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### Motivation

Introduction •0000000



Figure: Bennington Lake in Walla Walla, Washington after heavy rains in 2020.

# State Awareness and Media Coverage

Introduction



Figure: State advice for handling flood hazards to drinking water.

## State Awareness and Media Coverage

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Figure: News reports highlight these concerns.

# A Growing Problem

Introduction 00000000

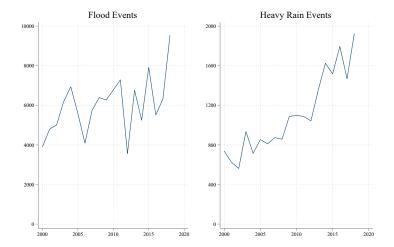


Figure: NOAA National Centers for Environmental Information (NCEI) Storm Events Database (2021). https://www.ncdc.noaa.gov/stormevents/

### Mechanisms

Introduction

How might extreme precipitation and flooding affect drinking water quality?

- Erosion and sediment mobilization impact source waters, decreasing water treatment effectiveness and increasing disinfectant byproduct formation (Liu et al., 2012) (Fakour et al., 2016).
- Human and animal waste runoff and less-effective disinfectants → risk of bacteriological contamination (Phan and Sherchan, 2020).
- Damage to infrastructure and capital equipment may cause leaks or drops in pressure → risk of bacteriological contamination. (Arrighi et al., 2017).
- $\bullet$  Unexpected and large flood events may cause industrial spills (Thomas et al., 2018) or mobilize legacy pollution (Zak et al., 2009) .

# Research Questions

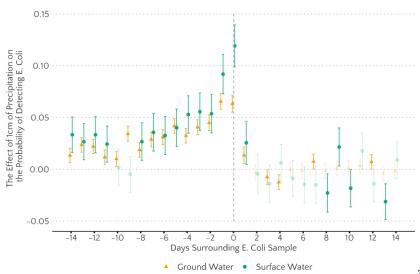
Introduction

### We ask three research questions:

- How does flooding and extreme precipitation affect the risk of drinking water bacteriological contamination?
- 2 How does flooding and extreme precipitation affect the concentration of disinfectant byproducts?
- 3 What water system types, states, and communities are most at risk?

# Preview of Findings

Introduction 00000000



# Preview of Findings

Introduction

- I How does flooding and extreme precipitation affect the risk of bacteriological contamination of drinking water?
  - A recent flood or heavy rain event increases the likelihood of Coliform detection by 14-25% and E. Coli detection by 17-26%.
- 2 How does flooding and extreme precipitation affect the concentration of disinfectant byproducts?
  - A recent flood or heavy rain event increases the concentration of TTHMs by 5-6% and HAA5s by 6%.
- 3 What states, water system types, and communities are most at risk?
  - Groundwater and surface-water sourcing systems both see increased bacteriological risk, although smaller systems and GW systems bear disproportionate risk.
  - Larger and surface water-sourcing systems have the largest increases in DBP formation.

Introduction

### Vast literature on flooding:

- Costs of floods to employment, displacement, property damage, crop damage, and other outcomes.
- FEMA Hazus model calculates flood costs to utility infrastructure and disruption of service but not drinking water quality changes from extreme precipitation or flood events (FEMA Hazus 2.1, 2018).

We quantify the average relationship between extreme precipitation and drinking water quality by leveraging:

- 20 year data time horizon  $\rightarrow$  impacts from rare events.
- Near national sampling coverage → distributional analysis across water system types, states, and communities.
- Use of drinking water samples escapes endogenous behavior relevant to SDWA violations (Bennear, Jessoe, Olmstead, 2009).

**Empirical Approach** 

### Data

### 1. Drinking Water Quality Samples (2000-2019)

- Publicly available samples over 30 states from 2000-2019.
- Six-year Review 4 Samples over 48 states from 2012-2019.
   Samples by County

## 2. National Oceanic and Atmospheric Association Storm Events Database (2000-2019)

All flood or heavy rain events causing loss of life, injuries, significant
property damage, and/or disruption to commerce.

Map of Floods in the US

## 3. Parameter-elevation Regressions on Independent Slopes Model (PRISM) (2000-2019)

• Daily county and HUC12 precipitation and temperature.

