

# Credit Risk under Noisy Asset Information

Thorsten Schmidt

University of Leipzig

[www.math.uni-leipzig.de/~tschmidt](http://www.math.uni-leipzig.de/~tschmidt)  
[thorsten.schmidt@math.uni-leipzig.de](mailto:thorsten.schmidt@math.uni-leipzig.de)

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## A Model with Noisy Asset Information

- Consider a structural model
- Fact: firm value  $(V_t)_t$  is not observable !
- Investors use other sources of information:
  - Accounting information (Duffie & Lando 2001)
  - Information collected from newspapers, etc

⇒ Incomplete information leads to a filter problem

Goal of this work:

1. Fundamental relationship of equity and debt under incomplete information (joint with R. Frey)
2. Tractable framework which incorporates sudden information updates using shot noise models

## Information Structure

What kind of information is available to the investor ?

- 1 Information on the default state
- 2 News on the company at random times
  - If the company is in trouble  $\Rightarrow$  more frequent news  
ie news intensity depends on healthiness of company
  - News is discrete, for example good / bad or rating
- 3 Dividends depend on the firm value

## Fundamental value of equity under full information

- Denote firm value by  $(V_t)_{t \geq 0}$ , default by  $\tau := \inf\{t > 0 : V_t \leq K\}$
- Dividends  $D_n$ , paid at  $T_n$ ,  $n = 1, 2, \dots$ , are noisy signals of  $V$ :

$$D_n = \delta_n V_{T_n-}$$

$\delta_n$  i.i.d.  $\sim \nu_\delta(dx)$

- Full information:  $\mathcal{F}_t = \sigma(V_s : s \leq t, (T_n, D_n) : T_n \leq t)$
- Set  $R_t = \sum_{T_n \leq t} \delta_n \cdot \kappa$ ,  $\kappa \in [0, 1]$ . Under some EMM  $Q$

$$dV_t = V_{t-} \mu dt + V_{t-} \sigma dW_t - V_{t-} dR_t$$

- **Equity** is discounted value of the paid dividends,

$$S_t = 1_{\{\tau > t\}} \mathbf{E}^Q \left( \int_t^\tau e^{-r(s-t)} V_{s-} dR_s \middle| \mathcal{F}_t \right). \quad (1)$$

Make it easy:

- Assume  $T_n$  are jump times from a Poisson process with intensity  $\lambda$
- Denote  $\bar{\delta} := \mathbb{E}^Q(\delta_1)$

### Proposition

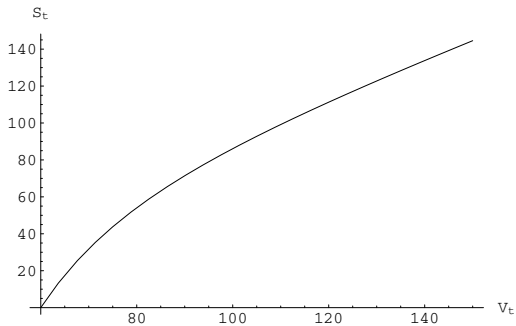
Under  $\mu < \lambda\bar{\delta} + r$ ,  $\kappa = 1$

$$S_t = \frac{\lambda\bar{\delta}}{r + \lambda\bar{\delta} - \mu} \left[ \frac{V_t}{K} - \left( \frac{V_t}{K} \right)^{\alpha^*} K \right], \quad (2)$$

where  $\alpha^* < 0$  such that

$$0 = \alpha\mu - r + \frac{1}{2}\sigma^2\alpha(\alpha - 1) + \lambda \left[ \int_0^1 (1-x)^{\alpha} \nu_{\delta}(dx) - 1 \right]$$

- In the simulation we use a Beta-distributed  $\delta$ .



Note that  $S$  is a concave function of  $V$

## Pricing Debt

- On  $\{\tau > T\}$  derive

$$\mathbb{P}(\tau > T | V_t) = \mathbb{P}\left(\inf_{s \in (t, T]} V_s > K | V_t\right)$$

- Default probability leads to prices for Bonds and CDS
- Summarizing, we can price equity and debt under full information

## The case of incomplete information

- News are modeled as  $(I_n)_{n=1,2,\dots}$ ,  $I_n$  occurring at  $\tilde{T}_n$ ,
- Specify  $\mathbb{P}(I_n = i | V_{T_n} = v) = \nu_I(i, v)$  for  $i \in \{i_1, \dots, i_M\}$
- Information **available to the investor** is

$$\mathcal{H}_t := \{1_{\{V_s > K\}} : s \leq t; (T_n, D_n) : T_n \leq t; (\tilde{T}_m, I_m) : \tilde{T}_m \leq t\}$$

- Value of equity is

$$S_t = 1_{\{\tau > t\}} \mathbb{E}^Q \left[ \underbrace{\mathbb{E}^Q \left( \int_t^\tau e^{-r(s-t)} V_{s-} dR_s \middle| \mathcal{F}_t \right)}_{(*)} \middle| \mathcal{H}_t \right].$$

