

Applications of Stochastic Calculus in Finance

Chapter 9: Intensity-based approach

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1 A simple reduced-form model with constant intensity

Intensity-based (or reduced-form) approach is silent about why a firm defaults, and instead default is exogenously given through a default rate, i.e. the default intensity of a single jump process.

A simple example would be assuming that the default time τ as an exponential random variable independent of the Brownian filtration $\{\mathcal{F}_t\}_{t \geq 0}$, and having constant intensity λ . Then

$$\mathbf{Q}(\tau > t | \mathcal{F}_t) = \mathbf{Q}(\tau > t) = e^{-\lambda t}.$$

Remark 1. (Calibration of default intensity λ). In an ideal situation where the CDS premium is paid continuously until the default time τ , the discounted payoff of the CDS buyer at time 0 is

$$\Pi_b(0) = \mathbf{1}_{\{\tau \leq T_n\}} P(0, \tau) LGD - \int_0^{\tau \wedge T_n} P(0, s) \kappa ds$$

Given the constant default intensity λ , the CDS spread at time $t = 0$ is the fixed rate κ such that $\mathbf{E}^{\mathbf{Q}}[\Pi_b(0)] = 0$, which is

$$R_{CDS}(0) = \frac{\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_{\{\tau \leq T_n\}} P(0, \tau)] LGD}{\mathbf{E}^{\mathbf{Q}}[\int_0^{\tau \wedge T_n} P(0, s) ds]}.$$

Note that τ has the intensity $\lambda e^{-\lambda s}$ for $s \geq 0$. Hence,

$$\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_{\{\tau \leq T_n\}} P(0, \tau)] = \int_0^{T_n} \lambda e^{-\lambda s} P(0, s) ds$$

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and

$$\begin{aligned} \mathbf{E}^Q[\int_0^{\tau \wedge T_n} P(0, s) ds] &= \int_0^\infty \lambda e^{-\lambda s} \int_0^{s \wedge T_n} P(0, u) du ds \\ &= - \int_0^\infty \left(\int_0^{s \wedge T_n} P(0, u) du \right) d e^{-\lambda s} \\ &= \int_0^{T_n} e^{-\lambda s} P(0, s) ds, \end{aligned}$$

which yields that

$$R_{CDS}(0) = \lambda LGD.$$

In practice, this is often used to calibrate the default intensity and therefore the default probability in terms of CDS spread.

However, the constant intensity λ is insufficient for most credit risk problems. It would be desirable that the default intensity λ is stochastic. Two types of stochastic intensities will be considered in this Chapter. The first one models the *systematic factor (common noise)* while the second one models the *systemic factor (correlated noise)*.

Moreover, we also need to calculate conditional default probability and conditional price at any time t , which is a delicate issue in the reduced-form framework. Note that in the structural models, conditional values can be simply obtained from the corresponding initial values by replacing V_0 with V_t .

2 \mathcal{F}_t -doubly stochastic stopping times

Fix a filtered probability space $(\Omega, \mathcal{G}, \{\mathcal{G}_t\}_{t \geq 0}, \mathbf{P})$, where the filtration $\{\mathcal{G}_t\}_{t \geq 0}$ represents the flow of the complete market information. Let τ be a default time which is a \mathcal{G}_t -stopping time. Throughout, we will assume that there exists a sub-filtration $\{\mathcal{F}_t\}_{t \geq 0}$, usually a Brownian filtration, such that $\mathcal{F}_t \subset \mathcal{G}_t$ and $\mathcal{G}_t = \mathcal{F}_t \vee \mathcal{H}_t$, where $\mathcal{H}_t = \sigma(\{\tau \leq u\} : u \leq t)$. When $\mathcal{F}_t = \{\emptyset, \Omega\}$, then $\mathcal{G}_t = \mathcal{H}_t$.

The following lemma shows that events in \mathcal{G}_t is actually \mathcal{F}_t -observable conditional on $\{\tau > t\}$.

Lemma 1. *For every $A \in \mathcal{G}_t$, there exists $B \in \mathcal{F}_t$ such that*

$$A \cap \{\tau > t\} = B \cap \{\tau > t\}. \quad (1)$$

Proof. Define σ -algebra:

$$\mathcal{G}_t^* = \{A \in \mathcal{G}_t : \text{there exists } B \in \mathcal{F}_t \text{ such that (1) holds}\}$$

It is obvious that $\mathcal{G}_t^* \subset \mathcal{G}_t$. We prove the other inclusion by showing $\mathcal{F}_t, \mathcal{H}_t \subset \mathcal{G}_t^*$.

Firstly $\mathcal{F}_t \subset \mathcal{G}_t^*$ by simply taking $B = A$ for any $A \in \mathcal{F}_t$. To show that $\mathcal{H}_t \subset \mathcal{G}_t^*$, note that any $A \in \mathcal{H}_t$, $A \cap \{\tau > t\}$ is either \emptyset or $\{\tau > t\}$, so we can take for B either \emptyset or Ω . \square

Assumption 1 *There exists a nonnegative \mathcal{F}_t -progressively measurable process λ such that*

$$\mathbf{P}(\tau > t | \mathcal{F}_t) = e^{-\int_0^t \lambda_s ds}.$$

From the above assumption, a market participant with access to the partial market information \mathcal{F}_t cannot observe whether default has occurred by time t . In other words, τ is NOT an \mathcal{F}_t -stopping time.

The following lemma generalizes Lemma 1 to random variables in the sense that for any r.v. $Y \in \mathcal{G}_t$, there exists $\tilde{Y}_t \in \mathcal{F}_t$ such that $\tilde{Y}_t = Y_t$ conditional on $\{\tau > t\}$. It is called filtration switching formula. When $\mathcal{F}_t = \{\emptyset, \Omega\}$, it is formula (10) in Chapter 8.