#### 4. Others

- Hydroshare U.S. Community Water Systems Service Boundaries, v2.0.0.
- National Hydrography Dataset Plus version 2.0.
- Safe Drinking Water Information System (SDWIS) for water intake locations (HUC12-level) and system characteristics.

# Summary Statistics

Mean	s.e.	Min	Max	Observations
2.19	14.63	0	100	33,129,011
1.13	10.58	0	100	9,642,038
30.12	27.92	0	1617.1	1,882,045
18.90	18.74	0	2900	1,612,642
ty				
55.24	129.83	0	2,084	2,999
13.06	31.00	0	586	2,999
ounty				
0.28	0.66	0.00	34.91	21,396,570
12.87	10.63	-34.56	39.76	$21,\!396,\!570$
	2.19 1.13 30.12 18.90 ty 55.24 13.06 county 0.28	2.19 14.63 1.13 10.58 30.12 27.92 18.90 18.74 ty 55.24 129.83 13.06 31.00 county 0.28 0.66	2.19 14.63 0 1.13 10.58 0 30.12 27.92 0 18.90 18.74 0 ty  55.24 129.83 0 13.06 31.00 0	2.19 14.63 0 100 1.13 10.58 0 100 30.12 27.92 0 1617.1 18.90 18.74 0 2900  ty 55.24 129.83 0 2,084 13.06 31.00 0 586  county 0.28 0.66 0.00 34.91

Let  $y_{imst}$  represent whether a drinking water sample detected a bacteriological contaminant or the concentration of TTHM or HAA5 in county i, day d, month m, state s, and year t.

$$y_{idmt} = \beta Disaster_{idmt} + X'_{idmt} \gamma + \eta_i + \eta_m + \eta_t + \nu_{idmt}$$

#### Core Terms:

- $Disaster_{idmt}$  is an indicator equal to one if a flooding event occurred within  $\delta \in 3, 10$  days prior in the county of the water system.
- In some specifications, Disaster<sub>idmt</sub> is the sum of precipitation in the HUC12 over the prior 3 days.
- X includes controls for daily temperature and two lagged daily temperature variables.
- Fixed effects for drinking water system, month, and year.

**Identifying assumption**: Conditional on controls for temperature and fixed effects for county, month, and year, drinking water quality is uncorrelated with extreme precipitation events except through the effect of interest.

► Event Study Identification Strategy

Results

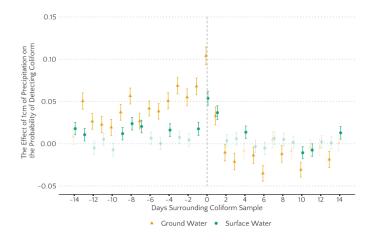
## Floods, Rainfall, and Total Coliform Detection

	Floods		Heavy Rains		Rainfall (cm)
	3-day	10-day	3-day	10-day	3-day
Precipitation Event	0.312*** (0.0528)	0.291*** (0.0336)	0.569*** (0.140)	0.351*** (0.0574)	0.0813*** (0.00354)
Dep Var Mean	2.22	2.22	2.22	2.22	2.22
Pct. Change	14.06	13.11	25.67	15.84	3.67
Water System FEs	✓	✓	✓	✓	✓
Year and Month FEs	✓	✓	✓	✓	✓
Observations	30,888,176	30,888,176	30,888,176	30,888,176	30,888,176

<sup>\*</sup> p < 0.1, \*\* p < 0.05, \*\*\* p < 0.01. Clustered standard errors at the water system.

Results by water source. Results by system size.

# Coliform Detection Event Study – Precipitation



## Floods, Rainfall, and E. Coli Detection

	Floods		Heavy Rains		Rainfall (cm)
	3-day	10-day	3-day	10-day	3-day
Precipitation Event	0.322*** (0.0645)	0.343*** (0.0457)	$0.0175 \\ (0.157)$	0.208*** (0.0748)	0.0749*** (0.00629)
Dep Var Mean	1.21	1.21	1.21	1.21	1.16
Pct. Change	26.73	28.45	1.45	17.26	6.45
Water System FEs	✓	✓	✓	✓	✓
Year and Month FEs	✓	✓	✓	✓	✓
Observations	8,069,499	8,069,499	8,069,499	8,069,499	8,903,963

<sup>\*</sup> p < 0.1, \*\* p < 0.05, \*\*\* p < 0.01. Clustered standard errors at the water system.

