

Financial Shock Transmission to Heterogeneous Firms: the Earnings-Based Borrowing Constraint Channel*

Livia Chițu¹

Magdalena Grothe¹

Tatjana Schulze²

Ine Van Robays¹

¹European Central Bank

²International Monetary Fund

European Economic Association

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*The views expressed in this paper are those of the authors and do not necessarily reflect the views of the European Central Bank nor of the IMF, its Executive Board, or IMF Management.

What we know:

- ▶ Firm heterogeneity and financial frictions play a role in the transmission of monetary policy shocks (e.g. Alder et al. 2024; Anderson & Cesa-Bianchi 2021; Cloyne et al. 2023; Gürkaynak et al. 2022; Jeenas, 2019; Ottonello & Winberry 2020; Palazzo & Yamarthy, 2022)

Motivation

What we know:

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What we understand less:

- ▶ Other types of shocks – global risk shocks – may matter and transmit heterogeneously across firms
- ▶ Limited firm-level literature on the transmission of global risk and MP shocks → challenge to disentangle the two shocks
- ▶ How the type of borrowing constraint affects transmission: asset-based vs. earnings-based borrowing constraint → Lian & Ma 2021: *Large US firms have 80% of their debt based on earnings, only 20% is collateralised by physical assets*

Research Agenda

Our contribution:

- ▶ Identification strategy to disentangle **global risk** and **monetary policy shocks** in an **integrated daily BVAR** exploiting cross-asset price movements
- ▶ We study two interrelated dimensions
 - (1) **firm heterogeneity**
 - (2) the **type of shocks**to understand how shocks transmit to firms' **financing conditions** (bonds & equity) and **default prospects**
- ▶ Tease out mechanisms by contrasting **asset-based** with **earnings-based** borrowing constraints, differentiating firms across **leverage** and **earnings**

Why does it matter?

- ▶ Shocks that **tighten firms' financing conditions** \Rightarrow possibly adverse consequences for investment and production
- ▶ Timely & policy relevant \Rightarrow US/global corporate sector recently hit simultaneously by these 2 shocks: **monetary policy tightening** and **global risk aversion**

Testable hypothesis

Heterogeneous effects across firms depending on the **type of borrowing constraint**:

- (1) **Asset-based collateral constraint**: Expect stronger responses from firms in the upper tail of the leverage distribution (i.e. higher leveraged firms)
- (2) **Earnings-based borrowing constraint**: Expect stronger responses from firms in the lower tail of the earnings distribution (i.e. less profitable firms)

Monetary policy and global risk shocks...

- ▶ Global risk shocks have **stronger** and **more heterogeneous effects** on corporate funding conditions which depend on firms' position within the **earnings distribution**

... and the earnings-based borrowing constraint transmission channel

- ▶ Responses of firms' funding costs to financial shocks are
 - (i) **more muted** for the upper tail of firm distribution by **leverage** (higher-leveraged firms)
 - (ii) **more pronounced** for the lower tail of firm distribution by **earnings** (less profitable firms)

Data

- ▶ Non-financial S&P 500 corporates (current and historical constituents)
- ▶ Sample period: **1,117 weeks** from 7-Jan-2000 to 17-Dec-2021
- ▶ **Bond-level:** Bloomberg [▶ Selection/Cleaning](#)
 - ▶ USD-denominated bonds
 - ▶ Option-adjusted spreads (OAS), composite ratings
 - ▶ Additional bond characteristics: duration, age, coupon, volume, embedded options, call dates
- ▶ **Firm-level:**
 - ▶ Datastream: equity prices and balance sheet
 - ▶ Moody's CreditEdge: expected default frequencies (EDFs)
 - ▶ Bloomberg: 5Y model-implied CDS spreads
- ▶ **Raw sample:** 436 firms, 12,996 bonds
- ▶ **Matched sample:** **407 firms, 7,825 bonds**
- ▶ **Aggregate-level:**
 - ▶ US 3m and 10Y yield, CESI, VIX, S&P 500 index, CAPE, US dollar NEER, US corp spread

Large firms exhibit considerable heterogeneity

Table 1: Firm characteristics – Summary statistics

	Mean	P25	Median	P75
EDF 1-Year (%)	0.41	0.03	0.05	0.19
Leverage ratio	47.68	30.70	42.46	57.51
Realized earnings per share	4.04	1.51	2.68	4.60
Expected earnings per share	4.52	1.69	2.94	5.03
Interest coverage ratio	12.92	3.45	7.25	13.72
S&P Issuer Rating	BBB+	BBB-	BBB+	A-

Note: Sample period: 2000/01/07 – 2021/12/17; Number of bond-week observations: 2,274,822; Number of bonds: 7,674; Number of firms: 407. The sample statistics are based on trimmed data.

