The Long-Lasting Effects of bombing on Environmental Management. Evidence from Kosovo

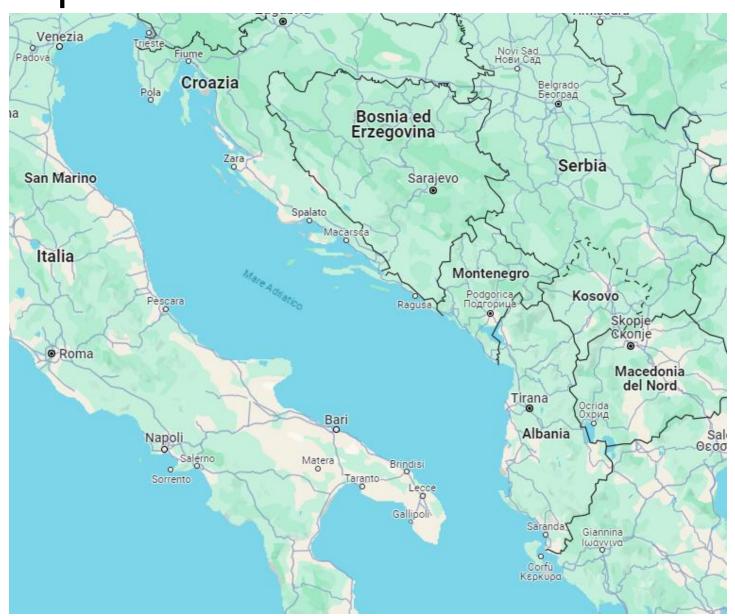
CHILDREN'S LIVES: INTERNATIONAL CONFERENCE ON CHILDREN AND THEIR FAMILIES USING THE MULTIPLE INDICATOR CLUSTER SURVEYS (MICS)

Elda Celislami,
University of Reading & Unito-Collegio Carlo Alberto
elda.celislami@carloalberto.org

Motivation

- ► Investigates long-term effects of the war on waste management and infant mortality rates in Kosovo.
- ► Extensive research has focused on socioeconomic impacts of the Yugoslav Civil War on labour, education and child development, but environmental and public health consequences remain underexplored.
- ► Enhances understanding of how environmental issues affect post-war recovery dynamics and public health, contributing to broader discussions on sustainable recovery.

In the map



Timeline of the conflict

- ▶ 1989 Tens of thousands of Kosovar Albanians are put out of work and the activities of their cultural organizations are restricted. Rioting ensue
- ▶ 1991/93 Federal Republic of Yugoslavia breaks up. After a secret vote, ethnic Albanians proclaim the creation of their own Republic of Kosovo, though it earns little international recognition. "Ethnic cleansing" idea spreads in the Balkans
- ▶ 1994 First NATO airstrike in history against Bosnian Serbs
- ▶ 1997 Violence escalates in Kosovo as Serbian security forces clamp down further on resistance and KLA steps up its attacks. Feb 1998 armed conflict starts within Kosovo
- ► March 1999 NATO intervened with an aerial bombing campaign, justifying it as a "humanitarian war"
- ▶ June 1999 End of war, Yugoslav and Serb forces agreeing to withdraw from Kosovo to make way for an international presence

Consequences

- ► 13,500 people were killed or went missing
- ► Between 1.2 million and 1.45 million Kosovo Albanians were displaced
- ▶ By November 1999, according to the UN High Commissioner for Refugees, 848,100 out of 1,108,913 had returned (76%)
- ▶ 364 bombs hit Kosovo, the most afflicted was the capital (54)



Kosovo Waste Management Issues

- ► Kosovo's waste management system faces major challenges with inadequate collection and treatment, affecting over 60% of the nation and causing significant environmental pollution and poor aesthetics.
- ▶ Insufficient cross-institutional cooperation, budget deficiencies, lack of trained staff, extensive informal sector activities, and weak law enforcement, contribute to Kosovo's missing 2020 waste management targets (European Environment Agency, 2021).
- ▶ Of the ten landfills built post-war, only five are state-managed. For the other ones, the management is still quite ambiguous.
- ► Example: the Municipal Landfill of Mitrovica, was established in 1998 by the Serb authorities and was legalised in 2002 by the Danish KFOR (NATO peacekeepers).

Still, no effort was made whatsoever to regulate the dump.

