Natural Disaster and Municipal Bonds

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Data and Methodology

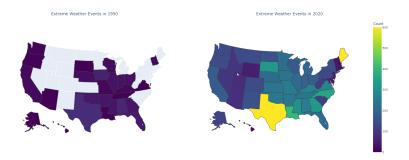
Baseline Results

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Natural Disasters Over Time

Frequency and severity of natural disasters are growing:

- Growth in the number of disaster and emergency declarations in the US averaged 7% in 1990–2020.
- Physical damage averaged over \$11 million per affected county.



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Asset Pricing Implication

- Widespread belief among professionals, academics, and regulators that asset prices underestimate climate risks.
 - Physical climate risks, however, rank as the top climate risk over next 30 years (Stroebel and Wurgler 2021).
- A potential reason for such underestimation of physical climate risk is that the effects are subject to:
 - 1. Confounding effect (e.g., stock prices of companies whose operations are geographically scattered).
 - 2. Statistical power to detect climate risks is low (e.g., slow-moving climate risks).
- Natural disasters provide an ideal setting to estimate the price implication of physical climate risk

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Our Strategy

- We focus on municipal bond returns (as opposed to bond yields or credit ratings).
- Why is studying municipal bond returns important?
 - 1. Can speak directly to the portfolio performance and wealth of bond investors.
 - 2. Allows to compare returns of the *same* bonds before and after disasters.
 - 3. Bond returns in high frequency reveal the pattern of investors' response in terms of intensity and immediacy.
- Using high-frequency return around exogenous events alleviates a concern for omitted variable problem without relying on an identification assumption (Painter (2020); Goldsmith-Pinkham et al. (2023)).

Impact and Mechanisms are Ex-Ante Unclear

- Can even have positive long-run effects on local economy:
 - "Creative destruction" or "build back better" (Strobl (2011); Deryungina et al. (2018))
- The price impact (if any) could be mainly driven by risk-aversion or liquidity premium, as opposed to physical damage.
 - Is it expected cashflow or discount rate?
- We dissect the sample in multiple dimensions that are correlated with physical damage but uncorrelated with other factors.
 - REV versus GO
 - Insured versus Uninsured
 - Direct measure of physical damage
 - Post-disater aid

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Data

Municipal Bond Data

- Municipal Securities Rulemaking Board (MSRB)
- Mergent Municipal Bond Database
- County-level Economic and Financial Data
 - Regional Economic Information System (REIS) from Bureau of Economic Analysis (BEA)
 - Census of Governments
- Natural Disaster Data
 - Spatial Hazard Events and Losses Database for the United States (SHELDUS)
 - Federal Emergency Management Agency (FEMA) Database

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Bond Returns

- Sparse trading makes it impossible to perform a high frequency analysis with raw data.
 - Municipal bonds are rarely traded (<3 times per year).
 - Hard to perform disaster-event studies with no available traded prices in high frequency.
- But many counties have hundreds of bonds outstanding at any given time.
- Solution: employ the repeat sales approach to obtain weekly bond returns for many US counties
 - Motivated by real-estate literature (e.g., Case and Shiller 1987)
 - Has been successfully applied to corporate bonds (Spiegel and Starks 2016; Robertson and Spiegel 2017)

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Repeat Sales

- To overcome scarce tradings in municipal bond markets, we use the repeat sales methodology to estimate county-level weekly returns.
- Estimate $\{R_t^c\}$ based on the following model:

$$R_{i,b:s} = \sum_{t=b+1}^{s} R_t^c + e_{i,b:s}$$

- $R_{i,b:s}$ is log return of bond *i* issued by county *c* from week *b* to *s*.
- R_t^c is county-level weekly log returns at week t.
- ► For each county c, we regress individual bond log returns on week-year indicators that equal 1 if b < t ≤ s.</p>
- ► We are effectively estimating time fixed effects of bonds issued by county *c*.

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Repeat Sales (Illustration)

Municipal bonds issued by Montgomery County in MD.

- Bond A: R1% log return from 2023w1 to 2023w4
- Bond B: R2% log return from 2023w1 to 2023w3
- Bond C: R3% log return from 2023w2 to 2023w4
- No subsequent prices, so no weekly returns.

