Packham, Natalie; Schlögl, Lutz; Schmidt, Wolfgang M.

Working Paper
Credit dynamics in a first passage time model with jumps

CPQF Working Paper Series, No. 21

Provided in Cooperation with:
Frankfurt School of Finance and Management

Suggested Citation: Packham, Natalie; Schlögl, Lutz; Schmidt, Wolfgang M. (2009) : Credit dynamics in a first passage time model with jumps, CPQF Working Paper Series, No. 21

This Version is available at:
http://hdl.handle.net/10419/40190

Terms of use:
Documents in EconStor may be saved and copied for your personal and scholarly purposes.
You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.
If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.
Credit dynamics in a first passage time model with jumps

Natalie Packham, Lutz Schlögl and Wolfgang M. Schmidt

September 2009

Authors:
Natalie Packham
Assistant Professor for Quantitative Finance
Frankfurt School of Finance & Management
Frankfurt/Main
n.packham@frankfurt-school.de

Lutz Schlögl
Fixed Income Quantitative Research
Nomura International Plc
London, UK
lutz.schloegl@nomura.com

Wolfgang M. Schmidt
Professor of Quantitative Finance
Frankfurt School of Finance & Management
Frankfurt/Main
w.schmidt@frankfurt-school.de

Publisher:
Frankfurt School of Finance & Management
Phone: +49 (0) 69 154 008-0 • Fax: +49 (0) 69 154 008-728
Sonnemannstr. 9-11 • D-60314 Frankfurt/M. • Germany
Credit dynamics in a first passage time model with jumps

Natalie Packham,* Lutz Schlögl† and Wolfgang M. Schmidt‡

September 2009

Abstract

The payoff of many credit derivatives depends on the level of credit spreads. In particular, the payoff of credit derivatives with a leverage component is sensitive to jumps in the underlying credit spreads. In the framework of first passage time models we extend the model introduced in [Overbeck and Schmidt, 2005] to address these issues. In the extended a model, a credit quality process is driven by an Itô integral with respect to a Brownian motion with stochastic volatility. Using a representation of the credit quality process as a time-changed Brownian motion, we derive formulas for conditional default probabilities and credit spreads. An example for a volatility process is the square root of a Lévy-driven Ornstein-Uhlenbeck process. We show that jumps in the volatility translate into jumps in credit spreads. We examine the dynamics of the OS-model and the extended model and provide examples.

Keywords: gap risk, credit spreads, credit dynamics, first passage time models, Lévy processes, general Ornstein-Uhlenbeck processes

JEL classification: G12, G13, G24, C69

MSC 2000 codes: 60G35, 60H05, 91B28

1 Introduction

Other than being subject exclusively to default risk, the payoff of some credit derivatives is determined explicitly by the level of CDS spreads. In this case, the dynamics of CDS spreads play a significant role in product valuation. Examples of such products are default swaptions and credit derivatives with a leverage component. The latter are in addition sensitive to gap risk, a risk that is linked to the occurrence of jumps in the evolution of credit spreads, even if such jumps do not lead to default. The model introduced here is suitable for valuing such credit derivatives.

To motivate this further, let us gather some stylised facts on credit spreads and let us consider the leveraged credit-linked note (LCLN) as an example of a product that features gap risk. We need the notion of a credit default swap (CDS) first. Given an underlying entity, such as a company, a CDS is a contract between two counterparties, the protection buyer and the protection seller, that insures the protection buyer against the default event (i.e., failure to fulfill a financial obligation) of the underlying entity. The protection buyer regularly pays a constant premium, the credit spread (or CDS spread), that is fixed at inception, up until maturity of the CDS or the default event, whichever occurs first. This stream of payments is termed the premium leg of the CDS. In return, the protection seller

---

* Natalie Packham, Centre for Practical Quantitative Finance, Frankfurt School of Finance & Management, Sonnemannstr. 9-11, 60314 Frankfurt am Main. n.packham@frankfurt-school.de
† Lutz Schlögl, Fixed Income Quantitative Research, Nomura International Plc, lutz.schoegl@nomura.com
‡ Wolfgang M. Schmidt, Centre for Practical Quantitative Finance, Frankfurt School of Finance & Management, Sonnemannstr. 9-11, 60314 Frankfurt am Main. w.schmidt@frankfurt-school.de
agrees to compensate the protection buyer for the loss incurred by default of the underlying entity at the time of default in case this occurs before maturity. This constitutes the protection leg of the CDS.

Following the principle of no-arbitrage, the value of a financial claim is considered to be the discounted expectation of its payoff under a risk-neutral measure. The fair credit spread (or fair CDS spread) is the CDS spread that makes the value of both legs equal. Since we only consider fair credit spreads in this paper, we just speak of credit spreads. The mapping of CDS spreads with respect to their maturity is called the term structure of CDS spreads.

CDS spreads and the term structure of credit spreads change over time. In general, two components of credit risk contribute to the behaviour of CDS spreads: jump-to-default risk and the risk of credit quality changes. We state some stylised facts about the shape and dynamics of the term structure. A wide variety of term structure shapes has been observed in the market, such as upward sloping, flat, hump-shaped and downward sloping curves [Helwege and Turner, 1999], [Zhou, 2001]. Another common observation is that credit spreads do not tend to zero as maturity tends to zero, but are significantly greater than zero [Duffie and Lando, 2001], [Zhou, 2001], [Duffie and Singleton, 2003] and [Lando, 2004]. This indicates that, for any time to maturity, market participants presume a positive probability of unexpected and instantaneous default.

There is also a significant amount of research that indicates that credit spreads are subject to sudden and unexpected jumps [Johannes, 2000], [Das, 2002], [Dai and Singleton, 2003], [Tauchen and Zhou, 2006] and [Zhang et al., 2009]. According to [Zhou, 2001], a more volatile jump component indicates that a firm is more likely to default on its short-maturity bonds than a firm with a more volatile diffusion component. In a recent empirical study on credit spreads, [Schneider et al., 2007] observe that CDS spreads exhibit frequent positive jumps, which typically affect CDS spreads of all maturities. These jumps are attributed to the arrival of bad news. Good news also affect the whole maturity spectrum, but tend to propagate gradually.

As an example of a credit derivative whose payoff is sensitive to the level of credit spreads, consider the leveraged credit-linked note. This note is particularly sensitive to jumps in CDS spreads, even if a jump does not lead to default. The principal idea is that an investor sells protection on an amount of default risk that is a multiple \( k \), the leverage factor, of his investment amount. The motivation for taking leveraged exposure is to earn a certain multiple \( \tilde{k} < k \) of the credit spread. Due to the leverage, his investment will most likely not suffice to compensate the loss incurred by default. Therefore, a trigger is agreed to terminate the structure while the cost of closing the position is still likely to be sufficiently covered by the investment amount. The time and cost of closing the position depends on the level of credit spreads, hence the investor is exposed mainly to spread risk, and to default risk only to a lesser extent. Furthermore, if a sufficiently high jump in the credit spread occurs, the cost of unwinding the transaction may exceed the investment amount, in which case the issuer of the note has to cover the missing amount required to close the position. This type of risk is called gap risk. Essentially, valuation of the note means determining the factor \((k - \tilde{k})\) that determines the spread received by the issuer for bearing gap risk.

There are generally two approaches to modelling credit risk: the structural and the reduced-form approach. In reduced-form models, default is not linked to economic variables, but is an unpredictable Poisson-type event, and the main object of the modeller’s attention is the hazard rate of the jump process describing default. This approach has been overwhelmingly popular with practitioners, its main advantage being its tractability: It is generally straightforward to fit a given term structure of CDS spreads and the techniques are very similar to those of interest rate modelling. The literature on this type of models is
vast, the papers by [Jarrow and Turnbull, 1995], [Lando, 1998] and [Duffie and Singleton, 1999] are only a few classic examples.

From the point of view of spread dynamics, modelling the default time as a totally unpredictable stopping time is not entirely satisfactory. Even with a low initial hazard rate, such a model will assign a non-negligible probability to the possibility of the credit defaulting without a prior movement in the credit spread. Defaults of this type are very uncommon in practice. The default swap market is efficient enough that default events are almost always preceded by a significant widening of credit spreads. It is this spread widening that is the real jump event that a market participant needs to worry about.

The ability of a model to assign probability mass to spread widening scenarios is constrained by the probability assigned to defaults in low spread scenarios, as the model must fit the initial credit spread. From a practical point of view, failure to assign enough probability to spread widening scenarios can lead to a dangerous underpricing of credit spread gap risk. The phenomenology of default that we are trying to capture is the following: a credit with a low default swap spread does not default “out of the blue”, but rather some kind of “regime switch” takes place, causing the credit spread to widen, after which the credit may either default or eventually recover.

We implement this idea via a first passage time model where a credit quality process exhibits stochastic volatility. In fact, the volatility process is a Levy-driven Ornstein-Uhlenbeck process. A jump in the volatility process is the “regime switch” we alluded to earlier. The current trend is to interpret the class of structural models in a wide sense to include any model where default is modelled as the first hitting time of a certain threshold by an abstract observable credit quality process. In this sense our model is structural, though the term “first passage time model” is technically more accurate. The structural approach, pioneered by [Merton, 1974], has been developed by [Black and Cox, 1976], [Longstaff and Schwartz, 1995], and many others.

As the name suggests, in a first passage time model, the computation of default probabilities is equivalent to computing the distributions of first passage times. The simplest case is that of a Brownian motion hitting a constant barrier, where a simple closed-form solution exists. This simple set-up, however, does not allow one to fit a given term structure of credit spreads. One approach to this problem in the past has been to use a time-dependent barrier, but already in the Brownian motion case this leads to tractability problems. Two further criticisms can be made of first passage time models in the simple diffusion case. Credit spreads only diffuse, whereas in reality they exhibit strong jump dynamics. Short term credit spreads in the model tend to zero, as a diffusion process needs a finite time to cross a barrier. [Zhou, 2001] proposes a model that overcomes both problems by modelling the credit quality process as a jump-diffusion. Computing first passage times for jump-diffusions is a very intractable problem, making the model by [Zhou, 2001] and a generalization by [Kiesel and Scherer, 2007] computationally very demanding. [Baxter, 2007] and [Cariboni and Schoutens, 2007] propose models where the credit quality process is a Lévy process. Again, the first passage time problem is difficult in this context.

[Overbeck and Schmidt, 2005] propose a simple solution to the problem of calibrating a first passage time model to a term structure of credit spreads. Instead of considering the hitting time of Brownian motion to a complicated, time-dependent barrier, they consider the first hitting time of a time-changed Brownian motion to a constant barrier. The time change is continuous, strictly increasing and deterministic. Because both the time change and the underlying Brownian motion are continuous, one can easily adapt the analytic formula from the simple Brownian case and obtain an analytic calibration to a term structure of default probabilities.

Our model builds directly on [Overbeck and Schmidt, 2005]. We consider a credit quality process $X$ to be a time-changed Brownian motion with a stochastic continuous and strictly
increasing time change independent of the Brownian motion. Our credit quality process \( X \) can also be represented as an Itô integral \( X = \int_0^\tau \sigma_u dW_u \) with a Brownian motion \( W \) and a càdlàg process \( \sigma \). Our standard example for the volatility process \( \sigma \) is the square root of an Ornstein-Uhlenbeck process driven by a subordinator, that is, a Lévy process with strictly increasing paths. In the time-changed Brownian motion interpretation, we can write \( X \) as

\[
X_t = B_t + \int_0^t \sigma_u dW_u.
\]

Because the time change \( \tau \) is continuous and independent of \( B \), we retain all the tractability of the deterministic case, while jumps in the volatility \( \sigma \) induce jumps in the credit spreads, even though the credit quality process \( X \) is continuous. This extended model is both tractable and at the same time allows for meaningful CDS spread dynamics. Furthermore, even though short-term credit spreads tend to zero (a problem which we consider to be secondary to achieving meaningful dynamics in a tractable manner), jump-to-default events can be approximated by introducing large jumps in the volatility process.

In a related paper [Packham et al., 2009], we show that the model can be calibrated to a wide set of term structures and that its implementation allows for efficient valuation of spread-risky claims. In this paper, we also give valuation examples for a leveraged credit-linked note. The structure of the time-changed Brownian motion may also prove powerful when valuing products that depend on the joint dynamics of several correlated underlyings. This has yet to be investigated.

The paper is structured as follows: We introduce some notation that is used throughout in Section 2. The Overbeck-Schmidt model and some of its properties are examined in Section 3. The extended model is introduced in Section 4. In Section 5, we show that jumps in the stochastic volatility of the credit quality process translate into jumps in default probabilities and credit spreads. Furthermore, we examine the dynamics by computing the distributions of conditional default probabilities and the averaged hazard rate (as a proxy for credit spreads). Some examples, including limiting cases, are also presented in this Section.

A result that is needed for computing conditional default probabilities, but that is not specific to the context of finance, is derived in Appendix B. Here, we are concerned with a particular class of continuous local martingales, called Ocone martingales, whose representation as a time-changed Brownian motion (by the famous Dambis, Dubins-Schwarz Theorem) features independent Brownian motion and time-change.