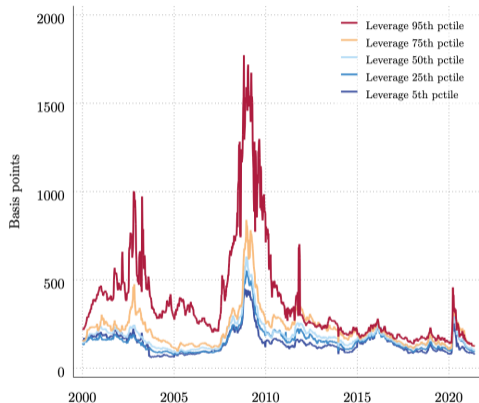
Heterogeneity in bond characteristics

Table 3: Bond characteristics – Summary statistics

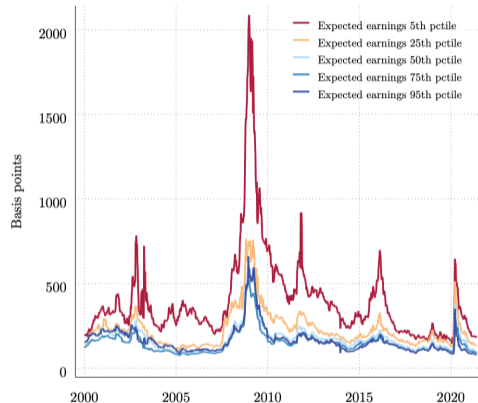
	Mean	P25	Median	P75
No. of bonds per firm/week	38.24	6.00	12.00	23.00
Bond volume (\$ mil)	640.72	250.00	500.00	800.00
Maturity at issue (years)	15.73	9.50	10.03	29.98
Term to maturity (years)	10.49	3.93	7.31	16.43
BB Composite Bond Rating	BBB+	BBB	BBB+	A
OAS spread (bsp)	174.25	85.56	138.24	209.36
Duration (years)	6.91	3.29	5.91	10.09
Coupon rate (pct)	5.18	3.75	5.05	6.62
Bond options (pct)	0.46			

Note: Sample period: 2000/01/07 – 2021/12/17; Number of bond-week observations: 2,274,822; Number of bonds: 7,674; Number of firms: 407. The sample statistics are based on trimmed data.

Heterogeneity in corporate bond spreads across the leverage and earnings distribution



(a) by leverage



(b) by earnings

Figure 1: Credit spreads across the firm distribution [▶ bond rating distribution](#)

Road map: Empirical approach

Step 1: Set up Bayesian VAR model

- ▶ Explore **daily** dynamics in **financial variables** which determine **financial conditions of firms** to identify US monetary policy and global risk shocks

Step 2: Explore firm heterogeneity

- ▶ Assess how firms' funding costs **react differently** depending on the type of borrowing constraint
- ▶ Panel local projections using the shocks identified in Step 1

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Shock Identification

- ▶ **Daily Bayesian VAR model** identified with sign, relative magnitude and narrative restrictions, similar to Brandt et al. (2021)

$$Ay_t = c + \sum_{l=0}^p B^l y_{t-l} + \varepsilon_t$$

- ▶ $y_t \equiv$ short-term and long-term UST yields, equity prices (CAPE), US dollar effective exchange rate, corporate spread
- ▶ Exploit **cross-asset price movements** that capture a wide range of financing costs for firms (Cieslak & Schrimpf 2019; Bobasu et al. 2023)
- ▶ Daily data Jan-1995 to Apr-2022, variables in (log first) differences, 4 lags
- ▶ Estimation follows Arias et al. (2018) and Antolin-Diaz & Rubio-Ramírez (2018)

Why a daily BVAR?

- ▶ **Simultaneous identification** of US monetary policy and global risk shock \Rightarrow consistent approach
- ▶ **Continuous shock series**: incorporates daily dynamics in market pricing of US monetary policy and global risk shock

▶ Model validation: global risk

Shock Identification: Sign, relative magnitude & narrative restrictions

Table 5: Sign restriction identification

	US policy	
Short-term rate	+	
Long-term rate	+*	
CAPE	($>$ foreign MP shock) -	US MP tightening → pushes up yields
Effective FX rate	+	→ depresses equity prices → USD appreciates
Corporate spread		

Note: A * denotes that relative magnitude or narrative restrictions are imposed in addition to the specified sign.
CAPE refers to cyclically-adjusted price to earnings ratio and policy to monetary policy shocks.

Shock Identification: Sign, relative magnitude & narrative restrictions

Table 5: Sign restriction identification

	US policy	US macro risk	
Short-term rate	+		
Long-term rate	+*	+	
CAPE	-	+*	Positive US macro shock
Effective FX rate	+	+	→ supports long-term yields and the USD
Corporate spread		-	→ boosts equity prices
			→ compresses corp. spreads

Note: A * denotes that relative magnitude or narrative restrictions are imposed in addition to the specified sign.
 CAPE refers to cyclically-adjusted price to earnings ratio and policy to monetary policy shocks.

Shock Identification: Sign, relative magnitude & narrative restrictions

Table 5: Sign restriction identification

	US policy	US macro risk	global risk	
Short-term rate	+			
Long-term rate	+*	+	-	
CAPE	-	+*	-*	Global risk shock
Effective FX rate	+	+	+*	→ flight to safety into US safe assets
Corporate spread		-		→ safe haven USD appreciation

(narrative Lehman)
(> foreign macro risk)

Note: A * denotes that relative magnitude or narrative restrictions are imposed in addition to the specified sign.

CAPE refers to cyclically-adjusted price to earnings ratio and policy to monetary policy shocks.

Shock Identification: Sign, relative magnitude & narrative restrictions

Table 5: Sign restriction identification

	US policy	US macro risk	global risk	foreign policy	
Short-term rate	+				
Long-term rate	+*	+	-	+	
CAPE	-	+*	-*	-	Foreign MP tightening → raises US LT yields
Effective FX rate	+	+	+*	-	→ pushes down US equities → USD depreciates
Corporate spread		-			

Note: A * denotes that relative magnitude or narrative restrictions are imposed in addition to the specified sign.