See results by water system source Y See results by water system size

# Floods, Rainfall, and Total Trihalomethanes $(\mu q/L)$

	Floods		Heavy Rains		Rainfall (cm)
	3-day	10-day	3-day	10-day	3-day
Precipitation Event	1.458*** (0.408)	1.787*** (0.244)	1.550*** (0.487)	1.648*** (0.257)	0.236*** (0.0210)
Dep Var Mean	30.40	30.40	30.40	30.40	30.40
Pct. Change	4.76	5.76	4.95	5.30	0.76
Water System FEs	✓	✓	✓	$\checkmark$	✓
Year and Month FEs	✓	✓	✓	$\checkmark$	✓
Observations	1,778,963	1,778,963	1,778,963	1,778,963	1,778,963

<sup>\*</sup> p < 0.1, \*\* p < 0.05, \*\*\* p < 0.01. Clustered standard errors at the water system.

See results by water system source See results by water system size

# Floods, Rainfall, and Haloacetic Acids ( $\mu q/L$ )

	Floods		Heavy Rains		Rainfall (cm)
	3-day	10-day	3-day	10-day	3-day
Precipitation Event	1.058*** (0.289)	1.165*** (0.143)	0.0273 $(0.462)$	$0.191 \\ (0.167)$	0.122*** (0.0143)
Dep Var Mean	18.90	18.90	18.90	18.90	18.90
Pct. Change	5.59	6.16	0.10	0.96	0.65
Water System FEs	✓	✓	✓	✓	✓
Year and Month FEs	✓	✓	✓	✓	✓
Observations	1,551,814	1,551,814	1,551,814	1,551,814	1,551,814

<sup>\*</sup> p < 0.1, \*\* p < 0.05, \*\*\* p < 0.01. Clustered standard errors at the county or water system.

See results by water system source Y See results by water system size

Conclusion

Damages from clinic visits, stomach illness purchases, sick children stay-at-home days, and bottled water purchases were \$7.7m from TCR violations in North Carolina (2005-2014) (Marcus, 2020):

- 424 detections of E. Coli and 16,482 detections of Coliform  $\rightarrow$  \$145 per E. Coli detection and \$461 per total coliform detection.
- 0.5 floods and 0.1 heavy rains per water system annually, 56,000 water systems, and risk of detection per extreme precipitation event.
- \$5.1m annually in costs from flooding and \$1.1m annually from heavy rains.

Un-quantified potential mortality costs from gastrointestinal illness as well as bladder cancer cost of illness and mortality.

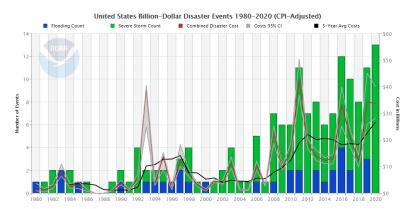
## Conclusion

Extreme precipitation and flooding events pose a risk to drinking water quality.

- Flooding events and extreme precipitation increase the likelihood of Coliform and E. Coli detection.
- Flooding events and extreme precipitation increase disinfectant byproduct concentrations.
- Combined, these results demonstrate the extent to which extreme precipitation complicates the water treatment process and poses a threat to public health.
- Flood mitigation and storm runoff collection strategies have economically meaningful and yet unquantified public health benefits.

Thank you!
Questions/comments?
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## **Appendix**



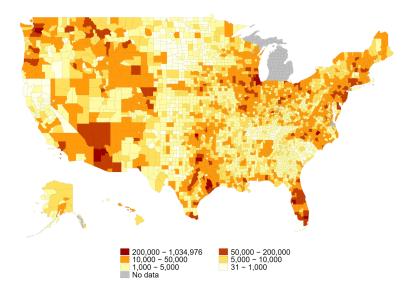


Figure: Number of Coliform, E. Coli, TTHM, and HAA5 samples by county (2000-2019) • Go back

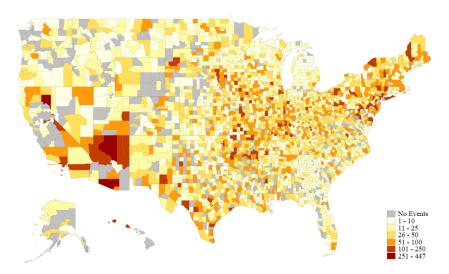


Figure: Number of NOAA Floods by County (2000 - 2019) Go back



# Event Study Empirical Approach

Let  $y_{imst}$  represent whether a sampling test detected a bacteriological contaminant or the continuous level of disinfectant byproduct detected by the sample in water system i, day d, month m, state s, and year t.