## 2 Notation and preliminaries

Throughout, let \((\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbf{P})\) be a complete probability space endowed with a filtration \((\mathcal{F}_t)_{t \geq 0}\), i.e., a family of sub-\(\sigma\)-algebras of \(\mathcal{F}\) such that \(\mathcal{F}_s \subseteq \mathcal{F}_t\), \(s \leq t\). The filtration represents the information available in the market, i.e., \(\mathcal{F}_t\) contains all events that are observable until time \(t\). In particular, \(\mathcal{F}_0\) is \(\mathbf{P}\)-trivial, i.e., for any \(A \in \mathcal{F}_0\), either \(\mathbf{P}(A) = 0\) or \(\mathbf{P}(A) = 1\). We assume that \((\mathcal{F}_t)_{t \geq 0}\) satisfies the usual hypotheses, i.e., \(\mathcal{F}_0\) contains all \(\mathbf{P}\)-null sets of \(\mathcal{F}\) and \((\mathcal{F}_t)_{t \geq 0}\) is right-continuous. We also assume that the probability space is rich enough to support any objects that we define. If not otherwise stated, all processes are \((\mathcal{F}_t)_{t \geq 0}\)-adapted. An \((\mathcal{F}_t)_{t \geq 0}\)-Brownian motion \(W\) is a Brownian motion \(W\) that is \((\mathcal{F}_t)_{t \geq 0}\)-adapted and such that for any \(s \leq t\), \(W_t - W_s\) is independent of \(\mathcal{F}_s\).

We omit explicit reference to the underlying entity when we speak of a default event, default time, etc.. Denote by \(\tau\) the random time of the default event. In our setup, \(\tau\) is an \((\mathcal{F}_t)_{t \geq 0}\)-stopping time. The distribution function of \(\tau\) conditional on the information flow \((\mathcal{F}_t)_{t \geq 0}\) is denoted by \(P(t,T) := \mathbf{P}(\tau \leq T | \mathcal{F}_t)\). For fixed maturity \(T\), \(P(t,T)\) is a process whose dynamics are driven by the information flow \((\mathcal{F}_t)_{t \geq 0}\). For each \(T\), the process \(P(t,T))_{t \leq T}\) is a martingale with a càdlàg modification. Furthermore, \(P(\cdot | \mathcal{F}_t)\) has a version that is a regular conditional probability, so we may treat \(P(\cdot | \mathcal{F}_t)(\omega)\) like a
probability measure for $P$-almost all $\omega \in \Omega$.

We assume that $P$ is a risk-neutral measure (presupposing existence of a risk-neutral measure), that is, $P$ is a probability measure equivalent to the real-world probability measure and such that discounted prices are $P$-martingales. It follows that the market is free of arbitrage.

Let $s(t, T)$ be the fair credit spread at time $t$ of a CDS with maturity $T$. Then $(s(t, T))_{t \leq T}$ is the spread process with maturity $T$ and $(s(t, T + t))_{t \geq 0}$ is the spread process with time to maturity $T$.

The link between conditional default probabilities and credit spreads is as follows: Let $r$ be the default-free interest rate, assumed to be constant for simplicity. Furthermore, assume that the payment at default is a fraction $(1 - R)$ of the notational amount, $R \in [0, 1)$. Entering into a CDS involves no initial cash-flow, that is, the market value of a CDS at inception is 0; in other words, the discounted fair values of the premium and the default legs are equal. From these considerations, one can show that the fair credit spread or fair CDS spread $s(t, T)$ at time $t$ is given by

$$s(t, T) = \frac{\int_t^T e^{-r(u-t)} dP(t,u)}{\int_t^T e^{-r(u-t)} (1 - P(t,u)) \, du} = e^{-r(T-t)} P(t,T) + \int_t^T r \, e^{-r(u-t)} P(t,u) \, du,$$

where the second line is obtained by integration by parts. On $\{\tau \leq t\}$ or for $t \geq T$, we set $s(t, T) = 0$.

3 Overbeck-Schmidt model

The model we propose is a generalisation of the Overbeck-Schmidt model (OS-model) [Overbeck and Schmidt, 2005], which we now introduce. The OS-model allows for straightforward analytic calibration to a given term structure of default probabilities. Although the model exhibits dynamics, these are fully determined by calibration to a given term structure of credit spreads. Our goal will be to extend the OS-model to allow for meaningful dynamics. As a precursor to the analysis of the more general model, we explore some characteristics of the OS-model dynamics. For a thorough analysis of the dynamics of the OS-model, see [Kammer, 2007] and [Packham, 2009].

3.1 Model specification

In the OS-model, the default time $\tau$ is determined as the first time that a credit quality process $X = (X_t)_{t \geq 0}$ hits a barrier $b < X_0$, that is,

$$\tau = \inf \{t \geq 0 : X_t \leq b\}.$$

The principal idea is to model $X$ as a time-changed Brownian motion. Given a Brownian motion $B$ and a deterministic, strictly increasing and continuous time transformation $\Lambda = (\Lambda_t)_{t \geq 0}$, with $\Lambda_0 = 0$, set

$$X_t := B_{\Lambda_t}, \quad t \geq 0.$$

Assume given the distribution $F(t)$, $t \geq 0$, of the default time. If the time-change $\Lambda$ is given by

$$\Lambda_t = \left(\frac{b}{N(-1) \left(\frac{F(t)}{2}\right)}\right)^2, \quad t \geq 0,$$

1It is straightforward to extend the setting to a non-constant interest rate by assuming that the short rate $r$ and the default indicator process $(1_{\{t \leq \tau\}})_{t \geq 0}$ are conditionally independent given $\mathcal{F}_t$, for any $t \geq 0$.
where $N^{(-1)}$ denotes the inverse of the Normal distribution function, then $\tau$ admits the distribution $F$. Furthermore, if the distribution of $\tau$ admits a density, then the time-change is absolutely continuous, and

$$
\Lambda_t = \int_0^t \sigma_s^2 \, ds,
$$

(3)

with $\sigma$ a nonnegative square-integrable function. The quadratic variation of $X$ is just $[X, X] = \Lambda$, so that there exists a representation of $X$ as a stochastic integral

$$
X_t = \int_0^t \sigma_s \, dW_s,
$$

for some Brownian motion $W$ (cf. Theorem 3.4.2 of [Karatzas and Shreve, 1998]). The volatility $\sigma$ can be interpreted as the default speed in the sense that the higher the default speed, the higher the likelihood of crossing the default barrier.

Overbeck and Schmidt use a multivariate extension of the model to value multi-name credit derivatives, such as first-to-default credit baskets. The payoff of these products does not depend on the level of credit spreads, only on whether default takes place or not. Nonetheless, by specification of the credit quality process, the model exhibits dynamics. These dynamics are fully determined by calibration to market-given default probabilities; it is not possible to assign different dynamics to the same term structure of default probabilities.

In the following, we explore some properties of the OS-model to gain some insight on how to extend the model to account for jumps in the evolution of default probabilities and credit spreads.

3.2 Conditional default probabilities

Recall that $P(t, T) = P(\tau \leq T|\mathcal{F}_t)$, $T \geq t$, has a version that is a regular distribution function and that $(1_{\{\tau > t\}} P(\tau \leq T|\mathcal{F}_t))_{t \geq 0}$ is a càdlàg semimartingale. Clearly, $\tau$ is an $(\mathcal{F}_t)_{t \geq 0}$-predictable stopping time.

**Proposition 3.1.** Let $X = (X_t)_{t \geq 0}$, with $X_t = B_{\Lambda_t}$ and $(\Lambda_t)_{t \geq 0}$ given by Equation (3), and let $\tau = \inf\{t \geq 0 : X_t \leq b\}$, $b < 0$. Then,

(i) On $\{\tau > t\}$, $P(t, T) = P(\tau \in (t, T]|\mathcal{F}_t)$; in particular, $P(t, T)$ is $\sigma(X_t)$-measurable on $\{\tau > t\}$.

(ii) For $t \leq T$, on $\{\tau > t\}$, the probability of default until $T$ conditional on $\mathcal{F}_t$ is given by

$$
P(t, T) = 2N\left(\frac{b - X_t}{\sqrt{\Lambda_T - \Lambda_t}}\right),
$$

(4)

where $N$ denotes the standard normal distribution function.

(iii) For any $T \geq 0$, the conditional default process $(P(t, T))_{t \geq 0}$ is continuous in $t$.

**Proof.** (i) Observe first that

$$
P(t, T) = 1_{\{\tau \leq t\}} + 1_{\{\tau > t\}} P(\tau \in (t, T]|\mathcal{F}_t).
$$

The events $\{\tau \in (t, T]\}$ and $\{\min_{t \leq u \leq T} X_u \leq b\}$ are equivalent, and the claim follows from the Markov property of $X$. 

7
(ii) The hitting time distribution of a Brownian motion starting at 0 is (see e.g. Section 2.6.A of [Karatzas and Shreve, 1998]),

\[ P \left( \min_{s \leq t} B_s < b \right) = 2 N \left( \frac{b}{\sqrt{t}} \right), \quad b < 0. \]  

(5)

On \( \{ \tau > t \} \), making use of the fact that \( X \) is a Markov process and taking into account that the time-change \( \Lambda \) is continuous,

\[
P(t, T) = P(\tau \in (t, T]|\mathcal{F}_t) = P \left( \min_{t<u \leq T} X_u \leq b | \mathcal{F}_t \right)
= P \left( \min_{t<u \leq T} B_{\Lambda_u} - B_{\Lambda_t} \leq b - B_{\Lambda_t} | B_{\Lambda_t} \right)
= P \left( \min_{0<u \leq \Lambda_T - \Lambda_t} B_{\Lambda_t+u} - B_{\Lambda_t} \leq b - B_{\Lambda_t} | B_{\Lambda_t} \right)
= 2 N \left( \frac{b - B_{\Lambda_t}}{\sqrt{\Lambda_T - \Lambda_t}} \right),
\]

where the last step follows since \((B_{\Lambda_t+u} - B_{\Lambda_t})_{u \geq \Lambda_t}\) is a Brownian motion independent of \( \mathcal{F}_t \) and \((b - B_{\Lambda_t})\) is \((\mathcal{F}_t)_{t \geq 0}\)-measurable.

(iii) Taking into account that \( N(\cdot), \sqrt{\cdot} \) and \( \Lambda \) are continuous, for any sequence \( t_n \rightarrow t \), as \( n \rightarrow \infty \), on \( \{ \tau > t \} \),

\[
\lim_{t_n \rightarrow t} P(t_n, T) = \lim_{t_n \rightarrow t} 2 N \left( \frac{b - X_{t_n}}{\sqrt{\Lambda_T - \Lambda_{t_n}}} \right) = 2 N \left( \frac{b - X_t}{\sqrt{\Lambda_T - \Lambda_t}} \right) = P(t, T).
\]

On \( \{ \tau < T \} \), to analyse the behaviour around default, observe that by the continuity of \( X \) and since \( X_\tau = b \), it follows easily that \( \lim_{t_n \uparrow \tau} P(t_n, T) = 1 \), as \( n \rightarrow \infty \) by Equation (4).

\[ \square \]

Inspection of Equation (4) reveals that \( P(t, T) \) depends on the time-change increment and on the distance of \( X_t \) to the barrier. The time-change, being deterministic, has limited impact on the dynamics of \( (P(t, T))_{t \geq 0} \). The last part of the proposition tells us that it is impossible to generate jumps in a default probability process when the time-change is deterministic and continuous. It is thus clear that the model is unsuitable for valuing gap risk – the probability of a gap event in the OS-model is zero. An important consequence of the proposition is that in the setting of the OS-model a necessary condition for jumps at random times in default probability processes is that the time-change be stochastic.

3.3 Distribution of conditional default probabilities

We compute the distribution of \( P(t, T) \). This will be insightful when we compare the OS-model with the extended model later.

**Proposition 3.2.** For \( T \geq 0 \) and \( t \leq T \), the distribution function of \( P(t, T) \) conditional on \( \{ \tau > t \} \) is

\[
P \left( P(t, T) \leq x | \tau > t \right) = \frac{P(P(t, T) \leq x)}{P(\tau > t)} = \frac{N(h_1(x)) - N(h_2(x))}{1 - 2N(b/\sqrt{\Lambda_t})}, \quad 0 \leq x < 1,
\]

with

\[
h_1(x) = \frac{N(-1) (\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} - b}{\sqrt{\Lambda_t}}
\]

\[
h_2(x) = \frac{N(-1) (\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} + b}{\sqrt{\Lambda_t}}.
\]
The conditional density of $P(t, T)$ is
\[
\frac{d}{dx} P(P(t, T) \leq x | \tau > t) = \frac{\sqrt{\Lambda_T - \Lambda_t}}{2 \sqrt{\Lambda_t} n \left( N^{-1}(x/2) \right)} (n(h_1(x)) - n(h_2(x))) \quad 0 < x < 1,
\]
where $n$ denotes the probability density function of a standard normal random variable.

Proof. Recall from part (i) of Proposition 3.1 that $P(t, T) = 1_{\{\tau \leq t\}} + 1_{\{\tau > t\}} P(\tau \in (t, T]|X_t)$. For $0 \leq x < 1$,
\[
\{P(t, T) \leq x\} = \{P(\tau \in (t, T]|X_t) \leq x, \tau > t\},
\]
which establishes
\[
P(P(t, T) \leq x | \tau > t) = \frac{P(P(t, T) \leq x)}{P(\tau > t)}, \quad x < 1.
\]
Using the fact that $\Lambda$ is continuous,
\[
P(P(t, T) \leq x) = P\left( 2N \left( \frac{b - X_t}{\sqrt{\Lambda_T - \Lambda_t}} \right) \leq x, \min_{0 < s \leq t} X_s > b \right)
\]
\[
= P\left( 2N \left( \frac{b - B_{\Lambda_t}}{\sqrt{\Lambda_T - \Lambda_t}} \right) \leq x, \min_{0 < s \leq \Lambda_t} B_s > b \right)
\]
\[
= P\left( B_{\Lambda_t} \geq -\left( N^{-1}(\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} - b \right), m_{\Lambda_t} > b \right),
\]
where $m_t = \min_{0 < s \leq t} B_s$ is the running minimum of $B$ at $t$. We have (see e.g. Section 2.8.A of [Karatzas and Shreve, 1998])
\[
P(B_t \geq a, m_t \leq c) = N(\sqrt{\frac{2c-a}{\sqrt{t}}}), \quad c \leq 0, a \geq c.
\]
Continuing above, yields
\[
P\left( B_{\Lambda_t} \geq -\left( N^{-1}(\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} - b \right), m_{\Lambda_t} > b \right)
\]
\[
= P\left( B_{\Lambda_t} \geq -N^{-1}(\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} + b \right)
\]
\[
- P\left( B_{\Lambda_t} \geq -N^{-1}(\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} + b, m_{\Lambda_t} \leq b \right)
\]
\[
= N\left( \frac{N^{-1}(\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} - b}{\sqrt{\Lambda_t}} \right) - N\left( \frac{N^{-1}(\frac{x}{2}) \sqrt{\Lambda_T - \Lambda_t} + b}{\sqrt{\Lambda_t}} \right).
\]
The density is obtained by differentiation.

