

# Illiquidity, Credit risk and Merton's model

*(joint work with J. Dong and L. Korobenko)*

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- A corporate bond is a contingent claim on the assets of a firm with pay-off  $\min(D, V_T)$ .  $D$  is the face value of the debt,  $T$  is the maturity.
- A geometric Brownian motion  $(V_t)_{t \geq 0}$  models the firm's assets:

$$dV_t = \mu V_t dt + \sigma V_t dW_t.$$

where  $\mu \in \mathbb{R}$ ,  $\sigma \in \mathbb{R}^+$ , and  $(W_t)_{t \geq 0}$  is a Brownian motion on some probability space  $(\Omega, \mathcal{F}, P)$ .

- The market is endowed with a money market account accumulating interest at a constant rate  $r$ .
- The firm's assets are liquidly traded in the market.

The liquidity assumption makes the pay-off replicable with the firm's assets and the money market account. Hence, the arbitrage-free price,  $B_t$ , of the corporate bond at time  $t$  is given by a variation of the Black-Scholes formula:

$$B_t = V_t N(d_1) - D \exp(-r(T-t)) N(d_2)$$

where  $N$  is the standard normal distribution function and

$$d_1 = \frac{\log(V_t/D) + (r + \frac{1}{2}\sigma^2)(T-t)}{\sigma\sqrt{T-t}}$$
$$d_2 = d_1 - \sigma\sqrt{T-t}$$

## Curious Observation:

$B_t$  does not depend on  $\mu$ !

# Default probabilities vs the price of the Merton style bond

**Remark:**  $P(V_T < D)$  is interpreted as the default probability of the firm.

## Question:

Consider two firms, Firm A and Firm B, which both issue Merton style bonds at time 0 with the same face value and maturity. Suppose that Firm A has a higher probability of default than Firm B. Shouldn't the premium of the Firm A bond be higher than the premium of the Firm B bond?

# Default probabilities vs the price of the Merton style bond

## Answer:

Not necessarily. Merton model is a counterexample:

- Suppose Firm A and Firm B have the same asset value at time 0, the same volatility, but different drifts with  $\mu_B > \mu_A$ . Then, the probability of default,  $P(V_T < D)$ , is higher for Firm A than Firm B.
- However, the prices of the Merton style bonds issued by the two firms are exactly the same.
- This happens because both bonds are replicable, and the prices of the replicating portfolios are exactly the same since the firm values are the same.

# Default probabilities vs the price of the Merton style bond

## Some insights

- Perfect replication of the pay-off  $\min(V_T, D)$  eliminates the effect of the increased default risk due to the drift term in the pricing of the zero coupon bond.
- In the absence of the liquidity assumption, perfect replication is no longer possible, hence the drift term may affect the price.

# A model where the firm's value is not a traded asset

## Assumptions:

- $V_t$  is not traded but observed.

$$dV_t = \mu_1 V_t dt + \sigma_1 V_t dW_t^1$$

- There is an asset in the market,  $S_t$  correlated with  $V_t$ . We assume

$$dS_t = \mu_2 S_t dt + \sigma_2 S_t dW_t^2,$$

- $(W_t^1, W_t^2)$  is a two dimensional Brownian motion with correlation  $\rho$ .
- All portfolios can be constructed using  $S_t$  and the money market account.
- For simplicity assume  $r = 0$ .
- The market information:  $\mathcal{F}_t = \sigma(V_s, S_s, s \leq t)$ .

