

Derivative Securities - Fall 2012 - Section 12

The relevant parts of Hull are:

- chapter 8, on structured credit products + the recent financial crisis
- sections 23.9-23.10, on Gaussian copula approach to default correlations
- section 24.3 about CDS indexes
- sections 24.8-24.10 on CDO's

In 211 we focused on bonds issued by a specific corporation, or CDS on a single entity. Therefore only the prob of default by this entity was relevant.

In multiname setting, correlation b/w defaults is also important.

Example: Suppose you hold bonds issued by 10 companies, & each has default prob of 1% in 1st yr. If independent, then

$$\begin{aligned} \text{prob of exactly } k \text{ defaults in 1}^{\text{st}} \text{ yr} \\ = \binom{10}{k} (0.01)^k (0.99)^{10-k} \end{aligned}$$

so, for example, prob that all default

is $(.01)^{10} = 10^{-20}$. (Exceedingly small!)

If fully correlated, then either there's no default or they all default; in this case

prob that all default in 1st year = .01 (not so small!)

Why do we care?

- (1) A bank with many credit exposures must estimate its "value at risk." This amounts to estimating the likelihood of a very large loss (due to defaults) within a specific time period.
- (2) An investor may seek to protect itself in a general, diversified way from deterioration of credit environment; or to protect by selling such protection. The standard tool is a basket CDS or index CDS contract.

Index CDS works like this:

- there's a specific list of companies (eg CDX NA IG corresp to list of 125 investment grade companies in North America)

- contract behaves like an (equally-weighted) CDO on each company. For example:

1st default \Rightarrow seller of protection pays $L(1-R) \cdot \frac{1}{125}$ to purchaser, and "spread payments" are reduced to $\frac{124}{125}$ of initial amount

2nd default \Rightarrow seller pays another $L(1-R) \cdot \frac{1}{125}$ to purchaser, + "spread payments" are reset to $\frac{123}{125}$ of initial amount.

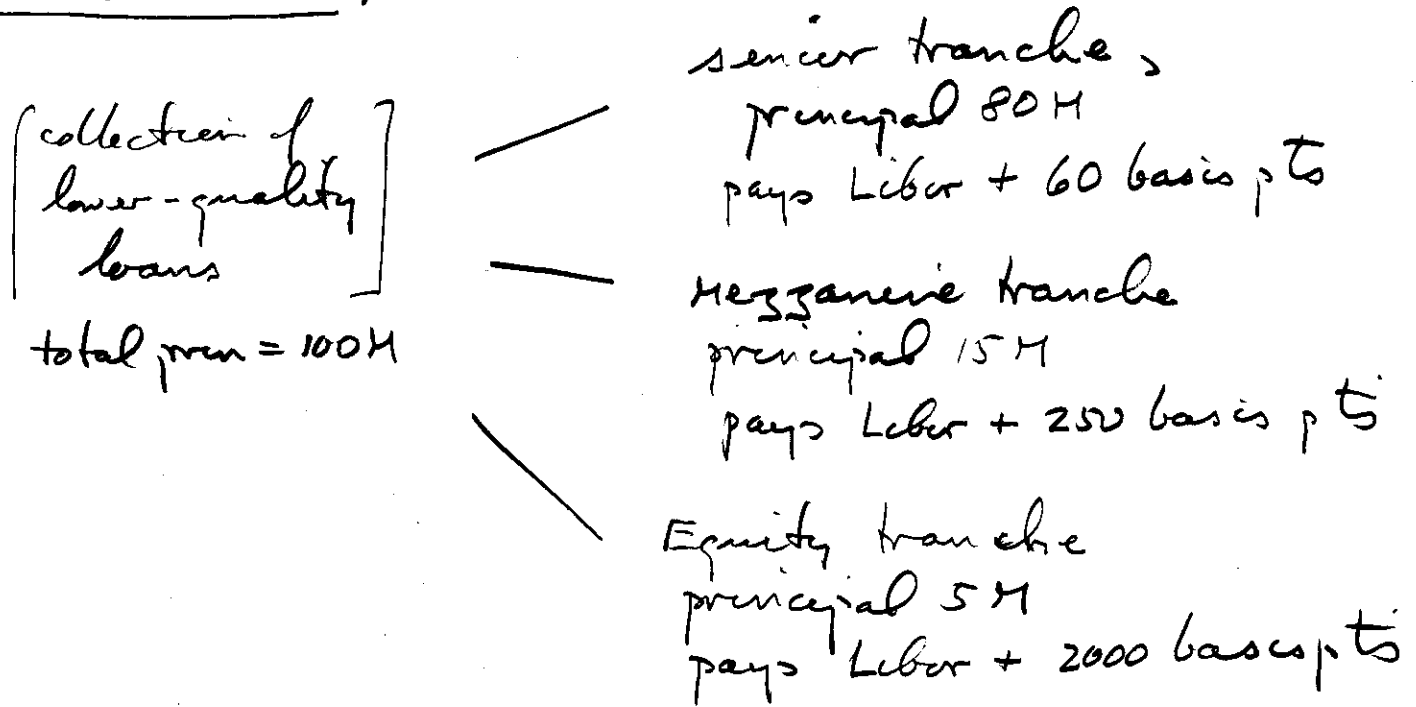
To value this, we need expected # defaults at each payment date.