Examples of conditional default densities are given in the left picture of Figure 1. Here, the conditional density of $P(\tau \in (1, 5]|X_1)$ is shown for initial default distributions $F(T) = 1 - e^{-hT}, \ T \geq 0, h \in \{0.03, 0.05, 0.07\}$. Consider for example the distribution of the default probability with $h = 0.03$. The initial 5-year default probability is approx. 0.14. It turns out that with a probability of roughly 1/2 the default probability in one year is below 0.05, and with a probability of roughly 1/5 it is above 0.2. Loosely speaking, with a high probability, the underlying entity will have either a very high or a very low credit quality in one year. This suggests a very volatile movement of the default probability process, which is explained as follows: In order to match the initial 5-year default probability, the credit
quality process must adopt a high default speed. The high volatility of the credit quality process then leads to a volatile default probability process.

One can also see (and show) that, as \( x \to 0 \), the density either converges to 0 or diverges. In fact, for sufficiently small remaining time to maturity, the density diverges. Loosely speaking, this effect can be explained by observing that, as \( t \) approaches maturity, the probability mass concentrates at 0 (conditional on no default).

The right picture of Figure 1 shows the conditional density of the term hazard rate \( \tilde{\lambda}(t, T) \), a proxy for the credit spread \( s(t, T) \). For details on the term hazard rate, see Appendix A. The conditional distribution and density are obtained easily by observing that \( P(t, T) \) is a monotone function of \( \tilde{\lambda}(t, T) \) (on \( \{ \tau > t \} \)).

Apart from the fact that the dynamics are fully specified by calibration to the given default probability distribution, we may ask how we can obtain distributions that are more realistic. The first issue will be solved by incorporating a stochastic volatility (i.e., stochastic time-change); the second issue will be addressed by incorporating jumps into the stochastic volatility.

**Remark 3.3.** An interesting point to note is that the dynamics are not influenced by the choice of the barrier \( b \), although at first sight of Proposition 3.2, one may be lead to think so. This follows from the relationship between the barrier and the time transformation. Indeed, using Equation (2), we can re-write \( h_1 \) and \( h_2 \) as follows: For any \( T \geq t \),

\[
h_{1,2}(x) = \frac{N(-1) \left( \frac{x}{2} \right)}{N(-1)(P(\tau \leq T)/2)} \left[ \left( \frac{N(-1)(P(\tau \leq t)/2)}{N(-1)(P(\tau \leq T)/2)} \right)^2 \right] \frac{1}{2} \pm \frac{1}{2} \\pm \frac{1}{2}.
\]

**4 First passage time model with jumps**

We have seen in the last section that the Overbeck-Schmidt model is unsuitable for valuing spread products, the principal reason being that the dynamics are completely determined by
calibrating the model to a given term structure of default probabilities (or credit spreads). We have further seen that the model is unable to produce random jumps in credit spreads, and that default probabilities and the hazard rate are quite volatile. On the other hand, the OS-model is easy to calibrate and conditional default probabilities are easy to compute.

The natural question is then: Can we extend the OS-model to allow for meaningful dynamics while at the same time maintaining its tractability?

More specifically, in the OS-model, if $X$ is a credit quality process, then on \{ $\tau > t$ \} we have $P(t, T) = F(t, T, X_t)$ for some function $F$ by the Markov property of $X$, and $(P(t, T))_{t \geq 0}$ is mainly driven by the credit quality process $X$. The dynamics of $X$ are completely determined by calibration to the spot curve, however. We rephrase our question as follows: Can we enrich the dynamics in the sense that we have $(X, \sigma)$ a credit quality process with stochastic volatility, as follows: Can we enrich the dynamics in the sense that $(X, \sigma)$ a credit quality process with variance $\sigma^2$ driven by a Lévy process, then, under some mild conditions, $(X, \sigma)$ is a Markov process and $\sigma$ plays the role of the process $Y$ mentioned above. We derive a formula for conditional default probabilities. In a more concrete setting, we specify the volatility process $\sigma$ as a square root of a Lévy-driven Ornstein-Uhlenbeck process. Typically, in applications, $\sigma$ is completely determined by calibration to the spot curve, however. We rephrase our question as follows: The credit quality process $X$ will be defined as a stochastic integral with respect to a Brownian motion, $X = \int_0^t \sigma_s \, dW_s$, where the integrand $\sigma$ is a càdlàg stochastic process independent of $W$. If, for example, $\sigma$ is the solution to an SDE driven by a Lévy process, then, under some mild conditions, $(X, \sigma)$ is a Markov process and $\sigma$ plays the role of the process $Y$ mentioned above. We derive a formula for conditional default probabilities. In a more concrete setting, we specify the volatility process $\sigma$ as the square root of a Lévy-driven Ornstein-Uhlenbeck process. Typically, in applications, the driving Lévy process will be a compound Poisson process with positive jumps, which reflects the occurrence of jumps as rare events.

4.1 Credit quality process with stochastic volatility

**Definition 4.1.** The credit quality process $X = (X_t)_{t \geq 0}$ of a risky entity is defined to be

$$X_t = \int_0^t \sigma_s \, dW_s, \quad t \geq 0,$$

where $W$ is an $(\mathcal{F}_t)_{t \geq 0}$-Brownian motion and $\sigma$ is a strictly positive $(\mathcal{F}_t)_{t \geq 0}$-adapted càdlàg process independent of $W$ with $P(\int_0^t \sigma_s^2 \, ds < \infty) = 1$, $t \geq 0$, and $\lim_{t \to \infty} \int_0^t \sigma_s^2 \, ds = \infty$ $\mathbf{P}$-a.s.

To emphasise the association with $\sigma$, we speak of $X$ as a credit quality process with volatility $\sigma$ or as a credit quality process with variance $\sigma^2$. Denote the quadratic variation process of $X$ by $\Lambda = (\Lambda_t)_{t \geq 0}$, with $\Lambda_t = \int_0^t \sigma_s^2 \, ds$. Observe that $\Lambda$ is continuous, strictly increasing and $(\mathcal{F}_t)_{t \geq 0}$-adapted.

As before, the default time $\tau$ of the risky entity associated with the credit quality process $X$ is the first time that $X$ hits a barrier $b < 0$:

$$\tau = \inf\{ t \geq 0 : X_t \leq b \}.$$

Analogous to the OS-model, we can express the credit quality process as a time-changed Brownian motion. Define the family of $(\mathcal{F}_{\tau_t})_{t \geq 0}$-stopping times $\tau_t = \inf\{ s \geq 0 : \Lambda_s > t \}$, $t \geq 0$. By application of the Theorem of Dambis, Dubins-Schwarz (see e.g. [Karatzas and Shreve, 1998, Chapter 3.4.B]) the process $B$, with $B_t = X_{\tau_t}$, $t \geq 0$, is an $(\mathcal{F}_{\tau_t})$-Brownian motion. Conversely, the credit quality process $X$ can be expressed as a time-changed Brownian motion, $X_t = B_{\Lambda_t}$, $t \geq 0$. We refer to $B$ as the DDS-Brownian motion of $X$. Furthermore, for every $t > 0$, $\Lambda_t$ is an $(\mathcal{F}_{\tau_t})_{t \geq 0}$-stopping time. Since $\Lambda$ is continuous, strictly increasing and $\lim_{t \to \infty} \Lambda_t = \infty$ $\mathbf{P}$-a.s., it follows that $\tau_{\Lambda_t} = \Lambda_{\tau_t} = t$. Clearly, the

---

2The requirement $\lim_{t \to \infty} \int_0^t \sigma_s^2 \, ds = \infty$ $\mathbf{P}$-a.s. ensures that $\tau < \infty$ $\mathbf{P}$-a.s., as will become clear later on.
credit quality process model of Definition 4.1 is a generalisation of the OS-model with an absolutely continuous time-change.

To compute conditional default probabilities, we shall use the fact that \( X \) is an Ocone martingale, that is, a martingale whose DDS-Brownian motion and quadratic variation (the time-change associated with the DDS-Brownian motion) are independent. That \( X \) is an Ocone martingale is proved in Proposition B.2 in Section B. We refer to Section B for a review and some results on time-changes and Ocone martingales.

4.2 Conditional default probabilities

We analyse the properties of conditional default probabilities as we did for the OS-model in Section 3.2. In particular, we are interested in deriving results analogous to Proposition 3.1. As in the OS-model, \( \tau \) is an \( (\mathcal{F}_t)_{t \geq 0} \)-predictable stopping time. For an analogue of the last part of Proposition 3.1, which states that the default probability process \( (P(t, T))_{t \geq 0} \) cannot jump, we have to be more specific about the volatility process. We shall show in Section 5.1 that, although the credit quality process is continuous, it is possible to specify a model in which default probability processes and credit spreads exhibit jumps. In the following, we state a formula for conditional default probabilities, analogous to Equation (4). Moreover, if \((X, \sigma)\) is a Markov process, then, on \( \{\tau > t\} \), \( P(t, T) = \sigma(X_t) \vee \sigma(\sigma_t)\)-measurable.

The fact that \( X \) is an Ocone martingale, together with Corollary B.5 of Appendix B are the key for establishing a formula for conditional default probabilities \( P(t, T) = P(\tau \leq T|\mathcal{F}_t), t \geq 0, T > t \).

**Proposition 4.2.** Let \( X \) be a credit quality process with volatility process \( \sigma \). Let \( \tau = \inf\{t \geq 0 : X_t \leq b\} \) be the associated default time. On \( \{\tau > t\} \), the probability of default until time \( T > t \), conditional on \( \mathcal{F}_t \), is given by

\[
P(\tau \leq T|\mathcal{F}_t) = \mathbb{E} \left( \frac{b - X_t}{\sqrt{\Lambda_T - \Lambda_t}} \bigg| \mathcal{F}_t \right) \quad \text{P-a.s..} \tag{7}
\]

**Proof.** Let \( B \) be the DDS-Brownian motion of \( X \), and recall that \( B_{\Lambda_t} = X_t, t \geq 0 \). By continuity of \( \Lambda \) and by properties of conditional expectation, \( \mathbb{P} \)-a.s., on \( \{\tau > t\}, \)

\[
P(\tau \leq T|\mathcal{F}_t) = P \left( \min_{0 < u \leq \Lambda_T} B_{\Lambda_t} \leq b | \mathcal{F}_t \right)
= P \left( \max_{0 < u \leq \Lambda_T} B_{\Lambda_t} - B_{\Lambda_t} \leq b - B_{\Lambda_t} | \mathcal{F}_t \right)
= P \left( \min_{0 < u \leq \Lambda_T} B_{\Lambda_t + u} - B_{\Lambda_t} \leq b - B_{\Lambda_t} | \mathcal{F}_t \right)
= P \left( \min_{0 < u \leq \Lambda_T - \Lambda_t} B_{\Lambda_t + u} - B_{\Lambda_t} \leq b - B_{\Lambda_t} | \mathcal{F}_t \vee \sigma(\Lambda_T) \right) \tag{8}
\]

The random time \( \Lambda_t \) is an \( (\mathcal{F}_t)_{t \geq 0} \)-stopping time, and with \( \mathcal{F}_{\tau_{\Lambda_t}} = \mathcal{F}_t \) it follows from Corollary B.5 that \( (B_{\Lambda_t + u} - B_{\Lambda_t})_{u \geq 0} \) is a Brownian motion independent of \( \mathcal{F}_t \vee \sigma(\Lambda_T) \subseteq \mathcal{F}_t \vee \mathcal{F}_{\tau_{\Lambda_t}} \).

On the other hand, the random variables \( \Lambda_T - \Lambda_t \) and \( b - B_{\Lambda_t} \) are \( \mathcal{F}_t \vee \sigma(\Lambda_T) \)-measurable. Hence, by the first passage time distribution of Brownian motion, cf. Equation (5), \( \mathbb{P} \)-a.s.,

\[
P \left( \min_{0 < u \leq \Lambda_T - \Lambda_t} B_{\Lambda_t + u} - B_{\Lambda_t} \leq b - B_{\Lambda_t} | \mathcal{F}_t \vee \sigma(\Lambda_T) \right) = 2N \left( \frac{b - B_{\Lambda_t}}{\sqrt{\Lambda_T - \Lambda_t}} \right).
\]

Inserting into Equation (8) yields Equation (7).

\[\square\]

**Remark 4.3.** It is easily verified that the conditional default distribution is well-defined, i.e., that the \( T \mapsto P(t, T) \) fulfills the properties of a distribution function, for every \( t \geq 0 \). This follows from the condition that \( \sigma \) be strictly positive, which implies that the time-change \( \Lambda \) is strictly increasing.

