

High-Frequency Cross-Sectional Identification of Military News Shocks

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OUTLINE

INTRODUCTION

HFXS FRAMEWORK & IDENTIFICATION

EMPIRICAL RESULTS

APPLICATION: US GDP XS-MULTIPLIERS

CONCLUSION

MOTIVATION

- Economists are interested in the effects of defense spending because it provides:
 - **Exogenous variation** in government spending (causal inference)
 - Multiplier estimates of **military build-ups** (policy relevance)
- Identification challenges:
 - Effects of government spending are anticipated (Ramey, [2011](#))
 - Measuring expectations is tricky

→ Research Question:

*“How does the economy respond to **anticipated** changes in defense spending?”*

&

*“How can we effectively measure **expectations** about future defense spending?”*

LITERATURE REVIEW

- Macro shocks need to be unanticipated (Ramey, 2016)
 - **Fiscal foresight**
Mertens and Ravn (2010) (gov. spending), Leeper et al. (2013) (taxes)
 - **Measurement delays** (Briganti et al., 2025)
 - Non-invertibility of fiscal shocks
- **Unanticipated measures** of government spending shocks:
 - **VAR Restrictions:** Blanchard and Perotti (2002) (short-run restrictions), Ben Zeev and Pappa (2017) (medium run restrictions (Barsky and Sims, 2011)), Ascari et al. (2023) (sign restrictions (Mountford and Uhlig, 2009))
 - **Narrative Instruments:**
Ramey and Shapiro (1998) (war dates), Ramey (2011)+Ramey and Zubairy (2018) (def. news)
 - **Bartik Instruments:** Nakamura and Steinsson (2014), Dupor and Guerrero (2017), Demyanyk et al. (2019), Auerbach et al. (2020), Muratori et al. (2023), Barattieri et al. (2023), Auerbach et al. (2024).
 - **Stock-Price-Based Instruments:**
Fisher and Peters (2010), McClure and Yding (2024)(narrative+high frequency).
 - **High Frequency Instruments:** Bandeira et al. (2025) (Brazil Deficit), Wiegand (2025)+Gomez-Cram et al. (2025)+Hazell and Hobler (2025)+Bi et al. (2025)(US Deficits)

CONTRIBUTION: HFXS IDENTIFICATION

- We introduce a novel method to quantify expectations of future military spending
 - I. Identify **HF**-fiscal events using narrative analysis augmented with LLM searches
 - II. Leverage stock price **XS**-variation to quantify expected shifts in defense expenditure
- Benefits:
 - I. Model consistent methodology grounded in asset pricing theory
 - II. Self-validating: it *estimates* and allows statistical validation of each event (*testing*)
 - III. Generalizable to contexts where units are heterogeneously impacted by aggregate shocks
 - IV. Parsimony and objectivity (i.e., minimizes subjectivity in narrative approaches)
- **Contributions:**
 - I. Novel LLM-augmented narrative analysis: key fiscal events (2001-2023)
 - II. Novel military news shocks series (2001-2023)
 - III. Novel XS-multiplier estimates (MSA / 2001-2023)

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A SIMPLE MODEL OF STOCK PRICES

- Profits $D_{i,t}$ of firm i at time t

$$D_{i,t} := (1 - \tau_t) \cdot \underbrace{(V_{i,t} + G_{i,t})}_{\text{TOTAL SALES}} \cdot \left(1 - \frac{1}{\mu_i}\right)$$

- $V_{i,t}$ is private sales
- $G_{i,t}$ is government sales
- μ_i is the markup and τ_t is a corporate tax

- Gordon (1959):

$$P_{i,t} := \sum_{h=0}^{\infty} \frac{D_{i,t+h}^e}{\prod_{\tau=0}^h (1 + i_{t,t+\tau}^e)}$$

- $P_{i,t}$ is the stock price of firm i
- $i_{t,t+\tau}^e$ is the expected $(t + \tau)$ -period ahead interest rate at time t

GOVERNMENT SALES: LINKED TO STOCK PRICES

- Under (i)-(ii):
 - I. Expected profits are proxied by current profits
 - II. Expectations hypothesis of the term structure holds

$$P_{i,t} = \frac{D_{i,t}}{1 - \frac{1}{1+i_t}} = \frac{1+i_t}{i_t} \cdot \underbrace{(1 - \tau_t) \cdot (V_{i,t} + G_{i,t})}_{D_{i,t}} \cdot \left(1 - \frac{1}{\mu_i}\right) \quad (1)$$

