Internet Appendix for
“Quantifying Liquidity and Default Risks of Corporate Bonds over the Business Cycle”

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1 Technical Details of Model Solution

First, we present the HJB equations for value functions for the 2-state case in the paper. Then, we present the HJB equations for the $n$-state case and present the solutions to the value functions, of which the $n = 2$ case is a special case. To this end, we split Proposition 1 in the main text into two propositions, one for the debt value function (Proposition 1), and one for the equity value function (Proposition 2).

Throughout, diag(·) is the diagonalization operator mapping any row or column vector into a diagonal matrix (in which all off-diagonal elements are identically zero).

1.1 State Transitions

As notational conventions, we use capitalized bold-faced letters (e.g., $X$) to denote matrices, lower case bold face letters (e.g. $x$) to denote vectors, and non-bold face letters denote scalars (e.g. $x$). The only exceptions are the value functions for debt and equity, $D,E$ respectively, which will be vectors, and the (diagonal) matrix of drifts, $\mu$. Dimensions for most objects are given underneath the expression. While we focus on 2-aggregate-state case where $s \in \{G, B\}$, the Appendix presents general results for an arbitrary number of (Markov) aggregate states.
Given the aggregate state \( s \), recall that we have assumed that the intensity of switching from state-\( H \) to state-\( L \) is \( \xi_s \), and the \( L \)-state is absorbing, i.e., those \( L \)-type investors leave the market forever. However, an \( L \)-type bond holder meets a dealer with intensity \( \lambda_s \) and sells the bond for \( B^s = \beta D^s_H + (1 - \beta) D^s_L \) that he himself values at \( D^s_L \). Then the \( L \)-type’s intensity-modulated surplus when meeting the dealer can be rewritten as

\[
\lambda_s (B^s - D^s_L) = \lambda_s \beta (D^s_H - D^s_L).
\]

As a result, for the purpose of pricing, the effective transitioning intensity from \( L \)-type to \( H \)-type is \( q_{L \to H} = \lambda_s \beta \) where \( \lambda_s \) is the state-dependent intermediation intensity and \( \beta \) is the investor’s bargaining power.

### 1.2 Debt

Recall that the midpoint price is

\[
P^s = \frac{A^s + B^s}{2} = \frac{(1 + \beta) D^s_H + (1 - \beta) D^s_L}{2} = w_H D^s_H + w_L D^s_L
\]

with \( w_H + w_L = 1 \) and that holding costs are

\[
hc^s = \chi_s [N - P^s] = \chi_s [N - w_H D^s_H - w_L D^s_L]
\]

Stacking the holding cost function, we have

\[
hc (y) = \chi [N - W \cdot D (y)]
\]
where $\chi = \text{diag} ([\chi_G, \chi_G, \chi_B, \chi_B])$, $N = [0, N, 0, N]^\top$ and

$$W = \begin{bmatrix}
0 & 0 & 0 & 0 \\
w_H & w_L & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & w_H & w_L \\
\end{bmatrix}$$

is the appropriate weighting matrix. Thus, $hc(y)$ has identically zero odd-numbered rows.

Then, debt follows, on $I_2$,

$$\hat{R} \cdot D^{(2)}(y) = \mu (D^{(2)})'(y) + \frac{1}{2} \Sigma (D^{(2)})''(y) + \hat{Q} \cdot D^{(2)}(y)$$

$$+ c14 + m [p14 - D^{(2)}(y)] - hc(y),$$

(1)

where $\hat{Q}$ is the effective transition matrix (accounting for the OTC market outcome via the recovery intensity $\lambda_s \beta$) for debt out on interval $I_2$, given by

$$\hat{Q} = \begin{bmatrix}
-x_G - \zeta & x_G & \zeta & 0 \\
\beta \lambda_G & -\beta \lambda_G - \zeta & 0 & \zeta \\
\zeta_B & 0 & -x_B - \zeta_B & -\zeta_B \\
0 & \zeta_B & \beta \lambda_B & -\beta \lambda_B - \zeta_B \\
\end{bmatrix},$$

and where

$$\hat{R} \equiv \text{diag} ([r_G, r_G, r_B, r_B]), \mu \equiv \text{diag} ([\mu_G, \mu_G, \mu_B, \mu_B]), \Sigma \equiv \text{diag} ([\sigma_G^2, \sigma_G^2, \sigma_B^2, \sigma_B^2]).$$

Substituting it in, and noting that $X = X^{(n)}$ where $n$ is the number of state, we have

$$\hat{R}^{(2)} \cdot D^{(2)}(y) = \mu^{(2)} (D^{(2)})'(y) + \frac{1}{2} \Sigma^{(2)} (D^{(2)})''(y) + \hat{Q}^{(2)} \cdot D^{(2)}(y)$$

$$+ c14 + m [p14 - D^{(2)}(y)] - \left[\chi^{(2)} N^{(2)} - \chi^{(2)} W \cdot D^{(2)}(y)\right],$$

(2)
For simulation purposes, we need to incorporate $W$ in such a way into $R$ and $Q$ that there is a unique discount rate in each state $(i,s)$, i.e. $R$ stays a diagonal discount matrix, and that $Q$ stays a true transition matrix.\(^1\) Thus, decompose $W = W_1 + W_2$ so that

\[
W_1 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1
\end{bmatrix}
\]

\[
W_2 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
w_H & -w_H & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & w_H & -w_H
\end{bmatrix}
\]

First, we augment $\hat{Q}$ to give the new transition matrix

\[
Q \equiv \hat{Q} + \chi W_2 = 
\begin{bmatrix}
-\xi_G - \zeta_G & \xi_G & \zeta_G & 0 \\
\beta \lambda_G + \chi_G w_H & -\beta \lambda_G - \zeta_G - \chi_G w_H & 0 & \zeta_G \\
\zeta_B & 0 & -\xi_B - \zeta_B & -\xi_B \\
0 & \zeta_B & \beta \lambda_B + \chi_B w_H & -\beta \lambda_B - \zeta_B - \chi_B w_H
\end{bmatrix}
\]

\(^1\)Let $Q$ be a transition matrix. Then define, for a finited time-interval $\Delta t$, the discrete-time transition matrix is given by

\[
Q_{\text{discrete}} = \exp (Q \Delta t)
\]

Note that $\Delta t = 0$ implies that $Q_{\text{discrete}} = I$ and thus there is total persistence of the states. Also, note that since $Q$ is a transition matrix that we have $Q_{\text{discrete}} 1 = 1$. This comes from the fact that

\[
\exp (Q \Delta t) 1 = I \cdot 1 + \sum_{k=1}^{\infty} \frac{Q^k \cdot 1 (\Delta t)^k}{k!} = 1
\]

as $Q^k 1 = 0$ for all $k \geq 1$.\]
Note that $Q$ is still a true transition matrix. Next, let us define the diagonal discount matrix $R^{(2)}$ as

$$R \equiv \hat{R} - \chi W_1 = \begin{bmatrix} r_G & 0 & 0 & 0 \\ 0 & r_G - \chi G & 0 & 0 \\ 0 & 0 & r_B & 0 \\ 0 & 0 & 0 & r_B - \chi B \end{bmatrix}$$

and we acknowledge the possibility of an individual entry being negative. Then, we have

$$\left[ R^{(2)} + m I_4 - Q^{(2)} \right] D^{(2)} (y) = \mu^{(2)}(D^{(2)})'(y) + \frac{1}{2} \Sigma^{(2)}(D^{(2)})''(y) + c_1 \cdot p_1^B + \chi^{(2)} N^{(2)} + m \cdot p_1^B,$$

the final form of the second-order ODE we are interested in.