12
The proof of the following corollary is straightforward.

**Corollary 4.4.** Let $X$ be a credit quality process with volatility process $\sigma$, and assume further that $(X, \sigma)$ has the Markov property. Let $\tau$ be the associated default time. Then, for $T > t$, on \{\tau > t\}, the conditional default distribution is

$$P(\tau \leq T|\mathcal{F}_t) = P(\tau \leq T|X_t, \sigma_t) = \mathbb{E}\left(2N\left(\frac{b - X_t}{\sqrt{\Lambda T - \Lambda t}}\right)\right|X_t, \sigma_t) \quad \text{P-a.s..} \quad (9)$$

Finally, by setting $t = 0$ we obtain a formula for unconditional default probabilities:

**Corollary 4.5.** Let $X$ be a credit quality process with volatility process $\sigma$, and let $\tau$ be the associated default time. Then, the default distribution is given by

$$P(\tau \leq T) = 2\mathbb{E}\left(N\left(\frac{b}{\sqrt{\Lambda T}}\right)\right), \quad T \geq 0. \quad (10)$$

It can also be shown that the conditional default distribution $P(t, T)$, $T \geq t$, is absolutely continuous and has a density, see Section 6.9 of [Packham, 2009].

### 4.3 Variance as Lévy-driven Ornstein-Uhlenbeck process

We put the model to work by specifying the variance process $\sigma^2$ to be a mean-reverting process with jumps. Candidates as drivers for the variance process are Lévy processes: they incorporate jumps, and we can build Markov processes by specifying the dynamics of the variance with respect to Lévy processes.

As an explicit example we model the variance process as a Lévy-driven Ornstein-Uhlenbeck process, which we now introduce.

**Definition 4.6.** Let $Z = (Z_t)_{t \geq 0}$ be a subordinator, that is, a Lévy process with non-decreasing paths $\text{P-a.s.}$ The process $Y = (Y_t)_{t \geq 0}$, defined by

$$Y_t = Y_0 e^{-at} + \int_0^t e^{-a(t-s)} dZ_s, \quad t \geq 0, \quad a \in \mathbb{R},$$

is a Lévy-driven Ornstein-Uhlenbeck process (LOU process).

The LOU process $Y$ is the solution of the SDE

$$dY_t = -aY_t dt + dZ_t, \quad t \geq 0.$$

It is easily seen that, if $Z$ is a pure-jump process with positive jumps, then $Y$ moves up by jumps and decays exponentially in-between the jumps. Models where an asset price’s variance is driven by an LOU process are considered for example by [Barndorff-Nielsen and Shephard, 2001]. For details on LOU processes, see also [Norberg, 2004], Chapter 5 of [Schoutens, 2003] and Chapter 15.3.3 of [Cont and Tankov, 2004].

We specify the variance process as an LOU process that incorporates in addition a deterministic function of time – denoted by $\theta$ below –, which will turn out to be useful for calibration. In the following, when we speak of an LOU process, we mean the more general process given by Equation (12) below.

**Proposition 4.7.** Let $Z$ be an $(\mathcal{F}_t)_{t \geq 0}$-subordinator, let $a \in \mathbb{R}_+$ and let $\theta$ be a bounded and càdlàg function. Then a solution $\sigma^2$ to the SDE

$$d\sigma^2_t = a(\theta(t) - \sigma^2_t) dt + dZ_t \quad (11)$$

13
exists and is given by
\[ \sigma_t^2 = e^{-at} \sigma_0^2 + \int_0^t e^{-a(t-u)} a \theta(u) \, du + \int_0^t e^{-a(t-u)} \, dZ_u, \quad t \geq 0. \] (12)

Moreover, for any \( T > t \), the increment of the integrated process is
\[ \int_t^T \sigma_u^2 \, du = \left( 1 - e^{-a(T-t)} \right) \frac{\sigma_t^2}{a} + \int_t^T \theta(u) \left( 1 - e^{-a(T-u)} \right) \, du + \int_t^T \left( 1 - e^{-a(T-u)} \right) \, dZ_u. \] (13)

The credit quality process model with LOU variance process is then:

**Proposition 4.8.** Let \( W \) be an \((F_t)_{t \geq 0}\)-Brownian motion, and let \( \sigma^2 = (\sigma_t^2)_{t \geq 0}, \sigma_0^2 > 0 \), be as in Proposition 4.7, with \( Z \) independent of \( W \) and with \( \theta \) such that \( \sigma_t^2 > 0, \ t \geq 0 \). Then the stochastic process \( X = (X_t)_{t \geq 0} \)
\[ X_t = \int_0^t \sigma_s \, dW_s, \quad t \geq 0, \] (14)
is a credit quality process in the sense of Definition 4.1. Moreover, \((X, \sigma)\) is a Markov process with respect to \((F_t)_{t \geq 0}\).

**Proof.** That \( P\left( \int_0^t \sigma_s^2 \, ds < \infty \right) = 1, t \geq 0 \), is a consequence of Equation (13) and the fact that \( Z_t < \infty \) \( \text{P-a.s.} \) for any \( t \geq 0 \), since \( Z \) is a subordinator. Finally, \( \lim_{t \to \infty} \int_0^t \sigma_s^2 \, ds = \infty \), since for a subordinator \( Z, \lim_{t \to \infty} Z_t = \infty \) \( \text{P-a.s.} \) (see e.g. Section 3.1 of [Bertoin, 1998]). That \((X, \sigma)\) is a Markov process follows from Theorem V.32 of [Protter, 2005]. \( \square \)

A sample path of \( \sigma^2 \) and \( \Lambda = \int_0^t \sigma_s^2 \, ds \) is given in Figure 2. Here, \( Z \) is a compound Poisson process with positive jumps.

## 5 Dynamics

In this section we establish some results concerning the dynamics of conditional default probabilities and credit spreads, resp. term hazard rates. We show that in the LOU variance process model, jumps in the variance propagate into jumps in default probabilities and credit spreads. We compute the distribution of conditional default probabilities and the term hazard rate. For specification of the dynamics of default probabilities and credit spreads using the Itô formula we refer to Chapter 7 of [Packham, 2009].

### 5.1 Jumps in default probabilities and credit spreads

The continuity of the credit quality process \( X \) and the associated time-change \( \Lambda \) were essential to derive the formula for conditional default probabilities, Equation (7), from which credit spreads can be computed. But recall that we wished to build a model that incorporates jumps in credit spreads. This is established by the following propositions.

**Proposition 5.1.** Let \( X \) be a credit quality process with LOU variance process \( \sigma^2 \) as in Proposition 4.8 with \( \sigma^2 \) driven by a subordinator. Let \( \tau = \inf \{ t > 0 : X_t \leq b \} \) be the associated default time. Fix \( T > 0 \) and let \( (P(t,T))_{t \leq T} \) be the associated conditional default probability process. Then, for \( P \)-almost all \( \omega \in \{ \tau > t \} \), \( (P(t,T))_{t \leq T} \) is a process whose jumps are positive and
\[ \Delta \sigma_t^2(\omega) = 0 \iff \Delta P(t,T)(\omega) = 0, \quad \text{for all } T > t. \]
Figure 2: Example of variance process and credit quality process. Top: variance process $\sigma^2$, Equation (12) (continuous line, left axis); time-change $\Lambda$, Equation (13) (dashed line, right axis).
Second from top: credit quality process $X$, Equation (14) (continuous line, left axis); volatility $\sigma$ (dashed line, right axis).
Second from bottom: 5-year default probability process, with decaying time-to-maturity (continuous line) and with fixed time-to-maturity (dashed line).
Bottom: Term hazard rate computed from default probabilities, i.e., $-\ln(1-P(t,T))/\left(T-t\right)$ (continuous line), $-\ln(1-P(t,T+t))/T$ (dashed line).
Parameters: $a = 2$, $\theta \equiv 0.25$, $\sigma^2 = 0.25$. $\sigma^2$ is driven by a compound Poisson process with jump intensity $\lambda = 2$ and discrete jump size distribution with jump sizes 0.05, 0.5 with probabilities 0.95, 0.05, respectively. The barrier is $b = -3$. 
Moreover, for any \( t < T \), \( P \)-a.s.,
\[
\Delta P(t, T) = 2\mathbb{E} \left( N \left( \frac{b - X_t}{\sqrt{\Lambda_T - \Lambda_t}} \right) |X_t, \sigma_t^2 \right) - 2\mathbb{E} \left( N \left( \frac{b - X_t}{\sqrt{\Lambda_T - \Lambda_t}} \right) |X_t, \sigma_t^2 \right).
\]

\( \Delta P(t, T) \)

Proof. Abbreviate
\[
h(t, T) = \int_t^T \theta(u) \left( 1 - e^{-a(t-u)} \right) du
\]
and
\[
L_{t,T} = \int_t^T \left( 1 - e^{-a(T-u)} \right) dZ_u,
\]
so that, by Equation (13),
\[
\Lambda_T - \Lambda_t = \left( 1 - e^{-a(T-t)} \right) \frac{\sigma_t^2}{\alpha} + h(t, T) + \frac{L_{t,T}}{\alpha}.
\]

On \( \{ \tau > t \} \) and by the Markov property of \((X, \sigma)\), the conditional default probability at \( t \) until \( T \) is given by Equation (9), and by the independence of \( L_{t,T} \) and \((X, \sigma)\), a version of this conditional probability is given by \( g_{t,T}(X_t, \sigma_t) \) with
\[
g_{t,T}(x, y) := \mathbb{E} \left( 2N \left( \frac{b - x}{\sqrt{(1 - e^{-a(T-t)}) y^2/a + h(t, T) + L_{t,T}/a}} \right) \right).
\]

To derive the claim of the Proposition we require the following:

(i) \( L_{t-T} = L_{t,T} \) \( P \)-a.s., for any \( t \leq T \),

(ii) for any sequence \((t_n, x_n, y_n) \to (t, x, y)\),

\[
g_{t_n,T}(x_n, y_n) \to g_{t,T}(x, y), \quad P \text{-a.s.,}
\]

(iii) for \((b - x) < 0\), \( g_{t,T}(x, y) \) is strictly increasing in \( y \).

Property (i) is well-known, see e.g. [Sato, 1999, p. 6]. For (ii) observe that
\[
\frac{b - x_n}{\sqrt{(1 - e^{-a(T-t_n)}) y_n^2/a + h(t_n, T) + L_{t_n,T}/a}} \to \frac{b - x}{\sqrt{(1 - e^{-a(T-t)}) y^2/a + h(t, T) + L_{t,T}/a}}, \quad P \text{-a.s., as } n \to \infty,
\]

by (i) and since all the terms in the sum of the denominator converge and the limit of the denominator is greater than 0, for \( t < T \). Equation (17) is obtained by continuity of the Normal distribution and Dominated Convergence.

For (iii) observe that the denominator in Equation (16) is strictly increasing in \( y \) and that for \((b - x) < 0\), \( t \mapsto N((b - x)/\sqrt{t}) \) is strictly increasing.

Fix \( g_{t,T}(X_t, \sigma_t) \) as the version of the conditional default probability \( P(t, T) \) on \( \{ \tau > t \} \). Then, taking into account that \( X \) is continuous \( P \)-a.s., and that on \( \{ \tau > t \} \) we have \((b - X_t) < 0\), we obtain \( P \)-a.s. for any sequence \( t_n \uparrow t \),
\[
P(t-, T) = \lim_{t_n \uparrow T} g_{t_n,T}(X_{t_n}, \sigma_{t_n}) = g_{t,T}(X_t, \sigma_{t-}) \left\{ \begin{array}{ll}
g_{t,T}(X_t, \sigma_t), & \text{if } \Delta \sigma_t = 0 \\< g_{t,T}(X_t, \sigma_t), & \text{if } \Delta \sigma_t > 0 \end{array} \right\} = P(t, T).
\]
A straightforward consequence of this Proposition is that a jump of \( \sigma^2_t \) triggers a jump in all conditional default probability processes \( (P(t, T))_{t \leq T}, T \geq 0 \).

The variance process in the example of Figure 2 has a large jump, which is also clearly visible in the default probability process.

**Remark 5.2.** Observe that in the proof we have not used properties specific to the mean-reversion feature of the LOU process. In order to establish the statement for other Markov processes \((X, \sigma^2)\) where \( \sigma^2 \) is driven by a Lévy process, essentially, one must show that a corresponding version of Equation (17) holds.

For CDS spreads we have the following result:

**Proposition 5.3.** Let \( (s(t, T))_{0 \leq t \leq T} \) be the CDS spread process for maturity \( T \). Then \( (s(t, T))_{0 \leq t \leq T} \) is càdlàg, and for \( t \leq T \) and for \( P \)-almost all \( \omega \in \{ \tau > t \} \),

\[
\Delta s(t, T)(\omega) > 0 \iff (\Delta P(t, u)(\omega) > 0, \text{ for some } u \in (t, T]).
\]

**Proof.** The claim is derived using the relationship between the CDS spread and conditional default probabilities, Equation (1). Consider first the integral of the numerator of Equation (1). For any sequence \( t_n \to t \), as \( n \to \infty \), we have

\[
\lim_{n \to \infty} \int_{t_n}^{T} r e^{-r(u-t_n)} P(t_n, u) \, du = \lim_{n \to \infty} \left[ \int_{t_n}^{t} r e^{-r(u-t_n)} P(t_n, u) \, du + \int_{t}^{T} r e^{-r(u-t_n)} P(t_n, u) \, du \right] = 0.
\]

Then, for the numerator of Equation (1), it follows by Dominated Convergence, for any sequence \( t_n \uparrow t \) as \( n \to \infty \),

\[
\lim_{n \to \infty} \left[ e^{-r(T-t_n)} P(t_n, T) + \int_{t}^{T} r e^{r(u-t_n)} P(t_n, u) \, du \right] = e^{-r(T-t)} P(t, T) + \int_{t}^{T} r e^{-r(u-t)} P(t, u) \, du.
\]

Similarly, we obtain for the denominator,

\[
\lim_{n \to \infty} \int_{t_n}^{T} e^{-r(u-t_n)} (1 - P(t_n, u)) \, du = \int_{t}^{T} e^{-r(u-t)} (1 - P(t, u)) \, du.
\]

It follows that

\[
\frac{\Delta s(t, T)}{1 - R} = \frac{e^{-r(T-t)} P(t, T) + \int_{t}^{T} r e^{-r(u-t)} P(t, u) \, du}{\int_{t}^{T} e^{-r(u-t)} (1 - P(t, u)) \, du} - \frac{e^{-r(T-t)} P(t, T) + \int_{t}^{T} r e^{-r(u-t)} P(t, u) \, du}{\int_{t}^{T} e^{-r(u-t)} (1 - P(t, u)) \, du}
\]

\[
= 0, \quad \text{if } \Delta P(t, u) = 0, \quad t < u \leq T
\]

\[
> 0, \quad \text{if } \Delta P(t, u) > 0, \quad t < u \leq T.
\]

In particular, this establishes that \( \Delta s(t, T) > 0 \iff (\Delta P(t, u) > 0, \text{ for some } u \in (t, T]). \) Finally, for the converse statement observe that \( \Delta P(t, u) > 0 \) for some \( u \in (t, T] \) if and only if \( \Delta P(t, u) > 0 \) for all \( u \in (t, T] \) by the previous Proposition and by Equation (18).  