Lemma 2. (Filtration switching formula) *Suppose that Assumption 1 holds. For any r.v. $Y \in \mathcal{G}_\infty$,*

$$\mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{G}_t] = \mathbf{1}_{\{\tau > t\}} \frac{\mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t]}{\mathbf{P}(\tau > t | \mathcal{F}_t)} = \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_s ds} \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t]. \quad (2)$$

Proof. We need to show that

$$\mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y \mathbf{P}(\tau > t | \mathcal{F}_t) | \mathcal{G}_t] = \mathbf{1}_{\{\tau > t\}} \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t].$$

That is, RHS of the above equality is the conditional expectation of $\mathbf{1}_{\{\tau > t\}} Y \mathbf{P}(\tau > t | \mathcal{F}_t)$ on \mathcal{G}_t : For any $A \in \mathcal{G}_t$,

$$\mathbf{E}[\mathbf{1}_A \mathbf{1}_{\{\tau > t\}} Y \mathbf{P}(\tau > t | \mathcal{F}_t)] = \mathbf{E}[\mathbf{1}_A \mathbf{1}_{\{\tau > t\}} \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t]].$$

From Lemma 1, there exists $B \in \mathcal{F}_t$ such that $\mathbf{1}_A \mathbf{1}_{\{\tau > t\}} = \mathbf{1}_B \mathbf{1}_{\{\tau > t\}}$. Therefore, it is equivalent to show that

$$\mathbf{E}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} Y \mathbf{P}(\tau > t | \mathcal{F}_t)] = \mathbf{E}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t]].$$

But this follows by taking the conditional expectations of both sides on \mathcal{F}_t :

$$\begin{aligned} RHS &= \mathbf{E}[\mathbf{E}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t] | \mathcal{F}_t]] \\ &= \mathbf{E}[\mathbf{1}_B \mathbf{E}[\mathbf{1}_{\{\tau > t\}} | \mathcal{F}_t] \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t]], \end{aligned}$$

and

$$\begin{aligned} LHS &= \mathbf{E}[\mathbf{E}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} Y \mathbf{E}[\mathbf{1}_{\{\tau > t\}} | \mathcal{F}_t] | \mathcal{F}_t]] \\ &= \mathbf{E}[\mathbf{1}_B \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_t] \mathbf{E}[\mathbf{1}_{\{\tau > t\}} | \mathcal{F}_t]]. \quad \square \end{aligned}$$

We have now the following expressions for conditional default probabilities.

Proposition 1. *Suppose that Assumption 1 holds. Then*

$$\mathbf{P}(\tau > T | \mathcal{G}_t) = \mathbf{1}_{\{\tau > t\}} \mathbf{E}[e^{-\int_t^T \lambda_s ds} | \mathcal{F}_t] \quad (3)$$

$$\mathbf{P}(t < \tau \leq T | \mathcal{G}_t) = \mathbf{1}_{\{\tau > t\}} \mathbf{E}[1 - e^{-\int_t^T \lambda_s ds} | \mathcal{F}_t] \quad (4)$$

$$\mathbf{P}(t < \tau \leq t + \Delta | \mathcal{G}_t) = \mathbf{1}_{\{\tau > t\}} \lambda_t \Delta \text{ for } \Delta = o(1). \quad (5)$$

Proof. We only establish the first one. The other two then follow from (3).

Note that $\mathbf{1}_{\{\tau > T\}} = \mathbf{1}_{\{\tau > t\}} \mathbf{1}_{\{\tau > T\}}$. Then from Lemma 2,

$$\begin{aligned} \mathbf{P}(\tau > T | \mathcal{G}_t) &= \mathbf{E}[\mathbf{1}_{\{\tau > t\}} \mathbf{1}_{\{\tau > T\}} | \mathcal{G}_t] \\ &= \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_s ds} \mathbf{E}[\mathbf{1}_{\{\tau > T\}} | \mathcal{F}_t] \\ &= \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_s ds} \mathbf{E}[\mathbf{E}[\mathbf{1}_{\{\tau > T\}} | \mathcal{F}_T] | \mathcal{F}_t] \\ &= \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_s ds} \mathbf{E}[e^{-\int_0^T \lambda_s ds} | \mathcal{F}_t]. \quad \square \end{aligned}$$

Remark 2. Note that in the simple example where τ is an exponential r.v. independent of \mathcal{F}_t , then (3) and (4) reduce to formulaes (11) and (12) in Chapter 8, respectively.

Based on Proposition 1, we have the following Doob-Meyer decomposition for the single jump process $\mathbf{1}_{\{\tau \leq t\}}$ for $t \geq 0$.

Proposition 2. *Suppose that Assumption 1 holds. Then the process*

$$N_t = \mathbf{1}_{\{\tau \leq t\}} - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds \text{ for } t \geq 0$$

is a \mathcal{G}_t -martingale.

Proof. From Proposition 1 and Lemma 2, we obtain that

$$\begin{aligned} \mathbf{E}[N_T | \mathcal{G}_t] &= 1 - \mathbf{E}[\mathbf{1}_{\{\tau > T\}} | \mathcal{G}_t] - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds - \int_t^T \mathbf{E}[\lambda_s \mathbf{1}_{\{\tau > s\}} | \mathcal{G}_t] ds \\ &= 1 - \mathbf{1}_{\{\tau > t\}} \mathbf{E}[e^{-\int_t^T \lambda_s ds} | \mathcal{F}_t] - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds \\ &\quad - \int_t^T \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_u du} \mathbf{E}[\lambda_s \mathbf{1}_{\{\tau > s\}} | \mathcal{F}_t] ds. \end{aligned}$$

But note that the last term can be further simplified as

$$\begin{aligned} \int_t^T \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_u du} \mathbf{E}[\lambda_s \mathbf{1}_{\{\tau > s\}} | \mathcal{F}_t] ds &= \int_t^T \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_u du} \mathbf{E}[\lambda_s \mathbf{E}[\mathbf{1}_{\{\tau > s\}} | \mathcal{F}_s] | \mathcal{F}_t] ds \\ &= \mathbf{1}_{\{\tau > t\}} \mathbf{E}\left[\int_t^T \lambda_s e^{-\int_t^s \lambda_u du} ds \middle| \mathcal{F}_t\right] \\ &= \mathbf{1}_{\{\tau > t\}} \mathbf{E}[1 - e^{-\int_t^T \lambda_s ds} | \mathcal{F}_t]. \end{aligned}$$

Hence, $\mathbf{E}[N_T | \mathcal{G}_t] = N_t$. \square

We further impose the following assumption, so called Hypothesis H in the literature. *This assumption will be used for the construction of the default time τ later on.*

Assumption 2 (Hypothesis H) Every \mathcal{F}_t -martingale is also a \mathcal{G}_t -martingale, or equivalently¹,

$$\mathbf{P}(\tau > t | \mathcal{F}_\infty) = \mathbf{P}(\tau > t | \mathcal{F}_t) \text{ for } t \geq 0.$$

The equivalent conditions of Hypothesis H can be established as follows.