$$y_{idmt} = \sum_{t \in -14}^{14} \beta_t Precipitation_{idmt} + X'_{idmt} \gamma + \eta_i + \eta_m + \eta_t + \nu_{idmt}$$
 (1)

#### Core Terms:

- $Precipitation_{idmt}$  is the sum of precipitation in the watershed of the water system intake on day t.
- X includes controls for daily temperature and two lagged daily temperature variables.
- Fixed effects for drinking water system, month, and year.

Identifying assumption: The drinking water sampling outcome is uncorrelated with precipitation except through the impact of precipitation itself conditional on controls for temperature and fixed effects for county, month, and year.

• Go back

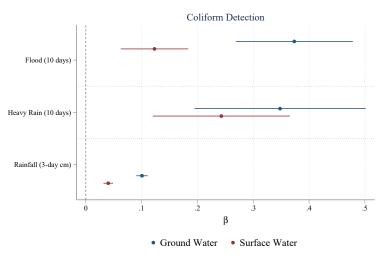


Figure: Extreme Precipitation and Coliform Detection by Source

• Go back

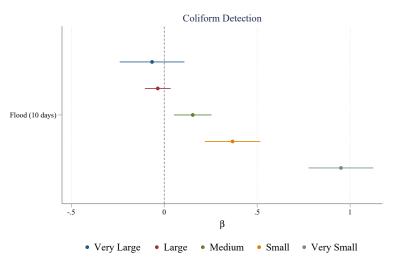


Figure: Flooding and Coliform Detection by System Size Go back

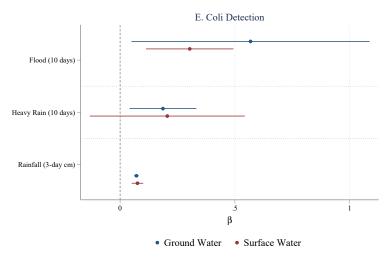


Figure: Extreme Precipitation and E. Coli Detection by System Source Goback

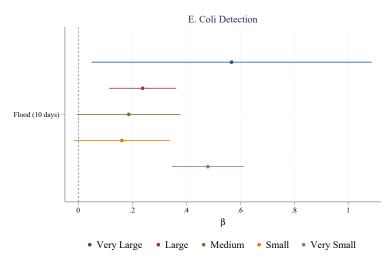


Figure: Flooding and E. Coli Detection by System Size Go back

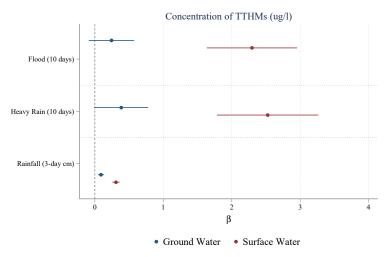


Figure: Extreme Precipitation and TTHM Concentrations by Source

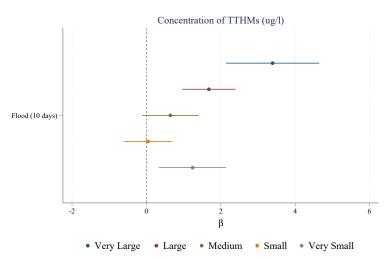


Figure: Flooding and TTHM Concentrations by System Size Go back

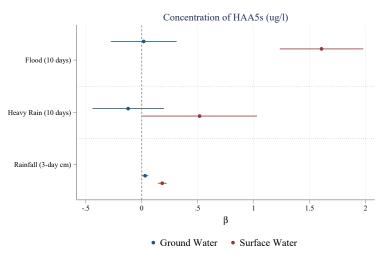


Figure: Extreme Precipitation and HAA5 Concentrations by Source

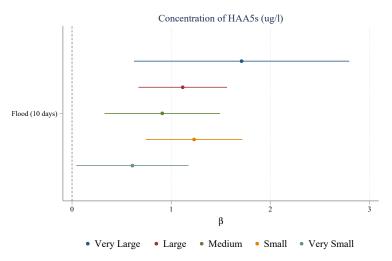


Figure: Flooding and HAA5 Concentrations by System Size Go back