

The Information-Driven Financial Accelerator

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- Empirically, it is well-known that credit spreads are large, volatile and countercyclical ([Gilchrist and Zakrajsek, 2012](#); [Greenwood and Hanson, 2013](#))
- What are the sources of credit market and macroeconomic fragility?
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- This paper shows that **imperfect information** in credit markets is a strong force behind credit cycles.
 - Debt investors are uninformed about firms' creditworthiness
 - Update beliefs using publicly-available forecasts of profit outlook

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Bond prices move in response to the arrival of noisy information, not just to changes in fundamentals.

- Policies that help to anchor investors' expectations could have substantial financial stability benefits.

New Fact

Changes in professional forecasters' expectations of **quarter-ahead** corporate profit

$$\text{rev}_t = E_t[\pi_{t+1}] - E_{t-1}[\pi_{t+1}]$$

jointly predict:

- excess corporate bond returns
 - macroeconomic aggregates
- } **at long horizons**

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$$R_{t \rightarrow t+k} = \alpha + \beta \text{rev}_t + \gamma \text{controls}_t + u_{t+k}$$

σ_t

The combined effect of $\downarrow \text{rev}_t$ and $\uparrow \sigma_t$ during 2007 financial crisis:

- spreads \uparrow 80 basis points

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The combined effect of $\downarrow \text{rev}_t$ and $\uparrow \sigma_t$ during 2007 financial crisis:

- spreads \uparrow 80 basis points
- investment \downarrow 1 percentage point and GDP \downarrow 40 basis points

Dynamic Model with Financing and Investment

Costly debt financing

- default risk

Dynamic Model with Financing and Investment

Costly debt financing + Imperfect information

- default risk
- investors do not observe firm's state

$$Z_t = \rho_z Z_{t-1} + \varepsilon_t^Z$$

- learn from a noisy public signal

$$S_t = \varepsilon_t^Z + U_t$$

using a Kalman filter

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$$q_t(s_t, s_{t-1}, \dots, s_0)$$

Dynamic Model with Financing and Investment

Costly debt financing + Imperfect information \implies Amplification

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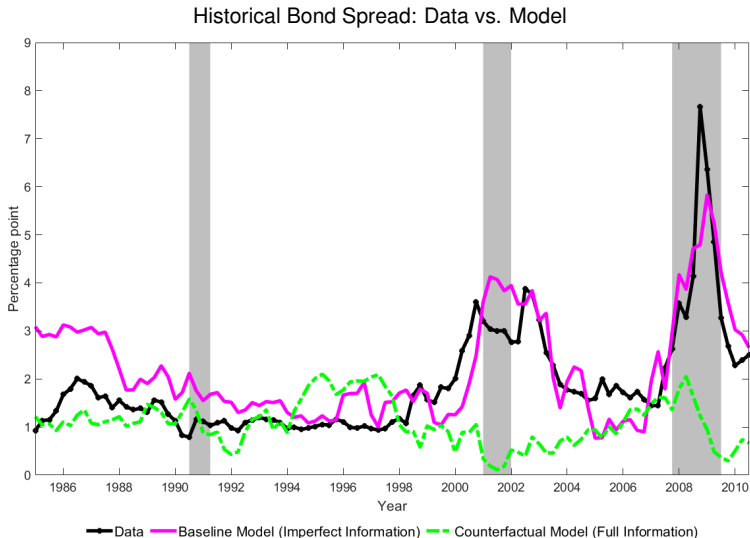
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Model-based counterfactual for 2007 financial crisis:

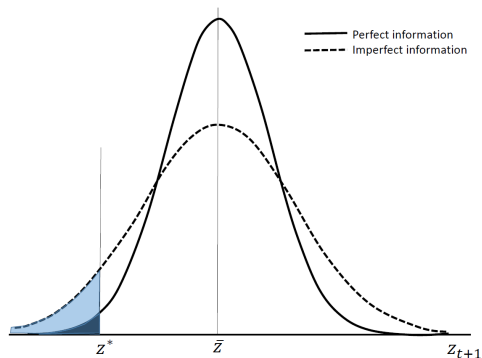
- 1/2 of increase in spread
 - 1/5 of contraction in aggregate investment
- } from noisy signals



Imperfect information model matches the **size** and **cyclical variation** of credit spreads

Information uncertainty:

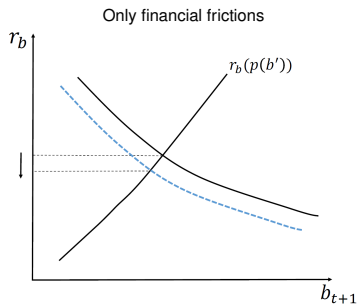
1. higher mean spread



subjective default probability > actual default probability

Information uncertainty:

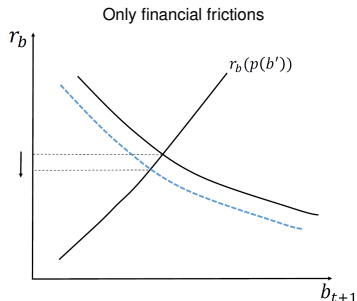
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2. countercyclical defaults and spreads



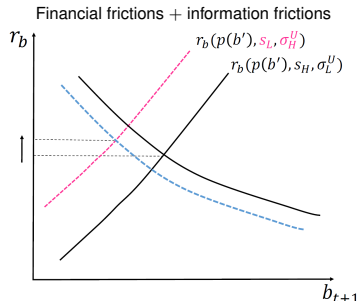
- default probability p increasing in b'
- lending schedule $r_b(p)$ increasing in b'
- recession $\Downarrow \rightarrow$ deleveraging $\rightarrow r_b \Downarrow b' \Downarrow$
(first-order effect)

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- lending schedule shifts to the left due to: (i) bad signal (s_L), (ii) noisy signal ($\sigma_u \uparrow$) in recession $\rightarrow r_b \uparrow$ & $b' \downarrow$

Microdata:

- IBES: firm-level estimates of earning forecasts
 - ICE/IDC and Warga: bond-level spreads
 - Compustat
- } 5,000 bonds & 10,000 firms
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 - **Quarter-ahead** forecast revisions are strongly and economically related to spreads and investment over **long horizons**

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⇒ Imperfect information in credit markets is a quantitatively important source of macroeconomic fragility.