- $(*) = f(V_t)$  already computed (full information case)
- ⇒ Need: conditional distribution of  $V_t$  given  $\mathcal{H}_t$
- Analogously for default probability, bond and CDS

## Find an approximate solution for the filter problem

- Consider a discrete time approximation,  $t_k = \Delta k$ , for fixed  $\Delta$
- Use discrete-time, discrete space Markov chain  $(V_k)$
- State space  $M^0 = \{m_1^0, \dots, m_N^0\}$ , transition probabilities  $p_{ij}^0$
- Discrete Information:

$$\mathcal{H}_k^\Delta := \sigma\left(V_i > K(t_i) : i \leq k, (T_n, D_n) : T_n \leq t_k, (\tilde{T}_m, I_m) : \tilde{T}_m \leq t_k\right).$$

## Conditional Distribution of $(V_k)$

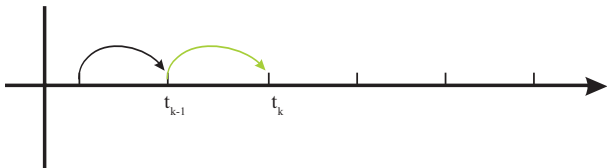
Information at time  $t_k$  after  $d$  dividends is

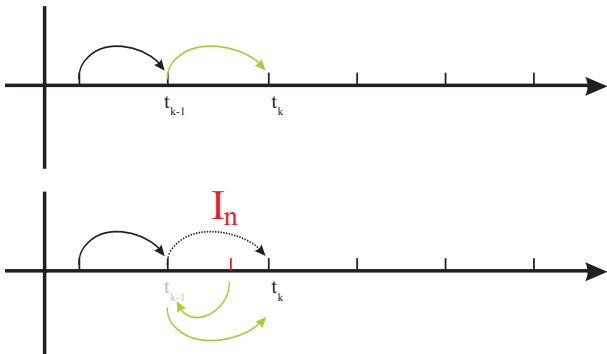
$$A_k := \{V_i > K(t_i) : i \leq k\}, \quad \mathcal{I}_k := \sigma((T_n, D_n) : T_n \leq t_k, (\tilde{T}_m, I_m) : \tilde{T}_m \leq t_k)$$

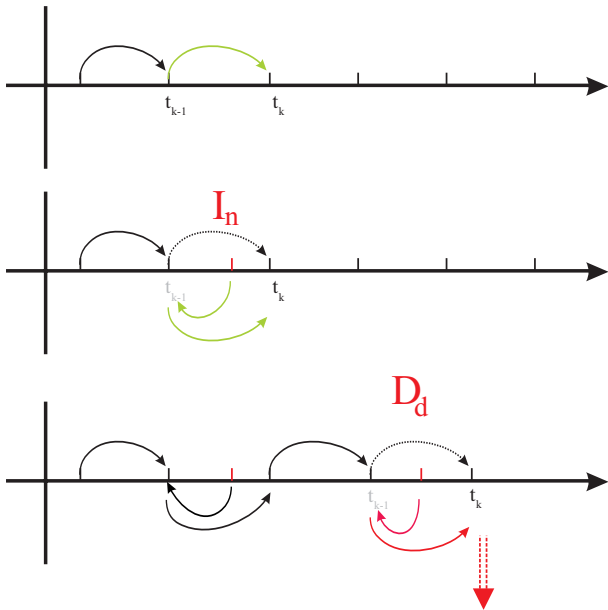
Then, on  $\{T_d \leq t_k, T_{d+1} > t_k\}$ ,

$$\begin{aligned} \mathbb{P}(V_k = m_j^0 | \mathcal{H}_k^\Delta) &= \mathbb{P}(V_k = m_j^0 | 1_{A_k}, \mathcal{D}_k) \\ &= \frac{\mathbb{P}(V_k = m_j^0, A_k | \mathcal{D}_k)}{\mathbb{P}(A_k | \mathcal{D}_k)} =: \frac{\pi_j^k}{\sum_{i=1}^N \pi_i^k}. \end{aligned}$$

Thus, we need only the **unnormalized probabilities**  $\pi_j^k$







## Proposition

If Information Intensity does not depend on the firm value, we have:

- ① If  $t_k$  is not a dividend date and no information arrives, then

$$\pi_j^k = \mathbf{1}_{\{m_j^k > K\}} \sum_{i=1}^N p_{ij}^k \pi_i^{k-1}. \quad (3)$$

- ② If  $t_k$  is not a dividend date but  $\{t_{k-1} < \tilde{T}_i \leq t_k\}$ , then

$$\pi_j^k = \mathbf{1}_{\{m_j^k > K\}} \tilde{\pi}_j^k \nu_l(l_i, m_j^k),$$

where  $\tilde{\pi}_j^k$  are computed according to (3)