▶ We run the following regression to apply repeat sales:

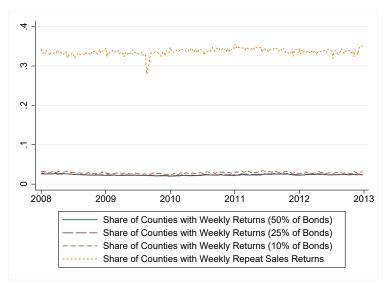
$$\begin{bmatrix} R1\%\\ R2\%\\ R3\% \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1\\ 1 & 1 & 0\\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} R_{2023w2}\\ R_{2023w3}\\ R_{2023w4} \end{bmatrix} + \varepsilon$$

▶ \hat{R}_{2023w2} , \hat{R}_{2023w3} , and \hat{R}_{2023w4} are estimated repeat sales returns in this example.

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Counties Covered by Repeat Sales vs. Actual Returns



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Event Study

- Using repeat sales estimation, we construct weekly cumulative abnormal returns.
- For each extreme weather event indexed by (c,t), weekly cumulative abnormal return at week τ from t-15 is:

$$WCAR_{c,t,\tau} = \sum_{s=-15}^{\tau} (R_{t+s}^c - R_{t+s}^b)$$

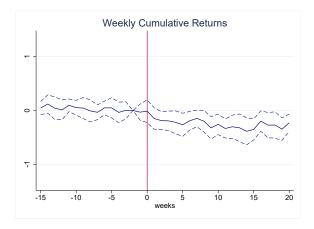
- R^b_{t+s} : Average repeat sales return of 20 benchmark counties.
- Benchmark counties chosen among disaster-unaffected counties 500+ miles away that most closely match based on lagged average coupon, credit rating, maturity, population, income per capita and unemployment rate.

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CAR Around Natural Disasters: All Uninsured Bonds



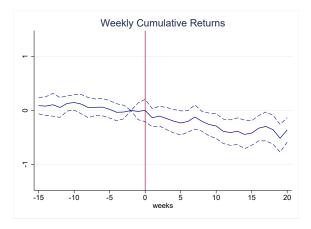
 Overall, uninsured muni bonds experience negative return (31bps over 10 weeks).

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CAR Around Natural Disasters: REV Bonds



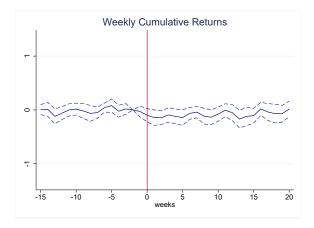
 REV bonds show more severe negative price drops (51bps over 10 weeks).

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CAR Around Natural Disasters: GO Bonds



 GO bonds do not exhibit significant price impact around natural disasters.

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CAR Around Natural Disasters

For brevity and statistical power, we focus on the difference between monthly CAR until t-2 versus CAR until t+4.

$$MCAR_{c,t,\tau} = \beta Post_{c,t,\tau} + \sum_{p=-1}^{12} \gamma(p) D_{c,t,\tau}^{M}(p) + \sum_{q=0}^{10} \delta(q) E_{c,t,\tau}^{M}(q) + \alpha_{c} + \epsilon_{c,t,\tau}$$

 \triangleright β estimates the difference:

Dep Var: CR	All Bonds	REV Bonds	GO Bonds
Post	-0.3144**	-0.5089**	-0.1277
	(-2.3279)	(-2.5602)	(-1.0594)
County FE	County FE YES		YES
No. of Obs.	1996	1185	1316
Adj. R-Squared	0.31	0.32	0.3

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With Raw Returns?

Dep Var: CR	Raw Returns		
	REV Bonds GO Bond		
Post	-1.2733	-0.0127	
	(-1.0952)	(-0.0015)	
County FE	YES	YES	
No. of Obs.	38	15	
Adj. R-Squared	0.22	0.1	

- The raw returns give us very little observation and no statistical power.
- Showcases the merit of the repeat sales approach.

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How About Insured Bonds?

Dep Var: CR	All Bonds	REV Bonds	GO Bonds
Post	-0.099	-0.1419	-0.06
	(-1.2862)	(-1.5319)	(-0.5195)
County FE	YES	YES	YES
No. of Obs.	3191	2052	1987
Adj. R-Squared	0.25	0.2	0.3

- Bond insurance effectively protects investors from natural disasters.
- Suggests such a price impact is caused by physical damage to cashflow as opposed to liquidity demand or behavioral resaons.