**Remark:** The above model is incomplete. There is no unique way to set the price of a Merton style bond.

- Let  $b_t$  be the cost at time  $t$  of building an optimal replicating portfolio for  $\min(V_T, D)$ . That is,

$$(b_t, (\theta_s^t)_{s \in [t, T]}) = \operatorname{argmin}_{P, \theta} \mathbb{E} \left[ \left( P + \int_t^T \theta_s dS_s - \min(V_T, D) \right)^2 \middle| \mathcal{F}_t \right]$$

- Let  $c_t = \min_{P, \theta} \mathbb{E} \left[ \left( P + \int_t^T \theta_s dS_s - \min(V_T, D) \right)^2 \middle| \mathcal{F}_t \right]$ .
- Let  $\kappa > 0$  and  $\tilde{c}_t = \frac{c_t}{V_t^2}$ . The proposed model for the price of the zero-coupon bond:

$$B_t = b_t \times e^{-\kappa \tilde{c}_t}$$



- $b_t$  is a benchmark price because it is the price of the closest traded instrument in the market.
- $e^{-\kappa\tilde{c}_t}$  is a discount factor. One can think of it as the compensation due to the extra variability in  $\min(V_T, D)$  that can not be hedged by the optimal replicating portfolio.
- $\tilde{c}_t$  is the relative replication error. The importance of the replication error depends on the firm value. Also, this way the price is a monotone function of the firm value which is important to have no-arbitrage.

# Finding the optimal replicating portfolio

- The optimization problem  $\min_{P,\theta} \mathbb{E} \left[ \left( P + \int_0^T \theta_s dS_s - H \right)^2 \right]$  for a given  $H \in \mathcal{L}^2(\mathcal{F}_T)$  is called the mean variance hedging (MVH) problem. Earlier works used martingale decomposition techniques.
- When the underlying processes (here  $S_t$  and  $V_t$ ) are Markov processes, the MVH problem can be formulated as a stochastic control problem.

# MVH as a stochastic control problem

Bertsimas, Kogan and Lo (2001) considered  $H = F(S_T, V_T)$  and formulated the MVH as the following stochastic optimal control problem:

$$\text{minimize } \mathbb{E}[(P_T - F(S_T, V_T))^2] \quad \text{over all } \theta \in \Theta$$

with the dynamics

$$\begin{aligned} dV_t &= \mu_1 V_t dt + \sigma_1 V_t dW_t^1, \\ dS_t &= \mu_2 S_t dt + \sigma_2 S_t dW_t^2, \\ dP_t &= \theta_t dS_t = \theta_t \mu_2 S_t dt + \theta_t \sigma_2 S_t dW_t^2, \\ P_0 &= p, S_0 = s, V_0 = v. \end{aligned}$$

where  $\Theta$  is the set of all  $\mathbb{R}$ -valued predictable  $S$ -integrable processes such that  $\int \theta dS$  is well-defined.

## Remark

In the above formulation the initial cost of the portfolio  $p$  is taken as fixed. If we can solve the above problem for any  $p$ , then we can optimize over  $p$ .

# MVH as a stochastic control problem: Dynamic version

$$\text{minimize } \mathbb{E}_{t,p,s,v}[(P_T - F(S_T, V_T))^2] \quad \text{over all } \theta \in \Theta_t$$

$P_t = p, S_t = s, V_t = v$ , and for  $s > t$

$$\begin{aligned} dV_s &= \mu_1 V_s ds + \sigma_1 V_s dW_s^1, \\ dS_s &= \mu_2 S_s ds + \sigma_2 S_s dW_s^2, \\ dP_s &= \theta_s dS_s = \theta_s (\mu_2) S_s ds + \theta_s \sigma_2 S_s dW_s^2. \end{aligned}$$

Let  $V(t, p, s, v)$  be the optimal value function of this control problem. It is well known that  $V(t, p, s, v)$  is characterized as the solution of the **Hamilton Jacobi and Bellman (HJB)** equation.