CAPE refers to cyclically-adjusted price to earnings ratio and policy to monetary policy shocks.

Shock Identification: Sign, relative magnitude & narrative restrictions

Foreign macro shock
 → similar to US macro shock
 → USD depreciates

Table 5: Sign restriction identification

	US policy	US macro risk	global risk	foreign policy	foreign macro risk
Short-term rate	+				
Long-term rate	+*	+	-	+	+
CAPE	-	+*	-*	-	+
Effective FX rate	+	+	+*	-	-
Corporate spread		-			

Note: A * denotes that relative magnitude or narrative restrictions are imposed in addition to the specified sign.
 CAPE refers to cyclically-adjusted price to earnings ratio and policy to monetary policy shocks.

The two shocks are large contributors to US financial conditions historically

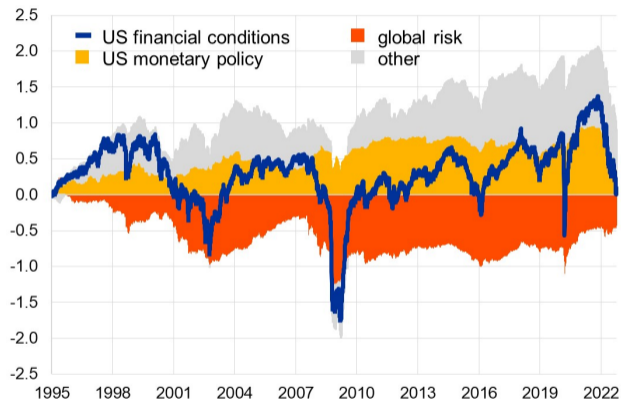


Figure 2: Model-based drivers of US financial conditions (cumulated contributions of shocks to standardized index, rebased to Jan 1995 = 0). US financial condition index based on Arrigoni et al. (2022).

Road map: Empirical approach

Step 1: Set up Bayesian VAR model

- ▶ Explore **daily** dynamics in **financial variables** which determine **financial conditions of firms** to identify US monetary policy and global risk shocks

Step 2: Explore firm heterogeneity

- ▶ Assess how firms financing conditions **react differently** depending on the type of borrowing constraint
- ▶ Panel local projections using the shocks identified in Step 1

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Empirical Approach: Exploiting **bond-** and **firm-**level heterogeneity

Step 2.1:

- ▶ Match equity and corporate bond spread indicators at the firm-level
- ▶ **Decompose corporate spreads** into credit risk component and excess premium (Gilchrist & Zakrajsek, 2012)

Step 2.2:

- ▶ Estimate the **response of corporate funding costs** to global risk and US monetary policy shocks using **panel local projections** à la Jordá (2005)
- ▶ Analyze if **tail of weak/strong firms** (by leverage, interest coverage ratio, earnings) is **more/less sensitive** to shocks

Step 2.1: Decompose corporate spreads into **fundamental risk** and **excess premium** components

- ▶ Following Gilchrist & Zakrajsek (2012), firm j 's option-adjusted spread $s_{j,t}[k]$ on bond k at time t is assumed to be linearly related to

$$s_{j,t}[k] = \underbrace{a_j}_{\text{industry FE}} + \underbrace{\Lambda^j EDF_{j,t}}_{\text{exp. default frequency}} + \underbrace{\Lambda^k \mathbf{X}_{j,t}[k]}_{\text{bond characteristics}} + \underbrace{u_{j,t}[k]}_{\text{pricing error}} \quad (1)$$

- ▶ $EDF_{j,t}$: Summary statistic of firm fundamentals
- ▶ $\mathbf{X}_{j,t}[k]$: $\text{Duration}_{t,j}[k]$, $\text{Age}_j[k]$, $\text{Volume}_j[k]$, $\text{CALL}_j[k]$
- ▶ $u_{j,t}[k]$: pricing error \Rightarrow aggregate to the firm-level:

$$u_{j,t} \equiv \underbrace{EBP_{j,t}}_{\text{excess bond premium}} = \frac{1}{N_j} \sum_k u_{j,t}[k] \quad (2)$$

Step 2.1: Excess bond premium **spikes, heterogeneously** across firms, during **tail events**

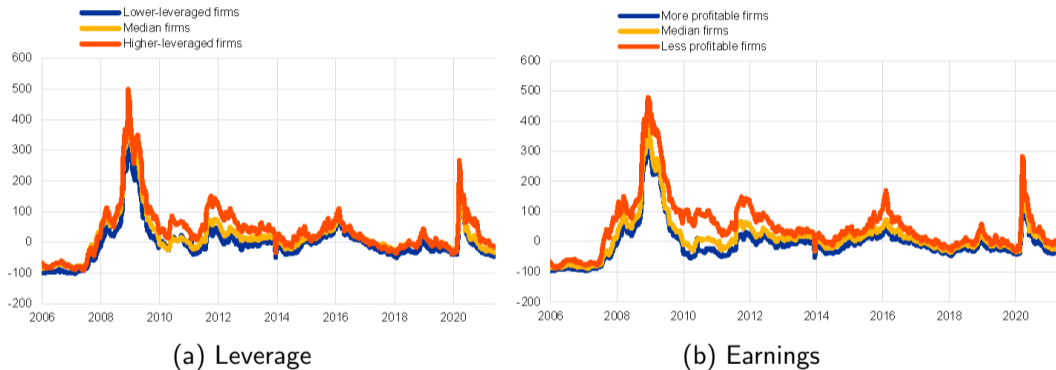


Figure 3: Excess Bond Premium (EBP) across firm percentiles

Step 2.2: Estimate **heterogeneous responses** to global risk and monetary policy shocks