→ THE STOCK PRICES ARE PROPORTIONAL TO GOVERNMENT SALES

STOCK RETURNS ARE PROPORTIONAL TO GOVERNMENT SALES

Let us focus on cross-section (XS) of contractors i

- Denote *Reliance on DoD* by $\lambda_i := \frac{G_i}{G_i + V_i}$; define $G_i := \theta_i \cdot G$
- Log-Differentiate (1) around a **HF fiscal event**:

$$\underbrace{d \log P_i}_{\text{STOCK RETURN}} = \alpha + \underbrace{\lambda_i}_{\text{RELIANCE}} \cdot \left(\underbrace{d \log G^e + d \log \theta_i^e - d \log V_i^e}_{\text{SHOCK}} \right) + \varepsilon_i \quad (2)$$

- α : time FEs (e.g., \mathbb{E} change in corporate taxes);
- ε_i : firm-specific FEs (e.g., \mathbb{E} change in markups)

EXTRACT NEWS SHOCKS FROM STOCK RETURNS

HFXS IDENTIFICATION: THEOREM

GENERALIZATION

Under weak **Assumptions**, regressing stock returns ($d \log P_i$) on reliance on DoD contracts (λ_i):

$$d \log P_i = \alpha + \gamma \cdot \lambda_i + e_i \quad (3)$$

yields

$$\hat{\gamma}_{\text{OLS}} \xrightarrow{P} d \log G^e$$

That is, $\hat{\gamma}_{\text{OLS}}$ consistently estimates expected changes in defense spending ($d \log G^e$)

“If Lockheed’s reliance on DoD (λ_i) is 71% and Boeing’s is 30%, a positive shock will affect Lockheed’s price more, mirroring its larger profit potential.”

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EVENTS THAT SHIFTED US EXPECTED MILITARY SPENDING (2000-2023)

Date	Sign	Description of the Event
11 September 2001	+	9/11 terrorist attacks + ensuing invasion of Afghanistan in October 2001
20 March 2003	+	U.S.-led invasion of Iraq opens a second major war
10 January 2007	+	President Bush's Iraq "Surge" address
4 November 2008	-	Barack Obama elected U.S. president after campaigning to end the Iraq War
2 August 2011	-	Budget Control Act of 2011 signed amid debt-ceiling crisis
1 March 2013	-	U.S. Government Sequestration takes effect after Congress fails to agree on deficit reductions
18 March 2014	+	Russia's illegal annexation of Crimea
22 September 2014	+	Extremist group ISIS seizes large parts of Iraq & Syria
8 November 2016	+	Trump wins 2016 U.S. Elections campaigning on military build-up
9 February 2018	+	Bipartisan Budget Act of 2018 lifts strict BCA spending caps for FY 2018–19
2 August 2019	+	Bipartisan Budget Act of 2019 raises defense spending caps + ends sequestration-era limits
24 February 2022	+	Russia invades Ukraine

DEFENSE CONTRACTORS DATA

- Annual official Top100 Report (available from 1958)
→ 430 Top100 Contractors from FY2001
- **Three conditions:**
 - I. **Publicly Traded** (NYSE or NASDAQ)
→ 57 contractors
 - II. **Salience**: investors associate contractors to “defense”
→ Appear at least four times in Top100 report (e.g., rules out **Moderna**)
 - III. **Relevance**: stock price non-negligibly affected by gov. contracts
→ Median **reliance** $\geq 1\%$ (e.g., rules out **BP**):

