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War makes the regime: regional rebellions and political militarization worldwide

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Online Appendix

Table A1: Summary statistics

	N	Mean	Sd	Min	Max
Military regime	8569	0.170	0.376	0	1
Military regime categorical	8569	0.241	0.560	0	2
Regional rebellion (decay)	16204	0.0442	0.199	0	1
Center-seeking rebellion (decay)	16150	0.0674	0.238	0	1
Previously experienced regional rebellion	16333	0.101	0.302	0	1
Previously experienced center-seeking rebellion	16333	0.226	0.418	0	1
Directly following regional rebellion	3834	0.0764	0.266	0	1
Directly following center-seeking rebellion	3834	0.184	0.388	0	1
Duration of regional rebellion	9619	0.891	5.057	0	66
Duration of ideological rebellion	9619	0.796	3.770	0	50
Regional rebellion (dummy)	9620	0.0681	0.252	0	1
Center-seeking rebellion (dummy)	9620	0.0984	0.298	0	1
GDP p.c. (log)	10416	8.365	1.096	5.7	11.5
Population size (log)	10119	15.22	2.070	9.2	21.0
Polity	8937	0.625	7.488	-10	10
Mil. capabilities (cube root)	8829	126.4	94.61	6.2	726.8
Unrest (log)	9705	0.356	0.676	0	4.5
War (dummy)	14600	0.0170	0.129	0	1
Rel. fractionalization	13277	0.442	0.257	0	0.9
Rents p.c. (log)	9335	2.815	3.013	0	11.4
Population density	12936	226.2	1395.1	0.5	25322.8
Trade	7556	76.46	48.32	0.3	531.7
Post Cold War	9207	0.391	0.488	0	1