Note: recent huge trading loss by JPMorgan Chase involved contracts of this type. (Why the big loss? Huge trades \Rightarrow they moved the market \Rightarrow price paid was artificially high \Rightarrow after a while they had a huge loss.)

(3) Before 2007, CDO's (collateralized debt obligations) backed by mortgages were widely

created + traded. Their purpose: to create high-quality bonds out of lower-quality loans (eg subprime). Also used for credit card debt, etc. Much less-used since 2008

How it works:



(Hull, fig 8.1); here

equity tranche bears 1st 5% of defaults (after which it's dead);

mezzanine tranche bears next 15% of defaults (then it too is dead)

senior tranche starts seeing losses if defaults exceed 20%.

in practice there could be many more tranches

Typically (based on hypoth of not-too-large correlation) senior tranche would be high quality even if underlying instruments aren't. So: it's a way to draw capital that otherwise wouldn't have been available.

Worked badly for many reasons:

- correlations in 2007-2008 were much greater than mkt assumed
- creator of underlying loans has no incentive to be honest or maintain quality

④ Also pre-2007: people also created synthetic CDO's, which were essentially tranch versions of CDS's on a portfolio.

eg a) equity tranche earns highest spread but bears full face of initial defaults (up to 5% say)

b) mezzanine tranche earns intermed spread + bears face of defaults beyond 5% up to 15% (say)

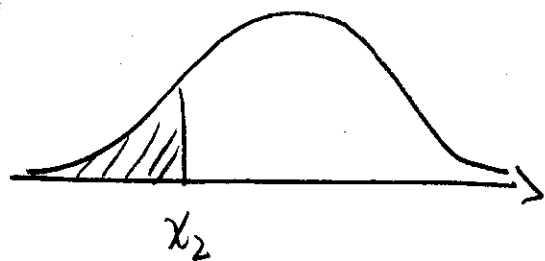
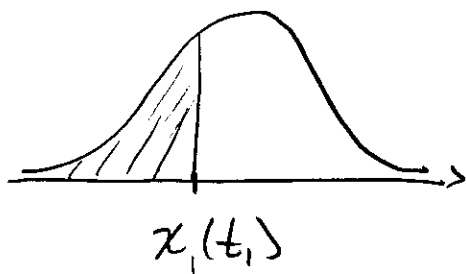
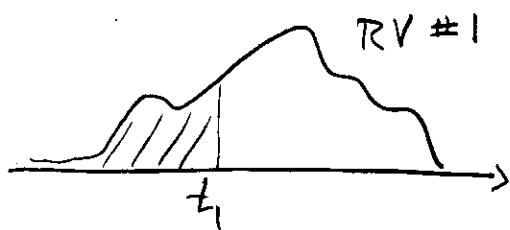
c) senior tranche gets smallest spread + bears face of defaults only after 20%.

why? Basically same reasons as regular CDS's in a portfolio. But much more complex to price or hedge.

Why were they popular? Mostly as a speculative tool, I think!

Stepping back: tranchet instruments are no longer so popular, but correlations b/w defaults are still important for basket or index CDS, or estimating VAR on a debt portfolio.

Basic tool: Gaussian copula permits disc'n of correlation b/w two unrelated, non-Gaussian RV's.



define $x_1(t_1)$ by
 $N(x_1(t_1)) = \text{prob that } RV\#1 \leq t_1$

+ similarly $x_2(t_2)$ (using RV#2). Now assume $x_1 + x_2$ are bivariate normal with corr. ρ .

Let's apply this viewpoint to estimate default risk involving a pool of entities, assuming for simplicity:

- homogeneity (all entities are equivalent)
- a one-factor model (to be explained below)

(Think of the common factor as the "state of the economy.")

- The pool is large (permitting us to use law of large nos)

Goal is an expression for fraction of the pool that has defaulted by time t . (or just the expected number of defaults by time t), given inputs

- prob that a single entity defaults by time t is given, say $D = D(t)$
- correlation ρ (and Gaussian copula).