Kosovo Waste Management – External financing

- ► After the war, the European Council established the European Agency for Reconstruction main mandate to manage the EU's aid programs in Kosovo.
- During its mandate the Agency managed a portfolio of almost 3 billion euros.
- ▶ The European Agency for Reconstruction was given a high degree of autonomy and delegated authority to streamline the delivery of EU aid to the Western Balkans region. This allowed the agency to cut bureaucratic corners and speed up the contracting and disbursement process compared to the EU's usual cumbersome procedures.
- ► The agency started operating in February 2000 then agreed to a dissolution right by the end of 2008.
- ► EU funded the design and construction of six landfills built to EU standards, aiming to replace the existing municipal dumpsites as soon as they were completed, ensuring maximum environmental protection.

Research Question

Has the violence endured during the war affected the management of infrastructures potentially creating negative externalities?

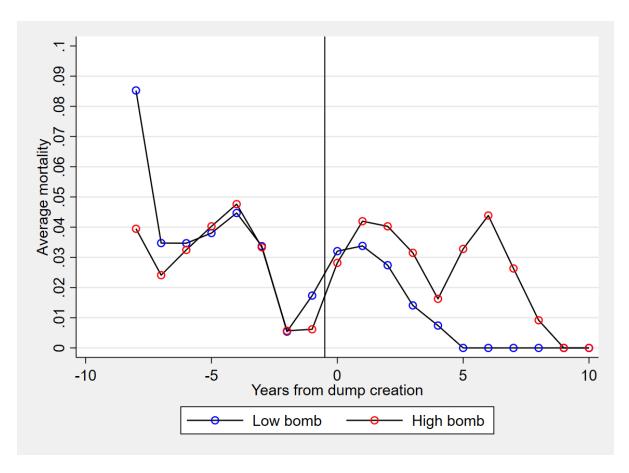
Data

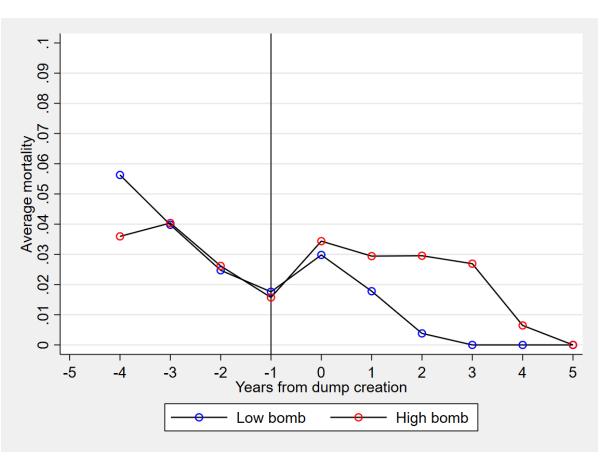
- ► Multiple Indicator Cluster Survey (MICS) from UNICEF, combining two surveys (2012-2018) focusing on infant mortality (children under one year old)
- ► Focus on the waste landfills that opened after the war (ie, after 2000)
- ► Landfills location and openings have been received through a collection of sources.
 - ► The exact location of the waste landfills constructed after the conflict has been provided by the Kosovo Environmental Protection Agency (KEPA).
 - ▶ Unfortunately, their list was not exhaustive and the other landfill information such as opening year and location was retrieved by reading a collection of reports, papers and news articles.
- ► Count of the number of bombs hitting the region is collected at the municipality level and is sourced by the Human Rights Watch.