Define $R^{(1)} = \begin{bmatrix} r_G + \zeta_G & 0 \\ 0 & r_G - \chi G + \zeta_G \end{bmatrix}$, $Q^{(1)} \equiv \begin{bmatrix} -\xi_G \\ \beta \lambda_G + \chi G w_H \end{bmatrix}$, and $\tilde{Q}^{(1)} \equiv \begin{bmatrix} \zeta_G \\ 0 \\ \zeta_G \end{bmatrix} = \zeta G I_2$. On interval $I_1 = [y_{\text{def}}(G) ; y_{\text{def}}(B)]$, the bond is “dead” in state $B$, and the alive bonds $D^{(1)} = \begin{bmatrix} D_H^{(G,1)}; D_L^{(G,1)} \end{bmatrix}^\top$ solve

$$\left[ R^{(1)} + m I_2 - Q^{(1)} \right] D^{(1)} = \mu^{(1)}(D^{(1)})' + \frac{1}{2} \Sigma^{(1)}(D^{(1)})'' + \tilde{Q}^{(1)} \begin{bmatrix} \alpha_H^B \\ \alpha_L^B \end{bmatrix} v^B_U (y) + (c_1 - \chi^{(1)} N^{(1)}) + m \cdot p_1^B$$

for

$$y \in I_1 = [y_{\text{def}}(G) ; y_{\text{def}}(B)],$$

where the last term is the recovery value in case of a jump to default brought about by a state jump.
The boundary conditions at $y = \infty$ and $y = y_{\text{def}}(G)$ are standard:

$$
\lim_{y \to \infty} |D^{(2)}(y)| < \infty, \quad \text{and} \quad D^{(1)}(y_{\text{def}}^G) = \begin{bmatrix} \alpha^G_H \\ \alpha^G_L \end{bmatrix} v^G_{\text{U}}(y_{\text{def}}^G)
$$

For the boundary $y_{\text{def}}^B$, we must have value matching conditions for all functions across $y_{\text{def}}(B)$:

$$
D^{(2)}(y_{\text{def}}^B) = \begin{bmatrix} D^{(1)}(y_{\text{def}}^B) \\ \alpha^B_H \\ \alpha^B_L \end{bmatrix} v^B_{\text{U}}(y_{\text{def}}^B)
$$

and smooth pasting conditions for functions that are alive across $y_{\text{def}}^B (\mathbf{x}_{[1:2]}$ selects the first 2 rows of vector $\mathbf{x}$):

$$
(D^{(2)})'(y_{\text{def}}^B)_{[1:2]} = (D^{(1)})'(y_{\text{def}}^B).
$$

1.3 Equity

For equity, we have

$$
\mathbf{RR} = \text{diag} ([r_G, r_B]), \quad \mathbf{\mu\mu} = \text{diag} ([\mu_G, \mu_B]), \quad \Sigma\Sigma = \text{diag} ([\sigma^2_G, \sigma^2_B]),
$$

and

$$
\mathbf{QQ}^{(2)} = \begin{bmatrix} -\zeta_G & \zeta_G \\ \zeta_B & -\zeta_B \end{bmatrix}
$$

so that

$$
\underbrace{\mathbf{RR}^{(2)} \mathbf{E}^{(2)}(y)}_{\text{Discounting,} 2\times 1} = \underbrace{\mathbf{\mu\mu}^{(2)} \mathbf{E}^{(2)}(y)}_{2\times 2} + \frac{1}{2} \underbrace{\Sigma\Sigma^{(2)} \mathbf{E}^{(2)}(y)}_{2\times 2}'' + \underbrace{\mathbf{QQ}^{(2)} \mathbf{E}^{(2)}(y)}_{2\times 1} + \underbrace{1_2 \exp(y)}_{\text{Cashflow,} 2\times 1} - \underbrace{(1 - \pi) c1_2}_{\text{Coupon,} 2\times 1} + \underbrace{m \cdot [\mathbf{S}^{(2)} \cdot D^{(2)}(y) - p 1_2]}_{\text{Rollover,} 2\times 1}
$$


where \( \mathbf{QQ}^{(2)} \) is the effective transition matrix for equity out on interval \( I_2 \) and \( \mathbf{RR}^{(2)} \) is the effective discount matrix.

For \( I_1 \), define \( \mathbf{RR}^{(1)} = r_G + \zeta_G; \mathbf{\mu}^{(1)} = \mu_G; \mathbf{SS}^{(1)} = \sigma_G^2; \mathbf{QQ}^{(1)} = 0, \mathbf{\overline{QQ}}^{(1)} = \zeta_G. \) Then

\[
\left[ \mathbf{RR}^{(1)} - \mathbf{QQ}^{(1)} \right] \mathbf{E}^{(1)} (y) = \mathbf{\mu}^{(1)} (\mathbf{E}^{(1)})' (y) + \frac{1}{2} \mathbf{SS}^{(1)} (\mathbf{E}^{(1)})'' (y) + \mathbf{\overline{QQ}}^{(1)} \mathbf{0}_1 \\
+ 1_1 \exp (y) - (1 - \pi) c 1_1 + m [\mathbf{S}^{(1)} \cdot \mathbf{D}^{(1)} (y) - \mathbf{P} 1_1] \quad (10)
\]

The particular solution is

\[
\begin{align*}
\mathbf{E}^{(2)} (y)_{2 \times 1} &= \mathbf{GG}^{(2)} \cdot \exp (\mathbf{\Gamma}^{(2)} y)_{4 \times 4} \cdot \mathbf{bb}^{(2)}_{4 \times 1} + \mathbf{KK}^{(2)} \cdot \exp (\mathbf{\Gamma}^{(2)} y)_{8 \times 8} \cdot \mathbf{b}^{(2)}_{4 \times 2} + \\
&\mathbf{kk}^{(2)}_{2 \times 1} + \mathbf{kk}^{(2)}_{2 \times 1} \exp (y) \text{ for } y \in I_2 \\
\mathbf{E}^{(1)} (y)_{1 \times 1} &= \mathbf{GG}^{(1)} \cdot \exp (\mathbf{\Gamma}^{(1)} y)_{2 \times 2} \cdot \mathbf{bb}^{(1)}_{2 \times 1} + \mathbf{KK}^{(1)} \cdot \exp (\mathbf{\Gamma}^{(1)} y)_{4 \times 4} \cdot \mathbf{b}^{(1)}_{4 \times 1} + \\
&\mathbf{kk}^{(1)}_{1 \times 1} + \mathbf{kk}^{(1)}_{1 \times 1} \exp (y) \text{ for } y \in I_1
\end{align*}
\]

where \( \mathbf{GG}^{(i)}, \mathbf{\Gamma}^{(i)}, \mathbf{bb}^{(i)}, \mathbf{KK}^{(i)}, \mathbf{kk}^{(i)}_0 \) and \( \mathbf{kk}^{(i)}_1 \) for \( i \in \{1, 2\} \) are given below. In particular, the constant vector \( \mathbf{bb}^{(i)} \) is determined by boundary conditions similar to those for debt.