$V(t, p, s, v)$  is quadratic in  $p$ , i.e. there are continuous functions  $a(t, s, v)$ ,  $b(t, s, v)$  and  $c(t, s, v)$  such that

$$V(t, p, s, v) = a(t, s, v) \cdot [p - b(t, s, v)]^2 + c(t, s, v), \quad 0 \leq t \leq T.$$

## Theorem (Bertsimas, Kogan and Lo (2001)) cont'd.

$$\begin{aligned}
 \frac{\partial a}{\partial t} &= \left(\frac{\mu_2}{\sigma_2}\right)^2 a + \mu_2 s \frac{\partial a}{\partial s} + \left[\frac{2\sigma_1 \rho v \mu_2}{\sigma_2} - \mu_1 v\right] \frac{\partial a}{\partial v} - \frac{1}{2} v^2 s^2 \frac{\partial^2 a}{\partial s^2} - \frac{1}{2} \sigma_1^2 v^2 \frac{\partial^2 a}{\partial v^2} \\
 &\quad - v^2 \sigma_1 s \rho \frac{\partial^2 a}{\partial s \partial v} + \frac{1}{a} v^2 s^2 \left(\frac{\partial a}{\partial s}\right)^2 + \frac{1}{a} \rho^2 \sigma_1^2 v^2 \left(\frac{\partial a}{\partial v}\right)^2 + 2v^2 \sigma_1 s \rho \frac{\partial a}{\partial s} \frac{\partial a}{\partial v}, \\
 \frac{\partial b}{\partial t} &= \left[\frac{\sigma_1}{\sigma_2} v \rho \mu_2 - \mu_1 v\right] \frac{\partial b}{\partial v} - \frac{1}{2} \sigma_1^2 v^2 \frac{\partial^2 b}{\partial v^2} - \frac{1}{2} \sigma_2^2 s^2 \frac{\partial^2 b}{\partial s^2} - \sigma_2 \sigma_1 v s \rho \frac{\partial^2 b}{\partial v \partial s} \\
 &\quad + \frac{\sigma_1^2 v^2}{a} (\rho^2 - 1) \frac{\partial a}{\partial v} \frac{\partial b}{\partial v}, \\
 \frac{\partial c}{\partial t} &= -\mu_1 v \frac{\partial c}{\partial v} - \mu_2 s \frac{\partial c}{\partial s} - \frac{1}{2} \sigma_1^2 v^2 \frac{\partial^2 c}{\partial v^2} - \sigma_2 \sigma_1 s v \rho \frac{\partial^2 c}{\partial s \partial v} - \frac{1}{2} \sigma_2^2 s^2 \frac{\partial^2 c}{\partial s^2} \\
 &\quad + a \sigma_1^2 v^2 (\rho^2 - 1) \left(\frac{\partial b}{\partial v}\right)^2,
 \end{aligned}$$

with boundary conditions

$$a(T, s, v) = 1, b(T, s, v) = F(s, v), c(T, s, v) = 0.$$

$a(t, s, v) > 0$ , hence

- Optimal initial wealth  $p^*(t, s, v)$  that minimizes the quadratic function is  $b(t, s, v)$ .
- The optimal-replication strategy is the  $\theta$  corresponding to this initial wealth  $p^*(t, s, v)$ .
- The optimal replication error is  $\sqrt{c(t, s, v)}$ .

Let  $F(S_T, V_T) = \min(V_T, D)$ . Because the pay-off is only a function of  $V_T$  but not  $S_T$ , it turns out that the functions  $a$ ,  $b$  and  $c$  are also only functions of  $t$  and  $v$  but not  $s$ . Thanks to this simplification the PDEs for  $a$ ,  $b$  and  $c$  become a linear system and can be solved explicitly.

## Theorem

(Dong, Korobenko, S.(2016))

$$\textcircled{1} \quad a(t, v) = e^{-\left(\frac{\mu_2}{\sigma_2}\right)^2 T-t};$$

$$\textcircled{2} \quad b(t, v) = ve^{\left(\mu_1 - \frac{\mu_2 \rho \sigma_1}{\sigma_2}\right)(T-t)} N(d1) + D(1 - N(d2)), \text{ where}$$

$$d1(t, v) = \frac{\ln \frac{D}{v} - \sigma_1^2 \left(1 - \left(\frac{1}{2} + \frac{1}{\sigma_1^2} \left[\frac{\sigma_1 \rho \mu_2}{\sigma_2} - \mu_1\right]\right)\right)(T-t)}{\sigma_1 \sqrt{T-t}},$$

$$d2(t, v) = d1 + \sigma_1 \sqrt{T-t};$$

$$\textcircled{3} \quad c(t, v) = \sigma^2(1 - \rho^2) \mathbb{E} \left( \int_t^T a_u V_u^2 \left( \frac{\partial b}{\partial v}(u, V_u) \right)^2 du \mid V_t = v \right).$$