Table A2: Countries included in main model

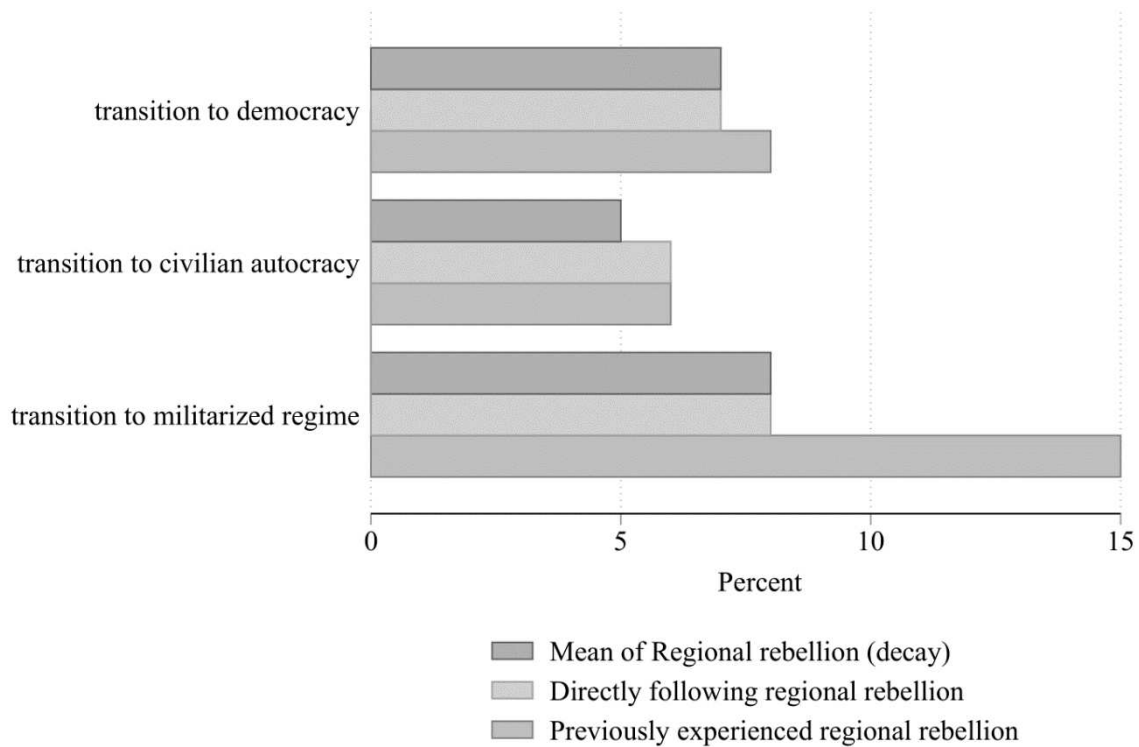
Country	N	Percent	Country	N	Percent
Afghanistan	25	0.53	Latvia	17	0.36
Albania	48	1.02	Lebanon	17	0.36
Algeria	9	0.19	Lesotho	34	0.72
Angola	32	0.68	Liberia	33	0.7
Argentina	33	0.7	Libya	9	0.19
Australia	48	1.02	Lithuania	17	0.36
Austria	48	1.02	Luxembourg	48	1.02
Bahrain	37	0.78	Macedonia	16	0.34
Bangladesh	35	0.74	Madagascar	45	0.95
Belarus	17	0.36	Malawi	44	0.93
Belgium	48	1.02	Malaysia	48	1.02
Benin	23	0.49	Mali	25	0.53
Bhutan	38	0.8	Mauritania	21	0.44
Bolivia	31	0.66	Mauritius	40	0.85
Botswana	42	0.89	Mexico	48	1.02
Brazil	27	0.57	Moldova	17	0.36
Bulgaria	48	1.02	Morocco	48	1.02
Burkina Faso	8	0.17	Mozambique	21	0.44
Burundi	9	0.19	Myanmar	2	0.04
Cambodia	29	0.61	Nepal	43	0.91
Cameroon	48	1.02	Netherlands	48	1.02
Cape Verde	33	0.7	New Zealand	48	1.02
Central African Republic	18	0.38	Nicaragua	29	0.61
Chad	22	0.47	Niger	26	0.55
Chile	31	0.66	Nigeria	11	0.23
China	48	1.02	Norway	48	1.02
Colombia	43	0.91	Oman	48	1.02
Comoros	27	0.57	Pakistan	12	0.25
Congo	13	0.28	Panama	28	0.59
Costa Rica	48	1.02	Papua New Guinea	33	0.7
Croatia	17	0.36	Paraguay	15	0.32
Cuba	48	1.02	Peru	27	0.57
Cyprus	37	0.78	Philippines	46	0.97
Czech Republic	15	0.32	Poland	38	0.8
Democratic Republic of the Congo	7	0.15	Portugal	45	0.95
Denmark	48	1.02	Qatar	37	0.78
Djibouti	31	0.66	Romania	47	0.99
Dominican Republic	43	0.91	Russia	16	0.34
East Timor	6	0.13	Rwanda	12	0.25
Ecuador	36	0.76	Saudi Arabia	48	1.02
Egypt	27	0.57	Senegal	48	1.02
El Salvador	22	0.47	Sierra Leone	34	0.72
Equatorial Guinea	11	0.23	Singapore	43	0.91

Estonia	17	0.36	Slovakia	15	0.32
Ethiopia	15	0.32	Slovenia	17	0.36
Fiji	24	0.51	Solomon Islands	29	0.61
Finland	48	1.02	Somalia	10	0.21
France	48	1.02	South Africa	47	0.99
Gabon	48	1.02	South Korea	20	0.42
Gambia	29	0.61	Spain	31	0.66
Germany	18	0.38	Sri Lanka	33	0.7
Ghana	26	0.55	Sudan	2	0.04
Greece	41	0.87	Suriname	24	0.51
Guatemala	23	0.49	Swaziland	40	0.85
Guinea	24	0.51	Sweden	48	1.02
Guinea-Bissau	18	0.38	Switzerland	48	1.02
Guyana	38	0.8	Syria	12	0.25
Haiti	43	0.91	Tanzania	47	0.99
Honduras	30	0.63	Thailand	30	0.63
Hungary	47	0.99	Togo	9	0.19
			Trinidad and		
India	48	1.02	Tobago	46	0.97
Indonesia	15	0.32	Tunisia	48	1.02
Iran	48	1.02	Turkey	41	0.87
Iraq	23	0.49	Uganda	31	0.66
Ireland	48	1.02	Ukraine	17	0.36
			United Arab		
Israel	48	1.02	Emirates	37	0.78
Italy	48	1.02	United Kingdom	48	1.02
Ivory Coast	39	0.83	Uruguay	36	0.76
Jamaica	47	0.99	Venezuela	48	1.02
Japan	48	1.02	Vietnam	17	0.36
Jordan	48	1.02	Yugoslavia	2	0.04
Kenya	45	0.95	Zambia	44	0.93
Kuwait	40	0.85	Zimbabwe	37	0.78
Laos	36	0.76			