Descriptive Statistics

All					High bon	nb		Close6			High bomba and close6			
Variable	Obs	Mean	Std. dev	Obs	Mean	Std. dev	Obs	Mean	Std. dev	Obs	Mean	Std. dev		
$infant_m$	5,373	0.028662	0.16687	2,601	0.032295	0.176817	2,945	0.028183	0.165525	1,455	0.030241	0.171307		
postdump	5,373	0.440908	0.496542	2,601	0.444444	0.497	2,945	0.41528	0.492854	1,455	0.418557	0.493492		
high_bomb	5,373	0.484087	0.499793	2,601	1	0	2,945	0.494058	0.50005	1,455	1	0		
close6	5,373	0.548111	0.497726	2,601	0.5594	0.496555	2,945	1	0	1,455	1	0		
rural	5,373	0.520007	0.499646	2,601	0.597847	0.490427	2,945	0.395925	0.489132	1,455	0.476976	0.499641		
ethnicity	5,373	1.227433	0.829473	2,601	1.324875	1.013084	2,945	1.240407	0.834015	1,455	1.335395	1.018511		
edu_none	5,373	0.026242	0.15987	2,601	0.024606	0.154951	2,945	0.027844	0.164553	1,455	0.026804	0.161566		
edu_primary	5,373	0.051368	0.220768	2,601	0.057286	0.232432	2,945	0.045501	0.208435	1,455	0.053608	0.22532		
$edu_low_se\sim d$	5,373	0.545505	0.497971	2,601	0.611688	0.48746	2,945	0.52326	0.499544	1,455	0.595189	0.491024		
$edu_up_sec\sim d$	5,373	0.262796	0.440193	2,601	0.226067	0.418363	2,945	0.279796	0.448975	1,455	0.24055	0.427564		
wealth_poor	5,373	0.215522	0.411222	2,601	0.244137	0.429657	2,945	0.20747	0.405564	1,455	0.238488	0.426306		
wealth_low \sim e	5,373	0.195422	0.396562	2,601	0.218378	0.413224	2,945	0.178608	0.383089	1,455	0.191065	0.393276		
wealth_up_ \sim e	5,373	0.194305	0.395701	2,601	0.205306	0.404003	2,945	0.188795	0.391412	1,455	0.209622	0.407179		
wealth_rich	5,373	0.193188	0.394836	2,601	0.167628	0.373607	2,945	0.194907	0.396196	1,455	0.183505	0.387213		
year_open	$5,\!373$	2005.084	2.776578	2,601	2004.199	1.889326	2,945	2005.792	3.054387	1,455	2004.772	2.042237		

Average infant mortality and rolling window





(a) 2 years average infant mortality

(b) 2 years rolling window mortality

Difference in Difference (in Difference)

DD (restricted to 6km)

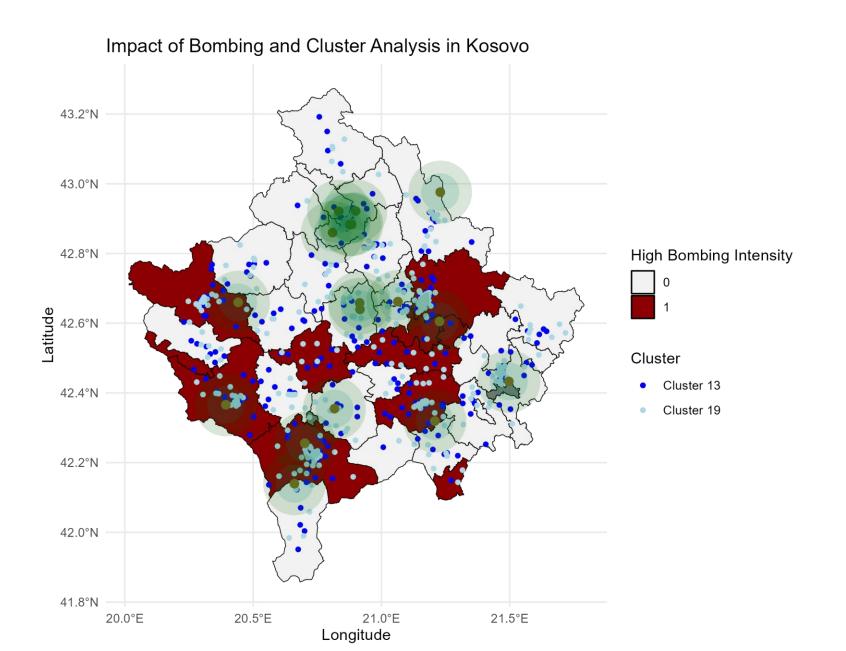
infant mortality_{it} =
$$\beta_0 + \beta_1 \text{postdump}_{it} + \beta_2 \text{high_bomb}_i$$

+ $\beta_3 \text{postdump}_{it} \times \text{high_bomb}_i + \beta_4 X_{it} + \epsilon$ (1)