- ▶ Panel local projections: firm-level regressions for firm j in week t (Jan-2000 to Dec-2021)

$$\Delta_h y_{j,t-1} = \beta_h \underbrace{\epsilon_t^i}_{\text{shock}} + \sum_{q \in \{H,L\}} \beta_{h,q} \underbrace{\epsilon_t^i \times \mathbb{1}_{j,q,t}}_{\text{tail firm}} + \underbrace{\phi_{j,h}(L) X_{j,t-1}}_{\text{controls}} + \epsilon_{j,t+h} \quad \text{for } h = 1, \dots, H \quad (3)$$

- ▶ $y_{j,t}$: credit spread, EBP, EDF, equity price (CDS spread)
- ▶ ϵ_t^i : monetary policy shock ϵ_t^m , global risk shock ϵ_t^r
- ▶ $\mathbb{1}_{j,q,t}$: dummy variable for $q = \{20^{\text{th}}, 80^{\text{th}}\}$ pct. of weak/strong firms by **leverage**, **interest coverage ratio**, **expected earnings**
- ▶ $\phi_{j,h}(L) X_t$: 4 lags of VIX, CESI, GFC dummy, Covid dummy, industry FE
- ▶ **Weak/strong firms**: percentiles of firms defined according to their balance sheet characteristics (every 2 years, starting 2005)

Results: impact of **monetary policy** shock ϵ_t^m on all firms

Table 7: Sensitivity of asset prices of tail firms upon impact of shocks.

	(1) Credit spread	(2) EBP	(3) Default risk	(4) ln(Equity Price)
ϵ_t^m	7.395***	5.889**	0.028*	-0.035***
Industry FE	YES	YES	YES	YES
Observations	222,060	219,513	220,710	220,964

Note: Estimates at horizon $h = 0$, i.e. upon impact of the identified monetary policy shock ϵ_t^m . Indicator variables for leverage (LEV), interest coverage ratio (ICR), and expected earnings (EPSE) are computed based on the tails of firms (20th and 80th percentiles). Shocks are calibrated to a 10 bsp increase in the US 10y yield. SEs are clustered along the time and industry dimension.

Results: heterogeneous impact of **monetary policy** shock ϵ_t^m by **leverage**

Table 7: Sensitivity of asset prices of tail firms upon impact of shocks.

	(1) Credit spread	(2) EBP	(3) Default risk	(4) ln(Equity Price)
ϵ_t^m	7.395***	5.889**	0.028*	-0.035***
ϵ_t^m	7.261***	6.220***	0.020**	-0.035***
LowLEV $\times \epsilon_t^m$	-1.167	-1.277	-0.003	0.000
HighLEV $\times \epsilon_t^m$	1.842	-0.537	0.045	-0.002
Industry FE	YES	YES	YES	YES
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Results: heterogeneous impact of **monetary policy shock** ϵ_t^m by **interest coverage ratio**

Table 7: Sensitivity of asset prices of tail firms upon impact of shocks.

	(1) Credit spread	(2) EBP	(3) Default risk	(4) ln(Equity Price)
ϵ_t^m	7.395***	5.889**	0.028*	-0.035***
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LowLEV $\times \epsilon_t^m$	-1.167	-1.277	-0.003	0.000
HighLEV $\times \epsilon_t^m$	1.842	-0.537	0.045	-0.002
ϵ_t^m	7.001***	6.145***	0.016**	-0.034***
LowICR $\times \epsilon_t^m$	2.670	-1.178	0.073	-0.007*
HighICR $\times \epsilon_t^m$	-0.358	-0.126	-0.005**	0.001
Industry FE	YES	YES	YES	YES
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Results: heterogeneous impact of **monetary policy** shock ϵ_t^m by **earnings**

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	(1) Credit spread	(2) EBP	(3) Default risk	(4) ln(Equity Price)
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LowICR $\times \epsilon_t^m$	2.670	-1.178	0.073	-0.007*
HighICR $\times \epsilon_t^m$	-0.358	-0.126	-0.005**	0.001
ϵ_t^m	7.140***	5.938***	0.021**	-0.034***
LowEPSE $\times \epsilon_t^m$	1.861	-0.652	0.048*	-0.004*
HighEPSE $\times \epsilon_t^m$	0.034	0.468	-0.006	-0.001*
Industry FE	YES	YES	YES	YES
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Results: heterogeneous impact of global risk shock ϵ_t^r

Table 9: Sensitivity of asset prices of tail firms upon impact of shocks

	(1) Spread	(2) EBP	(3) Default risk	(4) ln(Equity Price)
ϵ_t^r	18.628***	15.472***	0.056*	-0.069***
ϵ_t^r	17.502***	15.406***	0.039**	-0.068***
LowLEV $\times \epsilon_t^r$	-4.942	-5.107**	-0.006	-0.003
HighLEV $\times \epsilon_t^r$	10.456	5.000*	0.099	-0.002
ϵ_t^r	15.858***	14.366***	0.027**	-0.065***
LowICR $\times \epsilon_t^r$	18.773**	9.504**	0.176*	-0.022*
HighICR $\times \epsilon_t^r$	-3.616**	-2.995*	-0.011**	0.002
ϵ_t^r	16.439***	14.232***	0.037**	-0.065***
LowEPSE $\times \epsilon_t^r$	15.194***	8.416***	0.126**	-0.019***
HighEPSE $\times \epsilon_t^r$	-1.695	-0.788	-0.013	-0.005
Industry FE	YES	YES	YES	YES
Observations	222,060	219,513	220,710	220,964

Note: Estimates at horizon $h = 0$, i.e. upon impact of the identified global risk shock ϵ_t^r . Indicator variables for leverage (LEV), interest coverage ratio (ICR), and expected earnings (EPSE) are computed based on the tails of firms (20th and 80th percentiles). Shocks are calibrated to a 10 bps decrease in the US 10y yield. SEs are clustered along the time and industry dimension.