### 1.4 Defaulted bonds

Consider now a defaulted bond. Its value \( D^d_i (y) \) will be determined by the holding cost

\[
hc^{ddef} (y) = \chi_s^{ddef} [N^{ddef} - P^{ddef} (y)].
\]

Let \( \mathbf{V} (y) = \text{diag} ([v^{d}_G (y), v^{d}_U (y), v^{d}_L (y), v^{d}_R (y)]) \). Then, we can write out

\[
\dot{\mathbf{R}} \cdot \mathbf{D}^{ddef} (y) = \theta [\mathbf{V} (y) \alpha - \mathbf{D}^{ddef}] + \dot{\mathbf{Q}}^{ddef} \cdot \mathbf{D}^{ddef} (y) - \chi^{ddef} [N^{ddef} - \mathbf{W}^{ddef} \mathbf{D}^{ddef} (y)] \quad (11)
\]

which is solved by

\[
\mathbf{D}^{ddef} (y) = \left[ \dot{\mathbf{R}} + \theta \mathbf{I}_n - \dot{\mathbf{Q}}^{ddef} - \chi^{ddef} \mathbf{W}^{ddef} \right]^{-1} [\theta \mathbf{V} (y) \alpha - \chi^{ddef} N^{ddef}]
\]

7
Suppose now that $N^{def} = V(y)n^{def}$. Then, we have

$$D^{def}(y) = \left[ \hat{R} + \theta I_n - \dot{Q}^{def} - \chi^{def}\hat{W}^{def} \right]^{-1} V(y) \left[ \theta \alpha - \chi^{def}n^{def} \right]$$

(12)

where we used that fact that $\chi^{def}V(y) = V(y)\chi^{def}$ as both are diagonal matrices.

1.5 Generalization to $n$ aggregate states and solution

We follow the Markov-modulated dynamics approach of Jobert and Rogers (2006).

We note that there are multiple possible bankruptcy boundaries, $y^{def}(s)$, for each aggregate state $s$ one boundary. Order states $s$ such that $s > s'$ implies that $y^{def}(s) > y^{def}(s')$ and denote the intervals $I_s = [y^{def}(s), y^{def}(s + 1)]$ where $y^{def}(n + 1) = \infty$, so that $I_s \cap I_{s+1} = y^{def}(s + 1)$. Finally, let $y_b = [y^{def}(1), ..., y^{def}(n)]^T$ be the vector of bankruptcy boundaries.

It is important to have a clean notational arrangement to handle the proliferation of states. Let $D^{(s)}_l$ denote the value of debt for a creditor in individual liquidity state $l$ and with aggregate state $s$. We will use the following notation: $D^{(s,i)}_l \equiv D^{(s)}_l, y \in I_i$, that is $D^{(s,i)}_l$ is the restriction of $D^{(s)}_l$ to the interval $I_i$. It is now clear that $D^{(s,i)}_l = \text{recovery}$ for any $i < s$, as it would imply that the company immediately defaults in interval $I_i$ for state $s$. Let us, for future reference, call debt in states $i < s$ dead and in states $i \geq s$ alive.

Finally, let us stack the alive functions along states $s$ but still restricted to interval $i$ so that $D^{(i)} = \left[ D^{(1,i)}_H, D^{(1,i)}_L, ..., D^{(i,i)}_H, D^{(i,i)}_L \right]^T$ where $D^{(s,i)}_l$ has $s$ denoting the state, $i$ denotes the interval and $l$ denotes the individual liquidity state. The separation of $s$ and $i$ will clarify the pasting arguments that apply when $y$ crosses from one interval to the next. Let

$$I_i = \begin{bmatrix} 1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 1 \end{bmatrix}$$

(13)
be the $i$-dimensional identity matrix, and let
\[
\mathbf{1}_i = [1, \ldots, 1]^T \quad (14)
\]
be a column vector of ones.

**Fundamental parameters.** Let $\hat{\mathbf{R}} = \text{diag} ([r_1, r_1, \ldots, r_n, r_n])$ and let $\mathbf{R}\mathbf{R} = \text{diag} ([r_1, \ldots, r_n])$.

Let $\hat{\mathbf{Q}}$ summarize the OTC-augmented transition intensities. Define the building block matrices
\[
\mathbf{B}_s^{W_1} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \quad \text{and} \quad \mathbf{B}_s^{W_2} = \begin{bmatrix} 0 & 0 \\ w_H & -w_H \end{bmatrix}
\]
so that $\mathbf{B}_s^W = \begin{bmatrix} 0 & 0 \\ w_H & w_L \end{bmatrix} = \mathbf{B}_s^{W_1} + \mathbf{B}_s^{W_2}$.

Then, we have
\[
\mathbf{W}_1 = \text{diag} \left( [\mathbf{B}_1^{W_1}, \ldots, \mathbf{B}_n^{W_1}] \right), \quad \mathbf{W}_2 = \text{diag} \left( [\mathbf{B}_1^{W_2}, \ldots, \mathbf{B}_n^{W_2}] \right), \quad \chi = \text{diag} \left( [\chi_1, \chi_1, \ldots, \chi_n, \chi_n] \right)
\]
and we define the diagonal discount matrix $\mathbf{R}$ and the transition matrix $\mathbf{Q}$ as
\[
\mathbf{R} = \hat{\mathbf{R}} - \chi \mathbf{W}_1, \quad \mathbf{Q} = \hat{\mathbf{Q}} + \chi \mathbf{W}_2
\]

Let $\boldsymbol{\mu} = \text{diag} ([\mu_1, \mu_1, \ldots, \mu_n, \mu_n])$, $\Sigma = \text{diag} ([\sigma_1^2, \sigma_1^2, \ldots, \sigma_n^2, \sigma_n^2])$.

Define $\mathbf{X}^{(n)} = \mathbf{X}$, that is the restriction to the set $I_n = [\gamma_{\text{def}} (n), \infty)$ does not change any of the matrices. Let $\mathbf{Q}^{(i)}$ be the transition matrix of jumping into an alive state $s' \leq i$ when $y \in I_i$ and in an alive state $s \leq i$. Let $\tilde{\mathbf{Q}}^{(i)}$ be the transition matrix of jumping into a default state $s' > i$ when $y \in I_i$ and in an alive state $s \leq i$, so that $\mathbf{Q} = \tilde{\mathbf{Q}}^{(n)} = 0$. The decomposition for $y \in I_i$ takes the following form
\[
\begin{align*}
\mathbf{Q}_{[1:2i]}^{[1:2n]} &= \begin{bmatrix} \mathbf{Q}^{(i)} - \text{diag} \left( \tilde{\mathbf{Q}}^{(i)} \mathbf{1}_{2(n-i)} \right) \mid \tilde{\mathbf{Q}}^{(i)} \end{bmatrix}^{2i \times 2(n-i)} \\
\mathbf{R}_{[1:2i]}^{[1:2n]} &= \begin{bmatrix} \mathbf{R}^{(i)} - \text{diag} \left( \tilde{\mathbf{Q}}^{(i)} \mathbf{1}_{2(n-i)} \right) \mid \mathbf{0}_{2i \times 2(n-i)} \end{bmatrix}^{2i \times 2(n-i)}
\end{align*}
\]
so that $Q^{(i)}, \tilde{Q}^{(i)}, R^{(i)}, \mu^{(i)}, \Sigma^{(i)}$ are given by