Descriptive patterns for H1 and H2

In addition to presenting the findings from our quantitative models, we explore if our hypotheses are supported by patterns in descriptive data. Figure A1 displays the relative frequency of regional rebellions for three types of regime transitions: transitions toward military regimes, civilian autocracies, and democracies. In the case of transitions to military regimes, the mean of *Regional rebellion (decay)* over the measurement period is 15 and 25 percent higher than for transitions to civilian autocracy and democracy respectively, suggesting that military regimes emerge more often from regional rebellions; second, more transitions to military regimes are directly preceded by a regional rebellion than transitions to other regimes; third, about 15 percent of all transitions to military regimes occur in countries which at some point experienced a regional rebellion, while the equivalent shares in transitions to civilian autocracy and democracy lie around 6 and 8 percent respectively. Taken together, the data suggest that, in line with H1, regime transitions in the wake of regional rebellions are more likely to yield military rule.

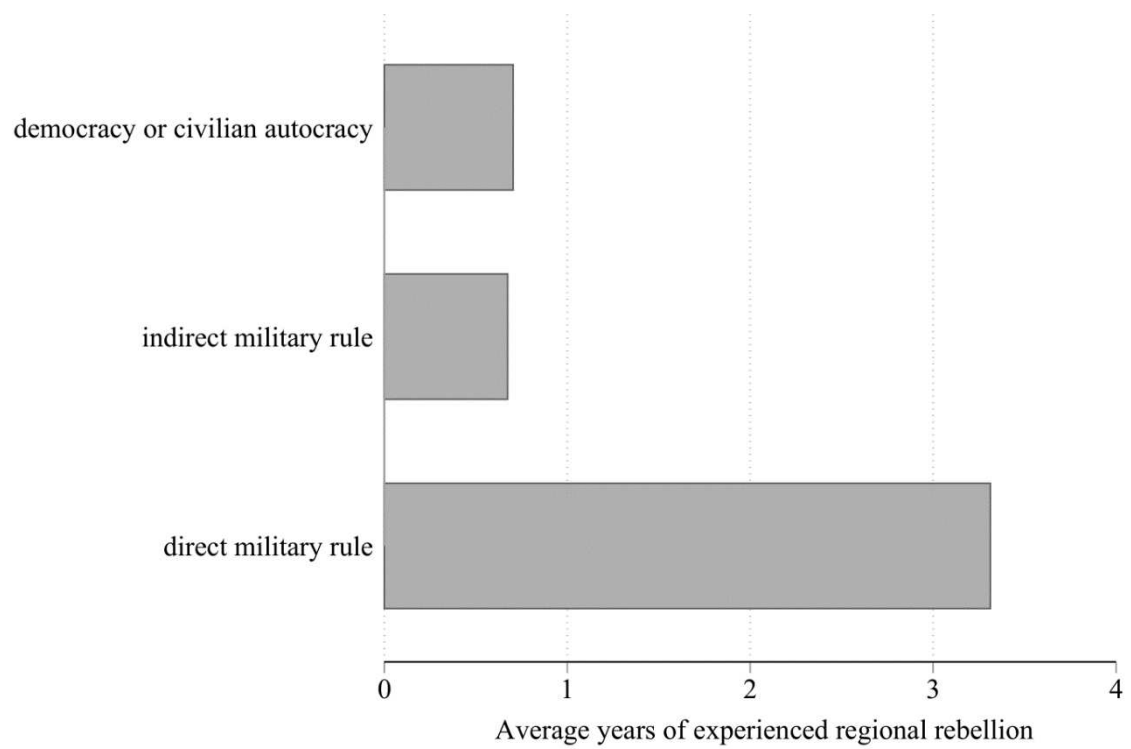
Figure A1: Regional rebellions and regime transitions