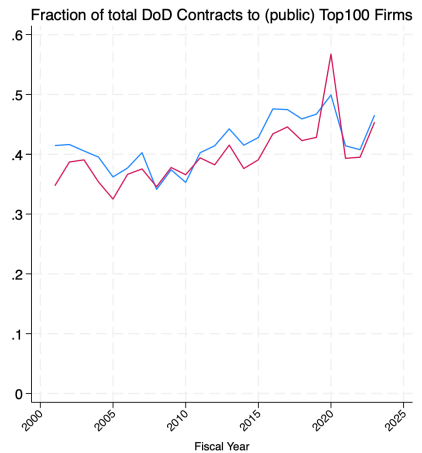
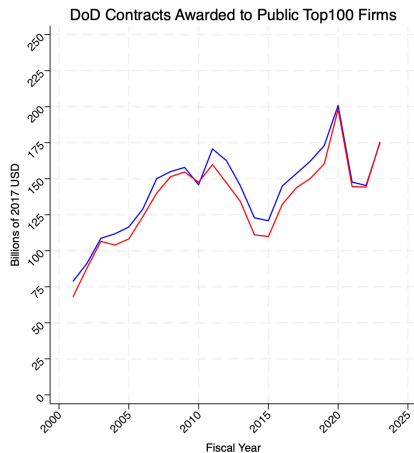
$$\text{Median}(\lambda_{i,t}) \geq 1\%, \quad \lambda_{i,t} := \frac{\text{DoD Contracts}_{i,t}}{\text{Tot. Sales}_{i,t}}$$

→ 33 Contractors meet conditions I-III

- Median reliance is 20%. Interquartile range is [3.7%,39.9%]

Descriptive Stats

33 COMPANIES = 40% TOTAL DoD PROCUREMENT SPENDING!



MODEL-IMPLIED XS-REGRESSION

- We have:
 - Set of **Narrative Dates** \mathcal{T}
 - Set of 33 **publicly traded, salient** and **relevant** defense contractors

→ Implement HFXS Regressions

- For each date $\tau \in \mathcal{T}$ estimate empirical analog of Equation (3):

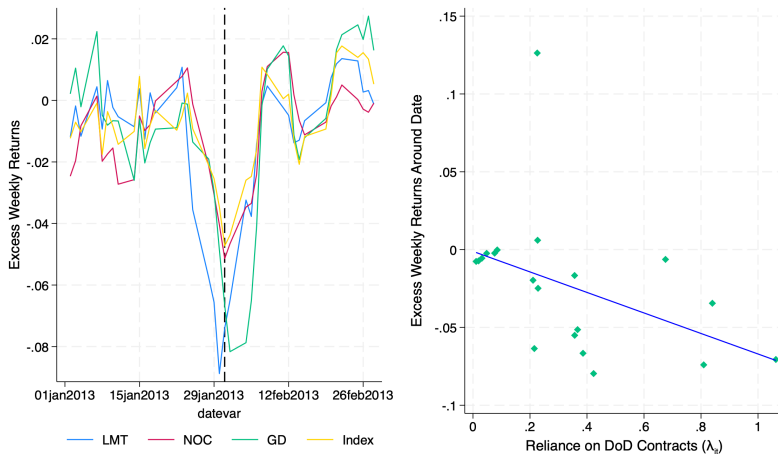
$$\underbrace{v_{i|t=\tau}}_{\approx d \log P_{i,t}} = \alpha + \gamma_{t=\tau} \cdot \underbrace{\lambda_{i|t=\tau}}_{\text{RELIANCE}} + \epsilon_i \quad \forall \tau \in \mathcal{T}, \forall i \in \mathcal{I}_\tau, \quad (4)$$

- $v_{i|t=\tau}$: weekly excess returns
 - Constructed using **Fama-French 3 factors model**
 - Frequency: five trading days
- $\lambda_{i|t=\tau}$: reliance on DoD purchases in the quarter of the event

- We have:
 - Set of **Narrative Dates** \mathcal{T}
 - Set of 33 **publicly traded, salient** and **relevant** defense contractors

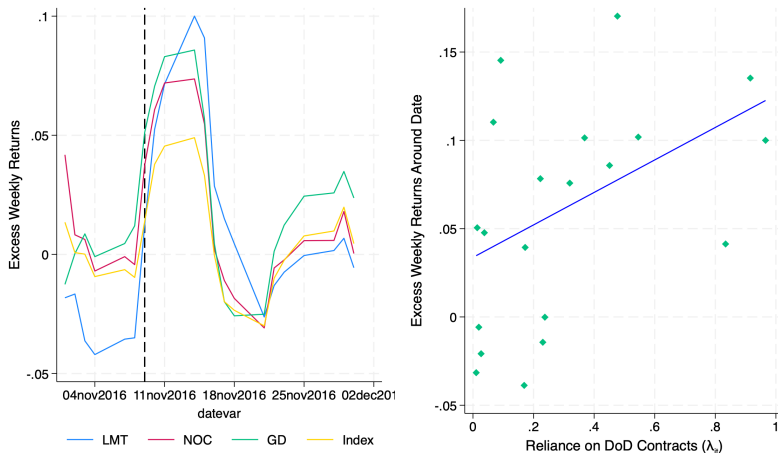
→ Implement HFXS Regressions

EXAMPLE 1: BUDGET SEQUESTRATIONS (2013Q1)