Results: heterogeneous impact of global risk shock ϵ_t^r

Table 11: Sensitivity of asset prices of tail firms upon impact of shocks

	(1) Spread	(2) EBP	(3) Default risk	(4) ln(Equity Price)
ϵ_t^r	18.628***	15.472***	0.056*	-0.069***
ϵ_t^r	17.502***	15.406***	0.039**	-0.068***
LowLEV $\times \epsilon_t^r$	-4.942	-5.107**	-0.006	-0.003
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Monetary policy and global risk shocks...

- ▶ Global risk shocks have stronger and more **heterogeneous** effects on corporate funding costs which depend on firms' position within the **earnings distribution**

... and the earnings-based borrowing constraint transmission channel

- ▶ Responses of firms' financing conditions are **more muted** for higher levered firms ...
- ▶ ... but **more pronounced** for less profitable firms

Investors price in **more risks** across the **earnings** distribution of firms

- ▶ No difference in corporate bond spreads between median and tail firm by *leverage* since GFC
- ▶ *Declining relevance of leverage* as an indicator of financial constraints

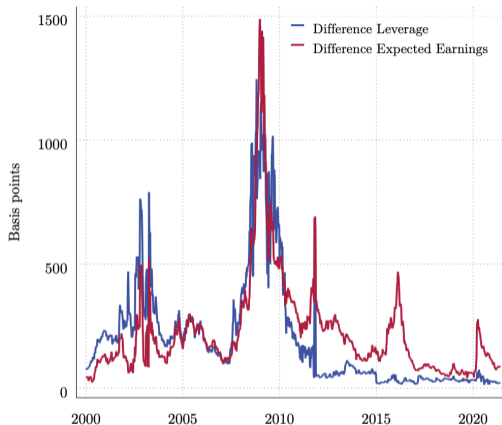


Figure 4: Difference in credit spreads between the tail of weakest firms and the median firm computed based on leverage and earnings

Results: impact of shock remains fairly **persistent** in the **non-fundamental bond spread component**

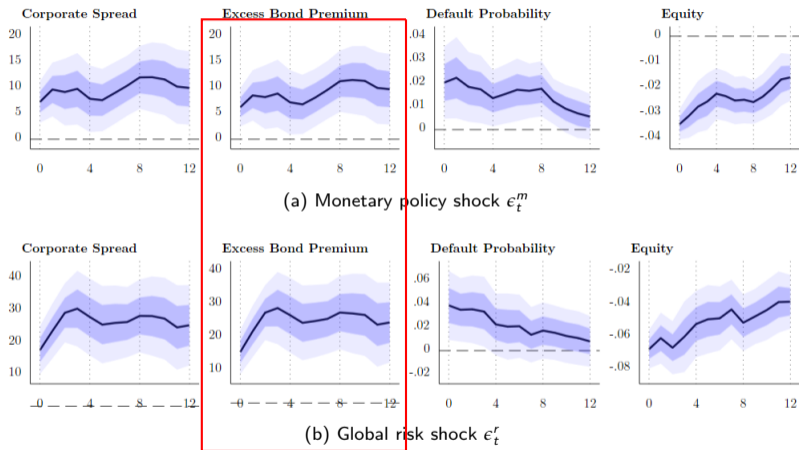


Figure 5: Cumulative responses to identified shocks equivalent to a 10 bps increase (decrease) in US 10y yield

Robustness of shocks:

- ▶ Additional robustness checks on sign restrictions in BVAR

Robustness of firm-level results:

- ▶ Sample period 2005-2021 to exclude earlier observations with fewer bonds outstanding
- ▶ Lagged dependent variables to account for autocorrelation in asset prices
- ▶ Week + week-industry FE → macro variables and time-varying industry-exposure
- ▶ Alternative measures of firm profitability [▶ Results earnings](#)
- ▶ Alternative definition of tails of firms (15th, 85th pct)
- ▶ Spread decomposition with log-spread, firm fundamentals as controls, only senior unsecured bonds
- ▶ Bond-level regressions

Conclusion

This paper:

- ▶ Proposes an integrated framework to **identify global risk and monetary policy shocks**
- ▶ Explores to which extent global risk and monetary policy shocks affect corporate financing conditions **heterogeneously** depending on **borrowing constraint**

Key Takeaway:

⇒ Global risk shocks have stronger and more heterogeneous effects on corporate financing conditions which depend on firms' position within the earnings distribution: the **earnings-based borrowing constraint transmission channel**

Policy implications:

- ▶ **Type of borrowing constraint** essential for transmission of shocks
- ▶ Enhancing the understanding of the type of borrowing constraint prevalent in other countries than the US

Appendix

Bond filters in Sample Selection

Download filters:

- (i) Active and matured bonds issued between 1 January 2000 and 28 May 2021
- (ii) denominated in USD
- (iii) issued by non-financial firms included in the S&P 500 between 2000 and 2021
⇒ download 13,233 bonds ⇒ cleaning 10,679 bonds

Additional filters/trimming:

- (iv) $1 \text{ mil} \leq \text{volume} \leq 5\text{bn}$
- (v) $1 \text{ year} \leq \text{term-to-maturity} \leq 30 \text{ years}$
- (vi) $-500 \leq \text{OAS} \leq 4,500$
- (vii) Drop OAS if illiquid > 26 weeks in a row
- (viii) Drop bond if there exist < 26 consecutive bond-week obs.