\[
\begin{align*}
\tilde{Q}^{(i)} &= Q_{[1:2i,2i+1:2n]} \\
Q^{(i)} &= Q_{[1:2i,1:2i]} + \text{diag} (\tilde{Q}^{(i)} 1_{2(n-i)}) \\
R^{(i)} &= R_{[1:2i,1:2i]} + \text{diag} (\tilde{Q}^{(i)} 1_{2(n-i)}) \\
\mu^{(i)} &= \mu_{[1:2i,1:2i]} \\
\Sigma^{(i)} &= \Sigma_{[1:2i,1:2i]}
\end{align*}
\]

where $X_{[1:k]}$ takes the first $k$ rows of matrix $X$ and $X_{[1:k,1:l]}$ takes the first $k$ rows and $l$ columns of matrix $X$. Note that in the $n = 2$ aggregate state case, we have $\tilde{Q}^{(1)} = \text{diag} [\tilde{Q}^{(1)} 1_2]$ as we have no joint jumps in liquidity state and in the aggregate state.

Let $\tilde{v}^{(i)} \exp (y)$ be the recovery or salvage value of the firm when default is declared in states $s > i$ for $y \in I$. Let $v^{(i)}$ be the vector unlevered values for states $(i + 1, ..., n) \times (H, L)$ (i.e., it is of dimension $2(n - i) \times 1$), and let $\alpha^{(i)}$ be the effective recovery ratios for those same states. Then, we have $\tilde{v}^{(i)} = \alpha^{(i)} \text{diag} (v^{(i)})$.

**Debt valuation within an interval $I_i$.** Debt valuation follows the following differential equation on interval $I_i$:

\[
(R^{(i)} + mI_i - Q^{(i)}) D^{(i)} = \mu^{(i)} (D^{(i)})' + \frac{1}{2} \Sigma^{(i)} (D^{(i)})'' + 1_{(i<n)} \tilde{Q}^{(i)} \tilde{v}^{(i)} \exp (y)
+ (c 1_{2i} - \chi^{(i)} N^{(i)}) + m \cdot p 1_{2i}
\]

(15)

where $\tilde{Q}^{(i)} \tilde{v}^{(i)} \exp (y)$ represents the intensity of jumping into default times the recovery in the default state and $m \cdot p 1_{2i}$ represents the intensity of randomly maturing times the payoff in the maturity state. Next, let us conjecture a solution of the kind $g \exp (\gamma y) + k^{(i)}_0 + k^{(i)}_1 \exp (y)$ where $g$ is a vector and $\gamma$ is a scalar. The particular part stemming from $c^{(i)}$ is solved by a term $k^{(i)}_0$ with

\[
\begin{align*}
k^{(i)}_0 &= (R^{(i)} + mI_i - Q^{(i)})^{-1} (c + m \cdot p) 1_{2i} - \chi^{(i)} N^{(i)}
\end{align*}
\]

(16)
and the particular part stemming from $\tilde{Q}^{(i)}\tilde{v}^{(i)}$ is solved by a term $k_1^{(i)} \exp(y)$ with

$$
k_1^{(i)} = \left( R^{(i)} + mI - Q^{(i)} - \mu^{(i)} - \frac{1}{2} \Sigma^{(i)} \right)^{-1} \tilde{Q}^{(i)} \tilde{v}^{(i)}
$$

(17)

It should be clear that $k_1^{(n)} = 0$ as on $I_n$ there is no jump in the aggregate state that would result in immediate default. Plugging in, dropping the $c^{(i)}$ and $\tilde{Q}^{(i)}\tilde{v}^{(i)} \exp(y)$ terms, canceling out $\exp(\gamma y) > 0$, we have

$$
0_{2i} = \left( Q^{(i)} + mI - R^{(i)} \right) g + \mu^{(i)} \gamma g + \frac{1}{2} \Sigma^{(i)} \gamma^2 g
$$

(18)

Following JR06, we premultiply by $2 \left( \Sigma^{(i)} \right)^{-1}$ and define $h = \gamma g$ to get

$$
\gamma g = h
$$

(19)

$$
\gamma h = -2 \left( \Sigma^{(i)} \right)^{-1} \mu^{(i)} h + 2 \left( \Sigma^{(i)} \right)^{-1} \left( R^{(i)} + mI - Q^{(i)} \right) g
$$

(20)

Stacking the vectors $j = \begin{bmatrix} g \\ h \end{bmatrix}$ we have

$$
\gamma j = \begin{bmatrix} 0_{2i} \\ 2 \left( \Sigma^{(i)} \right)^{-1} \left( R^{(i)} + mI - Q^{(i)} \right) -2 \left( \Sigma^{(i)} \right)^{-1} \mu^{(i)} \end{bmatrix}
$$

(21)

where $I$ is of appropriate dimensions. The problem is now a simple eigenvalue-eigenvector problem and each solution $j$ is a pair $\begin{bmatrix} \gamma_j^{(i)} \\ j_j^{(i)} \end{bmatrix}$ or rather $\begin{bmatrix} \gamma_j^{(i)} \\ g_j^{(i)} \end{bmatrix}$, as the vector $j_j^{(i)}$ contains the same information as $g_j^{(i)}$ when we know $\gamma_j^{(i)}$, so we discard the lower half of $j_j^{(i)}$. The number of solutions $j$ to this eigenvector-eigenvalue problem is $4i$. Let

$$
G^{(i)} = \begin{bmatrix} g_1^{(i)} , ..., g_{2i \times 2i}^{(i)} \end{bmatrix}
$$

(22)
be the matrix of eigenvectors, and let

\[ \gamma^{(i)} \equiv \begin{bmatrix} \gamma_{1}^{(i)} & \cdots & \gamma_{2 \times 2 \times i}^{(i)} \end{bmatrix}' \tag{23} \]

\[ \Gamma^{(i)} \equiv \text{diag} [\gamma^{(i)}] \tag{24} \]

be the corresponding vector and diagonal matrix, respectively, of eigenvalues.

The general solution on interval \( i \) is thus

\[ D^{(i)}_{2i \times 1} = G^{(i)}_{2i \times 4i} \cdot \exp \left( \Gamma^{(i)} y \right) \cdot c^{(i)}_{4i \times 1} + k_{6}^{(i)}_{2i \times 1} + k_{4}^{(i)}_{2i \times 1} \exp (y) \tag{25} \]

where the constants \( c^{(i)} = \begin{bmatrix} c_{1}^{(i)} & \cdots & c_{2i}^{(i)} \end{bmatrix}' \) will have to be determined via conditions at the boundaries of interval \( I_{i} \) (\textbf{NOTE:} \( c_{j}^{(i)} \neq c \) where \( c \) is the coupon payment).

**Boundary conditions.** The different value functions \( D^{(i)} \) for \( i \in \{1, \ldots, n\} \) are linked at the boundaries of their domains \( I_{i} \). Note that \( I_{i} \cap I_{i+1} = \{y_{\text{def}}(i + 1)\} \) for \( i < n \).