Suppose that every \mathcal{F}_t -martingale is a \mathcal{G}_t -martingale. Then, for $X \in \mathcal{F}_\infty$, $M_t := \mathbf{E}[X | \mathcal{F}_t]$ is an \mathcal{F}_t -martingale. It follows that

$$\mathbf{E}[X | \mathcal{G}_t] = \mathbf{E}[M_\infty | \mathcal{G}_t] = M_t = \mathbf{E}[X | \mathcal{F}_t]. \quad (6)$$

Since $\mathbf{1}_{\tau > t} \in \mathcal{G}_t$, by the definition of conditional expectation,

$$\mathbf{E}[\mathbf{1}_{\tau > t} X] = \mathbf{E}[\mathbf{1}_{\tau > t} \mathbf{E}[X | \mathcal{F}_t]].$$

However, conditioning on the RHS w.r.t \mathcal{F}_t yields

$$\begin{aligned} \mathbf{E}[\mathbf{1}_{\tau > t} \mathbf{E}[X | \mathcal{F}_t]] &= \mathbf{E}[\mathbf{E}[\mathbf{1}_{\tau > t} | \mathcal{F}_t] \mathbf{E}[X | \mathcal{F}_t]] \\ &= \mathbf{E}[\mathbf{E}[\mathbf{1}_{\tau > t} | \mathcal{F}_t] X]. \end{aligned}$$

By taking $X = \mathbf{1}_C$ for $C \in \mathcal{F}_\infty$, we have

$$\mathbf{E}[\mathbf{1}_{\tau > t} \mathbf{1}_C] = \mathbf{E}[\mathbf{E}[\mathbf{1}_{\tau > t} | \mathcal{F}_t] \mathbf{1}_C],$$

that is

$$\mathbf{P}(\tau > t | \mathcal{F}_\infty) = \mathbf{P}(\tau > t | \mathcal{F}_t).$$

On the other hand, suppose the above equality holds. We first prove (6). Note that for $A \in \mathcal{G}_t$, there exists $B \in \mathcal{F}_t$ such that $\mathbf{1}_A \mathbf{1}_{\tau > t} = \mathbf{1}_B \mathbf{1}_{\tau > t}$. Hence, it is sufficient to show that

$$\mathbf{E}[X \mathbf{1}_B \mathbf{1}_{\tau > t}] = \mathbf{E}[\mathbf{E}[X | \mathcal{F}_t] \mathbf{1}_B \mathbf{1}_{\tau > t}].$$

However, by conditioning w.r.t. \mathcal{F}_∞ and \mathcal{F}_t respectively, the LHS is equal to

$$\mathbf{E}[X \mathbf{1}_B \mathbf{1}_{\tau > t}] = \mathbf{E}[X \mathbf{1}_B \mathbf{E}[\mathbf{1}_{\tau > t} | \mathcal{F}_\infty]],$$

and the RHS is equal to

$$\mathbf{E}[\mathbf{E}[X | \mathcal{F}_t] \mathbf{1}_B \mathbf{1}_{\tau > t}] = \mathbf{E}[X \mathbf{1}_B \mathbf{E}[\mathbf{1}_{\tau > t} | \mathcal{F}_t]],$$

which are equal by the assumption. Next, we prove that every \mathcal{F}_t -martingale is a \mathcal{G}_t -martingale from (6). Let M be an \mathcal{F}_t -martingale. Then, for any $T > t$, since $M_T \in \mathcal{F}_\infty$, $\mathbf{E}[M_T | \mathcal{G}_t] = \mathbf{E}[M_T | \mathcal{F}_t] = M_t$, which means M is also a \mathcal{G}_t -martingale.

Definition 1. A \mathcal{G}_t -stopping time τ that satisfies Assumptions 1 and 2 is called an \mathcal{F}_t -doubly stochastic stopping time.

¹ One can make a comparison with Markov property with \mathcal{F}_∞ represents ‘future’, \mathcal{F}_t represents ‘present’, \mathcal{H}_t represents ‘past’ and $\mathcal{G}_t = \mathcal{F}_t \vee \mathcal{H}_t$ represents ‘present and past’. Then, the Hypothesis H also implies the following (1) $\mathbf{E}[X | \mathcal{G}_t] = \mathbf{E}[X | \mathcal{F}_t]$; (2) $\mathbf{E}[XY | \mathcal{F}_t] = \mathbf{E}[X | \mathcal{F}_t] \mathbf{E}[Y | \mathcal{F}_t]$ for $X \in \mathcal{F}_\infty$ and $Y \in \mathcal{H}_t$.

Proposition 3. *Suppose that Assumptions 1 and 2 hold. Then the process*

$$N_t = \mathbf{1}_{\{\tau \leq t\}} - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds \text{ for } t \geq 0$$

is an $(\mathcal{F}_\infty \vee \mathcal{H}_t)$ -martingale.

Proof. First, analogous to the proof of Lemma 2, it can be proved that the following generalized filtration switching formula holds

$$\mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_\infty \vee \mathcal{H}_t] = \mathbf{1}_{\{\tau > t\}} \frac{\mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_\infty]}{\mathbf{P}(\tau > t | \mathcal{F}_\infty)} = \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_s ds} \mathbf{E}[\mathbf{1}_{\{\tau > t\}} Y | \mathcal{F}_\infty].$$

Then similar to the proof of Proposition 2, we obtain that

$$\begin{aligned} & \mathbf{E}[N_T | \mathcal{F}_\infty \vee \mathcal{H}_t] \\ &= 1 - \mathbf{E}[\mathbf{1}_{\{\tau > T\}} | \mathcal{F}_\infty \vee \mathcal{H}_t] - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds - \int_t^T \mathbf{E}[\lambda_s \mathbf{1}_{\{\tau > s\}} | \mathcal{F}_\infty \vee \mathcal{H}_t] ds \\ &= 1 - \mathbf{1}_{\{\tau > t\}} e^{-\int_t^T \lambda_s ds} - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds - \int_t^T \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_u du} \mathbf{E}[\lambda_s \mathbf{1}_{\{\tau > s\}} | \mathcal{F}_\infty] ds \\ &= 1 - \mathbf{1}_{\{\tau > t\}} e^{-\int_t^T \lambda_s ds} - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds - \mathbf{1}_{\{\tau > t\}} \int_t^T e^{\int_0^t \lambda_u du} \lambda_s e^{-\int_0^s \lambda_u du} ds \\ &= 1 - \mathbf{1}_{\{\tau > t\}} e^{-\int_t^T \lambda_s ds} - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds - \mathbf{1}_{\{\tau > t\}} (1 - e^{-\int_t^T \lambda_s ds}) \\ &= \mathbf{1}_{\{\tau \leq t\}} - \int_0^t \lambda_s \mathbf{1}_{\{\tau > s\}} ds. \quad \square \end{aligned}$$

3 Intensity-based Approach

Construction of intensity-based models: We want to construct a default time τ which satisfies Assumptions 1 and 2, so it is a \mathcal{F}_t -doubly stochastic stopping time.