In a similar way, noting that $\Lambda$ is continuous, the corresponding result for CDS spread processes with a fixed time-to-maturity is obtained.

**Corollary 5.4.** Let $(s(t, t+T))_{t \geq 0}$ be the CDS spread process for time-to-maturity $T$. Then, under the assumptions of Proposition 5.3, $(s(t, t+T))_{t \geq 0}$ is càdlàg and for $t \leq T$ and $P$–a.a. $\omega \in \{\tau > t\}$,

$$\Delta s(t, t+T)(\omega) > 0 \iff \Delta P(t, u)(\omega) > 0, \text{ for some } u \in (t, T+t].$$

Obviously, the model excludes events where credit spreads jump for selected maturities only. However, this is compatible with the observation that credit spreads tend to jump together.

**Remark 5.5.** It is easily seen that a jump in the variance process $\sigma^2$ cannot lead to default $P$–a.s.. It suffices to recall that $\tau = \inf\{t > 0 : X_t \leq b\}$ is a predictable stopping time, whereas the jumps of the driving subordinator are totally inaccessible. As a consequence, one can show that the short-term credit spread vanishes, preventing jump-to-default events; see Section 7.2 of [Packham, 2009]. However, we shall see shortly that it is possible to specify the model in such a way that it allows for “near” jump-to-default events.

### 5.2 Distribution of conditional default probabilities

We compute the distribution of conditional default probabilities as we did for the OS-model in Section 3.3. Fixing $t \geq 0$ and $T \geq t$, the goal is to compute the distribution of $P(t, T)$.

Define

$$f_{\sigma^2_t}(X_t) := 2\mathbb{E}\left( \mathbb{N}\left( \frac{b - X_t}{\sqrt{\Lambda_t - \Lambda_t}} \right) \mid X_t, \sigma^2_t \right).$$

Observe that $f_{\sigma^2_t}$ is continuous and strictly decreasing so that its inverse $f_{\sigma^2_t}^{-1}$ exists.

**Proposition 5.6.** For $T \geq 0$ and $t \leq T$, the distribution function of $P(t, T)$ conditional on $\{\tau > t\}$ is

$$P(P(t, T) \leq x \mid \tau > t) = \mathbb{E}\left[ \mathbb{N}\left( -f_{\sigma^2_t}^{-1}(x) \sqrt{\Lambda_t} \right) - \mathbb{N}\left( \frac{2b - f_{\sigma^2_t}^{-1}(x)}{\sqrt{\Lambda_t}} \right) \right] / \left[ 1 - 2\mathbb{E}\left( \mathbb{N}\left( \frac{b}{\sqrt{\Lambda_t}} \right) \right) \right],$$

$$0 \leq x < 1.$$ 

**Proof.** We have

$$\{P(t, T) \leq x\} = \{P(\tau \in (t, T)\mid X_t, \sigma^2_t) \leq x, \tau > t\},$$

which establishes

$$P(P(t, T) \leq x \mid \tau > t) = \frac{P(P(t, T) \leq x)}{P(\tau > t)}, \quad x < 1.$$ 

By the independence of the DDS-Brownian motion $B$ and the pair $\sigma^2_t, \Lambda_t$, we obtain on
Figure 3: Distributions of $P(t, T)$ (left) and $\tilde{\lambda}(t, T)$ (right) conditional on $\{\tau > t\}$. The term hazard rate $\tilde{\lambda}(t, T)$ is an approximation of the credit spread, i.e., $s(t, T) \approx (1 - R)\tilde{\lambda}(t, T) \cdot 10^4$ (in basis points). As in Figure 1, we have $t = 1, T = 5$ and initial hazard rates of 3%, 5%, 7%. The diamonds mark the initial 5-year default probability $P(\tau \leq T)$. The model parameters are mean reversion $a = 3$, recovery rate $R = 0.4$, barrier $b = -3$; $\sigma^2$ is driven by compound Poisson processes with jump intensity $\lambda = 2$ and jump size distributions $\{0.1, 20\} \ (h = 0.03)$, $\{0.2, 50\} \ (h = 0.05)$, $\{0.4, 100\} \ (h = 0.07)$ with probabilities $\{0.95, 0.05\}$. In each case, the the initial variance $\sigma^2_0$ and the function $\theta$ were chosen to fit the given term structure.

$$\{\tau > t\},$$

$$P(P(t, T) \leq x | \Lambda_t, \sigma^2_t) = P\left(f_{\sigma^2_t}(X_t) \leq x, \min_{0 < s \leq t} X_s > b | \Lambda_t, \sigma^2_t\right)$$

$$= P\left(f_{\sigma^2_t}(B_{\Lambda_t}) \leq x, \min_{0 < s \leq \Lambda_t} B_s > b | \Lambda_t, \sigma^2_t\right)$$

$$= P\left(B_{\Lambda_t} \geq f_{\sigma^2_t}^{-1}(x), \min_{0 < s \leq \Lambda_t} B_s > b | \Lambda_t, \sigma^2_t\right)$$

$$= P\left(B_{\Lambda_t} \geq f_{\sigma^2_t}^{-1}(x) | \Lambda_t, \sigma^2_t\right) - P\left(B_{\Lambda_t} \geq f_{\sigma^2_t}^{-1}(x), \min_{0 < s \leq \Lambda_t} B_s \leq b | \Lambda_t, \sigma^2_t\right)$$

$$= N\left(-f_{\sigma^2_t}^{-1}(x) \sqrt{\Lambda_t}\right) - N\left(2b - f_{\sigma^2_t}^{-1}(x) \sqrt{\Lambda_t}\right),$$

using the joint distribution of a Brownian motion and its running minimum as in Section 3.3. The claim now follows by taking expectation.

Examples of the conditional distributions of $P(t, T)$ and the term hazard rate $\tilde{\lambda}(t, T)$, where the variance process follows an LOU process, are shown in Figure 3. The models are calibrated to the same term structures as the examples of the OS-model from Figure 1. Recall that the distributions in the OS-model suggested a very volatile movement of the credit quality process. By the introduction of jumps in the variance, the default speed in this example has been significantly reduced, while maintaining the default probabilities implied by the given term structures. Different distributions, and hence different dynamics, are obtained by changing the parameters of the LOU process. Further examples are discussed in the next section.
Figure 4: Distributions of $P(t,T)$ (left) and $\tilde{\lambda}(t,T)$ (right) conditional on $\{\tau > t\}$ for different dynamics. We choose $t = 1, T = 5$ and an initial hazard rate of 3%. The diamonds mark the initial 5-year default probability. The parameters are given in Table 1.

<table>
<thead>
<tr>
<th>Parameters</th>
<th>(a)</th>
<th>(b)</th>
<th>(c)</th>
<th>(d)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$a$</td>
<td>3</td>
<td>3</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>$b$</td>
<td>-3</td>
<td>-3</td>
<td>-3</td>
<td>-2</td>
</tr>
<tr>
<td>$\lambda$</td>
<td>0</td>
<td>2</td>
<td>1</td>
<td>0.0305</td>
</tr>
<tr>
<td>$F$</td>
<td>$[0.1 \ (0.95)]$</td>
<td>$[0.1 \ (0.95)]$</td>
<td>$1_{{(0.2 \cdot 10^{-4})}}$</td>
<td></td>
</tr>
<tr>
<td>$\sigma_0^2$</td>
<td>3.16</td>
<td>4.59</td>
<td>3.25</td>
<td>$10^{-6}$</td>
</tr>
<tr>
<td>$\theta$</td>
<td>$\in [0.23, 1.32]$</td>
<td>$\in [-0.22, 0.04]$</td>
<td>$\in [-1.10, 0.37]$</td>
<td>0</td>
</tr>
</tbody>
</table>

Table 1: Parameters of dynamics examples. In each case, the jump process $Z$ is a compound Poisson process with strictly positive jumps. The jump sizes are given by the first column and the corresponding probabilities in the second column of each matrix in the row of parameter $F$.

### 5.3 Examples

Calibration of default probabilities essentially means choosing the default barrier $b$ and the parameters of the variance process $\sigma^2$ of Equation (12), i.e., mean reversion $a$, jump intensity $\lambda$, jump size distribution $F$, deterministic function $\theta$ and initial variance $\sigma_0^2$. The last two parameters, $\theta$ and $\sigma_0^2$, will be chosen such as to minimize the root mean square error between given default probabilities and model default probabilities. The other parameters determine the dynamics of the model. It is important to note, that the deterministic function $\theta$ influences the dynamics and that the parameters for the dynamics influence the calibration of $\theta$. The overall calibration process is to assign parameters for the dynamics first and then to calibrate to the given spot curve. Again, details on calibration, calibration error, attainable term structures and examples are found in [Packham et al., 2009].

To illustrate the range of dynamics, Figure 4 shows the conditional distributions (conditional on no default) of the 5-year default probability and term hazard rate in one year for different parameter sets. The corresponding parameters are given in Table 1. In each case, the initial variance $\sigma_0^2$ and the function $\theta$ are calibrated to match default probabilities corresponding to an initial hazard rate of 3%.

Case (a) is a model with a deterministic time-change, and hence corresponds to the OS-model. Case (d) was chosen such that $\sigma_0^2 = 10^{-6}$ and $\theta \equiv 0$, so that the variance is very small.
until the first jump occurs. The jump size was chosen to be very large relative to the default barrier so that, heuristically, a single jump leads to default very quickly. Loosely speaking, case (d) can be considered an approximation of a reduced-form model with a deterministic and constant intensity: the credit quality process exhibits practically no movement, until the first jump occurs, which leads to default with a very high probability. This is also reflected in the jump intensity \( \lambda = 0.0305 \), which is approximately the initial hazard rate, and in \( P(\tau \in (1, 5]) | X_1, \sigma^2_1) \approx 1 - e^{-0.03^4} = 0.11308 \) conditional on no default until time 1.3 These two cases illustrate that the parameters can be classified into parameters that govern the jump part of the variance process, namely the jump intensity and jump size distribution, and parameters that control the continuous behaviour of the process in the sense that level of the function \( \theta \) determines the minimum volatility of the credit quality process at any time. By calibration to a term structure, a low level of jump activity leads to a higher minimum volatility and vice versa.

The characteristics of cases (b) and (c) are “in-between” cases (a) and (d): in both cases, the variance process exhibits jumps. However, the jump dynamics are moderate enough for the level of the variance process induced by \( \theta \) and \( \sigma^2_0 \), both of which are obtained by calibration to the given term structure, to be significantly above zero. In other words, the variance processes of both cases feature jumps and a significant minimum “default speed”.

6 Information flow and the pricing filtration

So far, we have made no assumptions about the filtration \( (\mathcal{F}_t)_{t \geq 0} \), other than that it is rich enough to carry \( (\mathcal{F}_t)_{t \geq 0} \)-adapted drivers of the credit quality process \( (X, \sigma^2) \). Following the principle of arbitrage pricing theory, the underlying filtration must be generated by the observable prices of traded assets, see e.g. Chapter 7 of [Hunt and Kennedy, 2004]. In general, a credit quality process \( (X, \sigma^2) \) is neither directly observable nor a traded asset.

Suppose now that we wish to price financial claims derived from credit spreads (or, equivalently, default, resp. survival probabilities). Application of a risk-neutral valuation formula with conditional default probabilities given by the model via Equation (7) (or credit spreads derived thereof) is justified only if \( (\mathcal{F}_t)_{t \geq 0} \) is generated by some observable information and if \( X \) and \( \sigma^2 \) are \( (\mathcal{F}_t)_{t \geq 0} \)-adapted. Otherwise, valuation of assets requires that prices are computed using a different – possibly coarser – filtration. One may think of a coarser filtration as the inavailability in the market of complete information about a company’s state.

Assuming independence of risk-free interest rates and the default indicator process, the filtration generated by risk-free zero-coupon bonds \( (B(t, T))_{T \geq t} \) and conditional default probabilities \( (P(t, T))_{T \geq t} \), \( t \geq 0 \), will be sufficient for this purpose; owing to the valuation formula for risky zero-coupon bonds, given by

\[
\mathbb{E} \left( e^{-\int_t^T r_s ds} 1_{\{\tau > T\}} \big| \mathcal{F}_t \right) = B(t, T) P(\tau > T | \mathcal{F}_t), \quad T \geq t,
\]

this filtration is equivalent to the filtration generated by risk-free and risky zero-coupon bonds (of all maturities).