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Physical Damage Matters for REV Bonds

	Below Med	Above Med	Below Med	Above Med
Dep Var: CR	REV Bonds	REV Bonds	GO Bonds	GO Bonds
Post	-0.3502	-0.6132*	-0.1243	-0.1986
	(-1.5429)	(-1.8343)	(-0.7799)	(-1.3201)
County FE	YES	YES	YES	YES
No. of Obs.	594	591	658	658
Adj. R-Squared	0.36	0.43	0.37	0.37

- Reiterates that the pattern is driven by rational investors' reaction to damaged cashflow.
- Above-Med: average per-capita damage is \$528
- Below-Med: average per-capita damage is \$6.14

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Undoing the Damage by Federal Disaster Aid: REV Bonds

Dep Var: CR	Zero Aid	Below Med	Above Med
	REV Bonds	REV Bonds	REV Bonds
Post	-1.1954*	-0.7086**	-0.2969
	(-1.9769)	(-2.3249)	(-1.2096)
County FE	YES	YES	YES
No. of Obs.	242	537	648
Adj. R-Squared	0.29	0.34	0.37

- Zero aid: average per-capita damage is \$54
- Some aid: average per-capita damage is \$336
- Federal disaster aid helps mitigate negative shocks on REV bonds.

Historical Disaster Frequency and Investor Reaction

Panel A: By Pre-2000 Historic Damage					
	Below Med	Above Med	Below Med	Above Med	
Dep Var: CR	REV Bonds	REV Bonds	GO Bonds	GO Bonds	
Post	-0.6014***	-0.4246	-0.3034**	0.0372	
	(-2.6991)	(-1.5729)	(-2.1214)	(0.2171)	
County FE	YES	YES	YES	YES	
No. of Obs.	610	575	658	658	
Adj. R-Squared	0.28	0.37	0.26	0.32	

Panel B: By Projected Flood Risk

	Below Med	Above Med	Below Med	Above Med
Dep Var: CR	REV Bonds	REV Bonds	GO Bonds	GO Bonds
Post	-0.6657**	-0.2441	-0.1464	-0.0834
	(-2.5070)	(-0.9536)	(-0.8607)	(-0.5136)
County FE	YES	YES	YES	YES
No. of Obs.	743	442	886	430
Adj. R-Squared	0.33	0.27	0.3	0.35

• Below Med \rightarrow Counties with lower disaster frequency.

The physical risk appears to be priced ex-ante.

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Municipalities' Leverage and GO Bond Returns

Dep Var: CR	Low Severity	High Severity	Low Severity	High Severity
	GO Bonds	GO Bonds	REV Bonds	REV Bonds
$Post \times Levered$	-0.0548	-0.5517**	-0.0635	0.2924
	(-0.2199)	(-2.3262)	(-0.1459)	(0.6155)
Post	-0.1518	-0.0142	-0.4773*	-0.5838*
	(-0.9086)	(-0.0600)	(-1.7731)	(-1.6831)
County FE	YES	YES	YES	YES
No. of Obs.	456	398	408	379
Adj. R-Squared	0.41	0.53	0.46	0.5

- If a municipality has high debt to tax revenue ratio (bad creditworthiness), it has less room to maneuver to make up for the cashflow damage.
- Severe disasters may negatively affect its GO bond returns.

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Revenue Concentration and GO Bond Returns

Dep Var: CR	Low Severity	High Severity	Low Severity	High Severity
	GO Bonds	GO Bonds	REV Bonds	REV Bonds
Post×Concentrated	0.2266	-0.5454*	0.2388	-0.01
	(0.7484)	(-1.9232)	(0.3992)	(-0.0176)
Post	-0.3643	-0.0652	-0.4649	-0.8143
	(-1.3679)	(-0.2497)	(-0.8535)	(-1.2748)
County FE	YES	YES	YES	YES
No. of Obs.	276	262	238	220
Adj. R-Squared	0.52	0.49	0.5	0.58

- If a municipality have a concentrated source of revenue, severe disasters may negatively affect GO bond returns.
- More direct evidence that GO bonds' resiliency is related to revenue diversification.

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Conclusion

- We use the repeat sales methodology to study how natural disasters affect municipal bond returns.
- Municipal bond market responds negatively but slowly to disasters:
 - indicative of underreaction by investors.
- An average disaster causes 31bps negative return, translating into \$9.2 million investor loss (average physical damage is \$19 mil.)
- Overall, our findings show that the post-disaster reaction is consistent with investors' rational reaction, rather than subjective perception changes.