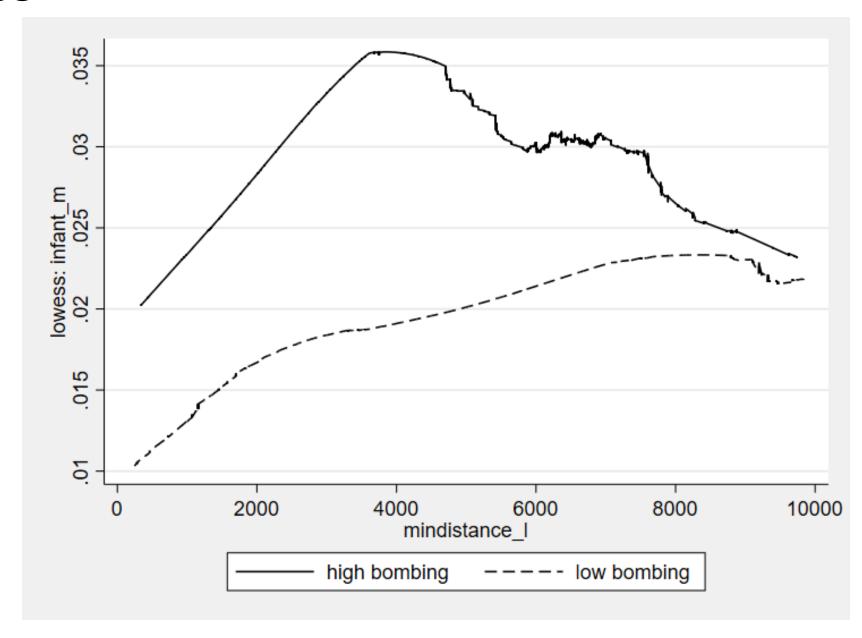
DDD

```
\begin{split} & \inf \mathsf{ant} \ \mathsf{mortality}_{it} = & \beta_0 + \beta_1 \mathsf{postdump}_{it} + \beta_2 \mathsf{high\_bomb}_i + \beta_3 \mathsf{close6}_{it} \\ & \beta_4 \cdot (\mathsf{postdump}_{it} \times \mathsf{high\_bomb}_i) + \beta_5 \cdot (\mathsf{postdump}_{it} \times \mathsf{close6}_{it}) + \\ & \beta_6 \cdot (\mathsf{high\_bomb}_i \times \mathsf{close6}_{it}) + \\ & \beta_7 \cdot (\mathsf{postdump}_{it} \times \mathsf{high\_bomb}_i \times \mathsf{close6}_{it}) + \\ & \beta_8 X_{it} + \epsilon_{it} \end{split}
```

Map



Lowess



Event Study Result - Callaway SantAnna

	Coefficient	Std. err	Z	P> z	95% conf.	interval
ATT	0.039337	0.008191	4.8	0.000	0.023282	0.055391
Pre_avg	-0.0299	0.008952	-3.34	0.001	-0.04745	-0.01236
$Post_avg$	0.038765	0.008302	4.67	0.000	0.022493	0.055036

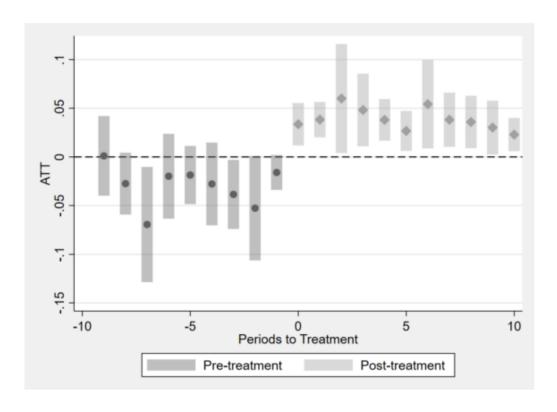
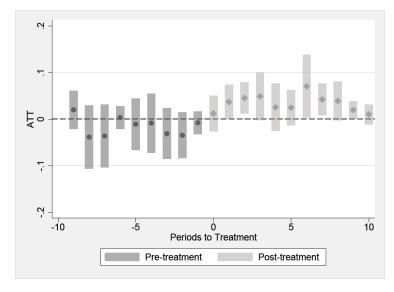
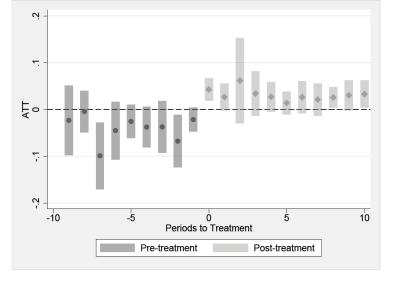
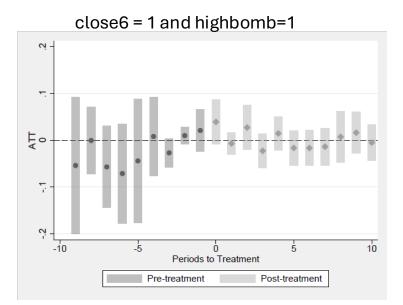