The assumption that there is indeed a process that drives a company’s credit quality via the information available about the company may be justified by the stylised facts recorded earlier, namely that the arrival of news about a company affects CDS spreads of

\footnote{If the initial hazard rate is not constant, then a calibration where the variance moves purely by jumps cannot be attained. This is due to the fact that the jump intensity of the variance’s compound Poisson process is constant, whereas a non-constant, deterministic hazard rate requires the jump intensity to be non-constant and deterministic. The former can be incorporated by specifying the jump process as an additive process, which is a process with independent increments that is stochastically continuous, but whose increments are not necessarily stationary.}
all matures in a similar fashion. We shall assume that the credit quality of a firm isindeed driven by a process \((X, \sigma^2)\) as in Proposition 4.8 with \(\sigma^2\) an LOU process driven bya pure-jump subordinator (with respect to the filtration \((\mathcal{F}_t)_{t \geq 0}\)). Furthermore, we assumethat the parameters of the credit quality process are known, i.e., \(\theta, \sigma_0^2, a, b, c, \lambda, F\) are\(\mathcal{F}_0\)-measurable. The following result establishes that, under these assumptions, the formulafor conditional default probabilities is suitable for valuation.

**Proposition 6.1.** Let \((X, \sigma^2)\) be a credit quality process as in Proposition 4.8, with \(\sigma^2\)an LOU process driven by a subordinator. Let \(\mathcal{F}^P_t = \sigma((P(s, T)_{T > s}), 0 \leq s \leq t)\). Then,\(\sigma(X_t, \sigma_t^2) \subseteq \mathcal{F}^P_t\). Moreover, \(P(\tau \leq T|\mathcal{F}^P_t) = P(\tau \leq T|X_t, \sigma_t^2)\).

**Proof.** As in the proof of Proposition 5.1, abbreviate
\[
h(t, T) = \int_t^T \theta(u) \left(1 - e^{-a(t-u)}\right) \, du
\]
\[
L_{t,T} = \int_t^T \left(1 - e^{-a(T-u)}\right) \, dZ_u,
\]
so that, by Equation (13),
\[
\Lambda_T - \Lambda_t = \left(1 - e^{-a(T-t)}\right) \frac{\sigma_t^2}{a} + h(t, T) + \frac{L_{t,T}}{a}.
\]
On \(\{\tau > t\}\), by the Markov property of \((X, \sigma^2)\) and by the independence of \(L_{t,T}\) and\((X_t, \sigma_t^2)\), a version of the conditional probability at \(t\) until \(T\) is given by \(g_{t,T}(X_t, \sigma_t)\) with
\[
g_{t,T}(x, y) := \mathbb{E}\left(2N\left(\frac{b-x}{\sqrt{(1-e^{-a(T-s)})y^2/a + h(t, T) + L_{t,T}/a}}\right)\right).
\]
Observe that for fixed \(x, y \mapsto g_{T}(x, y)\) is invertible, and that for fixed \(y, x \mapsto g_{T}(x, y)\)is invertible (both mappings being strictly monotone). By Proposition 5.1, \(\Delta \sigma_t^2 > 0\) ifand only if \(\Delta P(t, T) > 0, T > t\). Moreover, the only “random” movements of \(\sigma^2\) occur injumps, any movement between jumps, conditional on the value attained by the last jump, isdeterministic. Consequently, if \(\Delta P(t, T) = 0, T > t\), and if the value of \(\sigma^2\) at the lastjump time is known, then \(\sigma^2\) is deterministic, and \(X_t\) may be determined by inversion. Onthe other hand, if \(\Delta P(t, T) > 0, T > t\), then by Equation (15),
\[
\Delta P(t, T) = g_T(X_t, \sigma_t^2) - g_T(X_{t-}, \sigma_{t-}^2), \quad t < T, \quad \mathbb{P}\text{-a.s.,}
\]
which can again be inferred by inversion (in particular, since \(X_{t-} = X_t\)). Hence, \((X_u, \sigma_u^2)_{u \leq t}\)is \(\mathcal{F}^P_t\)-measurable, and the first claim follows.

For the second claim observe that \(\mathcal{F}^{W,Z}_t \subseteq \mathcal{F}^P_t \subseteq \mathcal{F}_t\), where \((\mathcal{F}^{W,Z}_t)_{t \geq 0}\) denotes thefiltration generated by \(W\) and \(Z\). It follows that \(W\) is an \((\mathcal{F}^P_t)_{t \geq 0}\)-Brownian motion and \(Z\)is a subordinator with respect to \((\mathcal{F}^P_t)_{t \geq 0}\), and the claim follows. \(\square\)

Under some further assumptions on the deterministic function \(\theta\), it can be shown that\(\sigma(X_t, \sigma_t^2) = \sigma(P(t, T)_{T > t})\), see Section 9.1 of [Packham, 2009].

The valuation of financial claims in the LOU variance model driven by a compoundPoisson process is done by a combination of Monte Carlo simulation and numericalcomputation. Conditional on \((X_t, \sigma_t^2)\), default probabilities \(P(t, T) = P(\tau \leq T|X_t, \sigma_t^2), T > t\),can be computed numerically, using e.g. Panjer recursion to compute the distribution of thetime-change. Monte Carlo simulation then reduces to simulating \(X_t\) and \(\sigma_t\). The advantageof such an algorithm is that valuation of a product involving \(P(t, T)\) or \(s(t, T)\) requires simulationonly until time \(t\) instead of \(T\). For example, valuation of a default swaption requires simulationuntil expiry of the option instead of simulation until maturity of the underlyingCDS. See [Packham et al., 2009] for details and the full algorithm.
A The term hazard rate

We sometimes consider the term hazard rate defined by

\[ \tilde{\lambda}(t, T) = -\frac{\ln(1 - P(t, T))}{T - t}, \quad T > t, \quad P(t, T) < 1, \]  

(19)

as a proxy for the credit spread \( s(t, T)/(1 - R) \). The use of the term hazard rate is motivated by the fact that \( \tilde{\lambda}(t, T) \) is a function of \( P(t, T) \) instead of \( (P(t, T))_{T \geq t} \) as is the case for the spread \( s(t, T) \). That it may be considered a proxy for the credit spread is explained as follows: If the default time admits a (conditional) density, then the hazard rate at time \( t \) is the mapping \( T \mapsto \lambda(t, T) \) defined by

\[ \lambda(t, T) = -\frac{d}{dT} \ln(1 - P(t, T)), \quad T \geq t. \]

It follows easily that

\[ 1 - P(t, T) = e^{-\int_t^T \lambda(t, u) du}, \]

which, together with Equation (1) yields the relationship

\[ \frac{s(t, T)}{1 - R} \int_t^T e^{-r(u-t)}(1 - P(t, u)) du = \int_t^T \lambda(t, u) e^{-r(u-t)}(1 - P(t, u)) du. \]

An approximation of the right-hand side is

\[ \tilde{\lambda}(t, T) \int_t^T e^{-r(u-t)}(1 - P(t, u)) du, \]

with

\[ \tilde{\lambda}(t, T) = \frac{\int_t^T \lambda(t, u) du}{T - t}, \]

which yields the well-known credit triangle

\[ \frac{s(t, T)}{1 - R} \approx \tilde{\lambda}(t, T). \]

B Ocone martingales

The famous Theorem of Dambis, Dubins-Schwarz (DDS-Theorem) (see e.g. [Karatzas and Shreve, 1998, Chapter 3.4.B]) states that a continuous local martingale has a representation as a time-changed Brownian motion. Ocone martingales are those continuous local martingales whose DDS-Brownian motion and associated time-change are independent, see [Ocone, 1993], [Dubins et al., 1993], [Vostrikova and Yor, 2000]. We study the relationship between stochastic integrals (with respect to Brownian motion) and Ocone martingales. The result is applied in Section 4.2 to derive conditional default probabilities in the credit quality process model.

Assume given a filtered probability space \( \Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, P \) that satisfies the usual hypotheses. For a continuous local martingale \( M \), let us call the Brownian motion in its representation as a time-changed Brownian motion the DDS-Brownian motion of \( M \), and let us denote denote the quadratic variation of \( M \) by \([M, M]\).

**Definition B.1.** A continuous local martingale \( M = (M_t)_{t \geq 0}, M_0 = 0 \), is an Ocone martingale, if its DDS-Brownian motion is independent of \([M, M]\).
We are interested in identifying Ocone martingales that are defined as stochastic integrals with respect to a Brownian motion. [Vostrikova and Yor, 2000] provide a characterisation for adapted, continuous integrands. We shall allow adapted càdlàg processes as integrands.

**Proposition B.2.** Let \( W = (W_t)_{t \geq 0} \) be an \( (\mathcal{F}_t)_{t \geq 0} \)-Brownian motion and let \( \sigma = (\sigma_t)_{t \geq 0} \) be an \( (\mathcal{F}_t)_{t \geq 0} \)-adapted, strictly positive càdlàg process such that \( \mathbf{P}(\int_0^t \sigma_s^2 \, ds < \infty) = 1, \ t \geq 0, \) and \( \mathbf{P}(\lim_{t \to \infty} \int_0^t \sigma_s^2 \, ds = \infty) = 1. \) Let \( M = (M_t)_{t \geq 0} \) be the continuous local martingale given by

\[
M_t = \int_0^t \sigma_s \, dW_s, \quad t \geq 0.
\]

Then, \( W \) and \( \sigma \) are independent if and only if \( M \) is an Ocone martingale.

Observe that \( [M, M] = \int_0^\infty \sigma_s^2 \, ds. \) For the proof, we need the following Lemma.

**Lemma B.3.** Let \( W, \sigma, M \) be as in Proposition B.2. Then, \( \sigma([M, M], 0 \leq t < \infty) = \sigma(\sigma_t^2, 0 \leq t < \infty). \)

**Proof.** Clearly, \( \sigma([M, M], 0 \leq t < \infty) \subseteq \sigma(\sigma_t^2, 0 \leq t < \infty). \) For the converse statement, observe first that \( [M, M] \) is differentiable a.e. (with respect to Lebesgue measure), and for every such \( t, [M, M]_t = \sigma_t^2. \) Conversely, at any point \( t \) where \( \sigma^2 \) is continuous, \( [M, M] \) is differentiable, for suppose that \( \sigma^2 \) is continuous at \( t, \) but

\[
\lim_{h \downarrow 0} \frac{[M, M]_{t+h} - [M, M]_t}{h} \neq \lim_{h \downarrow 0} \frac{[M, M]_{t+h} - [M, M]_t}{h},
\]

this is a contradiction to \( [M, M]' = \sigma^2 \) a.e. and the continuity of \( \sigma^2 \) at \( t. \) Now, let \( t \) be a point where \( [M, M] \) is not differentiable; i.e., \( \sigma^2 \) has a jump at \( t. \) From the right-continuity of \( \sigma^2, \) it follows that there exists \( h > 0 \) such that \( [M, M] \) is differentiable on \((t, t+h).\) By the Mean Value Theorem, there exists \( s_h \in (t, t+h) \) such that

\[
\frac{[M, M]_{t+h} - [M, M]_t}{h} = \sigma_{s_h}^2.
\]

Moreover, by the right-continuity of \( \sigma^2, \) taking the right limit,

\[
\lim_{h \downarrow 0} \frac{[M, M]_{t+h} - [M, M]_t}{h} = \lim_{h \downarrow 0} \sigma_{s_h}^2 = \sigma_t^2,
\]

and \( \sigma(\sigma_t^2, 0 \leq t < \infty) \subseteq \sigma([M, M], 0 \leq t < \infty) \) follows (see e.g. Lemma 1.13 of [Kallenberg, 2001]). \( \square \)

**Proof of Proposition B.2.** Define the family of stopping times \( (\tau_t)_{t \geq 0} \) by

\[
\tau_t = \inf \{ s \geq 0 : \int_0^s \sigma_u^2 \, du > t \}, \quad t \geq 0.
\]

Then \( B = (B_t)_{t \geq 0}, \) given by \( B_t = M_{\tau_t} = \int_0^{\tau_t} \sigma_s \, dW_s, \ t \geq 0, \) is the DDS-Brownian motion of \( M. \) It is an \( (\mathcal{F}_t)_{t \geq 0} \)-Brownian motion.

The “only if” part: Let \( C \) be the space of real-valued continuous functions on \( \mathbb{R}_+ \) and let \( D \) be the space of real-valued càdlàg functions on \( \mathbb{R}_+. \) Denote by \( (C, \mathcal{B}(C)), \) resp. \( (D, \mathcal{B}(D)), \) the measurable space of real-valued continuous, resp. càdlàg , functions on \( \mathbb{R}_+ \) endowed with the \( \sigma \)-algebra generated by the finite-dimensional cylinder sets of \( C, \) resp. \( D. \) For every \( \Gamma \in \mathcal{B}(C) \) and \( \Delta \in \mathcal{B}(D) \) we show that

\[
\mathbf{P}(B \in \Gamma, \sigma \in \Delta) = \mathbf{P}(B \in \Gamma) \mathbf{P}(\sigma \in \Delta). \quad (20)
\]

\footnote{Under a suitable metric on \( C, \) resp. \( D, \) the \( \sigma \)-algebra \( \mathcal{B}(C), \) resp. \( \mathcal{B}(D), \) corresponds to the \( \sigma \)-algebra generated by the open sets (with respect to the metric) of \( C, \) resp. \( D, \) see e.g. [Shiryaev, 1996, Section II.\S2] or [Karatzas and Shreve, 1998, Sections 2.4 and 6.2].}
That $B$ is independent of $[M, M]$ and hence an Ocone martingale then follows by Lemma B.3.