→ **Estimated Slope ($\hat{\gamma}$): -0.066 (0.015)**

EXAMPLE 2: TRUMP ELECTION (2016Q4)



→ **Estimated Slope ($\hat{\gamma}$): +0.092 (0.024)**

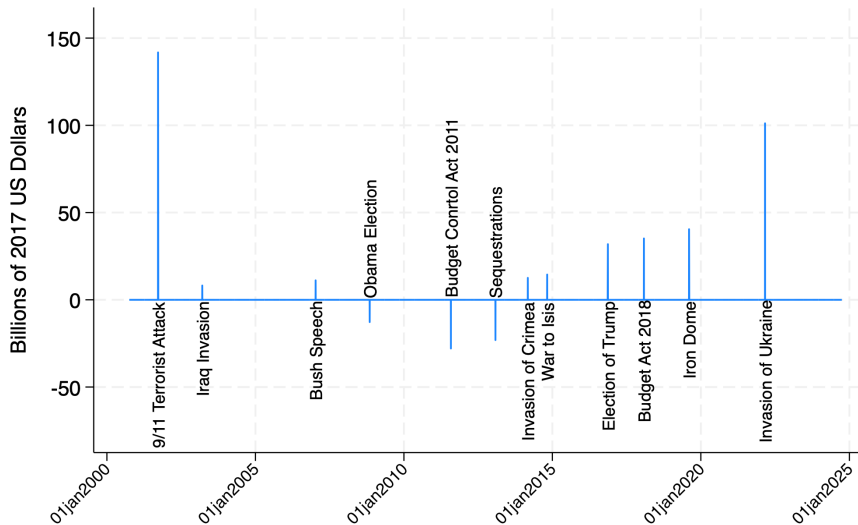
ESTIMATES OF HFXS MILITARY NEWS SHOCKS

<i>Event</i>	<i>Shock Trading Date</i>	<i>Expected Sign</i>	$d \log G_t (\gamma_{t=\tau})$	<i>pvalue</i>	<i>N</i>	<i>Defense Index</i>
<i>9/11 Terrorist Attack</i>	<i>September 21, 2001</i>	+	0.629 (0.133)	0.000	14	+5.2%
<i>Invasion of Iraq</i>	<i>March 19, 2003</i>	+	0.029 (0.035)	0.406	20	+ 6.4%
<i>Bush Speech on Iraq</i>	<i>January 11, 2007</i>	+	0.028 (0.017)	0.117	20	+3.1%
<i>Obama Election</i>	<i>November 6, 2008</i>	-	-0.031 (0.030)	0.327	18	-2.3%
<i>Budget Control Act 2011</i>	<i>August 2, 2011</i>	-	-0.065 (0.019)	0.002	23	-3.1%
<u>Sequestrations</u>	<i>January 31, 2013</i>	-	-0.066 (0.015)	0.000	21	-4.7%
<i>Russia's Invasion of Crimea</i>	<i>March 5, 2014</i>	+	0.038 (0.021)	0.086	21	+1.5%
<i>War to Isis</i>	<i>29 October 2014</i>	+	0.047 (0.024)	0.065	23	+3.3%
<u>Trump Election</u>	<i>November 14 2016</i>	+	0.092 (0.043)	0.042	23	+4.9%
<i>Bipartisan Budget Act 2018</i>	<i>January 31 2018</i>	+	0.091 (0.038)	0.024	23	+5.8%
<i>Bipartisan Budget Act 2019 + Iron Dome</i>	<i>9 August, 2019</i>	+	0.101 (0.028)	0.002	23	+3.7%
<i>Invasion of Ukraine</i>	<i>March 1, 2022</i>	+	0.273 (0.041)	0.000	23	+10.4%

Notes: Robust standard errors in parentheses. Last column (Defense Index), refers to the excess weekly returns of the Defense Index. Interquartile range of excess weekly returns of Defense Index is [-1.0%,+1.0%], 10th and 90th percentiles are -2.2% amd +2.1%.