Figure: Event Study without Covariate

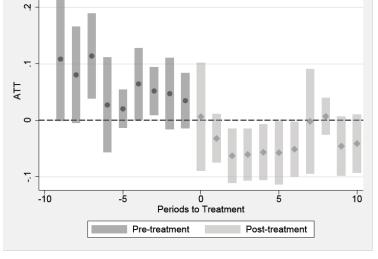
Callaway SantAnna on close and high bombed combied







close6 = 1 and highbomb=0



close6 = 0 and highbomb=0

close6 = 0 and highbomb=1

DDD and DD

Panel A: $D \times D \times D$									
infant mortality	(1)	(2)	(3)						
postdump	-0.005	-0.015	-0.003						
high bomb	$0.013) \\ 0.007$	$(0.015) \\ 0.004$	$(0.018) \\ 0.002$						
postdump×high bomb	$(0.010) \\ -0.015$	(0.016) -0.011	(0.017) -0.010						
close6	$(0.011) \\ 0.007$	$(0.016) \\ 0.000$	$(0.017) \\ 0.009$						
	(0.008)	(0.007)	(0.016)						
$postdump \times close6$	-0.003 (0.010)	$0.002 \\ (0.011)$	-0.005 (0.017)						
high bomb×close6	-0.018* (0.010)	-0.028* (0.015)	-0.001 (0.016)						
$postdump \times high bomb \times close6$	0.032*** (0.011)	0.040** (0.016)	0.016 (0.020)						
Covariates	X	X	X						
Birth Year FE	X	X	X						
N N	5373	3249	2564						
N_clust	26	17	18						

Panel B: $D \times D$										
infant mortality	(1)	(2)	(3)							
postdump	0.001 (0.011)	-0.004 (0.013)	-0.005 (0.009)							
high bomb	-0.012	-0.031***	-0.001							
$postdump \times high bomb$	$(0.008) \\ 0.018 \\ (0.011)$	(0.006) $0.040***$ (0.009)	$(0.007) \\ 0.009 \\ (0.010)$							
Covariates Child Birth Year FE	X X	X X	X X							
N N_clust	2944 15	1799 8	1383 11							

Notes: The table displays coefficients with p < 0.1, ** p < 0.05, *** p < 0.01, followed by standard errors in parentheses. The infant mortality mean for panel A is 0.0287 while for pane B is 0.0281

Panel A represents the triple difference in difference while the simple difference in difference is reported in Panel B. Panel B sample is obtained by subsampling the children living within 6 km to a waste landfill. Furthermore, it also includes weights to account for children born in places with more than one dump within 6 km. The table shows the results for infant mortality considering (1) all the waste landfills; (2) the landfills constructed with external funding (European Reconstruction Agency); (3) the waste landfills constructed with internal funding.

External Funding

Notes: The table displays coefficients with p < 0.1, ** p < 0.05, *** p < 0.01, followed by standard errors in parentheses. Panel A represents the triple difference in difference while the simple difference in difference is reported in Panel B. Panel B sample is obtained by subsampling the children living within 6 km to a waste landfill. Furthermore, it also includes weights to account for children born in places with more than one dump within 6 km. The table shows the results for infant mortality where (1) represents the triple difference in difference without covariates or fixed effects; (2) is the same as the previous one with covariates; (3) contains waste landfill opening years fixed effects; (4) contains birth year fixed effects; (5) has regional fixed effects: and (6) with waste landfill fixed effects