It is straightforward that Equation (20) holds for sets $\Delta$ with $P(\sigma \in \Delta) \in \{0, 1\}$. Choose $\Delta$ such that $P(\sigma \in \Delta) \in (0, 1)$, and denote by $\mathcal{D}$ the $\sigma$-algebra generated by $\{\sigma \in \Delta\}$. By properties of conditional expectation,

$$P(B \in \Gamma, \sigma \in \Delta) = E(1_{\sigma \in \Delta} P(B \in \Gamma | \mathcal{D})). \quad (21)$$

Writing $D_1 = \{\sigma \in \Delta\}$ and $D_2 = \{\sigma \notin \Delta\}$, it is easy to check that a version of the conditional probability of $A \in \mathcal{F}$ with respect to $\mathcal{D}$ is given by

$$P(A | \mathcal{D})(\omega) = \sum_{i=1,2} P(A \cap D_i) / P(D_i) 1_{\{D_i\}}(\omega), \quad \omega \in \Omega.$$ 

Fix this version of the conditional probability. For every $\omega \in \Omega$, $P(\cdot | \mathcal{D})(\omega)$ is a probability measure (and thus it is a variant of the regular conditional probability with respect to $\mathcal{D}$). Moreover, $P(\cdot | \mathcal{D})(\omega) \ll P$, i.e., $P(\cdot | \mathcal{D})(\omega)$ is absolutely continuous with respect to $P$. It follows, e.g., by Theorem 14 of [Protter, 2005, Section II.5], that $\int_0^s \sigma_t \, dW_t$ computed under the law $P(\cdot | \mathcal{D})(\omega)$ and $M$ are $P(\cdot | \mathcal{D})(\omega)$-indistinguishable.

By independence of $W$ and $\mathcal{D}$ it follows that $W$ is a Brownian motion under $P(\cdot | \mathcal{D})(\omega)$, and hence $M$ is a continuous local martingale under $P(\cdot | \mathcal{D})(\omega)$. The quadratic variation, as a limit in probability, is invariant to absolutely continuous changes in measure. Hence, by the Lévy-characterisation of Brownian motion, $B$ is an $(\mathcal{F}_t)$-Brownian motion under $P(\cdot | \mathcal{D})(\omega)$, in other words $P(B \in \cdot | \mathcal{D}) = P(B \in \cdot)$ is the Wiener measure on $(C, B(C))$. Finally, inserting into Equation (21) yields

$$E(1_{\sigma \in \Delta} P(B \in \Gamma | \mathcal{D})) = E(1_{\sigma \in \Delta} P(B \in \Gamma)) = P(B \in \Gamma) P(\sigma \in \Delta). \quad (22)$$

The “if” part: Now suppose that $M$ is an Ocone martingale, i.e., $B$ and $[M, M]$ are independent. We have

$$W_t = \int_0^t \frac{1}{\sigma_s} \, dM_s = \int_0^{[M, M]} \frac{1}{\sigma_s} \, dB_s, \quad P\text{-a.s.},$$

where the last part follows from [Karatzas and Shreve, 1998, Proposition 3.4.8]. Now it can be shown that, for $\Gamma, \Delta \in B(C)$,

$$P(W \in \Gamma, [M, M] \in \Delta) = P(W \in \Gamma) P([M, M] \in \Delta)$$

using the same technique as in the “only if” part of the proof. That $W$ and $\sigma$ are independent then follows from Lemma B.3. $\Box$

Remark B.4. For simplicity, we have required that the integrand $\sigma$ be strictly positive. It is not hard to extend Proposition B.2 to integrands that are $P$–a.s. nonzero, see e.g. [Vostrikova and Yor, 2000, Theorem 3].

In Proposition B.2 we started out with a continuous local martingale defined as a stochastic integral with respect to a Brownian motion. We may also consider the converse, where we define a continuous local martingale $M$ as a Brownian motion and an independent time-change. Then, a sufficient condition for $M$ to have a representation as a stochastic integral with respect to a Brownian motion is that $[M, M]$ be an absolutely continuous function of $t$ for $P$–almost every $\omega \in \Omega$, cf. [Karatzas and Shreve, 1998, Theorem 3.4.2].

We conclude this section by establishing some properties related to Proposition B.2 that are used in Section 4.2. For a stochastic process $X$, we denote by $(\mathcal{F}_t^X)_{t \geq 0}$ the filtration generated by $X$. 

25
Corollary B.5. Let $M = \int_0^\tau \sigma_u \, dW_u$ be as in Proposition B.2, and let $B$ the DDS-Brownian motion of $M$. Define the family of stopping times $(\tau_t)_{t \geq 0}$ by $\tau_t = \inf\{ s \geq 0 : \int_0^s \sigma_u^2 \, du > t \}$, $t \geq 0$. Furthermore, let $S$ be an $P$–a.s. finite $(\mathcal{F}_t)$-stopping time and define $B = (B_u)_{u \geq 0}$, with $\bar{B}_u := B_{S+u} - B_S$. If $M$ is an Ocone martingale, then

(i) $\bar{B}$ is an $(\mathcal{F}_t^B)_{t \geq 0}$-Brownian motion independent of $\sigma$;

(ii) $(\mathcal{F}_t^B)_{t \geq 0}$ and $\mathcal{F}_\tau$ are conditionally independent given $\mathcal{F}_\infty$, the $\sigma$-algebra generated by $\sigma$;

(iii) $\bar{B}$ is independent of $\mathcal{F}_\tau \vee \mathcal{F}_\sigma$, the smallest $\sigma$-algebra containing $\mathcal{F}_\tau$ and $\mathcal{F}_\sigma$.

Proof. (i) By the properties of Brownian motion, $\bar{B}$ is a Brownian motion independent of $\mathcal{F}_\tau$, cf. [Karatzas and Shreve, 1998, Theorem 2.6.16]. In the notation of the previous proof, let $\Delta \in B(D)$ and let $\emptyset$ be the $\sigma$-algebra generated by $\{ \sigma \in \Delta \}$. Since $B$ is a Brownian motion under $P(\cdot|\emptyset)(\omega)$, for $P$–almost all $\omega \in \Omega$, so is $\bar{B}$, and the first claim follows in analogy to Equation (22).

(ii) The second claim is a generalisation of the first claim. Since $B$ is independent of $\sigma$, we have $P(B \in \Gamma|\mathcal{F}_\infty^\sigma) = P(B \in \Gamma) P – a.s., \Gamma \in B(C)$. Moreover, there exists a version $Q$ of $P(B \in \cdot|\mathcal{F}_\infty^\sigma)$ that is a regular conditional distribution of $B$ with respect to $\mathcal{F}_\infty$. It follows that $\bar{B}$ is a Brownian motion under $Q$ and hence $\bar{B}$ is a Brownian motion independent of $\mathcal{F}_\tau$ under $Q$, which is the required result.

(iii) Since $\mathcal{F}_\infty^\sigma$ and $\mathcal{F}_\tau$ are conditionally independent given $\mathcal{F}_\infty$, $P$–a.s. (see Proposition 6.6 of [Kallenberg, 2001] for the first statement),

\[ P(\bar{B} \in \Gamma|\mathcal{F}_\infty^\sigma \vee \mathcal{F}_\tau) = P(\bar{B} \in \Gamma|\mathcal{F}_\infty^\sigma) = P(\bar{B} \in \Gamma), \quad \Gamma \in B(C). \]

References


L. Vostrikova and M. Yor. Some invariance properties (of the laws) of Ocone’s martingales. *In: Séminaire de Probabilités XXXIV*, Springer, pages 417–431, 2000. [One page of this article is misplaced: Page 421 should be read after page 425, not page 420.]


