Credit risk and credit sensitive instruments
Credit ratings and credit migration models
Introduction to structural credit models

Credit Risk Models

1. Credit Migration and Structural Models

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Outline

- Credit risk and credit sensitive instruments
- 2 Credit ratings and credit migration models
- 3 Introduction to structural credit models

Credit risk

The subject of this course is *credit risk* and its modeling.

- Credit risk is the potential that a borrower or a counterparty on a transaction will
 fail to meet its obligations in accordance with agreed terms. This failure is
 referred to as a default.
- The exact definition of what constitutes an event of default (bankruptcy, debt restructuring, failure to pay, etc.) is governed by strict legal language, and is constantly ammended.

Types of credit risk:

- Consumer (eg. a consumer fails to make a payment due on a mortgage loan, car loan, credit card, or line of credit).
- Corporate (eg. a company bond issuer is unable to make a payment on a coupon or principal payment when due).
- Sovereign (eg. a government bond issuer is unable to make a payment on a coupon or principal payment when due).



Credit risk

- Like many other types of risk (FX risk, interest rate risk, prepaymant risk, ...), credit risk can be traded through a variety of credit sensitive instruments.
- These instruments fall into two broad categories:
 - Cash instruments (e.g. corporate coupon bonds, sovereign bonds, asset backed securities).
 - Derivatives.

A basic credit sensitive instrument is a risky coupon bond.

- Bond holder receives a coupon in excess of the risk free rate in return for assuming the credit risk of the issuing entity
- The higher the default risk, the bigger the coupon.

Consider first a riskless coupon bond with face value \$1 and annual coupon C.

- A future cash flow at time T is discounted on the risk free interest rate curve with the discount factor P(T).
- The price of the bond is the sum of the present values of all coupon payments and the final principal repayment:

$$Price = \sum_{j=1}^{n} P(T_j)C + P(T_n).$$

The risk embedded in this bond is entirely interest rate risk (market risk).



This valuation formula has to be modified in the presence of credit risk:

- The issuer may stop making coupon payments.
- The issuer may fail to repay the face value of the bond.

In order to account for this we introduce the following concepts:

- By S(T) we denote the probability that the payment at time T will be made; we refer to it as the survival probability.
- The quantity Q(T) = 1 S(T) is the *default probability*; it is the probability that the default event will occur some time between now and T.
- By R(T) we denote the fraction of the face value that the bond holder will receive upon default at time T; we refer to it as the recovery value.

The bond valuation formula for a credit sensitive bond has to be modified as follows:

- Each coupon payment at time T_j is multiplied by the survival probability.
- The price of the bond is the sum of the present values of all coupon payments and the final principal repayment:

Price =
$$\sum_{j=1}^{n} (S(T_j)P(T_j)C + (S(T_{j-1}) - S(T_j))P(T_{j-1})R(T_{j-1})) + S(T_n)P(T_n).$$

- The risk has now two components: interest rate risk (market risk) and credit risk
- Note that the natural discount factor for cash flows in a risky bond is he risky discount factor, P(T) = S(T)P(T):

Price =
$$\sum_{j=1}^{n} (\mathcal{P}(T_j)C + (1 - S(T_j)/S(T_{j-1}))\mathcal{P}(T_{j-1})R(T_{j-1})) + \mathcal{P}(T_n).$$



- The quantity $1 S(T_j)/S(T_{j-1})$ is the *conditional* probability of default during $[T_{j-1}, T_j]$ provided that there was no default prior to T_{j-1} .
- The value of the risky discount factor takes into account the likelihood of default.
- If the discounting is done on a constant riskless rate, $P(T) = e^{-rT}$.
- Assume that $\mathcal{S}(T) = e^{-\lambda T}$, with λ constant. We see then that $\mathcal{P}(T) = e^{-(r+\lambda)T}$, and so λ is the extra discounting to accommodate for the probability of default.
- λ is referred to as the credit spread.

Credit derivatives

Credit derivatives are financial contracts that allow to transfer the *credit risk* of a *reference entity* from one counterparty to another.