HFXS MILITARY NEWS SHOCK SERIES

RZ18 COMPARISON



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MILITARY NEWS SHOCKS ARE (REGIONALLY) SALIENT!


Pasadena Star-News

News Local News Sports Things to Do Obituaries Opinion The T.E.A.

TRENDING: Broken building crumpled with family Man accused in kidnapping Nepher killer sentenced Big Bear Lake before/after Michelle's 'new discoverer'

LOCAL NEWS

Good news about a Trump economy? California's defense industry could benefit



MOST POPULAR

- Retired news anchor Chuck Henry's Pasadena home finds a buyer, fast
- California adopts one of nation's highest fixed-utility fees
- Alleged motive in Mexico snuff killings: 'They had a lot of money, devices and the pickup'
- Chinese food rises to another level at this San Gabriel restaurant
- An entire California town is for sale – again. This time for \$6.6M
- Kaiser Permanente's inaugural Pasadena medical school cohort set to graduate
- Here are the ultraprocessed foods you must need to avoid, according to a 30-year study
- Art Del Rey, an Arcadia pioneer in Spanish language barbershop quartet singing, has died
- USC faculty censures university president, provost over Gaza protest responses
- USC graduates relieved by smooth sailing on Day One of 100-plus ceremonies

Southern California is expected to benefit from a boost in defense spending under the Trump administration. Monrovia-based AeroSpaceport Inc. makes a variety of unmanned drones that are used by the U.S. military. One of the company's drones is shown here as a author prepares to test it in San Valley. (Andy Hochstadt/Los Angeles Daily News)

Full Text | Newspaper

THE NATION; Local defense firms expect boost; Trump's budget bodes well for California's aerospace and military contractors.

Herwig, W. J. [Los Angeles Times](#): Los Angeles, Calif. 23 Mar 2017: A-6.

[Preview publication details](#)

Full text

Details

Full Text

Southern California's defense industry, long the epicenter for high-flying aerospace technology and advanced weapons for the military, could get a major boost under President Trump's proposed new budget.

Los Angeles Times

BUSINESS

Southern California aerospace and defense contractors expecting boost from Trump budget



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KCAL NEWS December 6, 2016 3:28 AM PST | KCAL News

[f](#) [x](#) [v](#)

Los Angeles

LOS ANGELES (CBSLA.com) – The presidential election has thrown California's economic future into turmoil, but maybe for the better, according to the [UCLA Anderson Forecast](#) released Tuesday.

REGIONAL ECONOMIC EFFECTS

$$\frac{Y_{\ell,t+h} - Y_{\ell,t-1}}{Y_{\ell,t-1}} = \underbrace{\beta_h}_{\text{XS-MULTIPLIER}} \cdot \frac{G_{\ell,t+h} - G_{\ell,t-1}}{Y_{\ell,t-1}} + \alpha_{\ell}^h + \lambda_t^h + \varepsilon_{\ell,t+h}$$

- $Y_{\ell,t}$ real GDP, $G_{\ell,t}$ real DoD Contracts, α_{ℓ}^h & λ_t^h location & time FE; ($N = 377$; $T = 24$)

! **Endogeneity** of $G_{\ell,t}$: Reverse Causality (Mintz, 1992), Anticipation (Auerbach et al., 2020)

→ shift-share (**Bartik**) instrument:

→ WE REPLACE THE SHIFT:

$$Z_{\ell,t+h}^{\text{Bartik}} = \frac{s_{\ell} (G_{t+h} - G_{t-1})}{Y_{\ell,t-1}}$$

$$Z_{\ell,t+h}^{\text{HFXS}} = \frac{s_{\ell} \mathbb{E}_t(G_{t+1})}{Y_{\ell,t-1}}$$

- s_{ℓ} (Share): DoD contracts share MSA ℓ
- $\mathbb{E}_t(G_{t+1})$: **HFXS MILITARY NEWS SHOCKS**
- G_t (Shift): **National DoD contracts**

2-YEAR XS MULTIPLIER OF ≈ 1

Horizon	IV: HFXS Military News Shocks			IV: Standard Bartik		
	Coefficient	pvalue	Effective F	Coefficient	pvalue	Effective F
Impact	2.647 (2.307)	0.252	1.462	0.095 (0.044)	0.030	17.088
Year 1	1.352 (0.369)	0.000	14.939	0.539 (0.125)	0.000	95.193
Year 2	0.953 (0.271)	0.000	30.558	0.484 (0.148)	0.001	46.408
Year 3	0.614 (0.338)	0.070	6.257	0.639 (0.256)	0.013	15.239

Notes: 377 MSAs, 2001-23. GDP price deflator from BEA, base year 2017. Robust SE in parentheses, clustered at MSA level. Montiel Olea and Pflueger (2013) effective F is calculated with `weakivtest`.