For \( i = n \), we can immediately rule out all positive solutions to \( \gamma \) as debt has to be finite and bounded as \( y \to \infty \), so that the entries of \( C^{(n)} \) corresponding to positive eigenvalues will be zero:²

\[ \lim_{y \to \infty} |D^{(n)}(y)| < \infty \tag{26} \]

For \( i < n \), we must have value matching of the value functions that are alive across the boundary, and we must have value matching of the value functions that die across the boundary:

\[ D^{(i+1)}(y_{\text{def}}(i + 1)) = \begin{bmatrix} D^{(i)}(y_{\text{def}}(i + 1)) \\ \tilde{v}^{i+1}_{H} \\ \tilde{v}^{i+1}_{L} \end{bmatrix} \exp (y_{\text{def}}(i + 1)) \tag{27} \]

For \( i < n \), we must have mechanical (i.e. non-optimal) smooth pasting of the value functions

² According to JR06, there are exactly \( 2 \times |S| = 2n \) eigenvalues of \( A \) in the left open half plane (i.e. negative) and \( 2n \) eigenvalues in the right open half plane (i.e. positive).
that are alive across the boundary:

\[
(D^{i+1})' (y_{def}(i+1))_{[1:2i]} = (D^{(i)})' (y_{def}(i+1))
\]  

(28)

where \(x_{[1...2i]}\) selects the first 2i rows of vector \(x\).

Lastly, for \(i = 1\), we must have

\[
D^{(1)}(y_{def}(1)) = \begin{bmatrix} \tilde{v}_H^1 \\ \tilde{v}_L^1 \end{bmatrix} \exp(y_{def}(1))
\]

(29)

**Full solution.** We can now state the full solution to the debt valuation given cut-off strategies:

**Proposition 1.** The debt value functions \(D\) for a given default vector \(y_B\) are

\[
D(y) = \begin{cases} 
D^{(n)}(y) = G^{(n)} \cdot \exp(\Gamma^{(n)} y) \cdot c^{(n)} + k_0^{(n)} & y \in I_n \\
\vdots \\
D^{(i)}(y) = G^{(i)} \cdot \exp(\Gamma^{(i)} y) \cdot c^{(i)} + k_0^{(i)} + k_1^{(i)} \exp(y) & y \in I_i \\
\vdots \\
D^{(1)}(y) = G^{(1)} \cdot \exp(\Gamma^{(1)} y) \cdot c^{(1)} + k_0^{(1)} + k_1^{(1)} \exp(y) & y \in I_1
\end{cases}
\]
with the following boundary conditions to pin down vectors \( \mathbf{c}^{(i)} \):

\[
\lim_{y \to \infty} \left| \frac{\mathbf{D}^{(n)}(y)}{2n \times 1} \right| < \infty \tag{30}
\]

\[
\mathbf{D}^{(i+1)}(\text{y}_{\text{def}}(i+1)) = \begin{bmatrix}
\mathbf{D}^{(i)}(\text{y}_{\text{def}}(i+1)) \\
\tilde{v}_{i+1}^H \\
\tilde{v}_{i+1}^L
\end{bmatrix} \exp(\text{y}_{\text{def}}(i+1)) \tag{31}
\]

\[
(\mathbf{D}^{(i+1)})'_{2(i+1) \times 1} (\text{y}_{\text{def}}(i+1))_{[1:2i]} = (\mathbf{D}^{(i)})'_{2i \times 1} (\text{y}_{\text{def}}(i+1)) \tag{32}
\]

\[
\mathbf{D}^{(1)}(\text{y}_{\text{def}}(1)) = \begin{bmatrix}
\tilde{v}_{1}^H \\
\tilde{v}_{1}^L
\end{bmatrix} \exp(\text{y}_{\text{def}}(1)) \tag{33}
\]

where \( \mathbf{x}_{[1:2i]} \) selects the first \( 2i \) rows of vector \( \mathbf{x} \).

Note that the derivative of the debt value vector is

\[
(\mathbf{D}^{(i)})'_{2i \times 1} (y) = \mathbf{G}^{(i)} \mathbf{\Gamma}^{(i)} \cdot \exp(\mathbf{\Gamma}^{(i)}y) \cdot \mathbf{c}^{(i)} + \mathbf{k}^{(i)} \exp(y) \tag{34}
\]

where we note that \( \mathbf{\Gamma}^{(i)} \cdot \exp(\mathbf{\Gamma}^{(i)}y) = \exp(\mathbf{\Gamma}^{(i)}y) \cdot \mathbf{\Gamma}^{(i)} \) as both are diagonal matrices (although this interchangeability only is important when \( s = 1 \) as it then helps collapse some equations).

The first boundary condition (30) essentially implies that we can discard any positive entries of \( \gamma^{(n)} \) by setting the appropriate coefficients of \( \mathbf{C}^{(n)} \) to 0. The second boundary condition (31) implies that we have value matching at any boundary \( \text{y}_{\text{def}}(i+1) \) for \( i < n \), be it to a continuation state or a bankruptcy state. The third boundary condition (32) implies that we also have smooth pasting at the boundary \( \text{y}_{\text{def}}(i+1) \) for those states in which the firm stays alive on both sides of the boundary. Finally, the fourth boundary condition (33) implies value matching at the boundary \( \text{y}_{\text{def}}(1) \), but of course only for those states in which the firm is still alive.
1.6 Equity

The equity holders are unaffected by the individual liquidity shocks the debt holders are exposed to. The only shocks the equity holders are directly exposed to are the shifts in $\mu(s)$ and $\sigma(s)$, i.e. shifts to the cash-flow process.

However, as debt has finite maturity and has to be rolled over by assumption, equity holders are indirectly affected by liquidity shocks in the market through the effect it has on debt prices. Thus, when debt matures, it is either rolled over if the debt holders are of type $H$, or it is reissued to different debt holders in the case that the former debt holder is of type $L$. Either way, there is a cash flow (inflow or outflow) of $m[S^{(i)} \cdot D^{(i)}(y) - p1_i]$ at each instant as a mass $m \cdot dt$ of debt holders matures on $[t, t + dt]$.

For notational ease, we will denote by double letters (e.g. $xx$) a constant for equity that takes a similar place as a single letter (i.e. $x$) constant for debt. Then, the HJB for equity on interval $I_i$ is given by

$$
\left( RR^{(i)} - QQ^{(i)} \right) E^{(i)}(y) = \mu \mu^{(i)} \left( E^{(i)} \right)'(y) + \frac{1}{2} \Sigma \Sigma^{(i)} \left( E^{(i)} \right)''(y) + 1_{\left\{ i<n \right\}} \overline{QQ}^{(i)} 0_i + 1_i \exp(y) - (1 - \pi) c1_i + m \left[ S^{(i)} \cdot D^{(i)}(y) - p1_i \right] \quad (35)
$$

where

$$
\overline{QQ}^{(i)} = QQ_{[1:i,n:1]} \quad (36)
$$

$$
QQ^{(i)} = QQ_{[1:i,1:i]} + \text{diag} \left( \overline{QQ}^{(i)} 1_{n-i} \right) \quad (37)
$$

$$
RR^{(i)} = RR_{[1:i,1:i]} + \text{diag} \left( \overline{QQ}^{(i)} 1_{n-i} \right) \quad (38)
$$

$$
\mu \mu^{(i)} = \mu \mu_{[1:i,1:i]} \quad (39)
$$

$$
\Sigma \Sigma^{(i)} = \Sigma \Sigma_{[1:i,1:i]} \quad (40)
$$

Note that $\overline{QQ}^{(i)}$ is the transition matrix only between aggregate states that is also an $i \times i$ square matrix, and $S^{(i)}$ is a $i \times 2i$ matrix that selects which debt values the firm is able to issue (each row has to sum to 1), and $m$ is a scalar (NOTE: In contrast to $R$, the matrix
RR does not contain the maturity intensity \( m \)). For example, for \( i = 2 \), if the company is able to place debt only to \( H \) types, then \( S^{(2)} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \). It is important that for each row \( i \) only entries \( 2i - 1 \) and \( 2i \) are possibly nonzero, whereas all other entries are identically zero (otherwise, one would issue bonds belonging to a different state).