<table>
<thead>
<tr>
<th>No.</th>
<th>Author/Title</th>
<th>Year</th>
</tr>
</thead>
<tbody>
<tr>
<td>129.</td>
<td>Hankir, Yassin / Rauch, Christian / Umber, Marc</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>It’s The Market Power, Stupid! – Stock Return Patterns in International Bank M&amp;A</td>
<td></td>
</tr>
<tr>
<td>128.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Outline of a Darwinian Theory of Money</td>
<td></td>
</tr>
<tr>
<td>127.</td>
<td>Cremers, Heinz / Walzner, Jens</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Modellierung des Kreditrisikos im Portfoliofall</td>
<td></td>
</tr>
<tr>
<td>126.</td>
<td>Cremers, Heinz / Walzner, Jens</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Modellierung des Kreditrisikos im Einwertpapierfall</td>
<td></td>
</tr>
<tr>
<td>125.</td>
<td>Heidorn, Thomas / Schmaltz, Christian</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Interne Transferpreise für Liquidität</td>
<td></td>
</tr>
<tr>
<td></td>
<td>The economic function of credit rating agencies - What does the watchlist tell us?</td>
<td></td>
</tr>
<tr>
<td>123.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>A Neurolinguistic Approach to Performativity in Economics</td>
<td></td>
</tr>
<tr>
<td>122.</td>
<td>Winkler, Adalbert / Vogel, Ursula</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Finanzierungsstrukturen und makroökonomische Stabilität in den Ländern Südosteuropas, der Türkei und in den GUS-Staaten</td>
<td></td>
</tr>
<tr>
<td>121.</td>
<td>Heidorn, Thomas / Rupprecht, Stephan</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Einführung in das Kapitalstrukturmanagement bei Banken</td>
<td></td>
</tr>
<tr>
<td>120.</td>
<td>Rossbach, Peter</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Die Rolle des Internets als Informationsbeschaffungsmedium in Banken</td>
<td></td>
</tr>
<tr>
<td>119.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Diversity Management and diversitätsbasiertes Controlling: Von der „Diversity Scorecard“ zur „Open Balanced Scorecard“</td>
<td></td>
</tr>
<tr>
<td>118.</td>
<td>Hölscher, Luise / Clasen, Sven</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Erfolgsfaktoren von Private Equity Fonds</td>
<td></td>
</tr>
<tr>
<td>117.</td>
<td>Bannier, Christina E.</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Is there a hold-up benefit in heterogeneous multiple bank financing?</td>
<td></td>
</tr>
<tr>
<td>116.</td>
<td>Roßbach, Peter / Gießamer, Dirk</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Ein eLearning-System zur Unterstützung der Wissensvermittlung von Web-Entwicklern in Sicherheitsthemen</td>
<td></td>
</tr>
<tr>
<td>115.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Kulturelle Hybridisierung und Wirtschaftstransformation in China</td>
<td></td>
</tr>
<tr>
<td>114.</td>
<td>Schalast, Christoph:</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Staatsfonds – „neue“ Akteure an den Finanzmärkten?</td>
<td></td>
</tr>
<tr>
<td>113.</td>
<td>Schalast, Christoph / Alram, Johannes</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Konstruktion einer Anleihe mit hypothekarischer Besicherung</td>
<td></td>
</tr>
<tr>
<td>112.</td>
<td>Schalast, Christoph / Bolder, Markus / Radünz, Claus / Siepmann, Stephanie / Weber, Thorsten</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Transaktionen und Servicing in der Finanzkrise: Berichte und Referate des Frankfurt School NPL Forums 2008</td>
<td></td>
</tr>
<tr>
<td>111.</td>
<td>Werner, Karl / Moorman, Jürgen</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Efficiency and Profitability of European Banks – How Important Is Operational Efficiency?</td>
<td></td>
</tr>
<tr>
<td>110.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Moralphische Gefühle als Grundlage einer wohlstandschauffenden Wettbewerbsordnung: Ein neuer Ansatz zur erforschung von Sozialkapital und seine Anwendung auf China</td>
<td></td>
</tr>
<tr>
<td>109.</td>
<td>Heidorn, Thomas / Kaiser, Dieter G. / Roder, Christoph</td>
<td>2009</td>
</tr>
<tr>
<td></td>
<td>Empirische Analyse der Drawdowns von Dach-Hedgefonds</td>
<td></td>
</tr>
<tr>
<td>108.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2008</td>
</tr>
<tr>
<td></td>
<td>Neuroeconomics, Naturalism and Language</td>
<td></td>
</tr>
<tr>
<td>107.</td>
<td>Schalast, Christoph / Benita, Barten</td>
<td>2008</td>
</tr>
<tr>
<td></td>
<td>Private Equity und Familienunternehmen – eine Untersuchung unter besonderer Berücksichtigung deutscher Maschinen- und Anlagenbauunternehmen</td>
<td></td>
</tr>
<tr>
<td>106.</td>
<td>Bannier, Christina E. / Grote, Michael H.</td>
<td>2008</td>
</tr>
<tr>
<td>105.</td>
<td>Herrmann-Pillath, Carsten</td>
<td>2008</td>
</tr>
<tr>
<td></td>
<td>The Naturalistic Turn in Economics: Implications for the Theory of Finance</td>
<td></td>
</tr>
<tr>
<td>104.</td>
<td>Schalast, Christoph (Hrsg.) / Schanz, Kay-Michael / Scholl, Wolfgang</td>
<td>Aktionärsverschlag in der AG falsch verstanden? Die Leica-Entscheidung des LG Frankfurt am Main</td>
</tr>
<tr>
<td>103.</td>
<td>Bannier, Christina E. / Münch, Stefan</td>
<td>Die Auswirkungen der Subprime-Krise auf den deutschen LBO-Markt für Small- und MidCaps</td>
</tr>
<tr>
<td>100.</td>
<td>Schalast, Christoph (Hrsg.) / Schanz, Kay-Michael</td>
<td>Schaeffler KG/Continental AG im Lichte der CSX Corp.-Entscheidung des US District Court for the Southern District of New York</td>
</tr>
<tr>
<td>95.</td>
<td>Almer, Thomas / Heidorn, Thomas / Schmaltz, Christian</td>
<td>The Dynamics of Short- and Long-Term CDS-spreads of Banks</td>
</tr>
<tr>
<td>92.</td>
<td>Burger, Andreas</td>
<td>Produktivität und Effizienz in Banken – Terminologie, Methoden und Status quo</td>
</tr>
<tr>
<td>91.</td>
<td>Lochel, Horst / Pecher, Florian</td>
<td>The Strategic Value of Investments in Chinese Banks by Foreign Financial Institutions</td>
</tr>
<tr>
<td>89.</td>
<td>Schalast, Christoph / Stralkowski, Ingo</td>
<td>10 Jahre deutsche Buyouts</td>
</tr>
<tr>
<td>86.</td>
<td>Hölscher, Luise / Rosenthal, Johannes</td>
<td>Leistungsmessung der Internen Revision</td>
</tr>
<tr>
<td>85.</td>
<td>Bannier, Christina / Hänsel, Dennis</td>
<td>Determinants of banks' engagement in loan securitization</td>
</tr>
<tr>
<td>84.</td>
<td>Bannier, Christina</td>
<td>“Smoothing” versus “Timeliness” - Wann sind stabile Ratings optimal und welche Anforderungen sind an optimale Berichtsregeln zu stellen?</td>
</tr>
<tr>
<td>83.</td>
<td>Bannier, Christina E.</td>
<td>Heterogeneous Multiple Bank Financing: Does it Reduce Inefficient Credit-Renegotiation Incidences?</td>
</tr>
<tr>
<td>82.</td>
<td>Cremers, Heinz / Lühr, Andreas</td>
<td>Deskription und Bewertung strukturierter Produkte unter besonderer Berücksichtigung verschiedener Marktszenarien</td>
</tr>
<tr>
<td>80.</td>
<td>Cremers, Heinz / Walzner, Jens</td>
<td>Risikosteuerung mit Kreditderivaten unter besonderer Berücksichtigung von Credit Default Swaps</td>
</tr>
<tr>
<td>No.</td>
<td>Authors</td>
<td>Title</td>
</tr>
<tr>
<td>-----</td>
<td>---------</td>
<td>-------</td>
</tr>
<tr>
<td>78.</td>
<td>Gerdesmeier, Dieter / Roffia, Barbara</td>
<td>Monetary Analysis: A VAR Perspective</td>
</tr>
<tr>
<td>77.</td>
<td>Heidorn, Thomas / Kaiser, Dieter G. / Muschiol, Andrea</td>
<td>Portfoliooptimierung mit Hedgefonds unter Berücksichtigung höherer Momente der Verteilung</td>
</tr>
<tr>
<td>75.</td>
<td>Abrar, Kamyar / Schalast, Christoph</td>
<td>Fusionskontrolle in dynamischen Netzsektoren am Beispiel des Breitbandkabelsektors</td>
</tr>
<tr>
<td>74.</td>
<td>Schalast, Christoph / Schanz, Kay-Michael</td>
<td>Wertpapierprospekte: Marktinführungspublizität nach EU-Prospektverordnung und Wertpapierprospektgesetz 2005</td>
</tr>
<tr>
<td>70.</td>
<td>Baumann, Stefan / Löchel, Horst</td>
<td>The Endogeneity Approach of the Theory of Optimum Currency Areas - What does it mean for ASEAN + 3?</td>
</tr>
<tr>
<td>69.</td>
<td>Heidorn, Thomas / Trautmann, Alexandra</td>
<td>Niederschlagsderivates</td>
</tr>
<tr>
<td>65.</td>
<td>Gerdesmeier, Dieter / Polleit, Thorsten</td>
<td>Measures of excess liquidity</td>
</tr>
<tr>
<td>64.</td>
<td>Becker, Gernot M. / Harding, Perham / Hölscher, Luise</td>
<td>Financing the Embedded Value of Life Insurance Portfolios</td>
</tr>
<tr>
<td>63.</td>
<td>Schalast, Christoph</td>
<td>Modernisierung der Wasserwirtschaft im Spannungsfeld von Umweltschutz und Wettbewerb – Braucht Deutschland eine Rechtsgrundlage für die Vergabe von Wasserversorgungskonzessionen? –</td>
</tr>
<tr>
<td>62.</td>
<td>Bayer, Marcus / Cremers, Heinz / Klüß, Norbert</td>
<td>Wertsicherungssstrategien für das Asset Management</td>
</tr>
<tr>
<td>61.</td>
<td>Löchel, Horst / Polleit, Thorsten</td>
<td>A case for money in the ECB monetary policy strategy</td>
</tr>
<tr>
<td>58.</td>
<td>Heimer, Thomas / Köhler, Thomas</td>
<td>Auswirkungen des Basel II Akkords auf österreichische KMU</td>
</tr>
<tr>
<td>57.</td>
<td>Heidorn, Thomas / Meyer, Bernd / Pietrowiak, Alexander</td>
<td>Performanceeffekte nach Directors Dealings in Deutschland, Italien und den Niederlanden</td>
</tr>
<tr>
<td>56.</td>
<td>Gerdesmeier, Dieter / Roffia, Barbara</td>
<td>The Relevance of real-time data in estimating reaction functions for the euro area</td>
</tr>
<tr>
<td>55.</td>
<td>Barthel, Erich / Gierig, Rauno / Kühn, Ilmhart-Wofram</td>
<td>Unterschiedliche Ansätze zur Messung des Humankapitals</td>
</tr>
<tr>
<td>53.</td>
<td>Polleit, Thorsten</td>
<td>The Slowdown in German Bank Lending – Revisited</td>
</tr>
<tr>
<td>52.</td>
<td>Heidorn, Thomas / Siragusano, Tindaro</td>
<td>Die Anwendbarkeit der Behavioral Finance im Devisenmarkt</td>
</tr>
<tr>
<td>51.</td>
<td>Schütze, Daniel / Schalast, Christoph (Hrsg.)</td>
<td>Wider die Verschleuderung von Unternehmen durch Pfandversteigerung</td>
</tr>
</tbody>
</table>
50. Gerhold, Mirko / Heidorn, Thomas
Investitionen und Emissionen von Convertible Bonds (Wandelanleihen)
2004

49. Chevalier, Pierre / Heidorn, Thomas / Krieger, Christian
Temperaturderivate zur strategischen Absicherung von Beschaffungs- und Absatzrisiken
2003

48. Becker, Gernot M. / Seeger, Norbert
Internationale Cash Flow-Rechnungen aus Eigner- und Gläubigersicht
2003

47. Boenkost, Wolfram / Schmidt, Wolfgang M.
Notes on convexity and quanto adjustments for interest rates and related options
2003

46. Hess, Dieter
Determinants of the relative price impact of unanticipated Information in U.S. macroeconomic releases
2003

45. Cremers, Heinz / Kluß, Norbert / König, Markus
Incentive Fees. Erfolgssabhängige Vergütungsmodelle deutscher Publikumsfonds
2003

44. Heidorn, Thomas / König, Lars
Investitionen in Collateralized Debt Obligations
2003

43. Kahlert, Holger / Seeger, Norbert
Bilanzierung von Unternehmenszusammenschlüssen nach US-GAAP
2003

42. Beiträge von Studierenden des Studiengangs BBA 012 unter Begleitung von Prof. Dr. Norbert Seeger
Rechnungslegung im Umbruch - HGB-Bilanzierung im Wettbewerb mit den internationalen Standards nach IAS und US-GAAP
2003

41. Overbeck, Ludger / Schmidt, Wolfgang
Modeling Default Dependence with Threshold Models
2003

40. Balthasar, Daniel / Cremers, Heinz / Schmidt, Michael
Portfoliooptimierung mit Hedge Fonds unter besonderer Berücksichtigung der Risikokomponente
2002

39. Heidorn, Thomas / Kantwill, Jens
Eine empirische Analyse der Spreadunterschiede von Festzinsanleihen zu Floatern im Euroraum und deren Zusammenhang zum Preis eines Credit Default Swaps
2002

38. Böttcher, Henner / Seeger, Norbert
Bilanzierung von Finanzderivaten nach HGB, EstG, IAS und US-GAAP
2003

37. Moormann, Jürgen
Terminologie und Glossar der Bankinformatik
2002

36. Heidorn, Thomas
Bewertung von Kreditprodukten und Credit Default Swaps
2001

35. Heidorn, Thomas / Weier, Sven
Einführung in die fundamentale Aktienanalyse
2001

34. Seeger, Norbert
International Accounting Standards (IAS)
2001

33. Moormann, Jürgen / Stehling, Frank
Strategic Positioning of E-Commerce Business Models in the Portfolio of Corporate Banking
2001

32. Sokolovsky, Zbynek / Strohhecker, Jürgen
Fit für den Euro, Simulationsbasierte Euro-Maßnahmenplanung für Dresdner-Bank-Geschäftsstellen
2001

31. Roßbach, Peter
Behavioral Finance - Eine Alternative zur vorherrschenden Kapitalmarkttheorie?
2001

30. Heidorn, Thomas / Jaster, Oliver / Willeitner, Ulrich
Event Risk Covenants
2001

29. Biswas, Rita / Löchel, Horst
Recent Trends in U.S. and German Banking: Convergence or Divergence?
2001

28. Eberle, Günter Georg / Löchel, Horst
Die Auswirkungen des Übergangs zum Kapitaldeckungsverfahren in der Rentenversicherung auf die Kapitalmärkte
2001

27. Heidorn, Thomas / Klein, Hans-Dieter / Siebrecht, Frank
Economic Value Added zur Prognose der Performance europäischer Aktien
2000

26. Cremers, Heinz
Konvergenz der binomialen Optionspreismodelle gegen das Modell von Black/Scholes/Merton
2000

25. Löchel, Horst
Die ökonomischen Dimensionen der „New Economy“
2000

24. Frank, Axel / Moormann, Jürgen
Grenzen des Outsourcing: Eine Exploration am Beispiel von Direktbanken
2000

23. Heidorn, Thomas / Schmidt, Peter / Seiler, Stefan
Neue Möglichkeiten durch die Namensaktie
2000
22. Böger, Andreas / Heidorn, Thomas / Graf Waldstein, Philipp
   Hybrides Kernkapital für Kreditinstitute
   2000

21. Heidorn, Thomas
   Entscheidungsorientierte Mindestmargenkalkulation
   2000

20. Wolf, Birgit
   Die Eigenmittellkonzeption des § 10 KWG
   2000

19. Cremers, Heinz / Robé, Sophie / Thiele, Dirk
   Beta als Risikomaß - Eine Untersuchung am europäischen Aktienmarkt
   2000

18. Cremers, Heinz
   Optionspreisbestimmung
   1999

17. Cremers, Heinz
   Value at Risk-Konzepte für Marktrisiken
   1999

   Gründung einer deutschen Strombörse für Elektrizitätsderivate
   1999

15. Deister, Daniel / Ehrlicher, Sven / Heidorn, Thomas
   CatBonds
   1999

14. Jochum, Eduard
   Hoshin Kanri / Management by Policy (MbP)
   1999

13. Heidorn, Thomas
   Kreditderivate
   1999

12. Heidorn, Thomas
   Kreditrisiko (CreditMetrics)
   1999

11. Moormann, Jürgen
   Terminologie und Glossar der Bankinformatik
   1999

10. Löchel, Horst
   The EMU and the Theory of Optimum Currency Areas
   1998

09. Löchel, Horst
   Die Geldpolitik im Währungsraum des Euro
   1998

08. Heidorn, Thomas / Hund, Jürgen
   Die Umstellung auf die Stückaktie für deutsche Aktiengesellschaften
   1998

07. Moormann, Jürgen
   Stand und Perspektiven der Informationsverarbeitung in Banken
   1998

06. Heidorn, Thomas / Schmidt, Wolfgang
   LIBOR in Arrears
   1998

05. Jahresbericht 1997
   1998

04. Ecker, Thomas / Moormann, Jürgen
   Die Bank als Betreiberin einer elektronischen Shopping-Mall
   1997

03. Jahresbericht 1996
   1997

02. Cremers, Heinz / Schwarz, Willi
   Interpolation of Discount Factors
   1996

01. Moormann, Jürgen
   Lean Reporting und Führungsinformationssysteme bei deutschen Finanzdienstleistern
   1995

FRANKFURT SCHOOL / HfB – WORKING PAPER SERIES
CENTRE FOR PRACTICAL QUANTITATIVE FINANCE

No.  Author/Title                                                                 Year
20.  Reiswich, Dimitri / Wystup, Uwe                                         2009
     FX Volatility Smile Construction
19.  Reiswich, Dimitri / Tompkins, Robert                                    2009
     Potential PCA Interpretation Problems for Volatility Smile Dynamics
18.  Keller-Ressel, Martin / Kilin, Fiodar                                   2008
     Forward-Start Options in the Barndorff-Nielsen-Shephard Model
17.  Griebsch, Susanne / Wystup, Uwe                                       2008
     On the Valuation of Fader and Discrete Barrier Options in Heston’s Stochastic Volatility Model

Frankfurt School of Finance & Management
<table>
<thead>
<tr>
<th>No.</th>
<th>Author/Title</th>
<th>Year</th>
</tr>
</thead>
<tbody>
<tr>
<td>01.</td>
<td>Nicole Kahmer / Jürgen Moormann</td>
<td>2003</td>
</tr>
<tr>
<td></td>
<td>Studie zur Ausrichtung von Banken an Kundenprozessen am Beispiel des Internet</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(Preis: € 120,--)</td>
<td></td>
</tr>
</tbody>
</table>

**HfB – SONDERARBEITSBERICHTE DER HfB - BUSINESS SCHOOL OF FINANCE & MANAGEMENT**