We propose the following formula to calculate the price of  $\min(V_T, D)$ :

$$B(t, V_t) = b(t, V_t) \cdot e^{-\kappa \cdot \frac{c(t, V_t)}{V_t^2}},$$

- The price of the contingent claim  $B(t, V_t)$  converges to the payoff  $\min(V_T, D)$  as  $t$  approaches  $T$ .
- $\kappa$  is a preference parameter, a higher value indicates a higher level of risk aversion.
- The mean squared replication error  $c(t, V_t)$  has been normalized by  $V_t^2$ . This not only gives a better measure of the approximation of the replication (for example  $c(t, V_t) = 800$  would be more alarming if the the firm value were 10, as compared to 100), but also makes  $\kappa$  a unitless constant. In general, we need normalization for technical reasons, in particular, to show that the pricing formula is arbitrage free.

# A closer look at the function $c$

We derive a more explicit formula for the function  $c$  and analyze its qualitative properties:

## Theorem

(Dong, Korobenko, S. 2017)

$\frac{c(t,v)}{v^2}$  is monotone decreasing in  $v$ . In particular,

$$c(t, v) = \sigma_1^2 (1 - \rho^2) v^2 e^{[(2\mu_1 + \sigma_1^2)](T-t)} \quad (1)$$

$$\int_t^T e^{(-(\frac{\mu_2}{\sigma_2})^2 - \sigma_1^2 - 2\frac{(\mu_2)}{\sigma_2}\rho\sigma_1)(T-u)} E(N(d_1)^2) du \quad (2)$$

where  $d_1$  is normally distributed with mean

$$\mu(u, v) = \frac{\ln D - \ln v + (-\mu_1 - \frac{3}{2}\sigma_1^2)(u-t) - \sigma_1^2(1-\alpha)(T-u)}{\sigma_1\sqrt{T-u}}$$

with standard deviation  $\sqrt{\frac{u-t}{T-u}}$ .

The Fundamental Theorem of Asset Pricing (Dalbean and Schachermayer):

There is no arbitrage in the sense of *no free lunch with vanishing risk* if and only if there exists an equivalent probability measure  $Q$  rendering the price processes sigma martingales.

## Theorem

(Dong, Korobenko, S. 2017) There exist processes  $\lambda_1(t)$  and  $\lambda_2(t)$  such that

$$B_t = B_0 + \int_0^t N_t d\tilde{W}_t^1$$
$$S_t = S_0 + \int_0^t L_t d\tilde{W}_t^2$$

where  $\tilde{W}_t^i = \int_0^t \lambda_i(s) ds + W_t^i$ . Moreover, if

$$\mathbb{E}\left(\exp\left(\frac{1}{1-\rho^2} \int_0^T \left[\lambda_1^2(s) + \frac{1}{2}\lambda_2^2(s) - 2\rho\lambda_1(s)\lambda_2(s)\right] ds\right)\right) < \infty, \quad (3)$$

then there exists of a probability measure  $\mathbb{Q}$  equivalent to  $\mathbb{P}$  such that  $(\tilde{W}_t^1, \tilde{W}_t^2)_{0 \leq t \leq T}$  is a two dimensional Brownian motion with correlation  $\rho$ .

## Theorem

(Dong, Korobenko, S. 2017) Assume  $\frac{1}{2} + \frac{\mu_1 - \frac{\mu_2}{\sigma_2} \rho \sigma_1}{\sigma_1^2} > 0$ . We have  $\sup_{t \in [0, T]} |\lambda_i(t)| < K$  for some deterministic constant  $K$ .

