

## CQF Lecture 5.6 CDO and Copula Models

### Exercises

1. A **synthetic CDO**, based on balance sheet information, is structured as follows:

Assets:	125 single-name CDS
Principal:	0.8 million (per name)
Maturity:	5 years
CDS spread:	200 bps
Payments:	Act/360 quarterly in arrears

Tranche	Attachment point	Expected Loss	Fair Spread	Rating
Senior	7%-10%	0.002%	L+45	AAA
Class A	5%-7%	0.1%	L+70	AA-
Class B	2%-5%	2.3%	L+20	BBB-
Class C	0%-2%	26.27%	Excess spread	NR

Table 1: CDO Capital Structure

- (a) Holders of which tranche are long correlation and why? Which tranche is the most sensitive to changes in default correlation?
- (b) What about exposure of mezzanine noteholders to default correlation?
- (c) Assuming 0% recovery, how many defaults should occur before Senior tranche experiences a capital loss? If we assume 40% recovery how much more protection does that afford to Senior noteholder?
- (d) A downgrade is triggered when the entire Equity tranche is lost. Assuming 0% recovery, how many defaults should occur before the implied ratings are downgraded.

2. Consider a random variable  $X$  that provides information about default time, conditional on intensity parameter  $\theta$ .  $X$  follows the exponential distribution with *cdf*:

$$\Pr(X \leq x|\theta) \equiv F(x|\theta) = 1 - e^{-\theta x}$$

Assuming that intensity follows Gamma distribution, i.e.,  $\theta \sim \Gamma(\alpha, \beta)$  with *pdf*:

$$g(\theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta\theta}$$

Use the result below to show that the unconditional **marginal cdf** of  $X$  follows Pareto distribution—that is,

$$\Pr(X \leq x) \equiv F(x) = 1 - \left(1 + \frac{x}{\beta}\right)^{-\alpha}$$

**Hint:** In order to recover a *cdf* for the unconditional distribution, we need to integrate over the conditional distribution as follows:

$$F(x) = \int_0^\infty F(x|\theta) g(\theta) d\theta.$$

3. Consider two identically distributed variables  $X_1$  and  $X_2$ . When conditioned on  $\theta$ , same as in the previous exercise, they are independent. Their unconditional joint *cdf* is

$$\begin{aligned} F(x_1, x_2) &\equiv \Pr(X_1 \leq x_1, X_2 \leq x_2) \\ &= 1 - \Pr(X_1 > x_1) - \Pr(X_2 > x_2) + \Pr(X_1 > x_1, X_2 > x_2) \end{aligned}$$

In a practical context,  $X_1$  and  $X_2$  represent default times  $\tau_1$  and  $\tau_2$  respectively, so that  $F(t_1, t_2) = \Pr(\tau_1 \leq t_1, \tau_2 \leq t_2)$  but let's continue working in 'random variable X' notation.

- (a) Express the **joint cdf**  $F(x_1, x_2)$  as a function of the isolated marginal *cdfs*  $F(x_1)$  and  $F(x_2)$  (also called 'marginals').

**Hint:** We can spot  $F(x) = 1 - \Pr(X > x)$  but the unconditional term is unknown:  $\Pr(X_1 > x_1, X_2 > x_2)$ ? We need to calculate it by integration over the product of  $\Pr(X_1 > x_1 | \theta) \Pr(X_2 > x_2 | \theta) g(\theta)$ , treating *conditional*  $X_1$  and  $X_2$  as independent.

- (b) By substituting uniform variables  $u_1, u_2$  instead of marginals  $F(x_1)$  and  $F(x_2)$  show that the associated **copula function** is

$$\begin{aligned} C(u_1, u_2) &\equiv \Pr(U_1 \leq u_1, U_2 \leq u_2) \\ &= u_1 + u_2 - 1 + \left((1 - u_1)^{-\frac{1}{\alpha}} + (1 - u_2)^{-\frac{1}{\alpha}} - 1\right)^{-\alpha} \end{aligned}$$

4. Consider a **copula function** of the Archimedean family

$$C(u_1, u_2, \dots, u_n) = \phi^{-1}(\phi(u_1) + \phi(u_2) + \dots + \phi(u_n))$$

Given the copula generator

$$\phi(u) = -\ln\left(\frac{e^{-\alpha u} - 1}{e^{-\alpha} - 1}\right)$$

show that the copula function can be expressed explicitly as

$$C(u_1, u_2, \dots, u_n) = -\frac{1}{\alpha} \ln \left[ 1 + \frac{\prod_{i=1}^n (e^{-\alpha u_i} - 1)}{(e^{-\alpha} - 1)^{n-1}} \right]$$

Identify this copula function by name. What does parameter  $\alpha$  represent?

5. **Copula functions can price multi-asset options.** There is movement co-dependence among assets that can be modelled by a suitable copula. Copula function is equivalent to joint probability  $C(u_1, u_2) \equiv F(x_1, x_2)$  by Sklar theorem. Because of correlation, one cannot use multiplication rule of simple probabilities to find out the  $Pr$  of the joint event.

We need the risk-neutral probability of an option being *in the money* at maturity, which is  $u = N(d_2)$  for a European call. This risk-neutral probability can also be obtained by Monte Carlo.

For example, bi-variate European digital **put** pays one unit of currency if two underlying assets are both below a pair of strikes at maturity. Consider a simplified scenario of two identical assets and the same strikes:

$$T = 1, r = 0, K_{1,2} = 100, \sigma_{1,2} = 20\%, S_{1,2} = S_0 e^{\frac{1}{2}\sigma^2}, S_0 = 100,$$

Use Frank Copula function with  $\alpha = 5$  to calculate the price of the bi-variate digital put.

$$C(u_1, u_2, \dots, u_n) = -\frac{1}{\alpha} \ln \left[ 1 + \frac{\prod_{i=1}^n (e^{-\alpha u_i} - 1)}{(e^{-\alpha} - 1)^{n-1}} \right]$$