→ Military news shocks **have real economic effects**

→ Produce **higher XS-multipliers** than standard Bartik

Robustness: exclusion of 9/11

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CONCLUSION

- **New model-consistent methodology** to identify military news shocks
 - Estimate & Test shocks from the data!
 - A. Identify narrative events
 - B. Run model-implied HFXS-regressions around events
 - **Self-validating** (sign & significance)
 - **Generalizable** to contexts of aggregate shock/heterogeneous exposure
- Application: US Military Spending post-2000:
 - Document **novel series** of key US military events
 - Construct **new** (HFXS) defense news shocks
 - Defense news shocks have significant effects on regional GDP (2-year $\text{XS-}\mathcal{M} \approx 1$)

Thank You!

OUTLINE

APPENDIX

EXAMPLE OF PROMPT WITH NON-CONTROVERSIAL EVENTS BACK

"Compile a list of dates or events—from 2000 onward—that signal a potential shift in the expected path of US military procurement spending. Include both positive and negative shocks.

Examples: (a) September 11, 2001 terrorist attacks: widely seen as a precursor to higher defense spending; (b) Failure in February/March 2013 of President Obama and Congress to reach a budget agreement: triggered automatic cuts (sequestration) and reduced defense spending; (c) Unexpected election victory of Donald Trump, November 2016: he campaigned on increasing military outlays. Use a similar standard to identify and briefly justify each additional event you list."

- 9/11: defense news shock according to Ramey and Zubairy ([2018](#)).
- Budget Sequestrations: exogenous fiscal consolidation by Alesina et al. ([2014](#)).
- Trump's 2016 election: marginal win + campaign on "*peace through strength*"

PROMPT FOR NARROWER PERIODS WITHOUT EXAMPLES [BACK](#)

- Context: Iraq & Afghanistan wars followed from 9/11 and prompted increased spending
- Then, we ask:
“List the defining moments/events of the war in Iraq and Afghanistan that (a) had large media coverage in the US around the years 2004-2008 and (b) which also gave the impression of an expected increase in military spending in the US.”

ASSUMPTIONS FOR IDENTIFICATION BACK

- **Assumption 1.** $\lambda_i \perp d\log V_i^e$
- **Assumption 2.** $\mathbb{E}[d\log V_i^e] = 0$
Private sales may move in both directions:
 - Lee (2024): new contracts crowd-in private sales via learning-by-doing.
 - Ilzetzki (2023): capacity constraints during WWII may have limited the ability of contractors to expand private
 - di Giovanni et al. (2023): crowding-out on impact, and crowding-in after one year after winning a contract.
- **Assumption 3.** $\lambda_i \perp d\log \pi_i^e := \frac{d\log \mu_i^e}{\mu_i - 1}$
If investors form expectations about future contractors' profitability, those expectations must be independent of reliance
- **Assumption 4.** $\lambda_i \perp d\log \theta_i^e$
- **Assumption 5.** $\mathbb{E}[d\log \theta_i^e] = 0$
If investors form expectations about future contractors' profitability, those expectations must be independent of reliance and average out to zero

FRAMEWORK GENERALIZABLE TO BROADER MACRO CONTEXTS

BACK

- It is possible to show that:

$$d \log P_{i,t} = \underbrace{\lambda_{i,t}}_{(i)} \cdot \underbrace{\xi_i}_{(ii)} \cdot \underbrace{d\varepsilon_t}_{(iii) - \text{Shock}}$$

Heterogeneous Exposure

- I. $\lambda_{i,t}$: fraction of sales exposed to the news shock
- II. ξ_i : elasticity of sales with respect to the shock
- III. $d\varepsilon_t$: shock you want to identify

PROPOSITION: GENERALIZATION

Let units experience a common shock ε_t with heterogeneous loadings captured by observable (or parametrizable) terms $(\lambda_{i,t}, \xi_i)$. Then, estimating the cross-sectional regression around the event yields an estimate of the shock magnitude.