- ③ On  $\{t_{k-1} < T_d \leq t_k\}$ , but no information we have

$$\pi_j^k = \mathbf{1}_{\{m_j^{k-1} > K + D_d\}} \tilde{\pi}_j^k \frac{1}{m_j^{k-1}} \nu_\delta\left(\frac{D_d}{m_j^{k-1}}\right),$$

where  $\tilde{\pi}_j^k$  are computed according to (3)

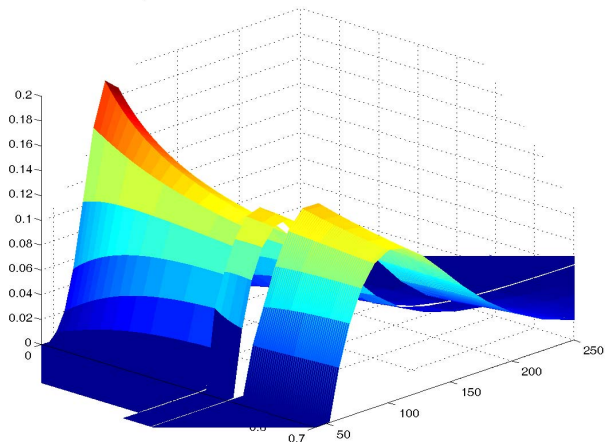
- ④ Combination of information and dividends ...

Incorporating additional information is also possible (see paper):

- News intensity depends on the firm value
- Correlated index is equivalent to time dependent boundary

## Analysis of the filter in practice

Conditional Density for Firm Value



We have the following result:

### Lemma

*If the stock price is a concave function of the firm value (under full information) then neglecting incomplete information leads to a underestimation of the firm value.*

### Proof.

We have from Proposition 1, that

$$S_t = f(V_t),$$

where  $f$  is a concave function. Using the stock price as statistic for the firm value and neglecting incomplete information leads to

$$\hat{V}_t = f^{-1}(S_t) = f^{-1}(\mathbf{E}(f(V_t)|\mathcal{H}_t)).$$

$f$  is monotonically increasing, so  $\hat{V}_t \leq V_t$  is equivalent to

$$f(\hat{V}_t) = \mathbf{E}(f(V_t)|\mathcal{H}_t) \leq f(\mathbf{E}(V_t|\mathcal{H}_t)),$$

which follows using Jensen's inequality. ■

## A Shot Noise Model for Pricing Credit

Motivated by the previous results, assume the intensity follows

$$\lambda_t = \tilde{\lambda}_t + J_t,$$

where  $\tilde{\lambda}$  is some diffusion component (eg general quadratic) and

$$J_t = \sum_{i=1}^{\tilde{N}_t} Y_i \cdot h(t - \tilde{\tau}_i),$$

$\tilde{N}$  is Poisson process with jumps  $\tilde{\tau}_i$ .

Typical examples:

$$h(t) = e^{-at}$$

$$h(t) = 1_{[0,a]}$$

or more general forms. Markovianity only for the 1st case!

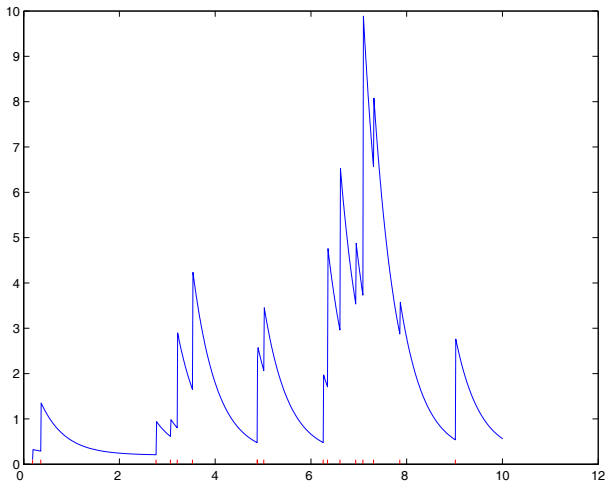


Figure: Possible realization of the process  $J$  with  $h(x) = e^{-0.5x}$  and  $\chi_2^2$ -distributed  $Y_i$ .

## Credit Portfolios with Clustering

- In Gaspar, S (2006) we apply this approach to credit PFs and CDOs.
- Consider  $k = 1, \dots, n$  credit risky with individual intensities

$$\lambda_t^k = \mu_t^k + \epsilon^k \mu_t^c.$$

Here  $\mu^k$  and  $\mu^c$  have the shot noise feature and the diffusions are general quadratic.

- ⇒ Obtain Markovian model with closed-form solutions for lots of credit derivatives, etc.
- Generalizes Duffie, Garleanu (2001) and allows for default clustering, ie high default correlation.



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Ann. Appl. Prob. 2004



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Financial Analysts Journal 2001