(1) Given a probability space $(\Omega, \mathcal{G}, \mathbf{Q})$, start with a filtration $\{\mathcal{F}_t\}_{t \geq 0}$ (usually Brownian filtration) satisfying the usual conditions, and $\mathcal{F}_\infty \subset \mathcal{G}$.

(2) Let λ_t for $t \geq 0$ be a nonnegative \mathcal{F}_t -progressively measurable process such that $\int_0^t \lambda_s ds < \infty$, a.s. for $t \geq 0$.

(3) Fix an exponential r.v. E with intensity 1 and independent of \mathcal{F}_∞ , and define

$$\tau = \inf\{t \geq 0 : \int_0^t \lambda_s ds \geq E\}.$$

(4) Finally define $\mathcal{G}_t = \mathcal{F}_t \vee \mathcal{H}_t$ where $\mathcal{H}_t = \sigma(\{\tau \leq u\} : u \leq t)$.

Proposition 4. *Under the above canonical construction, the random time τ is a \mathcal{F}_t -doubly stochastic stopping time.*

Proof.

$$\mathbf{Q}(\tau > t | \mathcal{F}_\infty) = \mathbf{Q}(E > \int_0^t \lambda_s ds | \mathcal{F}_\infty) = \mathbf{Q}(E > x) |_{x=\int_0^t \lambda_s ds} = e^{-\int_0^t \lambda_s ds}.$$

Conditioning both sides on \mathcal{F}_t then yields

$$\mathbf{Q}(\tau > t | \mathcal{F}_t) = e^{-\int_0^t \lambda_s ds}.$$

Hence, both Assumptions 1 and 2 hold. \square

Assumption 3 Suppose that under the spot measure \mathbf{Q} , the short rate r is \mathcal{F}_t -progressively measurable, and there exists a nonnegative \mathcal{F}_t -progressively measurable intensity λ such that

$$\int_0^t (|r_s| + \lambda_s) ds, \text{ a.s. for } t \geq 0.$$

and Assumptions 1 and 2 hold.

Conditional default probability can be easily computed under intensity-based models.

$$\begin{aligned} \mathbf{Q}(\tau \leq T | \mathcal{G}_t) &= 1 - \mathbf{Q}(\tau > T | \mathcal{G}_t) \\ &= 1 - \mathbf{1}_{\{\tau > t\}} \mathbf{E}[e^{-\int_t^T \lambda_s ds} | \mathcal{F}_t]. \end{aligned}$$

Consider a corporate bond which is supposed to pay 1 at maturity T . When default occurs for such a corporate bond, the recovery could be delivered either at maturity T (correspond to Merton's model), or at the default time τ (correspond to first-passage-time model).

If the recovery is delivered at maturity T , the discounted payoff at time t is then $\frac{B_t}{B_T}(\mathbf{1}_{\{\tau > T\}} + \delta \mathbf{1}_{\{t < \tau \leq T\}})$ for some recovery rate $\delta \in [0, 1]$

Proposition 5. Under the intensity-based model, if the recovery is delivered at maturity T , the conditional value of the corporate bond at time t is

$$P_t = \delta \mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[e^{-\int_t^T r_s ds} | \mathcal{F}_t] + (1 - \delta) \mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[e^{-\int_t^T (r_s + \lambda_s) ds} | \mathcal{F}_t].$$

where $P(t, T)$ is the time t value of T -bond.

Proof. The arbitrage price the corporate bond at any time $t \leq T$ is

$$\begin{aligned} P_t &= \mathbf{E}^{\mathbf{Q}}[\frac{B_t}{B_T}(\mathbf{1}_{\{\tau > T\}} + \delta \mathbf{1}_{\{t < \tau \leq T\}}) | \mathcal{G}_t] \\ &= \mathbf{E}^{\mathbf{Q}}[\frac{B_t}{B_T}(\delta \mathbf{1}_{\{\tau > t\}} + (1 - \delta) \mathbf{1}_{\{\tau > T\}}) | \mathcal{G}_t] \end{aligned}$$

For the first term, by the Hypothesis H, we have

$$\begin{aligned}
\mathbf{E}^{\mathbf{Q}}\left[\frac{B_t}{B_T}\delta\mathbf{1}_{\{\tau>t\}}|\mathcal{G}_t\right] &= \delta\mathbf{1}_{\{\tau>t\}}\mathbf{E}^{\mathbf{Q}}\left[\frac{B_t}{B_T}|\mathcal{G}_t\right] \\
&= \delta\mathbf{1}_{\{\tau>t\}}\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T r_s ds}|\mathcal{F}_t\right].
\end{aligned}$$

Hence, we only need to compute

$$\begin{aligned}
\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T r_s ds}\mathbf{1}_{\{\tau>T\}}|\mathcal{G}_t\right] &= \mathbf{1}_{\{\tau>t\}}e^{\int_0^t \lambda_s ds}\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T r_s ds}\mathbf{1}_{\{\tau>T\}}|\mathcal{F}_t\right] \\
&= \mathbf{1}_{\{\tau>t\}}e^{\int_0^t \lambda_s ds}\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T r_s ds}\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_{\{\tau>T\}}|\mathcal{F}_T]|\mathcal{F}_t\right] \\
&= \mathbf{1}_{\{\tau>t\}}\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T (r_s+\lambda_s)ds}|\mathcal{F}_t\right],
\end{aligned}$$

where the first equality follows from the filtration switching formula in Lemma 2. \square

If the recovery is delivered at default time τ , the discounted payoff at time t is then $\frac{B_t}{B_T}\mathbf{1}_{\{\tau>T\}} + \frac{B_t}{B_\tau}\delta\mathbf{1}_{\{t<\tau\leq T\}}$.