Panel A: $D \times D \times D$											
infant mortality	(1)	(2)	(3)	(4)	(5)	(6)					
postdump	-0.024**	-0.021**	-0.019*	-0.015	-0.019*	-0.019*					
	(0.009)	(0.010)	(0.010)	(0.015)	(0.009)	(0.010)					
high bomb	0.007	0.005	-0.015	0.004	-0.017						
	(0.016)	(0.015)	(0.014)	(0.016)	(0.014)						
$postdump \times high\ bomb$	-0.008	-0.010	-0.012	-0.011	-0.011	-0.012					
	(0.017)	(0.016)	(0.016)	(0.016)	(0.015)	(0.016)					
close6_era	-0.003	-0.001	-0.006	0.000	-0.009	-0.006					
	(0.009)	(0.008)	(0.007)	(0.007)	(0.007)	(0.008)					
$postdump \times close6_era$	0.005	0.003	-0.000	0.002	-0.000	-0.000					
	(0.010)	(0.010)	(0.011)	(0.011)	(0.009)	(0.011)					
high bomb \times close6 _era	-0.027*	-0.027*	-0.020	-0.028*	-0.018	-0.020					
	(0.015)	(0.015)	(0.013)	(0.015)	(0.013)	(0.013)					
postdump \times high bomb \times close6 _era	0.038**	0.040**	0.042***	0.040**	0.042***	0.042***					
	(0.015)	(0.015)	(0.014)	(0.016)	(0.013)	(0.014)					
Covariates		X	X	X	X	X					
Dump Opening year FE			X								
Child Birth Year FE				\mathbf{X}							
Region FE					X						
Waste landfill FE						X					
N	3249	3249	3249	3249	3249	3249					
N_clust	17	17	17	17	17	17					
	Panel B	: D×D									
infant mortality	(1)	(2)	(3)	(4)	(5)	(6)					
postdump	-0.019***	-0.023***	-0.021***	-0.004	-0.022***	-0.021***					
F	(0.005)	(0.005)	(0.005)	(0.013)	(0.005)	(0.005)					
high bomb	-0.019***	-0.024***	-0.027***	-0.031***	-0.027***	(01000)					
	(0.004)	(0.003)	(0.006)	(0.006)	(0.005)						
postdump×high bomb	0.030***	0.035***	0.033***	0.040***	0.033***	0.033***					
PesseamP81	(0.006)	(0.007)	(0.006)	(0.009)	(0.006)	(0.006)					
Covariates		X	X	X	X	X					
Dump Opening year FE			X								
				X							
Child Birth Year FE											
Child Birth Year FE Region FE					X						
					X	X					
Region FE	1800	1935	1800	1799	X 1800	X 1800					

Water test

```
water_test<sub>i</sub> = \beta_0 + \beta_1 high_bomb<sub>i</sub> + \beta_2 close6<sub>i</sub> + \beta_3 high_bomb<sub>i</sub> × close6<sub>i</sub> + \beta_4 piped_water<sub>i</sub> + \beta_5 publictap_water<sub>i</sub> + \beta_6 well_water<sub>i</sub> + \beta_7 spring_water<sub>i</sub> + \gamma_{Nr} + \epsilon_i
```

$$\log(\text{water_test}_i) = \beta_0 + \beta_1 \text{ high_bomb}_i + \beta_2 \text{ close6}_i + \beta_3 \text{ high_bomb}_i \times \text{close6}_i +$$
$$\beta_4 \text{ piped_water}_i + \beta_5 \text{ publictap_water}_i + \beta_6 \text{ well_water}_i +$$
$$\beta_7 \text{ spring_water}_i + \gamma_{Nr} + \epsilon_i$$