Note: “Previously experienced regional rebellions” means occurrence of a regional rebellion at some point in the past; “directly following regional rebellions” means a regional rebellion occurred in the previous year. Data on regime types taken from Svobik (2012); rebellion data from Gleditsch et al. (2002).

Figure A2 presents descriptive evidence for our second hypothesis on the intensity of regional conflict and the level of regime militarization. It shows that in regimes under direct military rule the average years of experienced regional rebellions – the mean of our duration measure – is more than three times higher than in democracies, civilian autocracies, or regimes under indirect military rules. This is a first indication that more intense regional conflicts leads to greater militarization of regimes in the form of direct military rule.

Figure A2: Intensity of regional conflict and regime militarization



Note: Data on regime types taken from Svobik (2012); rebellion data from Gleditsch et al. (2002).

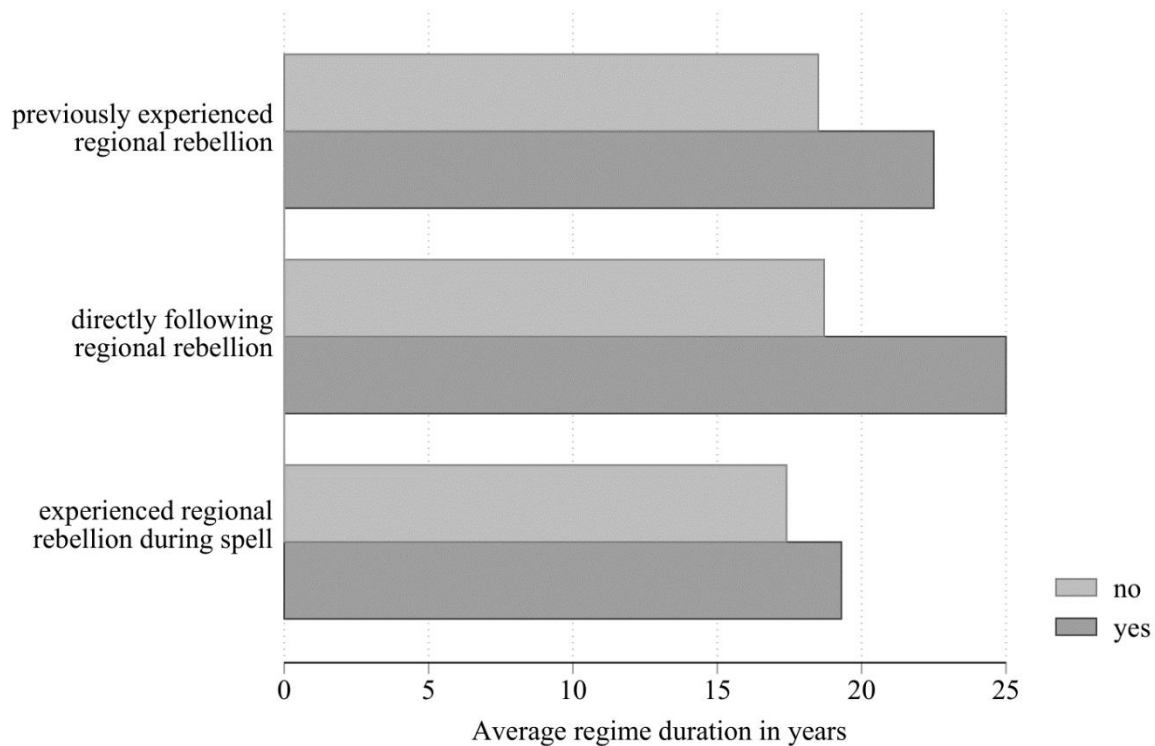
Modelling military regime democratization