- The counterparties on a credit derivative are a protection seller (the party that assumes the credit risk) and a protection buyer.
- Credit risk of the counterparties on credit derivatives (as well as on any other financial transaction) came to the forefront of interest after the spectacular failure of AIG Financial Products in 2008 followed up by the US government bailout.
- We will address the counterparty risk later in the course, for now assume no counterparty risk.
- A precise definition of a default event is largely a legal matter and it will not concern us here.

Credit derivatives

Two main categories of credit derivatives are:

- Funded credit derivatives. These are bilateral contracts between two
 counterparties, each of which is responsible for making its contractual payments
 (i.e. payments of premiums and any cash or physical settlement amount) itself
 without recourse to other assets.
- Unfunded credit derivatives. Under a contract of this type, the protection seller
 makes an initial payment that is used to settle any potential future credit events.
 Note that the protection buyer is exposed to the credit risk of the protection seller
 this is an example of counterparty credit risk that we will discuss in detail later.

Credit derivatives

Unfunded credit derivative products include the following products:

- Single name credit default swap (CDS)
- Total return swap (TRS)
- CDS index products (CDX, iTraxx)
- Constant maturity credit default swap (CMCDS)
- First to default credit default swap
- Portfolio credit default swap
- Credit default swap on asset backed securities (ABS CDS)
- Credit default swaption
- Recovery lock
- Credit spread option

Funded credit derivative products include the following products:

- Credit-linked note (CLN)
- Synthetic collateralized debt obligation (SCDO)
- Constant proportion debt obligation (CPDO)
- Synthetic constant proportion portfolio insurance (Synthetic CPPI)



Credit modeling

- Default events are relatively infrequent.
- Historical default data are available for past events.
- Moody's Investor Services has an extensive database of corporate defaults, various companies collect data on consumer credit.
- The central questions of credit risk modeling are:
 - Probability of default
 - Credit spread
 - Recovery rate

Approaches to credit risk modeling:

- Scoring. Each entity is assigned a score. The scores can be produced on the basis of a mathematical model, purely judgmentally, or through a combination of both. Typically, credit scores are not associated with specific forecasts of default probabilities. Examples are Altman's Z-score (for corporations) and FICO scores (for consumers).
- Credit migration models. Each entity is assigned a credit rating from "riskless" to "defaulted", and probabilities of transitions between these ratings are estimated. This approach is taken by credit rating agencies such Moody's, S&P, and Fitch.
- Structural models. In this approach, the default probability is modeled through a Black-Scholes style stochastic model.
- Reduced form models. This approach uses the conditional probability of default, called the intensity or hazard rate, as the starting point.

Credit migration matrix

- Credit rating agencies periodically review the credit quality of each issuer, and may change their rating.
- Such a change will likely impact the perception of credit worthiness of the name.
- Historical observations of such changes are summarized in the form of a credit migration matrix.
- For example, here is a one-year credit migration matrix from Moody's data [1]:

	Grade	Aaa	Aa	Α	Baa	Ва	В	D
Π=	Aaa Aa A Baa	91.027% 7.003% 2.000% 0.299%	6.998% 85.823% 10.865% 0.999%	1.003% 5.997% 80.251% 3.798%	0.650% 0.704% 6.159% 90.624%	0.238% 0.266% 0.397% 3.680%	0.059% 0.147% 0.238% 0.400%	0.025% 0.060% 0.090% 0.200%
	Ba B D	0.151% 0.007% 0.000%	0.902% 0.047% 0.000%	3.701% 0.217% 0.000%	7.002% 0.405% 0.000%	72.855% 8.898% 0.000%	12.889% 78.849% 0.000%	2.500% 11.576% 100.000%

 The probabilities in Moody's transition matrix are historical ("P-measure") probabilities. A matrix with risk neutral ("Q-measure") probabilities would typically lead to higher default probabilities.