Water test - Results

	All			Ex	ternal fundi	ng	Internal funding			
E.Coli count	OLS	Poisson	Neg Bin	OLS	Poisson	Neg Bin	OLS	Poisson	Neg Bin	
high bomb	2.038	0.148***	-0.583	-2.806**	-1.206***	-2.278***	-1.945	0.323***	-0.317	
	(1.373)	(0.052)	(0.597)	(1.267)	(0.097)	(0.675)	(1.977)	(0.063)	(0.537)	
close6	0.211	-0.214***	1.849***	-2.431*	0.235***	0.464	-5.508***	-0.540***	2.935***	
	(1.347)	(0.054)	(0.625)	(1.347)	(0.078)	(0.621)	(2.063)	(0.081)	(0.640)	
$high_bomb \times close6$	-0.344	0.536***	0.460	5.451***	2.518***	1.380	5.103*	0.195*	-1.449**	
	(1.641)	(0.070)	(0.672)	(1.796)	(0.132)	(0.858)	(2.635)	(0.111)	(0.735)	
w_{drink_piped}	-1.383	-0.812***	-0.846*	-1.390	-0.942***	1.981***	0.958	1.300**	-0.325	
	(1.355)	(0.088)	(0.481)	(1.504)	(0.112)	(0.759)	(2.524)	(0.507)	(0.993)	
$w_drink_publictab$	-2.486	-1.615***	-0.024	-7.762*	-4.118***	1.101	-2.156	-0.339	-0.701	
	(3.427)	(0.316)	(1.101)	(4.032)	(1.011)	(2.175)	(4.334)	(0.597)	(1.194)	
$w_{drink_{well}}$	11.581***	1.475***	3.061***	7.160***	0.455***	4.523***	14.168***	3.116***	3.104***	
	(1.712)	(0.089)	(0.582)	(2.131)	(0.124)	(1.115)	(2.945)	(0.507)	(0.934)	
w_{drink_spring}	1.228	0.320***	4.272***	24.802***	1.918***	9.530***	6.955*	2.884***	2.654**	
	(2.151)	(0.102)	(0.896)	(3.746)	(0.153)	(1.528)	(3.761)	(0.511)	(1.175)	
rural	1.870**	0.864***	1.411***	3.874***	2.320***	2.626***	0.755	2.068***	2.966***	
	(0.875)	(0.051)	(0.303)	(1.035)	(0.086)	(0.453)	(1.630)	(0.135)	(0.541)	
lnalpha			2.517***			2.255***			1.893***	
			(0.081)			(0.121)			(0.112)	
N	1298	1298	1298	720	720	720	630	630	630	
Mean	3.220			2.465			3.914			
Predicted value	3.176	3.178	19.647	3.453	3.082	278.695	3.738	3.553	31.498	

Notes: The table displays coefficients with standard errors in parentheses.

^{*}p < 0.1 * *p < 0.05, *** p < 0.01.

Corruption

- UNDP and USAID biannual survey June 2010 to March 2020
- Focus on perception of corruption in local government, central government and international organizations
- The variables are coded as zero if the respondent perceives the institution to have no or minimal corruption, and as one if the institution is perceived to experience moderate to severe corruption.
- Wastelands can be of two types: internally funded or externally funded
- X are individual controls such as ethnicity, employment, age, schooling years, marital status and if they live in an urban or rural area.

corruption_institution_{it} =
$$\beta_0 + \beta_1 \text{high_bomb}_i + \beta_2 \text{wasteland}_i + \beta_3 (\text{high_bomb}_{it} \times \text{wasteland}_{it}) + X_{it} + \gamma_t + \epsilon_{it}$$

Corruption result

	corruption	$n_localgov$	corruption	.central_gov	$corruption_international_or$		
	(1)	(2)	(3)	(4)	(5)	(6)	
Panel A: 1.high_bomb							
high_bomb	-0.030***	-0.026**	0.022**	0.031***	0.107***	0.110***	
	(0.010)	(0.010)	(0.009)	(0.009)	(0.015)	(0.015)	
wastelansfill	-0.053***	-0.045***	-0.016**	-0.003	-0.074***	-0.066***	
	(0.009)	(0.009)	(0.008)	(0.008)	(0.013)	(0.013)	
1.high_bomb×wasteland	0.072***	0.074***	-0.031**	-0.032**	-0.067***	-0.068***	
_	(0.015)	(0.015)	(0.013)	(0.013)	(0.022)	(0.022)	
N	18967	18967	18108	18108	9996	9996	
Controls	X	X	X	X	X	X	
Year FE		X		X		X	
Panel B: extern vs intern fund							
high_bomb	-0.030***	-0.025**	0.022**	0.031***	0.110***	0.113***	
	(0.010)	(0.010)	(0.009)	(0.009)	(0.015)	(0.015)	
wl_ext_fund	-0.028***	-0.021**	-0.013	-0.001	-0.061***	-0.052***	
	(0.010)	(0.010)	(0.009)	(0.009)	(0.016)	(0.016)	
wl_{intern_fund}	-0.095***	-0.085***	-0.022*	-0.005	-0.099***	-0.091***	
	(0.013)	(0.013)	(0.012)	(0.012)	(0.020)	(0.020)	
$\begin{array}{c} \operatorname{high_bomb} \times \operatorname{wl_ext_fund} \end{array}$	0.045***	0.052***	-0.042***	-0.032**	-0.045*	-0.041	
	(0.017)	(0.017)	(0.016)	(0.015)	(0.027)	(0.027)	
${ m high_bomb}{ imes wl_intern_fund}$	0.115***	0.108***	-0.016	-0.031*	-0.076***	-0.081***	
	(0.019)	(0.019)	(0.018)	(0.018)	(0.029)	(0.028)	
N	18967	18967	18108	18108	9996	9996	
Controls	X	X	X	X	X	X	
Year FE		X		X		X	
outcome mean	0.698	0.698	0.793	0.793	0.603	0.603	