While military regimes have, on average, had shorter spells than other regime types (Geddes 1999a), there is actually considerable variation across military regimes, for which our theory provides a potential explanation (Croissant and Kuehn 2016; Kim and Kroeger 2018). Our third hypothesis is thus as follows: Our theoretical argument implies that regional rebellions prompt militaries to hold on to power longer rather than handing it back quickly to civilian elites as their motivation and capacity to do so is stronger. Specifically, we hypothesize the following:

H3: Military regimes arising in the wake of regional rebellions or exposed to regional rebellions during their spell should withstand pressures for democratization longer than other military regimes.

In terms of descriptive statistics, Figure A3 depicts the average duration of military regimes in years differentiated by whether they experienced regional rebellions. In line with our expectations in H3, the average duration of all military regimes that have experienced some form of regional rebellion is noticeably higher: military regimes that previously experienced a regional rebellion last on average four years longer; regimes that directly emerge in the wake of a regional rebellion last over six years longer; and military regimes facing a regional rebellion during their spell last about two years longer. On the whole, the patterns emerging from the descriptive statistics are in line with our theory and inspire confidence in our theory.

Figure A3: Regional rebellions and military regime duration



Note: “Previously experienced regional rebellions” means occurrence of a regional rebellion at some point in the past; “directly following regional rebellions” means a regional rebellion occurred in the previous year; “experienced regional rebellion during spell” means at least one regional rebellion occurred during the regime spell. Data on regime types taken from Svobik 2012; rebellion data from Gleditsch et al. 2002.

To model the effect of regional rebellions on military regimes’ ability to withstand democratization (H3) we use a variant of our Markov transition model presented above. Based on the theory-generating Southeast Asian cases, we hypothesize two scenarios for military regimes to become more resistant to democracy through regional rebellion: first, military regimes that either directly emerge out of a regional rebellion – that is, experience one in the year preceding regime onset – or that were preceded by regional rebellions at some point in the past become more resilient in the face of democratizing forces as the preceding conflict has galvanized military elites into a cohesive ruling group; second, military regimes that experience

regional rebellions during their spell should also be more resistant to democracy as the cohesion of the ruling group is enhanced by conflict. Note that in the latter case military regimes could have come to power for reasons other than regional rebellions but then subsequently become more resilient as a result of regional rebellion.

To capture the foundational effect of preceding regional rebellions, we create two new variables, henceforth called origin variables: *Previously experienced regional rebellion* is a binary variable taking the value of 1 if a military regime was preceded by a regional rebellion at some point before its onset; *Directly following regional rebellion* is also a binary variable indicating whether or not a military regime was preceded by a regional rebellion in the period immediately prior to regime onset. Both variables are time-invariant for the duration of a military regime spell. The effect of rebellions occurring during a spell are captured by our standard decay variables. In terms of the model specification, we follow the setup of the Markov transition model with the exception that we now model the transition from a military regime to democracy. The equation is as follows:

$$\Pr(Democracy_{i,t} = 1 | Mil. regime_{i,t-1} = 1) = \text{Logit}(\alpha_0 + \beta_1 R_{i,t-1} + \beta_2 X_{i,t-1} + \delta_3 + \beta_4 \vartheta_{i,t} + \mu_{i,t}), (3)$$

with α designating the constant, R our measures of regional rebellions (origin and during spell), X a vector of covariates, and δ regional fixed effects. The event counter ϑ measures the number of previous regime breakdowns, while μ designates cubic polynomials of the time since military regime onset. As before, we lag all variables except for the event counter and the regional fixed effects,¹ and estimate a pooled as well as a fixed effects logit