Corollary: (Dong, Korobenko, S. 2017) Assume that  $\frac{1}{2} + \frac{\alpha}{\sigma_1^2} > 0$ . Let  $\tilde{c}(t, v) = c(t, v)/v^2$ . For any  $\kappa > 0$ ,  $B_t = b(t, V_t)e^{-\kappa \tilde{c}(t, V_t)}$  gives an arbitrage free price for the Merton style bond  $\min(V_T, D)$  in the sense of NFLVR.

# How do the parameters affect the price?

Overall there are three sets of parameters

- ① Parameters of the pay-off:  $T, D$ .
- ② Parameters of the underlying processes:  $\mu_1, \theta := \frac{\mu_2}{\sigma_2}, \rho, \sigma_1$ .
- ③ Risk aversion parameter  $\kappa$ .

We are most interested in  $\mu_1, \theta := \frac{\mu_2}{\sigma_2}, \rho, \sigma_1$  and  $\kappa$  and their effects on the yield of the bond.

# How do the parameters affect the price?

Some highlights:

- A key quantity is  $\alpha = (\mu_1) - \theta\rho\sigma_1$ .  $b(t, v)$  depends on  $\mu_1$ ,  $\theta$  and  $\rho$  only through the term  $\alpha$ .
- $b(t, v)$  increases with  $\alpha$  (when everything else is fixed.).  
Interpretation: The larger the  $\mu_1$ , the higher the price of the optimal replicating portfolio. The higher the risk premium on the underlying asset  $S_t$  the higher the premium of the optimal replicating portfolio.
- If  $\rho\theta > 0$  then  $b(t, v)$  decreases with  $\sigma_1$ . Numerical calculations show that  $\tilde{c}(t, v)$  should be increasing with  $\sigma_1$  as well. This would mean the yield of the bond increases with  $\sigma_1$ .
- Surprisingly  $\tilde{c}(t, v)$  is not monotone in  $\rho$  (except in the case  $\theta = 0$ ).
- The yield at date  $t$  of a Merton style bond with maturity  $T$  is defined as the function

$$y(t, T) = \frac{\log(D) - \log(B_t)}{T - t}$$

Yields increase with  $\kappa$ , most noticeably when  $\alpha \geq 0$ .

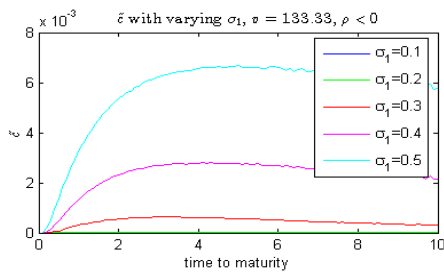
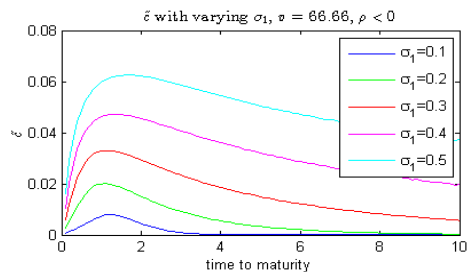
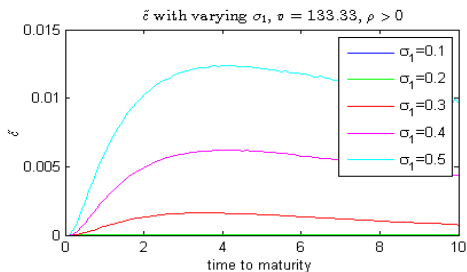
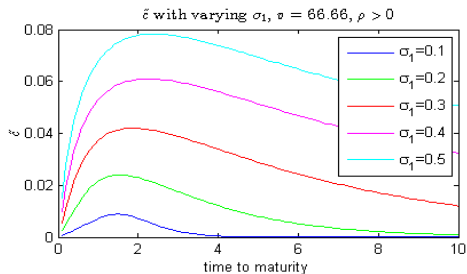