Let x_j be the standard Gaussian assoc to the j th entity. Then

- j^{th} entity defaults by $t \Leftrightarrow x_j < N^{-1}(D)$
- $x_j = \sqrt{\rho} F + \sqrt{1-\rho} Z_j$ (the one-factor hypothesis)

where $F + Z_j$ are indep standard Gaussians
 (Note that $E x_j = 0$, $E(x_j^2) = 1$, $E(x_j x_k) = \rho$.)

So: if value of F is fixed, then

$$\text{name } j \text{ defaults} \Leftrightarrow x_j < N^{-1}(D)$$

$$\Leftrightarrow Z_j < \frac{N^{-1}(D) - \sqrt{\rho} F}{\sqrt{1-\rho}}$$

$$\Leftrightarrow \text{prob of default is } N\left(\frac{N^{-1}(D) - \sqrt{\rho} F}{\sqrt{1-\rho}}\right)$$

To find prob that any indiv. credit defaults we're thus left to do a numerical integration

$$\text{prob} = \int N\left(\frac{N^{-1}(D) - \sqrt{\rho} f}{\sqrt{1-\rho}}\right) \cdot \frac{1}{\sqrt{2\pi}} e^{-f^2/2} df$$

If pool is large, it's a decent approxn to identify this with the fraction of the pool that has defaulted by time t .

Uses of this:

a) for estimating VAR:

what is the likelihood that 10% of the pool defaults by t ?

\Updownarrow (under large-pool approx)

what is the likelihood that

$$N\left(\frac{N^{-1}(D) - \sqrt{p} F}{\sqrt{1-p}}\right) > .1$$

\Updownarrow
what is the likelihood that a Gaussian F has

$$F < \frac{N^{-1}(D) - \sqrt{1-p} N^{-1}(.1)}{\sqrt{p}}$$

b) for pricing a CDS on an index:

protection purchaser pays:

$$L \frac{p}{f} E\left[\frac{\text{fraction of names}}{\text{still alive at } t_i}\right] \quad \text{at } t_i$$

and receives

$$L(1-R) \left(E\left[\frac{\text{fraction alive}}{\text{at } t_{i-1}}\right] - E\left[\frac{\text{fraction alive}}{\text{at } t_i}\right] \right)$$

due to defaults, plus "accrued interest"

$$L \frac{1}{2\delta} (E[\text{fraction alive at } t_{j-1}] - E[\text{fraction alive at } t_j])$$

(at time $(t_{j-1} + t_j)/2$, by the usual fudge).

Note that calcs of $E[\text{fraction alive at } t_j]$ are all "the same" (as explained above), except for use of different values of

$D_t = \text{prob that any single name has defaulted by time } t$.

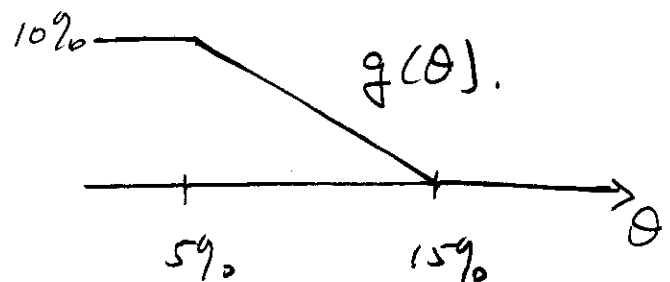
c) valuation of a tranched CDO is only a little different. Let

$$\theta_t(F) = N\left(\frac{N^{-1}(D_t) - \sqrt{\rho} F}{\sqrt{1-\rho}}\right)$$

= fraction of defaults by time t , given F (under large-pool approx)

Then eg for mezzanine tranche in our example, protection purchaser's payment at time t_j is

$$L \frac{1}{2\delta} g(\theta_{t_j})$$



which gets valued as

$$L \frac{1}{T} E[g(\theta_{t_i})] B(0, t_i)$$

↑
Gaussian expectation,
since θ_t is a ln of the Gaussian RV F

and protection purchaser receives (at $\frac{t_i + t_{i+1}}{2}$)
 $L(1-R) + L \frac{1}{2T}$ times $g(\theta_{t_{i-1}}) - g(\theta_{t_i})$,
 valued using

$$E[g(\theta_{t_{i-1}})] - E[g(\theta_{t_i})]$$

(were expectations wrto Gaussian F).

For all these calcs, choice of f is crucial. Lesson of JPM Chase blowup and also 2007-8 crisis: inferring f from mkt data is dangerous (The market could have wrong beliefs, or be skewed by supply/demand issues, for example.)