- We consider a *finite state* space E = {E₁,..., E_K}, and a sequence of random variables X_n, with X_n ∈ E. Such a sequence is called a *discrete-time stochastic process*.
- A Markov chain is a discrete-time stochastic process such that

$$P(X_{n+1} = e | X_n = e_n, X_{n-1} = e_{n-1}, \dots, X_0 = e_0) = P(X_{n+1} = e | X_n = e_n).$$

- The property above is called the *Markov property*.
- A Markov chain is called *homogeneous*, if the conditional probability $P(X_{n+1} = e | X_n = e_n)$ is independent of the time n.
- We call the matrix

$$\Pi_{ij} = \mathsf{P}(X_{n+1} = E_i | X_n = E_i)$$

the transition matrix. For a homogeneous process, Π is independent of n.



- The probability distribution of the initial state is denoted by $p_0(i)$ and is given by $p_0(i) = P(X_0 = E_i)$.
- Using Bayes' rule and the Markov property we find that the m-step transition probability

$$\Pi(m)_{ij} = \mathsf{P}(X_{n+m} = E_j | X_n = E_i)$$

is given by

$$\Pi(m) = \Pi^m$$
.

• The probability distribution p_n at time n is given by

$$p_n^{\mathrm{T}} = p_0^{\mathrm{T}} \Pi^n$$
.

- The credit migration matrix is a starting point to a Markov chain model of credit ratings.
- Assume that there are K "live" (non-default) states, and a (K + 1)-st default state
 D. The transition probabilities between the states are given by the entries of the
 credit migration matrix.
- Notice that *D* is an absorbing state: the probability of transitioning out of it is zero.
- By the Markov property, the *n*-year transition matrix is given by:

$$\Pi_n = \Pi^n$$
,

for integer n (of course, $\Pi^1 = \Pi$).

 The transition matrix does not have to be built out of historical data (it can be produced e.g. by a model).



 For convenience, one may choose to work with a continous-time Markov chains X_t. We than have:

$$P(X_{t_{n+1}} = e | X_{t_n} = e_n, X_{t_{n-1}} = e_{n-1}, \dots, X_{t_0} = e_0) = P(X_{t_{n+1}} = e | X_{t_n} = e_n),$$

for all choices of observation times $t_0 < \ldots < t_{n-1} < t_n < t_{n+1}$.

 In the continuous time framework, it is convenient to work with a so called generator matrix Λ. For a homogeneous Markov chain, it is defined by

$$\Pi(t) = e^{\Lambda t}$$

$$\triangleq \sum_{k=0}^{\infty} \frac{(t\Lambda)^k}{k!}$$

(we use the notation $\Pi(t)$ rather than Π_t).

- Notice that $\Pi(1) = e^{\Lambda} = \Pi$.
- Calibrating the generator matrix to historical data (or to a model) allows us to calculate transition probabilities for all time horizons.

 For example, the generator matrix used to produce the credit migration matrix above is given by:

$\Lambda =$	Grade	Aaa	Aa	Α	Baa	Ва	В	D
	Aaa Aa A	-0.0971753 0.0788 0.0182	0.0788 -0.16071 0.13035	0.0087 0.072 -0.22666	0.0065 0.0051 0.0718	0.0026 0.0029 0.0031	0.0004 0.00143 0.0025	0.000175 0.000482 0.000705
	Baa	0.0025	0.0083	0.0433	-0.10179	0.0452	0.001	0.001491
	Ba	0.0009	0.0079	0.0463	0.0846	-0.32923	0.1711	0.018426
	В	0	0	0	0	0.1182	-0.24746	0.129255
	D	0	0	0	0	0	0	0

 Notice that the transition matrix satisfies the following ODE (forward Kolmogorov equation):

$$\frac{d\Pi\left(t\right)}{dt}=\Lambda\Pi\left(t\right),$$

with $\Pi(0) = I$.



It is possible (and, sometimes, necessary) to extend the Markov chain model to time dependent generators $\Lambda(t)$.

The starting point is the ODE:

$$\frac{d\Pi(t)}{dt} = \Lambda(t)\Pi(t).$$

- In general (when the matrices Λ (t) do not commute for different values of t), it is impossible to solve it in closed form.
- Numerical solution, satisfying $\Pi(0) = I$, is given by:

$$\Pi(t) = \operatorname{T} \exp \int_0^t \Lambda(s) \, ds$$

$$\triangleq \lim_{n \to \infty} \prod_{j=1}^n \left(I + \Lambda(\frac{(j-1)t}{n}) \frac{t}{n} \right).$$

Structural credit models

- Structural models of credit risk historically emerged first, they originated with Merton's paper [4] applying the Black-Scholes framework to the problem of default.
- The structural approach allows for relating the economic context, such as balance sheet information, to understand the definition of the event of default.
- We begin with the discussion of Merton's model.
- It is a single period (extremely simplified) stochastic model of a firm with time horizon T.