[◀ Go back](#)

Bond spread trimming

	No. of remaining bond-week obs.
No. OAS observations before trimming	2,272,931
(i) Drop volume < 1mil and volume > 5bn	2,262,348
(ii) Drop term-to-maturity < 1 year and > 30 years	2,194,404
(iii) Drop OAS < -500 and OAS > 4,500	2,186,060
(iv) Drop OAS if illiquid > 26 weeks in a row	2,186,060
(v) Drop bond if there exist < 26 consecutive bond-week obs.	2,185,478
No. OAS observations after trimming	2,185,478

[◀ Go back](#)

Heterogeneity in corporate bond spreads across bond rating categories

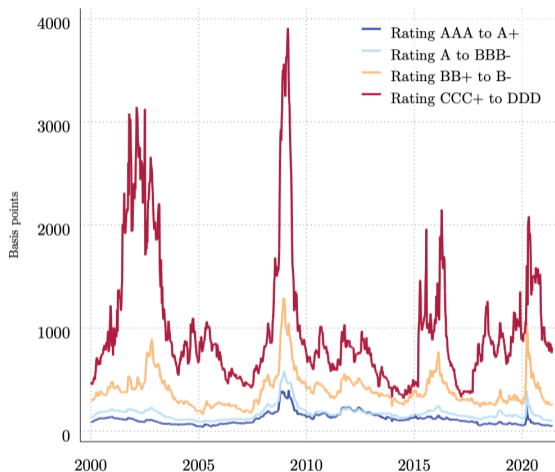


Figure 6: Average credit spread across bond ratings. Ratings are based on the Bloomberg composite bond rating. [◀ Go back](#)

BVAR model validation: Global risk shock correlated with VIX around major events

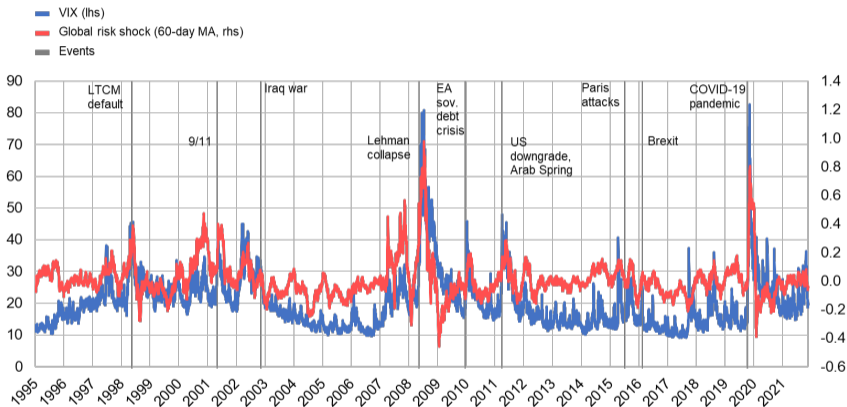


Figure 7: Comovement of global risk shock with the VIX and selected narrative events

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BVAR model validation: Global risk shock correlated with other indicators

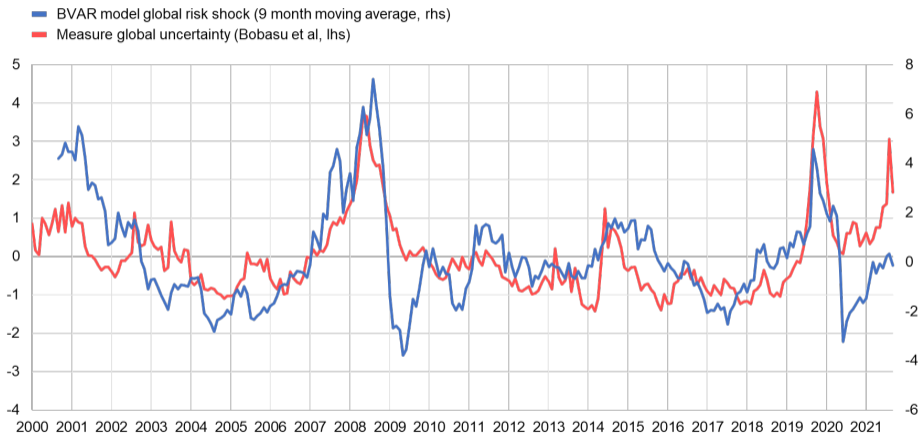


Figure 8: Comovement of global risk shock with global uncertainty measure by Bobasu et al. 2023

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Excess bond premium estimation results

Table 13: Spread decomposition following Gilchrist & Zakrajsek (2012) with a level-dependent variable, $S_j[k]$, i.e the spread of bond k of firm j .