LARGEST FIRMS IN THE SAMPLE [BACK](#)

- Median reliance is 20%. Interquartile range is [3.7%,39.9%].
- Top 3 firms by (median) reliance:
 - **VSE Corp** (86%) (Aviation Services)
 - **L3 Harris Technologies** (82%) (Avionics)
 - **Huntington Ingalls Industries** (73%) (Ship building)
- Top 3 firms by fraction of DoD Contracts (FY23):
 - **Lockheed Martin** (14.7%) (Aerospace)
 - **Raytheon (RTX)** (6.5%) (Weapons and Electronics)
 - **General Dynamics** (5.0%) (Aerospace, Submarines, Vehicles)
- Data cross-validation:
 - We match these companies with universe of micro-contracts from FPDS
 - We compare FPDS data with Top100 Report data
 - **The two data sources match!**

CONSTRUCTION OF EXCESS RETURNS BACK

- Need to “*clean*” the returns → extract *excess returns*
- Fama and French (1993) three factor model:

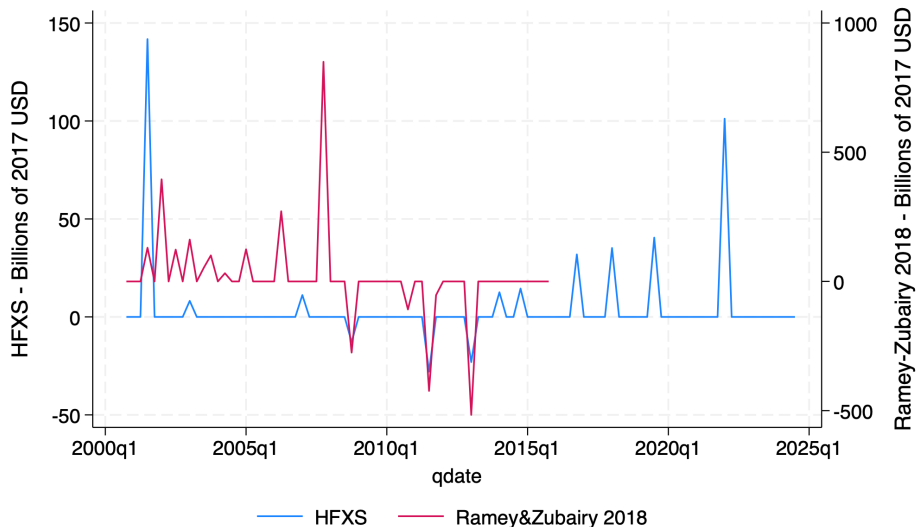
$$r_{i,t} = \alpha_i + \beta_i^1 \cdot \text{MKT}_t + \beta_i^2 \cdot \text{SML}_t + \beta_i^3 \cdot \text{HML}_t + v_{i,t} \quad \forall i \in \mathcal{I}$$

- $r_{i,t}$: contractors' weekly returns (WRDS)
- Three factors: MKT (market), SML (size) and HML (value)

→ $v_{i,t}$: OLS residuals ~ weekly excess returns

HFXS AND RZ18 SHOCKS ARE SIMILAR BUT NOT IDENTICAL

BACK



ROBUSTNESS: EXCLUSION OF 9/11 BACK

Robustness - Sample: 2002-2023 (Without 9/11) - 377 MSAs								
Horizon	IV: HFXC Military News Shocks			IV: Standard Bartik			OLS	
	Coefficient	pvalue	Effective F	Coefficient	pvalue	Effective F	Coefficient	pvalue
Impact	-0.112 (0.209)	0.594	9.428	0.124 (0.047)	0.008	17.575	0.009 (0.018)	0.622
Year 1	0.609 (0.301)	0.044	17.868	0.494 (0.120)	0.000	100.184	0.052 (0.025)	0.042
Year 2	0.571 (0.268)	0.033	12.293	0.437 (0.142)	0.002	42.991	0.078 (0.046)	0.090
Year 3	0.620 (0.427)	0.147	6.656	0.638 (0.271)	0.019	10.163	0.123 (0.069)	0.074

Notes: 377 MSAs, 2001-23. GDP price deflator from BEA, base year 2017. Robust standard errors in parentheses, clustered at MSA level. Montiel Olea and Pflueger (2013) effective F is calculated with `weakivtest`, coincides with Kleibergen and Paap (2006) statistic for single instrument