- The capital structure of the firm consists only of
 - (i) debt whose value is B, and
 - (ii) equity whose value is S.
- The debt is a zero coupon bond with face value F which matures at T.
- The equity consists of non-dividend paying shares.
- The total value V (t) of the firm at time t is thus

$$V(t) = B(t) + S(t).$$

• At time T, the firm is solvent if $V(T) \ge F$, and it is in default, if V(T) < F.



The debt holders' payoff at T is

$$B(T) = \min(F, V(T))$$

= $V(T) - (V(T) - F)^+,$

where we have used the notation $x^+ = \max(x, 0)$.

• The equity holders' payoff at T is

$$S(T) = (V(T) - F)^+$$
$$= V(T) - B(T).$$

We assume that the value of the firm follows a lognormal diffusion:

$$dV(t) = V(t) (\mu dt + \sigma_V dW(t)),$$

where μ is the firm's rate of return, and σ_V is the volatility of the value process.



- We assume that the pricing is done in the risk neutral measure Q associated with a riskless rate r.
- From the Black-Scholes theory we get

$$S(t,T) = V(t) N(d_1) - Fe^{-r(T-t)} N(d_2),$$

$$B(t,T) = Fe^{-r(T-t)} N(d_2) + V(t) N(-d_1).$$

• Here N(x) is the cumulative normal distribution function, and

$$\begin{aligned} d_1 &= \frac{\log \frac{V(t)}{F} + \left(r + \frac{1}{2}\sigma_V^2\right)(T-t)}{\sigma_V \sqrt{T-t}}, \\ d_2 &= d_1 - \sigma_V \sqrt{T-t}. \end{aligned}$$

We can now calculate the credit metrics defined earlier within Merton's model.

Survival probability:

$$S(T) = Q(V(T) \ge F)$$

= $N(d_2)$.

Recovery rate:

$$R(t) = \frac{V(t) N(-d_1)}{Fe^{-r(T-t)}N(-d_2)}$$
.

Implied volatility for the stock process:

$$\sigma_{S} = \frac{\partial S(t)}{\partial V(t)} \frac{V(t)}{S(t)} \sigma_{V}$$
$$= N(d_{1}) \frac{V(t)}{S(t)} \sigma_{V}.$$

Credit spread: Recall that the credit spread s is defined by

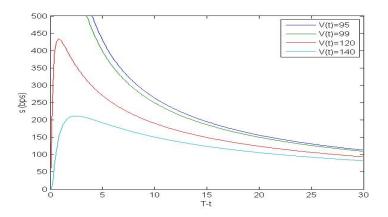
$$B(t) = Fe^{-(r+s)(T-t)}.$$

Hence

$$s = -\frac{1}{T - t} \log \frac{B(t, T)}{F} - r$$

= $-\frac{1}{T - t} \log \left(N(d_2) + e^{r(T - t)} \frac{V(t)}{F} N(-d_1) \right).$

The graph below shows the credit spreads for different values $V\left(t\right)$ implied by Merton's model.



- One of the problems with Merton's original model is that the default can occur
 only at one time, namely the time horizon T.
- A natural extension, due to Black and Cox [3] assumes that there is a *safety* covenant in place, and the firm is obliged to reimburse its debt holders, if V(t) falls below a predefined safety level.
 - Such a safety level is given by a continuous barrier H(t) and the firm defaults as soon as V(t) ≤ H(t). From the point of view of option pricing, the price of the bond is related to the price of a barrier option.
 - lacktriangle To describe this, let au be the default time of the firm. Then

$$\tau = \inf\{t > 0 : V(t) \le H(t)\}.$$

• The choice of the barrier H(t) should be financially meaningful. Observe that if H(t) > F, for all t, then the debt holders would always be protected from losses. Therefore, we require that at the very least $H(T) \le F$.



