u_1, \dots, u_n)$$

$$= C(F_1(x_1), \dots, F_n(x_n))$$

Why  $Y = F(X)$  is uniform

$$\textcircled{1} Y \in [0, 1]$$

$$Y = F(X) = \Pr(X \leq x) \in [0, 1]$$

$$\textcircled{2} f_Y(y) = 1$$

$$\begin{aligned} f_X(x) dx &= f_Y(y) dy & Y &= F(X) \\ f_Y(y) &= f_X(x) \frac{dx}{dy} & \Pr(n \leq X \leq n+dy) & \\ &= f_X(x) \frac{1}{\frac{dy}{dx}} = f_X(x) \cdot \frac{1}{F'(x)} = 1 \end{aligned}$$

Sklar's Theorem  $C_{\text{copula}} \Rightarrow F(x_1, \dots, x_n)$

$$C(u_1, \dots, u_n) = \Pr\{U_1 \leq u_1, \dots, U_n \leq u_n\}$$

$$= \Pr\{F_1(X_1) \leq u_1, \dots, F_n(X_n) \leq u_n\} \quad (u_i = F_i(x_i))$$

$$= \Pr\{X_1 \leq F_1^{-1}(u_1), \dots, X_n \leq F_n^{-1}(u_n)\}$$

$$= \Pr\{X_1 \leq x_1, \dots, X_n \leq x_n\}$$

$$= F_n(x_1, \dots, x_n)$$

Copula Density  $f^C$

$$f(x_1, \dots, x_n) = \frac{\partial F_n(x_1, \dots, x_n)}{\partial x_1 \dots \partial x_n}$$

$$= \frac{\partial C(F_1(x_1), \dots, F_n(x_n))}{\partial x_1 \dots \partial x_n} \quad (u_i = F_i(x_i))$$

$$= \frac{\partial C(u_1, \dots, u_n)}{\partial u_1 \dots \partial u_n} \frac{\prod_{i=1}^n \frac{1}{f_i(x_i)} \frac{\partial u_i}{\partial x_i}}$$

$$= C^d(u_1, \dots, u_n) \prod_{i=1}^n \frac{1}{f_i(x_i)}$$

$$C^d(u_1, \dots, u_n) = \frac{f(x_1, \dots, x_n)}{\prod_{i=1}^n f_i(x_i)}$$

## Bi-Variate Gaussian Copula

$$C(u_1, u_2) = \Phi_2(\Phi^{-1}(u_1), \Phi^{-1}(u_2); \rho)$$

$$= \int \int_{-\infty}^{\Phi^{-1}(u_1)} \frac{1}{2\pi\sqrt{1-\rho^2}} \exp\left\{-\frac{x^2-2\rho xy+y^2}{2(1-\rho^2)}\right\} dx dy$$

A  $p_A = 1\%$   $\Rightarrow \Pr(A \text{ B defaults}; \vec{p})$

B  $p_B = 5\%$   $u_1 = 1\% \quad \Phi^{-1}(u_1) = -2.3$   
 $u_2 = 5\% \quad \Phi^{-1}(u_2) = -1.96$

## Simulate default times using Gaussian Copula

$$C(u_1, \dots, u_n) = \Phi_n(\Phi^{-1}(u_1), \dots, \Phi^{-1}(u_n))$$

$$\rightarrow u_i = F_i(z_i) \quad \text{-- marginal dist for } z_i$$

$$\rightarrow x_i = \Phi^{-1}(u_i) = \Phi^{-1}(F_i(z_i))$$

$$C(u_1, \dots, u_n) = \Phi_n(x_1, \dots, x_n; \rho)$$

① generate  $x_i$

$$\left. \begin{array}{l} \text{② } u_i = \Phi(x_i) \\ \text{③ } z_i = F_i^{-1}(u_i) \end{array} \right\} \Rightarrow z_i = F_i^{-1}(\Phi(x_i))$$

Example: Simulate Bi-Variate Gaussian Copula

$$z_i \sim \exp(\lambda_i)$$

$$u_i = F_i(t) = 1 - e^{-\lambda_i t}$$

$$\rightarrow t = -\frac{\ln(1 - u_i)}{\lambda}$$

$\rho$ : Correlation

$z_1, z_2 \sim \text{i.i.d.}$   
 $N(0,1)$

$$\left\{ \begin{array}{l} x_1 = z_1 \\ x_2 = \rho z_1 + \sqrt{1-\rho^2} z_2 \end{array} \right.$$

$$A = \begin{pmatrix} 1 & 0 \\ \rho & \sqrt{1-\rho^2} \end{pmatrix}$$

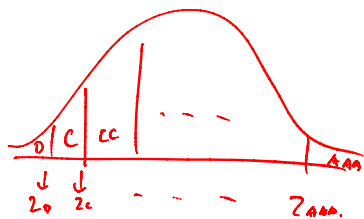
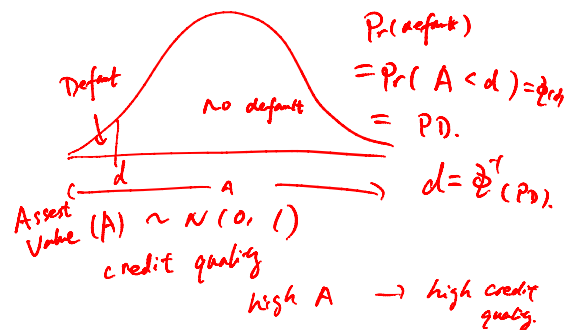
$$\Sigma \rho = A A^T$$

t-copula

$$t = \frac{z}{\sqrt{\frac{x_v}{v}}}$$

$$C(u_1, \dots, u_n) = t_v(t_v^{-1}(u_1), \dots, t_v^{-1}(u_n))$$

$x_v$  - chi-square dis  
d.f. v.




Copula

$$\begin{aligned} C(u_1, u_2) &= \Phi_2(\Phi^{-1}(u_1), \Phi^{-1}(u_2); \rho) \\ &= \Phi_2(\Phi^{-1}(PD_1), \Phi^{-1}(PD_2); \rho) \\ &= \Phi_2(d_1, d_2, \rho) \end{aligned}$$

factor model

$$\begin{aligned} &Pr(A_1 < d_1, A_2 < d_2, \rho_A') \\ &= \Phi_2(d_1, d_2, \rho_A') \\ &\text{if } \rho_A' = \rho \Rightarrow \text{Copula} \equiv \text{factor model} \end{aligned}$$



premium

$$\text{premium} = E[\text{Loss}]$$

$r = 0.$