Writing out \( D^{(i)}(y) = G^{(i)} \exp (\Gamma^{(i)} y) \) and conjecturing a solution to the particular, non-constant part \( KK^{(i)} \exp (\Gamma^{(i)} y) c^{(i)} \), we have

\[
\begin{align*}
(\widetilde{RR}^{(i)} - \widetilde{QQ}^{(i)}) \quad & KK^{(i)} \exp (\Gamma^{(i)} y) c^{(i)} \\
= & \left[ \mu \mu^{(i)} \cdot KK^{(i)} \cdot \Gamma^{(i)} + \frac{1}{2} \Sigma \Sigma^{(i)} KK^{(i)} \cdot (\Gamma^{(i)})^2 + m \cdot S^{(i)} \cdot G^{(i)} \right] \exp (\Gamma^{(i)} y) c^{(i)} \quad (41)
\end{align*}
\]

We can solve this by considering each \( \gamma_j^{(i)} \) separately — recall that \( c^{(i)} \) is a vector and \( \exp (\Gamma^{(i)} y) \) is a diagonal matrix and in total there are \( 4i \) different roots. Consider the part of the particular part \( S^{(i)} \cdot g_j^{(i)} \exp (\gamma_j^{(i)} y) \cdot c_j^{(i)} \) and our conjecture gives \( KK_j^{(i)} \exp \left( \frac{\gamma_j^{(i)} y}{4i} \right) \cdot c_j^{(i)} \)

for each root \( j \in [1, ..., 4i] \). Plugging in and multiplying out the scalar \( \exp (\gamma_j^{(i)} y) c_j^{(i)} \), we find that

\[
(\widetilde{RR}^{(i)} - \widetilde{QQ}^{(i)}) \quad KK_j^{(i)} = \mu \mu^{(i)} \cdot KK_j^{(i)} \cdot \gamma_j^{(i)} + \frac{1}{2} \Sigma \Sigma^{(i)} KK_j^{(i)} \cdot (\gamma_j^{(i)})^2 + m \cdot S^{(i)} \cdot g_j^{(i)} \quad (42)
\]

Solving for \( KK_j^{(i)} \), we have

\[
KK_j^{(i)} = \left[ (\widetilde{RR}^{(i)} - \widetilde{QQ}^{(i)} - \mu \mu^{(i)} \cdot \gamma_j^{(i)} - \frac{1}{2} \Sigma \Sigma^{(i)} \cdot (\gamma_j^{(i)})^2 \right]^{-1} m \cdot S^{(i)} g_j^{(i)} \quad (43)
\]

Finally, for the homogenous part we use the same approach as above, but now we have less states as the individual liquidity state drops out. Thus, we conjecture \( gg \exp (\gamma y) \) to get

\[
0_i = (\widetilde{QQ}^{(i)} - \widetilde{RR}^{(i)}) gg + \mu \mu^{(i)} \gamma \gamma gg + \frac{1}{2} \Sigma \Sigma^{(i)} \gamma \gamma gg \quad (44)
\]
so that, again, we have the following eigenvector eigenvalue problem

\[
\gamma_{jj} = \begin{bmatrix}
0_i & I_i \\
2 (\Sigma^i)^{-1} (\overline{RR} - \overline{QQ}) & -2 (\Sigma^i)^{-1} \mu^i
\end{bmatrix}
\begin{bmatrix}
jj \\
jj
\end{bmatrix} = \frac{1}{2i \times 2i} \mathbf{A} \mathbf{A}^{(i)} (jj)
\] (45)

which gives \((\gamma_j^{(i)}, \mathbf{g}_j^{(i)})\) for \(j \in [1, \ldots, 2i]\) solutions. We stack these into a matrix of eigenvectors \(\mathbf{G}^{(i)}\) and a vector of eigenvalues \(\gamma^{(i)}\), from which we define the diagonal matrix of eigenvalues \(\mathbf{\Gamma}^{(i)} \equiv \text{diag}(\gamma^{(i)})\). What remains is to solve for \(k^0_i\) and \(k^1_i\). We have

\[
k^0_i = \frac{1}{\left(\overline{RR} - \overline{QQ}\right)^{-1} \left[- (1 - \pi) c \mathbf{1}_i + m \left(\mathbf{S}^{(i)} k^0_i - \mathbf{p}^{(i)}_1\right)\right]}
\] (46)

and

\[
k^1_i = \frac{1}{\left(\overline{RR} - \overline{QQ} - \mu^i - \frac{1}{2} \Sigma^i\right)^{-1} \left(\mathbf{1}_i + m \cdot \mathbf{S}^{(i)} k^1_i\right)}
\] (47)

with \(k^1_i = 0\).

We are left with the following proposition.

**Proposition 2.** The equity value functions \(\mathbf{E}\) for a given default vector \(\mathbf{y}_B\) are

\[
\mathbf{E}(y) = \begin{cases}
\mathbf{E}^{(n)}(y) = \mathbf{G}^{(n)} \cdot \exp \left(\mathbf{\Gamma}^{(n)} y\right) \cdot \mathbf{c}^{(n)} + \mathbf{K}^{(n)} \exp \left(\mathbf{\Gamma}^{(n)} y\right) \mathbf{c}^{(n)} + k^0_n + k^1_n \exp (y) & y \in I_n \\
\vdots & \vdots \\
\mathbf{E}^{(i)}(y) = \mathbf{G}^{(i)} \cdot \exp \left(\mathbf{\Gamma}^{(i)} y\right) \cdot \mathbf{c}^{(i)} + \mathbf{K}^{(i)} \exp \left(\mathbf{\Gamma}^{(i)} y\right) \mathbf{c}^{(i)} + k^0_i + k^1_i \exp (y) & y \in I_i \\
\vdots & \vdots \\
\mathbf{E}^{(1)}(y) = \mathbf{G}^{(1)} \cdot \exp \left(\mathbf{\Gamma}^{(1)} y\right) \cdot \mathbf{c}^{(1)} + \mathbf{K}^{(1)} \exp \left(\mathbf{\Gamma}^{(1)} y\right) \mathbf{c}^{(1)} + k^0_1 + k^1_1 \exp (y) & y \in I_1
\end{cases}
\]
with the following boundary conditions to pin down the vector $cc^{(i)}$: 

\[
\lim_{y \to \infty} \left| E^{(n)}(y) \exp (-y) \right|_{n \times 1} < \infty \quad (48)
\]

\[
E^{(i+1)}(y_{def}(i+1)) = \begin{bmatrix} E^{(i)}(y_{def}(i+1)) \\ 0 \end{bmatrix} \quad (49)
\]

\[
(E^{(i+1)})'(y_{def}(i+1))_{[1:i]} = (E^{(i)})'(y_{def}(i+1))_{[1:i]} \quad (50)
\]

\[
E^{(i)}(y_{def}(1))_{[i]} = 0 \quad (51)
\]

where $x_{[1:i]}$ selects the first $i$ rows of vector $x$.