A natural choice could thus be

$$H(t) = H_0 e^{at},$$

where $H_0 < Fe^{-aT}$. In the following we use this form of the barrier.

As before, we assume that V (t) follows the lognormal process

$$dV(t) = V(t) (\mu dt + \sigma_V dW(t)),$$

and so, under the risk neutral measure Q,

$$V(t) = V_0 \exp\left(\sigma_V W(t) + \left(r - \frac{\sigma_V^2}{2}\right)t\right).$$

As a consequence,

$$\left\{V\left(t\right) \leq H\left(t\right)\right\} = \left\{\sigma_{V}W\left(t\right) + \left(r - \sigma_{V}^{2}/2 - a\right)t \leq \log\frac{H_{0}}{V_{0}}\right\}.$$



Consequently,

$$\begin{aligned} \mathsf{Q}(\tau \leq t) &= \mathsf{Q}\Big(\min_{s \leq t} \frac{\mathsf{V}(s)}{\mathsf{H}(s)} \leq 1\Big) \\ &= \mathsf{Q}\Big(\min_{s \leq t} \big(\mathsf{W}(s) + \mathsf{b}s\big) \leq \frac{1}{\sigma_{\mathsf{V}}} \log \frac{\mathsf{H}_0}{\mathsf{V}_0}\Big), \end{aligned}$$

where
$$b = (r - \sigma_V^2/2 - a)/\sigma_V$$
.

 From the mathematical point of view, this is the classic problem of the first passage time for the Brownian motion with drift. We will digress now to briefly present the key ideas of the solution to this problem.

Consider the following stochastic processes:

$$M(t) = \max_{s \le t} (W(s) + bs),$$

$$m(t) = \min_{s < t} (W(s) + bs).$$

- We are concerned with determining the probability distributions of M(t) and m(t).
- To this end, we define the following function:

$$\Phi(\alpha, \beta, b, t) = N\left(\frac{\alpha - bt}{\sqrt{t}}\right) - e^{2b\beta}N\left(\frac{\alpha - 2\beta - bt}{\sqrt{t}}\right),\,$$

for $\alpha < \beta$.



- The probability distribtions of M(t) and m(t) can be expressed in terms of the function Φ as follows.
- For any α and $\beta \geq 0$,
 - (i) $P(M(t) \le \beta, W(t) + bt \le \alpha) = \Phi(\min(\alpha, \beta), \beta, b, t),$
 - (ii) $P(m(t) \ge -|\beta|, W(t) + bt \ge \alpha) = \Phi(-\min(\alpha, \beta), \beta, -b, t),$
 - (iii) $P(M(t) \leq \beta) = \Phi(\beta, \beta, b, t),$
 - (iv) $P(m(t) \ge -|\beta|) = \Phi(\beta, \beta, -b, t)$.

Note first that (i) implies the other three formulas by changing the sign of $W\left(t\right)$ and noting that

$$m(t) \leq W(t) + bt \leq M(t)$$
.

The proof of (i) proceeds in two steps.

• Consider first b = 0, and $\alpha \le \beta$. We have:

$$P(M(t) \leq \beta, W(t) \leq \alpha) = P(W(t) \leq \alpha) - P(M(t) \geq \beta, W(t) \leq \alpha)$$

$$= P(W(t) \leq \alpha) - P(M(t) \geq \beta, W(t) \geq 2\beta - \alpha)$$

$$= P(W(t) \leq \alpha) - P(W(t) \geq 2\beta - \alpha)$$

$$= N(\alpha/\sqrt{t}) - N((\alpha - 2\beta)/\sqrt{t})$$

$$= \Phi(\min(\alpha, \beta), \beta, b, t).$$

 The case of b ≠ 0 is now a consequence of Girsanov's theorem, since the Radon-Nikodym derivative is equal to