Proposition 6. *Under the intensity-based model, if the recovery is delivered at the default time τ , the conditional value of the corporate bond at time t is*

$$\begin{aligned}
P_t &= \mathbf{1}_{\{\tau>t\}}\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T (r_s+\lambda_s)ds}|\mathcal{F}_t\right] \\
&\quad + \delta\mathbf{1}_{\{\tau>t\}}\int_t^T \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s+\lambda_s)ds}|\mathcal{F}_t]du.
\end{aligned}$$

Proof. The arbitrage price the corporate bond at any time $t \leq T$ is

$$\begin{aligned}
P_t &= \mathbf{E}^{\mathbf{Q}}\left[\frac{B_t}{B_T}\mathbf{1}_{\{\tau>T\}} + \frac{B_t}{B_\tau}\delta\mathbf{1}_{\{t<\tau\leq T\}}|\mathcal{G}_t\right] \\
&= \mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^T r_s ds}\mathbf{1}_{\{\tau>T\}}|\mathcal{G}_t\right] + \delta\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^\tau r_s ds}\mathbf{1}_{\{t<\tau\leq T\}}|\mathcal{G}_t\right]
\end{aligned}$$

The first conditional expectation has been computed in the last proposition. Hence, we only need to compute the second one:

$$\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^\tau r_s ds}\mathbf{1}_{\{t<\tau\leq T\}}|\mathcal{G}_t\right] = \mathbf{E}^{\mathbf{Q}}\left[\mathbf{E}^{\mathbf{Q}}\left[e^{-\int_t^\tau r_s ds}\mathbf{1}_{\{t<\tau\leq T\}}|\mathcal{F}_\infty \vee \mathcal{H}_t\right]|\mathcal{G}_t\right]$$

Similar to (4) in Proposition 1, it can be shown that for $t \leq u$,

$$\mathbf{Q}(t < \tau \leq u | \mathcal{F}_\infty \vee \mathcal{H}_t) = \mathbf{1}_{\{\tau>t\}}(1 - e^{-\int_t^u \lambda_s ds})$$

so the conditional density of τ on $\mathcal{F}_\infty \vee \mathcal{H}_t$ is

$$\mathbf{1}_{\{\tau>t\}}\lambda_u e^{-\int_t^u \lambda_s ds}$$

for $u \geq t$. Therefore,

$$\begin{aligned}
\mathbf{E}^{\mathbf{Q}}[e^{-\int_t^\tau r_s ds} \mathbf{1}_{\{t < \tau \leq T\}} | \mathcal{G}_t] &= \mathbf{E}^{\mathbf{Q}}[\int_t^T \mathbf{1}_{\{\tau > t\}} \lambda_u e^{-\int_t^u \lambda_s ds} e^{-\int_t^u r_s ds} du | \mathcal{G}_t] \\
&= \mathbf{1}_{\{\tau > t\}} \int_t^T \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{G}_t] du
\end{aligned}$$

We are left to show that

$$\mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{G}_t] = \mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t]$$

which follows from Assumption 2. Indeed, it is sufficient to show that for any $A \in \mathcal{G}_t$,

$$\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_A \mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t]] = \mathbf{E}^{\mathbf{Q}}[\mathbf{1}_A \mathbf{1}_{\{\tau > t\}} \lambda_u e^{-\int_t^u (r_s + \lambda_s) ds}]$$

From Lemma 1, there exists $B \in \mathcal{F}_t$ such that $\mathbf{1}_A \mathbf{1}_{\{\tau > t\}} = \mathbf{1}_B \mathbf{1}_{\{\tau > t\}}$. Therefore, it is equivalent to show that

$$\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t]] = \mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} \lambda_u e^{-\int_t^u (r_s + \lambda_s) ds}]$$

Take conditional expectations of RHS first on \mathcal{F}_∞ and then on \mathcal{F}_t , and of LHS on \mathcal{F}_t :

$$\begin{aligned}
RHS &= \mathbf{E}^{\mathbf{Q}}[\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} \lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_\infty]] \\
&= \mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{Q}(\tau > t | \mathcal{F}_\infty) \lambda_u e^{-\int_t^u (r_s + \lambda_s) ds}] \\
&= \mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{Q}(\tau > t | \mathcal{F}_t) \lambda_u e^{-\int_t^u (r_s + \lambda_s) ds}] \\
&= \mathbf{E}^{\mathbf{Q}}[\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{Q}(\tau > t | \mathcal{F}_t) \lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t]] \\
&= \mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{Q}(\tau > t | \mathcal{F}_t) \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t]],
\end{aligned}$$

and

$$\begin{aligned}
LHS &= \mathbf{E}^{\mathbf{Q}}[\mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{1}_{\{\tau > t\}} \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t] | \mathcal{F}_t]] \\
&= \mathbf{E}^{\mathbf{Q}}[\mathbf{1}_B \mathbf{Q}(\tau > t | \mathcal{F}_t) \mathbf{E}^{\mathbf{Q}}[\lambda_u e^{-\int_t^u (r_s + \lambda_s) ds} | \mathcal{F}_t]]. \quad \square
\end{aligned}$$

Remark 3. Note that in the above calculation, using filtration switching formula directly on $\mathbf{E}^{\mathbf{Q}}[e^{-\int_t^\tau r_s ds} \mathbf{1}_{\{t < \tau \leq T\}} | \mathcal{G}_t]$ does not help:

$$\mathbf{E}^{\mathbf{Q}}[e^{-\int_t^\tau r_s ds} \mathbf{1}_{\{t < \tau \leq T\}} | \mathcal{G}_t] = \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \lambda_s ds} \mathbf{E}^{\mathbf{Q}}[e^{-\int_t^\tau r_s ds} \mathbf{1}_{\{t < \tau \leq T\}} | \mathcal{F}_t],$$

as this needs the conditional density of τ on \mathcal{F}_t , which is difficult to obtain.

4 Reduced-form models with correlated intensities

We model a pair of interactive default times τ and $\bar{\tau}$, where τ represents the default time of the reference entity, and $\bar{\tau}$ represents the default time of the CDS seller. Another interpretation is $(\tau, \bar{\tau})$ are a pair of default times for a basket CDS with two reference names.

