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1 2 3 4 5 6 7 **Supplementary Information for** A comparison of political violence by left-wing, right-wing and Islamist 8 9 extremists in the United States and the world 10 11 Katarzyna Jasko 12 Institute of Psychology, Jagiellonian University, Poland 30-060 13 Gary LaFree 14 Department of Criminology and Criminal Justice, University of Maryland, College Park, MD, 15 20742 16 James Piazza 17 Department of Political Science, The Pennsylvania State University, University Park, PA 16801 18 Michael H. Becker 19 Department of Justice, Law, and Criminology, American University, Washington, DC, 20016 20 21 Katarzyna Jasko Email: katarzyna.jasko@uj.edu.pl 22 23 24 25 26 27 This PDF file includes: Tables S1 to S5 28 29 30 31 32

Table S1. Comparison of individuals representing different ideological groups.

Variable	Categories	Right-	Islamists	Left-	p (Chi-	Cramer'
		wing		wing	square	s V
					test)	
Education	Up to high school (1)	53%	42%	21%	<.001	.29
	College or vocational	40%	47%	58%		
	education (some or complete					
	degree; (2)					
	Post-graduate education	7%	11%	21%		
	(some or complete degree;					
	(3)					
Marital status	Yes (married)	62%	59%	76%	<.001	.14
	No	38%	41%	24%		
Immigrant	Yes	0.2%	45%	3%	<.001	.60
background						
	No	99.8%	55%	97%		
Military	Yes	23%	11%	10%	<.001	.18
experience						
	No	77%	89%	90%		
Gender	Male	94%	91%	75%	<.001	.26
	Female	6%	9%	25%		
Age ¹		37.56 ^a	31.22 ^b	29.68 ^b	<.001	
Ethnicity (white)	Yes	95%	9%	70%	<.001	.67

	No	5%	81%	30%		
Previous violent criminal experience	Yes	27%	17%	16%	<.001	.13
	No	73%	83%	84%		
Decades	40s-60s	5%	0%	17%	<.001	
	70s	4%	0%	26%		
	80s	12%	0%	8%		
	90s	21%	1%	10%		
	00s	22%	38%	21%		
	10s	36%	61%	18%		

³⁴ Theans with different superscripts are significantly different from each other.

Table S2. Logistic Regression Model (Study 1; Left-wing ideology as a reference category).

	b	SE	OR	p	b	SE	OR	p
(Intercept)	-0.72	0.11	0.49	<.001	-0.71	0.44	0.49	.111
Islamist ideology	1.20	0.17	3.32	<.001	2.15	0.31	8.55	<.001
Right-wing ideology	1.15	0.13	3.17	<.001	1.70	0.21	5.50	<.001
Education (2)					-0.08	0.52	0.92	.877
Education (3)					-0.56	0.47	0.57	.252
Marital status					-0.30	0.19	0.74	.116
Immigrant background					-0.68	0.29	0.51	.020
Military experience					-0.10	0.21	0.91	.643
Gender (male)					0.18	0.24	1.20	.449
Age					-0.01	0.01	0.99	.089
Ethnicity (white)					-0.43	0.21	0.65	.045
Previous violent criminal					1.08	0.21	2.95	<.001
experience								
60s					0.56	0.31	1.74	.072
70s					1.39	0.30	4.01	<.001
80s					0.94	0.28	2.56	.001
90s					0.24	0.22	1.27	.291
00s					-0.50	0.18	0.60	.006

Note: The reference category for these models are individuals who identified as left-wing, achieved a high school or less education (Education 1), were not married, not an immigrant, did not have any military experience, did not engage in any previous criminal violence, and whose exposure event occurred in the 2010's. Significance here is determined using a two-tailed t-test based upon 50 pooled samples with the multivariate imputation through chained equations method.

Table S3. Distribution of Right-wing, Left-wing, and Islamist Attacks Over Time (Study 2).

Decades		70s	80s	90s	00s	10s
Right-wing	N	232	2159	1055	273	416
ideology						
	% within ideology	5.6%	52.2%	25.5%	6.6%	10%
Left-wing	N	2 982	12 525	6 531	2 675	7 531
ideology		2 902	12 323	0 331	2073	/ 551
	% within ideology	9.2%	38.8%	20.3%	8.3%	23.4%
Islamist	N	70	244	2.160	4.044	20.004
ideology		50	344	2 168	4 944	28 094
	% within ideology	0.1%	1%	6.1%	13.9%	78.9%

44 Table S4. Analysis of Fatalities in Global Terrorism Database (Study 2; after excluding

45 generic groups from analyses)

			Zero-inflated n	egative binomial		
	Logistic	regression	model			
	Logistic regression		Conditional	Zero-inflation		
				model		
	b	OR	b	b		
Intercept	-0.39	0.67	-0.26	-2.13		
Left-wing ideology	-0.69	0.50***	-0.39	0.66***		
Islamist ideology	0.50	1.64*	1.27***	-17.62		
Random part						
Intercept Variance	1.29		1.79			
(groups)						
Intercept Variance	0	72	1	02		
(countries)	0.73		1.03			
Deviance	721	33.9	244917.7			
Observations	N: 60,547 Groups: 468 Countries: 111		N: 60,547			
			Groups: 468 Countries: 111			

p < .05; p < .01; p < .001

Table S5. Analysis of Injuries in Global Terrorism Database (Study 2)

Logistic regr	ression	mo Conditional model	del Zero-inflation	
			Zero-inflation	
b		model		
b			model	
	OR	b	b	
1.30	0.27	0.25	-0.36	
0.24	0.79	-0.39	-0.22*	
1.18	3.25***	1.36***	-17.67	
0.58		1.27		
75567.	8	208576.4		
N: 63,83	10	N: 63,810		
Groups: 523		Groups: 523		
Countries	:128	Countries:128		
	0.24 1.18 0.58 0.18 75567. N: 63,83 Groups: 3	0.24 0.79 1.18 3.25*** 0.58 0.18 75567.8 N: 63,810	0.24 0.79 -0.39 1.18 3.25*** 1.36*** 0.58 1.3 0.18 0.4 75567.8 2085 N: 63,810 N: 63 Groups: 523 Group	

p < .05; **p < .01; ***p < .001

Note: Due to the way fatalities and injuries are coded in GTD we consider fatalities as the primary variable. Specifically, if the source noted that the attack had resulted in fatalities but it did not mention any injuries, the variable "injuries" was coded as 0 (suggesting that no people were injured). Similarly, when the source noted that the attack had resulted in injuries but it did not mention any fatalities, "0" was assigned to fatalities. Given the different level of visibility and clarity as to who is counted as killed vs. injured, we think that it is less likely that the data miss real fatalities (and inaccurately assign "0") than it is that the data miss real injuries (and inaccurately assign "0"). In support of this interpretation, 37% of cases in GTD report no fatalities or injuries and 25% of cases report both fatalities and injured victims. However, among the inconsistent cases (i.e., fatalities but 0 injuries or injuries but 0 fatalities), 11% are attacks that

mention injuries but 0 fatalities, but 27% are attacks that mention fatalities but 0 injuries. Such selective attacks that result only in fatalities but no injuries seem to be less likely. In short, we suggest that the number of injuries might be underestimated (and the number of true zero injuries overestimated) and misrepresent the actual numbers more so in the case of injuries than fatalities.