Note first the dimensionalities: $\Gamma^{(i)}_{2i \times 2i}$, $G^{(i)}_{i \times 2i}$, and $\Gamma^{(i)}_{4i \times 2i}$, $G^{(i)}_{2i \times 4i}$. Note second the derivative of the equity value vector is

\[
(E^{(i)})'(y) = GG^{(i)} \Gamma^{(i)} \cdot \exp \left( \Gamma^{(i)} y \right) \cdot cc^{(i)} + KK^{(i)} \Gamma^{(i)} \exp \left( \Gamma^{(i)} y \right) c^{(i)} + kk^{(i)} \exp (y) \quad (52)
\]

where we note that $\Gamma^{(i)} \cdot \exp \left( \Gamma^{(i)} y \right) = \exp \left( \Gamma^{(i)} y \right) \cdot \Gamma^{(i)}$ and $\Gamma^{(i)} \cdot \exp \left( \Gamma^{(i)} y \right) = \exp \left( \Gamma^{(i)} y \right) \cdot \Gamma^{(i)}$ as both are diagonal matrices (although this interchangeability only is important when $s = 1$ as it then helps collapse some equations).

The optimality conditions for bankruptcy boundaries $\{y_{def}(i)\}_{i}$ are given by

\[
(E^{(i)})'(y_{def}(i))_{[i]} = 0 \quad (53)
\]

i.e., a smooth pasting condition at the boundaries at which default is declared.

Finally, we summarize all the different matrices and vectors involved in the derivation of the propositions in Table 1.
Table 1: Matrix & Vector Dimensions.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Interpretation</th>
<th>Dimension</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\mathbf{D}^{(i)}(y))</td>
<td>Debt Value Function</td>
<td>(2i \times 1)</td>
</tr>
<tr>
<td>(\mu^{(i)})</td>
<td>(Log)-Drifts</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(\Sigma^{(i)})</td>
<td>Volatilities</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(R^{(i)})</td>
<td>Discount rates and maturity</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(\chi^{(i)})</td>
<td>Holding cost slopes</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(N^{(i)})</td>
<td>Holding cost intercepts</td>
<td>(2i \times 1)</td>
</tr>
<tr>
<td>(Q^{(i)})</td>
<td>Transition to cont. states</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(\tilde{Q}^{(i)})</td>
<td>Transition to default states</td>
<td>(2i \times 2(n - i))</td>
</tr>
<tr>
<td>(\tilde{v}^{(i)})</td>
<td>Vector of recovery values</td>
<td>(2(n - i) \times 1)</td>
</tr>
<tr>
<td>(W^{(i)})</td>
<td>Mid-point weighting matrix</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(\Gamma^{(i)})</td>
<td>Diag matrix of eigenvalues</td>
<td>(4i \times 4i)</td>
</tr>
<tr>
<td>(\mathbf{G}^{(i)})</td>
<td>Matrix of eigenvectors</td>
<td>(2i \times 4i)</td>
</tr>
<tr>
<td>(k^{(i)}_0, k^{(i)}_1)</td>
<td>Coeff. of particular sol.</td>
<td>(2i \times 1)</td>
</tr>
<tr>
<td>(c^{(i)})</td>
<td>Vector of constants</td>
<td>(4i \times 1)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Interpretation</th>
<th>Dimension</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\mathbf{E}^{(i)}(y))</td>
<td>Equity Value Function</td>
<td>(i \times 1)</td>
</tr>
<tr>
<td>(\mu\mu^{(i)})</td>
<td>(Log-)Drifts</td>
<td>(i \times i)</td>
</tr>
<tr>
<td>(\Sigma\Sigma^{(i)})</td>
<td>Volatilities</td>
<td>(i \times i)</td>
</tr>
<tr>
<td>(RR^{(i)})</td>
<td>Discount rates</td>
<td>(i \times i)</td>
</tr>
<tr>
<td>(\pi)</td>
<td>Tax rate</td>
<td>(1 \times 1)</td>
</tr>
<tr>
<td>(QQ^{(i)})</td>
<td>Transition to cont. states</td>
<td>(i \times i)</td>
</tr>
<tr>
<td>(\mathbf{A}A^{(i)})</td>
<td>Matrix to be decomposed</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(\Gamma\Gamma^{(i)})</td>
<td>Diag matrix of eigenvalues</td>
<td>(2i \times 2i)</td>
</tr>
<tr>
<td>(\mathbf{G}G^{(i)})</td>
<td>Matrix of eigenvectors</td>
<td>(i \times 2i)</td>
</tr>
<tr>
<td>(k^{(i)}_0, k^{(i)}_1)</td>
<td>Coeff. of particular sol.</td>
<td>(i \times 1)</td>
</tr>
<tr>
<td>(\mathbf{S}^{(i)})</td>
<td>Issuance matrix</td>
<td>(i \times 2i)</td>
</tr>
<tr>
<td>(KK^{(i)})</td>
<td>Coeff. of particular sol.</td>
<td>(i \times 4i)</td>
</tr>
<tr>
<td>(cc^{(i)})</td>
<td>Vector of constants</td>
<td>(2i \times 1)</td>
</tr>
</tbody>
</table>

2 Computing the Model-implied Moments by Ratings

We follow the steps below in computing the model-implied aggregate moments on default probability, credit spreads, and others. These are also described, in more concise form, at the beginning of Section 3.3 in the paper. Due to a proliferation of material in the appendix in this revision, we have so far opted to leave the detailed description given below out of the appendix.

**Step 1: Construct the model-implied mapping between market leverage and credit spread.** We use Monte-Carlo simulation method to compute the total credit spread for a given maturity and aggregate state on a grid of log cash flow level. We then translate the cash flow level to its model-implied market leverage level through the analytical solution to the model. This gives us a scattered plot of credit spread over market leverage. We then fit a piece-wise hyperbolic curve to the scatters to derive the model-predicted credit spread over the continuous range of market leverage. The result is presented in Figure 3 as the solid line. Consistent with the findings in the previous literature as David (2008), Bhamra, Kuehn, and Strebulaev (2010); credit spread is a convex function of the market leverage in the model.
This means, due to the dispersion in market leverage for a given rating class (See figure 2), average credit spread of the firms for a given rating class will be bigger than the credit spread of the average firm in that group; and the former is empirically observed and therefore shall be the target of our calibration.

Step 2: Matching with the empirical distribution of market leverage in Compustat data. We compute the market leverage (i.e. book debt over the sum of market equity and book debt) of each Compustat firm-quarter observation where we have the rating information. We report the leverage distribution by rating class in Figure 2. We then compute the model-implied credit spread for each empirical firm-quarter observation using the fitted curve we constructed in Step 1. This gives a credit spread value the model predicts for each firm-quarter observation in the data.