Corruption descriptive stats

		all			high bomb	b		low bomb			highbomb	& no wl	highbo	omb& wl	
Variable	Obs	Mean	Std.dev.	Obs	Mean	Std.dev.	Obs	Mean	Std.dev.	Obs	Mean	Std.dev.	Obs	Mean	Std.dev.
high_bomb	22,805	0.338084	0.473068	7,710	1	0	15,095	0	0	3,433	1	0	4,277	1	0
wasteland	23,663	0.386299	0.486911	7,710	0.554734	0.497027	15,095	0.322226	0.467345	3,433	0	0	4,277	1	0
rural	23,663	0.514854	0.49979	7,710	0.463684	0.498712	15,095	0.543889	0.498087	3,433	0.342849	0.474731	4,277	0.560673	0.496363
$ethn_alb$	23,657	0.664581	0.472147	7,708	0.68137	0.465975	15,091	0.681068	0.466078	3,433	0.794349	0.404235	4,275	0.590643	0.491773
$ethn_serb$	23,657	0.177072	0.381738	7,708	0.135963	0.342771	15,091	0.167385	0.373331	3,433	0.087387	0.282443	4,275	0.174971	0.379987
$ethn_other$	23,663	0.157419	0.364203	7,710	0.18249	0.386273	15,095	0.150315	0.357391	3,433	0.117973	0.322623	4,277	0.234276	0.423595
$empl_public$	23,510	0.130668	0.337044	7,694	0.131921	0.338427	14,983	0.121271	0.326453	3,431	0.146605	0.353763	4,263	0.120103	0.32512
$empl_private$	23,510	0.186814	0.389771	7,694	0.216142	0.411639	14,983	0.175332	0.380264	3,431	0.232294	0.422357	4,263	0.203143	0.402386
$empl_noregular$	23,510	0.044832	0.206939	7,694	0.037822	0.190777	14,983	0.045719	0.208881	3,431	0.031478	0.17463	4,263	0.042928	0.202718
$empl_pension$	23,510	0.089366	0.285278	7,694	0.087341	0.282352	14,983	0.09117	0.28786	3,431	0.074614	0.262806	4,263	0.097584	0.296786
$empl_homecare$	23,510	0.158954	0.365641	7,694	0.156746	0.363584	14,983	0.165521	0.371662	3,431	0.129117	0.335378	4,263	0.178982	0.383382
$empl_student$	23,510	0.082731	0.275481	7,694	0.0902	0.286487	14,983	0.078823	0.269471	3,431	0.097639	0.296869	4,263	0.084213	0.27774
$empl_other$	23,510	0.008082	0.089536	7,694	0.007668	0.087238	14,983	0.007809	0.088025	3,431	0.006995	0.083356	4,263	0.00821	0.090248
age	$23,\!563$	40.49315	15.94584	7,704	40.11397	15.8582	15,022	40.7927	16.03593	3,431	39.25823	15.59813	4,273	40.80108	16.03276
$schooling_years$	23,663	11.40126	3.571747	7,710	11.55824	3.449342	15,095	11.27605	3.605888	3,433	11.79085	3.387719	4,277	11.37152	3.487203
$marital_status$	23,541	2.447729	9.710182	7,693	2.156116	8.108545	15,001	2.488701	9.959894	3,422	2.118352	7.835548	4,271	2.186373	8.321604
income	22,641	105.9069	227.8784	7,495	98.35304	197.5032	14,361	111.7483	245.7264	3,282	117.1767	217.0476	4,213	83.68906	179.4835
male	23,663	0.510755	0.499895	7,710	0.511025	0.499911	15,095	0.50891	0.499937	3,433	0.525488	0.499423	4,277	0.499416	0.500058