$$\frac{d\mathsf{P}^b}{d\mathsf{P}}\Big|_t = e^{-bW(t) - \frac{1}{2}\,b^2t}.$$



Therefore,

$$\begin{split} \mathsf{E}^{\mathsf{P}} \big[\mathbf{1}_{M(t) \leq \beta} \mathbf{1}_{W(t) + bt \leq \alpha} \big] \\ &= \mathsf{E}^{\mathsf{P}^{b}} \left[\frac{d^{\mathsf{P}}}{d^{\mathsf{P}^{b}}} \mathbf{1}_{M(t) \leq \beta} \mathbf{1}_{W(t) + bt \leq \alpha} \right] \\ &= \mathsf{E}^{\mathsf{P}^{b}} \left[e^{b(X(t) - bt) + \frac{1}{2} b^{2}t} \mathbf{1}_{M(t) \leq \beta} \mathbf{1}_{X(t) \leq \alpha} \right] \\ &= \int_{-\infty}^{\alpha} e^{b(y + bt) - \frac{1}{2} b^{2}t} \int_{y^{+}}^{\beta} \frac{\partial^{2}}{\partial x \partial y} \, \Phi(y, x, 0, t) dx \, dy \\ &= \frac{1}{\sqrt{t}} \int_{-\infty}^{\alpha} e^{b(y + bt) - \frac{1}{2} b^{2}t} \big(n(-|y|/\sqrt{t}) - n((y - 2\beta)/\sqrt{t}) \big) dx \, dy, \end{split}$$

where
$$n(x) = (2\pi)^{-1/2} \exp(-x^2/2)$$
.

Carrying out the integration we obtain formula (i).



Let us now go back to the Black and Cox model. We have

$$Q\left(\min_{s \le t} (W(s) + bs) \le d\right) = 1 - \Phi(-d, -d, -b, t)$$
$$= 1 - N\left(\frac{-d + bt}{\sqrt{t}}\right) + e^{2bd}N\left(\frac{d + bt}{\sqrt{t}}\right),$$

where $d = \frac{1}{\sigma_V} \log \frac{H_0}{V_0}$.

• Using the fact that 1 - N(-x) = N(x), we see that the risk neutral probability of default is given by

$$Q(\tau < t) = N(\frac{d-bt}{\sqrt{t}}) + e^{2bd} N(\frac{d+bt}{\sqrt{t}}).$$

The equity holders' payoff at maturity T is given by

$$(V(T)-F)^+ 1_{\tau \geq T} = (V_0 e^{aT} e^{\sigma_V(W(T)+bT)} - F)^+ 1_{m(T) \geq d}.$$

- This is the payoff of a down and out call option.
- Its expected value gives the equity value of the company.
- As a consequence, the equity value in the Black and Cox model is smaller than the equity value in Merton's model.

- In the event of default, the debt holders' payoff for is $V(\tau) = H(\tau)$. The recovery value can be computed by integrating the discounted H(s) through τ .
- The recovery value is thus

$$B^{\text{rec}}(t,T) = -\int_{t}^{T} e^{-r(s-t)} H(s) \frac{\partial}{\partial s} \Phi(-d(t),-d(t),-b,s-t) ds,$$

where $d(t) = \frac{1}{\sigma_V} \log \frac{H(t)}{V_0}$. This integral can be calculated explicitly.

- The value of the bond at time t prior to default is the sum of the payment at maturity (B^{mat}) and the recovery value (B^{rec}).
- The payment at maturity term can be written as the risk neutral price of a difference of two barrier call options:

$$B^{\text{mat}}(t,T) = \mathsf{E}^{\mathsf{Q}} \big[e^{-r(T-t)} \big(V(T) - (V(T) - F)^+ \big) \mathbb{1}_{\tau > T} \, | \, \mathscr{F}_t \big]$$

= $e^{-r(T-t)} F \Phi(-d(t), -d(t), -b, T-t).$



References



Andersen, L.: Credit models, lecture notes (2010).



Bielecki, T. R., and Rutkowski, M.: Credit Risk: Modeling, Valuation and Hedging, Springer (2002).



Black, F., and Cox, J. C.: Valuing corporate securities: Some effects of bond indenture provisions, *J. Finance*, **31**, 351 - 367 (1976).



Merton, R.: On the pricing of corporate debt: the risk structure of interest rates, *J. Finance*, **29**, 449 - 470 (1974).