Step 3: Aggregation by macro state, maturity and rating class. For a given rating class and maturity, we aggregate all model-implied spreads first by quarter then by aggregate state (Note: We define a quarter to be in the B state if any of its three months is classified as a NBER recession month) to arrive at the model implied moments reported in Table 2.

We used the same method in computing the model-implied moments on default probability and structural liquidity-default decomposition.

3 Calibration With Lower Liquidity Shock Intensities

While we think the broader interpretation of liquidity shocks helps justify the approach of matching liquidity shock intensity with corporate bond market turnover to some extent, we also consider an alternative calibration to demonstrate the sensitivity of our results to this concern. In the new calibration, both $\xi_G$ and $\xi_B$ are reduced by half. Then, we recalibrate the three free parameters $N, \chi_G, \chi_B$ by targeting the 6 bid-ask spread moments (as discussed in Point 1).

The results from this calibration are reported in Table 2. The re-calibrated model still does a good matching the moments of bid-ask spreads, but the total credit spreads and bond-CDS
spreads become significantly lower. The reason is that lowering liquidity shock intensities, all else equal, raises bid-ask spreads (see also Table 4 in the main text on this point). To reduce the bid-ask spreads, the calibration then reduces the holding costs by lowering \( N, \chi_G, \chi_B \), which, together with lower liquidity shock intensities, reduce the bond-CDS spreads and total credit spreads. We also show the decomposition of these re-calibrated total credit spreads in Table 3. The pure-default components of the spreads under the calibration is similar to the original calibration, while the pure liquidity component, default-driven liquidity component, and the liquidity-driven default components all roughly drop by half.

These findings show that our model’s performance is indeed sensitive to the calibration of \( \xi_G, \xi_B \). Put differently, they show that in order to jointly explain the bid-ask spreads and bond-CDS spreads, our model calls for relatively high levels of liquidity shock intensities.

4 Dollar Decomposition

We decompose corporate bond prices into 4 components in a similar way as we decompose the spreads. Specifically, consider the following 5 different bond prices:

1. \( B^{NL} \): The price of a default-free bond in a setting without any liquidity frictions, with maturity \( T \), face value \( F \), and annualized coupon rate \( c \) equal to the risk-free rate \( r \). This bond will be priced at par, \( B^{NL} = F \).

2. \( B \): The price of the same default-free bond in the presence of liquidity frictions.

3. \( P \): The price of the \( T \)-year defaultable bond in the presence of liquidity frictions, using the default policy computed from the benchmark model.

4. \( P^{NL} \): The price of the \( T \)-year defaultable bond in a setting without any liquidity frictions, but keeping the default policy the same as the benchmark model.

5. \( P^{DEF} \): The price of the \( T \)-year bond in a setting without any liquidity frictions, with an updated default policy that is optimal for the firm in the absence of liquidity frictions.
Table 2: **Comparison of benchmark to re-calibrated $\xi = 0.35, 0.5$ case (10 year bonds).**

The “benchmark” case is our benchmark calibration. The “$\xi = 0.35, 0.5$” is when we halve the liquidity shock intensities in both states, and recalibrate the remaining parameters. Thus, this is not a comparative statics exercise.

<table>
<thead>
<tr>
<th></th>
<th>Panel A. Credit spreads (bps)</th>
<th>Panel B. Bid-Ask Spreads (bps)</th>
<th>Panel C. Bond-CDS Spreads (bps)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>State $G$</td>
<td>State $B$</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Aaa/Aa A A Baa Baa</td>
<td>Aaa/Aa A A Baa Baa</td>
<td>Aaa/Aa A A Baa Baa</td>
</tr>
<tr>
<td>benchmark</td>
<td>86 122 182 301</td>
<td>136 185 261 404</td>
<td>48 53 61 60</td>
</tr>
<tr>
<td>$\xi = 0.35, 0.5$</td>
<td>58 89 143 252</td>
<td>96 136 201 327</td>
<td>22 24 26 61</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>State $G$</td>
<td>State $B$</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Superior Investment Junk</td>
<td>Superior Investment Junk</td>
<td></td>
</tr>
<tr>
<td>benchmark</td>
<td>40 47 61</td>
<td>114 137 186</td>
<td>48 53 61 60</td>
</tr>
<tr>
<td>$\xi = 0.35, 0.5$</td>
<td>40 48 62</td>
<td>109 135 186</td>
<td>22 24 26 61</td>
</tr>
</tbody>
</table>

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Table 3: Structural Decomposition for 10-Year Bonds Across Ratings for re-calibrated \( \xi = 0.35, 0.5 \) case. We perform the structural liquidity-default decomposition for a 10-year bond, given rating and aggregate state, and then aggregate over the empirical leverage distribution in Compustat. The reported credit spreads are relative to the risk-free rate.

<table>
<thead>
<tr>
<th>Rating</th>
<th>State</th>
<th>Credit Spread</th>
<th>Structural Decomposition</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>Pure Def</td>
</tr>
<tr>
<td>Aaa/Aa</td>
<td>G</td>
<td>43</td>
<td>20</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(45)</td>
<td>(1)</td>
</tr>
<tr>
<td></td>
<td>B</td>
<td>56</td>
<td>22</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(39)</td>
<td>(2)</td>
</tr>
<tr>
<td>A</td>
<td>G</td>
<td>74</td>
<td>47</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(63)</td>
<td>(2)</td>
</tr>
<tr>
<td></td>
<td>B</td>
<td>96</td>
<td>56</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(58)</td>
<td>(3)</td>
</tr>
<tr>
<td>Baa</td>
<td>G</td>
<td>128</td>
<td>94</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(74)</td>
<td>(2)</td>
</tr>
<tr>
<td></td>
<td>B</td>
<td>161</td>
<td>110</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(68)</td>
<td>(3)</td>
</tr>
<tr>
<td>Ba</td>
<td>G</td>
<td>237</td>
<td>194</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(82)</td>
<td>(3)</td>
</tr>
<tr>
<td></td>
<td>B</td>
<td>287</td>
<td>219</td>
</tr>
<tr>
<td></td>
<td>(%)</td>
<td>(76)</td>
<td>(3)</td>
</tr>
</tbody>
</table>
The price gap between the default-free bond and defaultable bond $B^{NL} - P$ can be viewed as the “value lost” due to default and liquidity. With the set of 5 prices $(B^{NL}, B, P^{DEF}, P^{NL}, P)$, we can now decompose the total gap $B^{NL} - P$ into 4 components:

$$B^{NL} - P = \Delta P^{pureDEF} + \Delta P^{LIQ\rightarrow DEF} + \Delta P^{pureLIQ} + \Delta P^{DEF\rightarrow LIQ}$$  \hspace{1cm} (54)$$

where

$$\Delta P^{pureDEF} = B^{NL} - P^{DEF}$$  \hspace{1cm} (55)$$

$$\Delta P^{LIQ\rightarrow DEF} = (B^{NL} - P^{NL}) - \Delta P^{pureDEF}$$  \hspace{1cm} (56)$$

$$\Delta P^{pureLIQ} = B^{NL} - B$$  \hspace{1cm} (57)$$

$$\Delta P^{DEF\rightarrow LIQ} = (P^{NL} - P) - \Delta P^{pureLIQ}$$  \hspace{1cm} (58)$$

After conducting the decomposition at firm level (separately for the good and bad aggregate states), we average the results across firms within each rating class to get the average decompositions by ratings.
References

