IEOR E4706 Foundations of Financial Engineering Martin Haugh

### Assignment 8 (Mandatory)

## 1. (Pricing a CDS in a Structural Model)

- (a) Replicate the equity and debt lattices on page 3 of the *Credit Modeling and Credit Derivatives* lecture notes. These lattices give the price processes for the debt and equity components of the capital structure under the Black-Cox default framework. (You will of course need to first construct the firm value lattice as given on page 2 of the lecture notes.)
- (b) Consider a CDS on the debt component of part (a) where the premium leg makes a payment in every period and the default leg pays the difference between the face value of the bond, i.e. \$800, and the firm value upon default. Use separate lattices to compute the time t = 0 value of the premium and default legs. Now use trial and error to determine the fair dollar premium per period.

### 2. (Estimating Ratings Transitions)

Consider the ratings models as described in Section 2 of the *Credit Modeling and Credit Derivatives* lecture notes. Given empirical transitions data, how should we estimate, for example, the 1-year transition matrix  $\mathbf{P}$ ? There are a few approaches and we will consider two of them here. In order to simplify matters let us assume that there are just four possible ratings, A, B, C and D (default). In addition we have the following data from the past year:

- At the beginning of the year, there were 10 firms in each of the four classes.
- In the 1<sup>st</sup> month, one *B*-rated firm migrated to a *C*-rating and one *B*-rated firm migrated to an *A*-rating.
- In the  $2^{nd}$  month, two A-rated firms migrated to a B-rating.
- In the 6<sup>th</sup> month, one C-rated firm migrated to a B-rating and one C-rated firm migrated to a D-rating.
- In the 8<sup>th</sup> month, one C-rated firm migrated to a D-rating.
- There were no other transitions during the year.

In the following questions you can assume that transitions took place at the *end* of the month in question.

(a) What is  $\widehat{\mathbf{P}}_{emp}$ , the empirical estimator of  $\mathbf{P}$  that we obtain by simply counting the empirical transitions between ratings in the given data-set?

(b) An alternative estimator is based on the so-called *generator* matrix,  $\Lambda$ , of the Markov chain. The generator matrix for a Markov chain with n states has the form

$$\Lambda = \begin{pmatrix} -\lambda_{1,1} & \lambda_{1,2} & \cdots & \lambda_{1,n} \\ \lambda_{2,1} & -\lambda_{2,2} & \cdots & \lambda_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ \lambda_{n,1} & \lambda_{n,2} & \cdots & -\lambda_{n,n} \end{pmatrix}$$

Note that each row in  $\Lambda$  sums to 0. Moreover,  $\lambda_{i,j}$  for  $i \neq j$ , represents the transition rate from state *i* to state *j*. We can estimate it according to

$$\hat{\lambda}_{i,j} = \frac{M_{i,j}}{N_i}$$

where  $N_i$  is the total number (across all firms) of time units spent in state *i* and  $M_{i,j}$  is the total number of transitions from *i* to *j*. Using the data-set given above, we would estimate

$$\hat{\lambda}_{A,B} = \frac{2}{8 \times \frac{12}{12} + 2 \times \frac{2}{12} + 1 \times \frac{11}{12}}.$$
(1)

Note that the time unit here is "years" and a term such as  $\frac{2}{12}$  in the denominator of (1) corresponds to 2 months =  $\frac{2}{12}$  of a year. The final piece we need is a well-known and very important property of  $\Lambda$  is that it can be used to determine the *m*-step transition matrix for any *m*. In particular, we have

$$\mathbf{P}^m = \exp\left\{\Lambda m\right\}$$

where exp denotes the *matrix exponential* function. This function is defined according to

$$\exp\left\{\mathbf{A}\right\} := \sum_{i=0}^{\infty} \frac{\mathbf{A}^{i}}{i!}$$

and it can be evaluated quickly using diagonalization methods. (You don't need to do that, however, as the Matlab function expm will calculate exp {A} for you.)

Estimate **P** using the methodology / theory described here.

(c) Which of the methods in (a) and (b) yields a superior estimator in your opinion? Justify your answer.

### 3. (Intensity Calculations)

We can define the default intensity,  $\lambda_t$  to be

$$\lambda_t := \frac{f_\tau(t)}{\mathbf{P}(\tau > t)}$$

where  $\tau$  is the time of default and  $f_{\tau}(\cdot)$  is the PDF of  $\tau$ . It therefore follows that  $\lambda_t dt$  is equal to the probability of defaulting in the interval (t, t + dt) given that default has not occurred in [0, t]. Show that

$$P(\tau > s) = \exp\left\{-\int_0^s \lambda_t \, dt\right\}.$$

# 4. (Constructing Intensity-Based Lattice Models)

Reproduce the three lattices on page 10 of the *Credit Modeling and Credit Derivatives* lecture notes. (The details are described in Section 4.1 of the notes.)

### 5. (Pricing a CDS in a Lattice Intensity Model)

Compute the fair spread of a CDS on the bond of Question 4 except now you should assume that the recovery upon default is 40% of face value. To simplify calculations you should assume that the premium leg makes a payment every period, i.e. year, and that the default leg pays the par amount, i.e. the face value of \$100, minus the recovery upon default.

*Hint:* See the final paragraph of Section 4.1 of the *Credit Modeling and Credit Derivatives* lecture notes to see how the fair spread can be calculated very quickly.