Assumption 4 *Let $(\tau, \bar{\tau})$ be a pair of non-negative random variables defined on a complete probability space $(\Omega, \mathcal{G}, \mathbf{Q})$, and $\{\mathcal{G}_t\}_{t \geq 0}$ be the natural filtration of $(H_t, \bar{H}_t) = (\mathbf{1}_{\{\tau \leq t\}}, \mathbf{1}_{\{\bar{\tau} \leq t\}})$, $t \geq 0$. i.e. $\mathcal{G}_t = \sigma(H_s, \bar{H}_s, s \leq t)$, such that*

$$M_t := H_t - \int_0^t \mathbf{1}_{\{\tau > s\}} \lambda_s ds, \quad t \geq 0,$$

and

$$\bar{M}_t := \bar{H}_t - \int_0^t \mathbf{1}_{\{\bar{\tau} > s\}} \bar{\lambda}_s ds, \quad t \geq 0,$$

are both $(\mathcal{G}_t, \mathbf{Q})$ -martingales. Moreover, λ and $\bar{\lambda}$ are given by

$$\lambda_s = a_1 + a_2 \mathbf{1}_{\{\bar{\tau} \leq s\}}, \quad (7)$$

$$\bar{\lambda}_s = \bar{a}_1 + \bar{a}_2 \mathbf{1}_{\{\tau \leq s\}}, \quad (8)$$

for constants $a_1, \bar{a}_1 > 0$ and $a_2, \bar{a}_2 \geq 0$. Finally, we assume that $\mathbf{Q}(\tau > 0) = \mathbf{Q}(\bar{\tau} > 0) = 1$.

We aim to calculate the joint distribution of $(\tau, \bar{\tau})$ using the Girsanov's theorem introduced in Chapter 8. The difficulty herein is the looping feature of the two intensity processes. If τ happens (the reference entity defaults), the intensity of $\bar{\tau}$ will increase from \bar{a}_1 to $\bar{a}_1 + \bar{a}_2$. Similarly, if $\bar{\tau}$ happens (the CDS seller defaults), the intensity of τ will increase from a_1 to $a_1 + a_2$. By changing the probability measure \mathbf{Q} to another probability measure, we open this loop, which will in turn facilitate the calculation of the joint distribution of $(\tau, \bar{\tau})$.

Theorem 1. *Suppose that Assumption 4 holds. Then, for $T, U \geq 0$,*

$$\mathbf{Q}(\tau > T) = \frac{a_2 e^{-(a_1 + \bar{a}_1)T} - \bar{a}_1 e^{-(a_1 + a_2)T}}{a_2 - \bar{a}_1}; \quad (9)$$

$$\mathbf{Q}(\bar{\tau} > U) = \frac{\bar{a}_2 e^{-(\bar{a}_1 + a_1)U} - a_1 e^{-(\bar{a}_1 + \bar{a}_2)U}}{\bar{a}_2 - a_1}. \quad (10)$$

Proof. We first prove (9). For $t \in [0, T]$, let

$$Z_t = \mathbf{1}_{\{\tau > t\}} e^{\int_0^t \mathbf{1}_{\{\tau > s\}} \lambda_s ds}$$

Then, Theorem 8 in Chapter 8 (with $\mu = 1$) implies that Z is an $(\mathcal{G}_t, \mathbf{Q})$ -martingale. Define a new probability measure \mathbf{Q}^1 on \mathcal{G}_T by

$$\frac{d\mathbf{Q}^1}{d\mathbf{Q}} = Z_T.$$

Then, Theorem 9 in Chapter 8 (with $\mu = 1$) implies that $H_t = \mathbf{1}_{\{\tau \leq t\}}$, $t \in [0, T]$, is an $(\mathcal{G}_t, \mathbf{Q}^1)$ -positive martingale, which is 0, \mathbf{Q}^1 -a.e.

By Bayes' formula, we have

$$\begin{aligned} \mathbf{Q}(\tau > T) &= \mathbf{E}^{\mathbf{Q}}[Z_T e^{-\int_0^T \mathbf{1}_{\{\tau > s\}} \lambda_s ds}] \\ &= \mathbf{E}^{\mathbf{Q}}[Z_T e^{-\int_0^T \lambda_s ds}] \\ &= \mathbf{E}^{\mathbf{Q}^1}[e^{-\int_0^T \lambda_s ds}] \\ &= \mathbf{E}^{\mathbf{Q}^1}[e^{-\int_0^T a_1 + a_2 \mathbf{1}_{\{\bar{\tau} \leq s\}} ds}] \end{aligned}$$

Note that under \mathbf{Q}^1 , for $s \in [0, T]$, $\bar{\lambda}_s = \bar{a}_1 + \bar{a}_2 \mathbf{1}_{\{\bar{\tau} \leq s\}} = \bar{a}_1$, \mathbf{Q}^1 -a.e., so we further have

$$\begin{aligned} \mathbf{Q}(\tau > T) &= e^{-a_1 T} \left(\mathbf{E}^{\mathbf{Q}^1}[\mathbf{1}_{\{\bar{\tau} > T\}}] + \mathbf{E}^{\mathbf{Q}^1}[\mathbf{1}_{\{\bar{\tau} \leq T\}} e^{-a_2(T-\bar{\tau})}] \right) \\ &= e^{-a_1 T} \left(e^{-\bar{a}_1 T} + \int_0^T e^{-a_2(T-s)} \bar{a}_1 e^{-\bar{a}_1 s} ds \right) \\ &= e^{-a_1 T} \frac{(a_2 - \bar{a}_1) e^{-\bar{a}_1 T} + \bar{a}_1 e^{-\bar{a}_1 T} - \bar{a}_1 e^{-a_2 T}}{a_2 - \bar{a}_1}, \end{aligned}$$

which proves (9).

To prove (10), with

$$\bar{Z}_t = \mathbf{1}_{\{\bar{\tau} > t\}} e^{\int_0^t \mathbf{1}_{\{\bar{\tau} > s\}} \bar{\lambda}_s ds}$$

for $t \in [0, U]$, we define a new probability measure $\bar{\mathbf{Q}}^1$ on \mathcal{G}_T by

$$\frac{d\bar{\mathbf{Q}}^1}{d\mathbf{Q}} = \bar{Z}_U.$$

Then, $\bar{H}_t = \mathbf{1}_{\{\bar{\tau} \leq t\}}$, $t \in [0, U]$, is an $(\mathcal{G}_t, \bar{\mathbf{Q}}^1)$ -positive martingale, which is 0, $\bar{\mathbf{Q}}^1$ -a.e. The rest of the proof follows along the same argument as the first case. \square

An interesting observation is that the reference entity's survival probability $\mathbf{Q}\{\tau > T\}$ is independent of \bar{a}_2 . This is because for $s \in [0, T]$, $\{\tau \leq s\}$ is a zero event on the set $\{\tau > T\}$, so $\bar{\lambda}_s$ will keep as the constant \bar{a}_1 and is independent of \bar{a}_2 . Similarly, the CDS seller's survival probability $\mathbf{Q}\{\bar{\tau} > U\}$ is independent of a_2 .