	(1) Est.	SE	(2) Est.	SE
EDF _{<i>j,t</i>}	59.396***	9.207	54.617***	17.333
Duration _{<i>j,t</i>} [<i>k</i>]	3.316***	0.448	4.680***	0.681
Coupon _{<i>j</i>} [<i>k</i>]	27.350***	2.312	20.905***	2.901
Age _{<i>j,t</i>} [<i>k</i>]	-1.986***	0.565	-0.684	0.601
Volume _{<i>j</i>} [<i>k</i>]	-5.979	4.838	-9.527	6.120
CALL _{<i>j</i>} [<i>k</i>]	3.599	4.657	-34.402**	13.505
EDF _{<i>j,t</i>} × CALL _{<i>j</i>} [<i>k</i>]			7.258	16.264
Duration _{<i>j,t</i>} [<i>k</i>] × CALL _{<i>j</i>} [<i>k</i>]			-2.257***	0.674
Coupon _{<i>j</i>} [<i>k</i>] × CALL _{<i>j</i>} [<i>k</i>]			12.778***	2.881
Age _{<i>j,t</i>} [<i>k</i>] × CALL _{<i>j</i>} [<i>k</i>]			-4.463***	0.871
Volume _{<i>j</i>} [<i>k</i>] × CALL _{<i>j</i>} [<i>k</i>]			7.611	5.961
Industry FE	YES		YES	
Observations	2,207,373		2,207,373	
Adjusted R ²	0.424		0.430	

BVAR **robustness**: Shocks remain highly correlated across various tests

Table 15: Robustness BVAR: correlation between US MP and global risk shock of the benchmark BVAR model and shocks identified in alternative versions

	US mon policy shock	global risk shock
Test 1: no rel. restriction on US NEER	0.9993	0.9910
Test 2: corp. spread not restricted after US macro	0.9944	0.9865
Test 3: US policy and global risk shock only	0.9634	0.9671
Test 4: 1 lag	0.9972	0.9792
Test 5: 2 lags	0.9991	0.9817
Test 6: 3 lags	0.9988	0.9838
Test 7: 5 lags	0.9979	0.9834
Test 8: 6 lags	0.9973	0.9838

Notes: Test 1 does not impose the relative magnitude restriction that a global risk shock should have larger effects on the US nominal effective exchange rate than a foreign macro risk shock; test 2 leaves out the sign restriction on corporate bond spreads following a US macro risk shock; test 3 only identifies the US monetary policy and the global risk shock in the BVAR model (using the same restrictions for the two shocks as in Table 5), leaving the other shocks unidentified; test 4-8 test the structural shock correlations for different lag lengths of the endogenous variables in the BVAR. [◀ Go back](#)

How our estimation differs from Gilchrist & Zakrajsek (2012)

- ▶ **Firm sample:** We use S&P500 firms, GZ use a broader set of firms
- ▶ **Time period:** Our time period is 2000-2021, GZ 1973-2010 (in the initial paper);
- ▶ **Frequency:** GZ use month-end credit spreads and estimate at a monthly frequency. We use weekly frequency.
- ▶ **Updating:** unclear how often GZ rebalance their sample of bonds used to compute the EBP. We include bonds issued before 28 May 2021.
- ▶ **Measure of credit risk:** we use Moody's EDF, GZ own estimates of distance to default
- ▶ **Measure of spreads:** we use option adjusted spreads (OAS) in bps, GZ use log-spreads (and log explanatory variables) and construct themselves the so-called "GZ spread" - by constructing a synthetic risk-free security that mimics exactly the cash flows of the corresponding corporate debt instrument (following Gürkaynak, Sack and Wright, 2007)
- ▶ **Magnitude:** GZ spreads range from 5 bps to 3500 bps. Our spreads range from -500 bps to 4500 bps due to the nature of OAS spreads.

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How our estimation differs from Gilchrist & Zakrajsek (2012)

Differences in the regression model:

- ▶ GZ uses bond characteristics and, additionally, interaction terms of callable bond dummy with bond characteristics; we use only bond characteristics as the option adjusted spreads correct for pricing effects of embedded options.
- ▶ GZ control for call options and liquidity premia by interacting regressors with a CALL dummy and with the slope/level/curvature of the yield curve. We account for this through our OAS spread measure
- ▶ GZ include firm-level ratings fixed effects (S&P rating). We use for some specifications bond-level ratings fixed effects (Bloomberg composite rating), although not in the baseline.
- ▶ GZ use industry fixed effects based on three-digit NAICs industry codes, which is very granular. We use industry fixed effects based on Bloomberg industry classifications which is much less granular.

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Robustness: results remain robust to different measures of earnings

Table 17: Earnings-based tails of firms: estimated **monetary policy shock** impact on corporate spreads (1), their predicted and excess bond premium components (2-3), CDS spreads (4), equity prices (5), and default probabilities (6).