Figure:  $\tilde{\epsilon}$  with varying  $\sigma_1$ ,  $\mu_1 = 0$ ,  $\theta = 0.66$



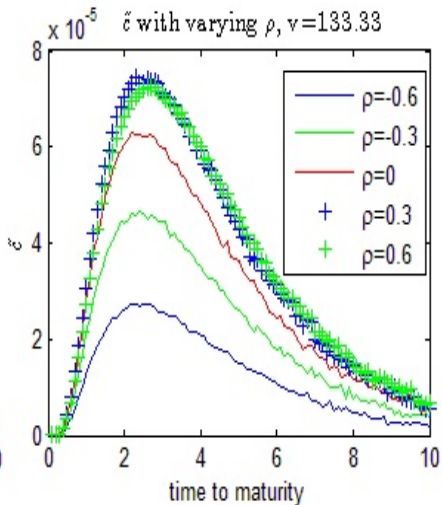
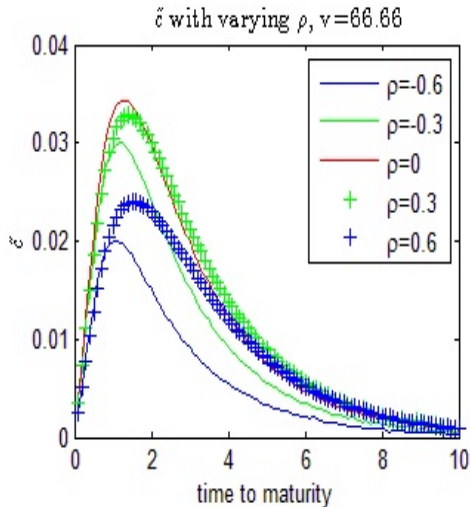


Figure:  $\tilde{c}$  with varying  $\rho$ ,  $\mu_1 = 0.2$ ,  $\sigma_1 = 0.2$ ,  $\theta = 1.5$

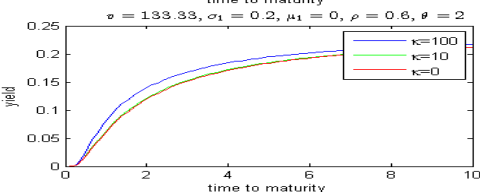
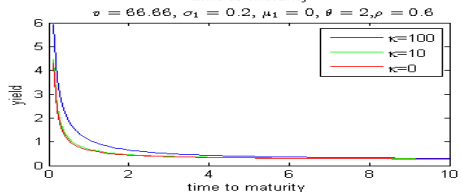
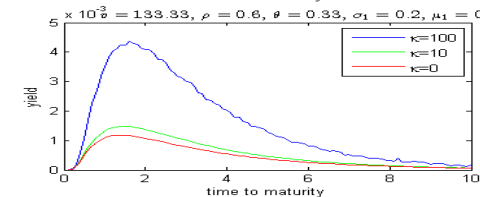
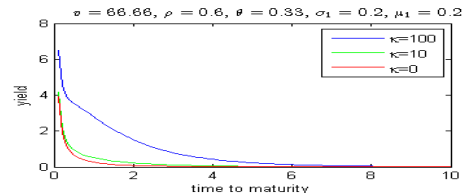
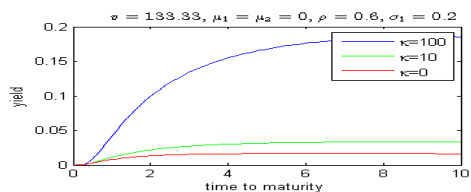
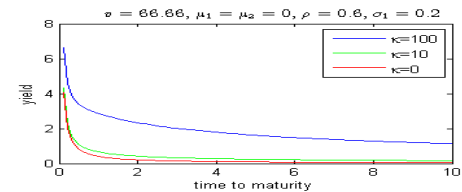


Figure: yield with varying  $\kappa$  for various configurations