When $a_2 = 0$, i.e. the CDS seller's default will not affect the reference entity, then (9) reduces to $\mathbf{Q}(\tau > T) = e^{-a_1 T}$. When $a_2 = \bar{a}_1$, then the L'Hopital's rule implies that

$$\mathbf{Q}(\tau > T) = (1 + \bar{a}_1 T) e^{-(a_1 + \bar{a}_1)T}.$$

Theorem 2. Suppose that Assumption 4 holds. Then, for $U > T \geq 0$,

$$\mathbf{Q}(\tau > T, \bar{\tau} > U) = e^{-(a_1 + \bar{a}_1)T} \frac{\bar{a}_2 e^{-(a_1 + \bar{a}_1)(U-T)} - a_1 e^{-(\bar{a}_1 + \bar{a}_2)(U-T)}}{\bar{a}_2 - a_1}. \quad (11)$$

Proof. We take conditional expectation on \mathcal{G}_T and get

$$\mathbf{Q}(\tau > T, \bar{\tau} > U) = \mathbf{E}^{\mathbf{Q}} [\mathbf{1}_{\{\tau > T\}} \mathbf{Q}(\bar{\tau} > U | \mathcal{G}_T)].$$

Note that

$$\mathbf{Q}(\bar{\tau} > U | \mathcal{G}_T) = \mathbf{1}_{\{\bar{\tau} > T\}} \left(\mathbf{1}_{\{\tau > T\}} \mathbf{Q}(\bar{\tau} > U - T) + \mathbf{1}_{\{\tau \leq T\}} e^{-(\bar{a}_1 + \bar{a}_2)(U-T)} \right).$$

Hence,

$$\begin{aligned} \mathbf{Q}(\tau > T, \bar{\tau} > U) &= \mathbf{E}^{\mathbf{Q}} [\mathbf{1}_{\{\tau > T\}} \mathbf{1}_{\{\bar{\tau} > T\}}] \times \mathbf{Q}(\bar{\tau} > U - T) \\ &= \mathbf{E}^{\mathbf{Q}} [\mathbf{1}_{\{\tau > T\}} \mathbf{1}_{\{\bar{\tau} > T\}} e^{\int_0^T \mathbf{1}_{\{\bar{\tau} > s\}} \bar{\lambda}_s ds} e^{-\int_0^T \mathbf{1}_{\{\bar{\tau} > s\}} \bar{\lambda}_s ds}] \times \mathbf{Q}(\bar{\tau} > U - T) \\ &= \mathbf{E}^{\mathbf{Q}} [\mathbf{1}_{\{\tau > T\}} \bar{Z}_T e^{-\int_0^T \bar{\lambda}_s ds}] \times \mathbf{Q}(\bar{\tau} > U - T) \\ &= \mathbf{E}^{\bar{\mathbf{Q}}^1} [\mathbf{1}_{\{\tau > T\}} e^{-\int_0^T \bar{a}_1 + \bar{a}_2 \mathbf{1}_{\{\tau \leq s\}} ds}] \times \mathbf{Q}(\bar{\tau} > U - T). \end{aligned}$$

Note that under $\bar{\mathbf{Q}}^1$, for $s \in [0, T]$, $\bar{\lambda}_s = a_1 + a_2 \mathbf{1}_{\{\tau \leq s\}} = a_1$, $\bar{\mathbf{Q}}^1$ -a.e., so we further have

$$\mathbf{Q}(\tau > T, \bar{\tau} > U) = \bar{\mathbf{Q}}^1(\tau > T) e^{-\bar{a}_1 T} \times \mathbf{Q}(\bar{\tau} > U - T) = e^{-(a_1 + \bar{a}_1)T} \times \mathbf{Q}(\bar{\tau} > U - T),$$

which completes the proof. \square

Note that the above joint probability is independent of a_2 . This is because on the set $\{\bar{\tau} > U\}$, we also have $\bar{\tau} > T$, so $\{\bar{\tau} \leq s\}$ for $s \in [0, U]$ is a zero event on $\{\bar{\tau} > U\}$. Thus, $\bar{\lambda}$ will keep as the constant a_1 , and is independent of a_2 .

Exercise 1. For $T > U \geq 0$, show the joint probability

$$\mathbf{Q}(\tau > T, \bar{\tau} > U) = e^{-(a_1 + \bar{a}_1)U} \mathbf{Q}(\tau > T - U).$$

Thus, $(\tau, \bar{\tau})$ admit the joint density

$$f_{(\tau, \bar{\tau})}(T, U) = \begin{cases} (\bar{a}_1 + \bar{a}_2) e^{-(\bar{a}_1 + \bar{a}_2)U} a_1 e^{(\bar{a}_2 - a_1)T}, & \text{if } U > T; \\ (a_1 + a_2) e^{-(a_1 + a_2)T} \bar{a}_1 e^{(a_2 - \bar{a}_1)U}, & \text{if } U < T. \end{cases} \quad (12)$$

Exercise 2. (Total hazard construction of correlated default times [1])

In general, when there are M default times $(\tau^i)_{1 \leq i \leq M}$ as in the basket CDS, we may simulate those default times via the so called *total hazard construction*. It is based on the following simple observation: For a single jump process $H_t = \mathbf{1}_{\{\tau \leq t\}}$, $t \geq 0$, with a constant intensity $a > 0$, we can simulate its associated jump time τ as

$$\hat{\tau} = \inf\{t \geq 0 : \Lambda_t \geq E\}$$

where E is a standard exponential r.v. and $\Lambda_t := \int_0^t a ds$, the so called hazard of τ . Then, $\hat{\tau} = \tau$ in distribution. Indeed, for any $T > 0$,

$$\mathbf{Q}(\hat{\tau} > T) = \mathbf{Q}(\Lambda_T < E) = \mathbf{Q}(aT < E) = e^{-aT}.$$