	(1)	(2)	(3)	(4)	(5)	(6)
	ΔSpread	$\Delta\widehat{\text{Spread}}$	ΔEBP	ΔCDS	$\Delta\ln(\text{PI})$	ΔEDF
<i>Panel (a): Monetary policy shock</i>						
s_t^m	7.007***	1.048*	5.982***	2.790***	0.019**	-0.035***
LowROE $\times s_t^m$	2.026	3.227**	-0.988	2.531*	0.057**	-0.004*
HighROE $\times s_t^m$	0.351	-0.037	0.367	-0.427	0.000	0.002*
s_t^m	6.838***	0.782**	6.018***	2.770***	0.014**	-0.034***
LowPE $\times s_t^m$	4.269*	4.455**	0.018	2.281	0.076**	-0.009**
HighPE $\times s_t^m$	0.680	1.348	-0.549	0.662	0.022	-0.002*
s_t^m	7.051***	1.097*	6.036***	2.767***	0.020*	-0.034***
LowEPSEgrowth $\times s_t^m$	3.275	3.080**	-0.084	3.196*	0.052***	-0.007*
HighEPSEgrowth $\times s_t^m$	1.229	2.412*	-1.385	1.075	0.044*	-0.008***
s_t^m	6.900***	1.271*	5.695***	2.852***	0.024*	-0.034***
LowEPSTgrowth $\times s_t^m$	5.016*	1.915**	2.947	1.972*	0.032**	-0.005*
HighEPSTgrowth $\times s_t^m$	0.357	1.086*	-0.818	0.886	0.016*	-0.006***

Robustness: results remain robust to different measures of earnings

Table 19: Earnings-based tails of firms: estimated **global risk shock** impact on corporate spreads (1), their predicted and excess bond premium components (2-3), CDS spreads (4), equity prices (5), and default probabilities (6).

	(1) ΔSpread	(2) ΔSpread	(3) ΔEBP	(4) ΔCDS	(5) $\Delta\ln(\text{PI})$	(6) ΔEDF
<i>Panel (b): Global risk shock</i>						
s_t^r	16.414***	1.979*	14.521***	5.686***	0.035*	-0.068***
LowROE $\times s_t^r$	15.342***	7.769***	8.024**	7.423**	0.135***	-0.015***
HighROE $\times s_t^r$	-1.553	-0.028	-1.568	-1.543*	0.003	0.010*
s_t^r	14.898***	1.294**	13.516***	5.329***	0.023**	-0.064***
LowPE $\times s_t^r$	25.218**	11.454***	14.385**	8.297**	0.194***	-0.036***
HighPE $\times s_t^r$	5.048**	2.856	2.364	2.143	0.048	-0.006
s_t^r	16.652***	2.134*	14.653***	5.605***	0.038*	-0.065***
LowEPSEgrowth $\times s_t^m$	18.337***	7.637***	10.039**	8.501***	0.128***	-0.025***
HighEPSEgrowth $\times s_t^m$	6.455	5.267*	0.889	3.607	0.093*	-0.020**
s_t^r	15.756***	2.447*	13.431***	5.810***	0.044*	-0.065***
LowEPSTgrowth $\times s_t^m$	21.520***	5.583***	15.593**	6.133***	0.093***	-0.028***
HighEPSTgrowth $\times s_t^m$	9.726*	2.656	6.816*	2.673	0.041	-0.018***

Results: shock responses across high/low leverage firms

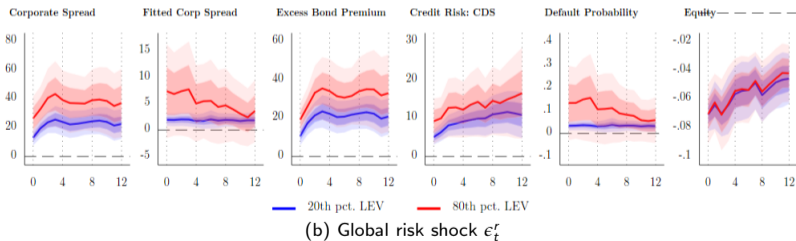
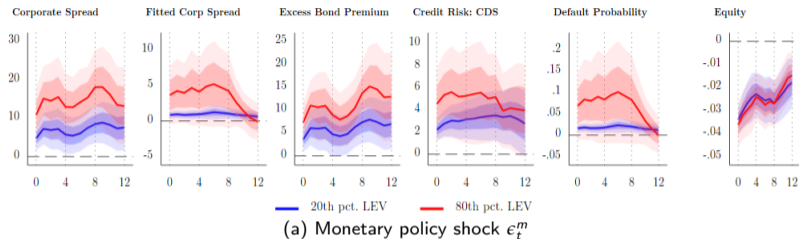


Figure 9: Cumulative responses for weak/strong firms by leverage (LEV)

Results: shock responses across high/low earnings firms

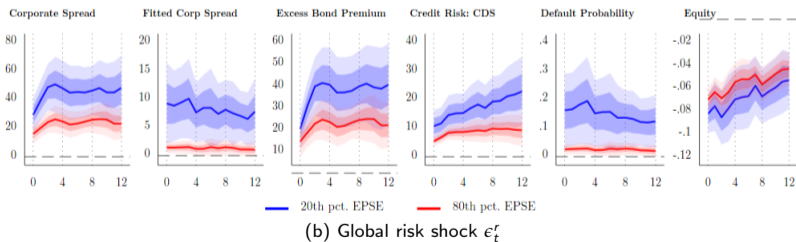
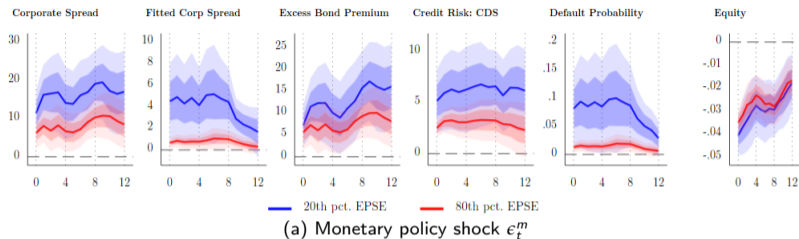


Figure 10: Cumulative responses for weak/strong firms by earnings (EPSE)