Assumption 5 Let $(\tau^i)_{1 \leq i \leq M}$ be a sequence of non-negative random variables defined on a complete probability space $(\Omega, \mathcal{G}, \mathbf{Q})$, and $\{\mathcal{G}_t\}_{t \geq 0}$ be the natural filtration of $H_t^i = \mathbf{1}_{\{\tau^i \leq t\}}$, $t \geq 0$, $1 \leq i \leq M$, i.e. $\mathcal{G}_t = \sigma(H_s^i, s \leq t, 1 \leq i \leq M)$, such that

$$M_t^i := H_t^i - \int_0^t \mathbf{1}_{\{\tau^i > s\}} \lambda_s^i ds, \quad t \geq 0, \quad 1 \leq i \leq M,$$

are $(\mathcal{G}_t, \mathbf{Q})$ -martingales. Moreover, λ^i , $1 \leq i \leq M$, are given by

$$\lambda_s^i = a_{ii} + \sum_{j \neq i} a_{ij} \mathbf{1}_{\{\tau^j \leq s\}}, \quad (13)$$

with a constant matrix $A = (a_{ij})_{1 \leq i, j \leq M}$ satisfying $a_{ij} > 0$ and $a_{ij} \geq 0$ for $i \neq j$. Finally, we assume that $\mathbf{Q}(\tau^i > 0) = 1$ for $1 \leq i \leq M$.

Note that there are $M!$ orders of $(\tau^i)_{1 \leq i \leq M}$. Let us consider one of the orders: $\{\tau^1 < \tau^2 < \dots < \tau^M\}$. In such a case, we construct the M default times as follows.

Step 0: Let $(E^i)_{1 \leq i \leq M}$ be a sequence of mutually independent standard exponential random variables.

Step 1: Since there is no default prior to τ^1 , its intensity is a_{11} and, therefore,

$$\hat{\tau}^1 = \inf \left\{ t \geq 0 : \int_0^t a_{11} ds \geq E^1 \right\} = \frac{E^1}{a_{11}}.$$

Step 2: Prior to τ^2 , τ^1 has already occurred, so the intensity of τ^2 is $a_{22} + a_{21} \mathbf{1}_{\{\tau^1 \leq s\}}$ and it can be constructed by

$$\begin{aligned} \hat{\tau}^2 &= \inf \left\{ t \geq 0 : \int_0^{\hat{\tau}^1} a_{22} ds + \int_{\hat{\tau}^1}^t (a_{22} + a_{21}) ds \geq E^2 \right\} \\ &= \frac{E^1}{a_{11}} + \frac{a_{22}}{a_{22} + a_{21}} \left(\frac{E^2}{a_{22}} - \frac{E^1}{a_{11}} \right). \end{aligned}$$

Step k : In general, we construct τ^k for $1 \leq k \leq M$ recursively by

$$\hat{\tau}^k = \inf \left\{ t \geq 0 : \sum_{i=1}^{k-1} \int_{\hat{\tau}^{i-1}}^{\hat{\tau}^i} (a_{kk} + \sum_{j=1}^{i-1} a_{kj}) ds + \int_{\hat{\tau}^{k-1}}^t (a_{kk} + \sum_{j=1}^{k-1} a_{kj}) ds \geq E^k \right\},$$

where $\hat{\tau}^0 := 0$ and $\sum_{i=1}^0 := 1$.

Then, on the set $\{\tau^1 < \tau^2 < \dots < \tau^M\}$, we have $(\tau_i)_{1 \leq i \leq M} = (\hat{\tau}_i)_{1 \leq i \leq M}$ in distribution. The same result holds for the other $(M! - 1)$ situations via permutation.

Consider the intensity model in section 4. Thus, $M = 2$, $a_{11} = a_1$, $a_{12} = a_2$, $a_{22} = \bar{a}_1$ and $a_{21} = \bar{a}_2$. Show that $(\tau, \bar{\tau})$ have the expressions (equal in distribution)

$$\tau = \begin{cases} \frac{E^1}{a_{11}}, & \text{if } \frac{E^1}{a_{11}} < \frac{E^2}{a_{22}}; \\ \frac{E^2}{a_{22}} + \frac{a_{11}}{a_{11} + a_{12}} \left(\frac{E^1}{a_{11}} - \frac{E^2}{a_{22}} \right), & \text{if } \frac{E^1}{a_{11}} > \frac{E^2}{a_{22}}, \end{cases}$$

and

$$\bar{\tau} = \begin{cases} \frac{E^1}{a_{11}} + \frac{a_{22}}{a_{22} + a_{21}} \left(\frac{E^2}{a_{22}} - \frac{E^1}{a_{11}} \right), & \text{if } \frac{E^1}{a_{11}} < \frac{E^2}{a_{22}}; \\ \frac{E^2}{a_{22}}, & \text{if } \frac{E^1}{a_{11}} > \frac{E^2}{a_{22}}. \end{cases}$$

Equivalently, on $\{\tau < \bar{\tau}\}$,

$$\begin{cases} E^1 = a_{11} \tau; \\ E^2 = a_{22} \tau + (\bar{\tau} - \tau)(a_{22} + a_{21}), \end{cases}$$

and on the set $\{\tau > \bar{\tau}\}$,

$$\begin{cases} E^2 = a_{22} \bar{\tau}; \\ E^1 = a_{11} \bar{\tau} + (\tau - \bar{\tau})(a_{11} + a_{12}). \end{cases}$$

The above total hazard construction procedure also yields the joint density of default times. To see this, since (E^1, E^2) have the joint density $f_{(E^1, E^2)}(t, u) = e^{-(t+u)}$, and the Jacobi determinant of (E^1, E^2) with respect to $(\tau, \bar{\tau})$ is

$$\left| \frac{\partial(E^1, E^2)}{\partial(\tau, \bar{\tau})} \right| = \begin{cases} a_{11}(a_{22} + a_{21}), & \text{if } \tau < \bar{\tau}; \\ a_{22}(a_{11} + a_{12}), & \text{if } \tau > \bar{\tau}. \end{cases}$$

Hence, we can obtain the joint density of $(\tau, \bar{\tau})$ by

$$f_{(\tau, \bar{\tau})}(T, U) = f_{(E^1, E^2)}(t, u) \left| \frac{\partial(E^1, E^2)}{\partial(\tau, \bar{\tau})} \right|,$$

which yields exactly the same joint density formula (12).

References

1. Yu, Fan. Correlated defaults in intensity-based models. *Mathematical Finance* 17(2) (2007): 155-173.