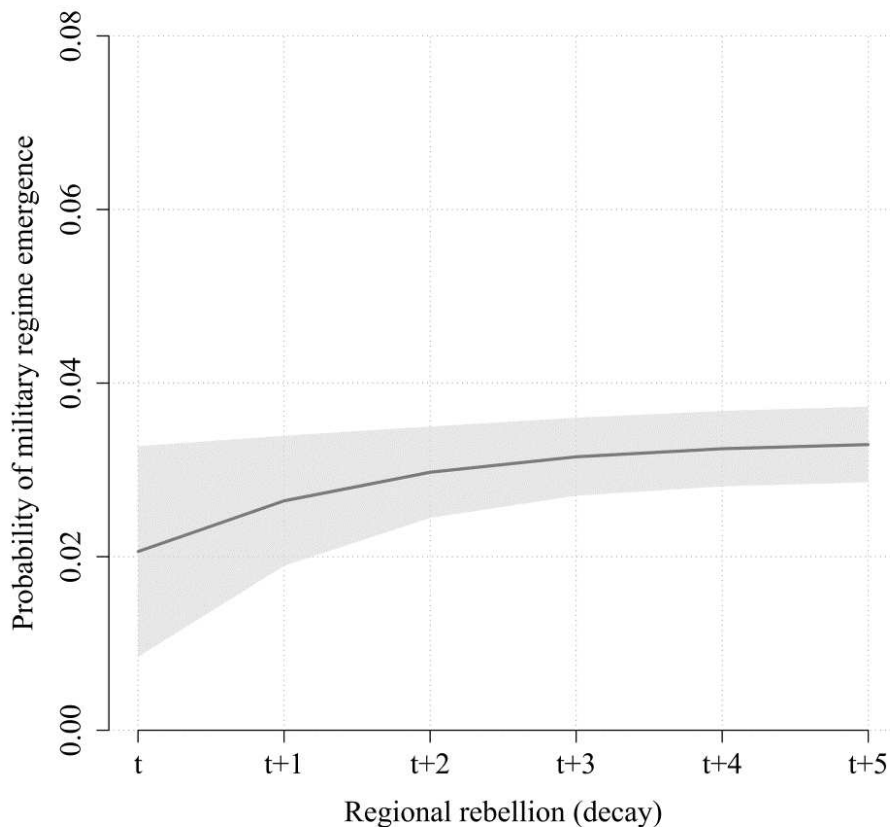
¹ Although our origin variables are time-invariant, we use the one-year lag to bring them in line with the lag structure of the model. Using the contemporaneous version of these variables yields nearly identical results.

model. In the latter, the origin variables are not estimated as they are time-variant so that the model uniquely focuses on the effect of regional rebellions occurring during a military spell.²

Our regressions results are displayed in Table A4. Columns 1 and 2 report the pooled logit estimates of the origin variables, while column 3 shows the estimates for rebellions occurring during a regime spell only. Columns 4 and 5 show the results when both measures of rebellion are added simultaneously into the regression. The conditional logit results are shown in Column 6. On the whole, the findings partially support H3. On the one hand, rebellions occurring during regime spells significantly reduce the likelihood of a transition to democracy. Given a baseline probability of democratization of 3.3 percent in the pooled model, the occurrence of a regional rebellions reduces this probability on average by 1.5 percent absolutely and 45 percent relatively speaking. The effect then fades out over the five following periods (see Figure A4). The conditional logit model which focuses solely on within regime variation yields a similarly, slightly stronger finding as significance levels increase.

² We also tested another variant of the Markov model where we keep *Military regime* as our dependent variable and condition on $Military\ regime_{t-1} = 1$. While this model slightly shifts the focus from democratization to military regime breakdown more generally, the main finding is the same in that regional rebellions during regime spells increase the durability of military regimes. Results are available upon request.

Figure A4: Effect of regional rebellion on military regime democratization



Note: Marginal effects derived from Table A4, column 5. *Regional rebellion* set at 1, 0.5, 0.25, 0.125, and 0.0625 according to decay function. All other covariates set at their observed values (Hanmer and Ozan Kalkan 2013).

On the other hand, our expectation that regimes emerging out of, or in contexts preceded by regional rebellions are more stable, does not bear out. In fact, all signs are positive except in column 1. The results of column 5 suggest that regimes directly following a regional rebellion are more likely to break down than regimes without previous rebellions. This finding is somewhat surprising in light of the suggestive descriptive statistics presented at the outset of this section. We see two possible reasons for this: First, with a total of three cases (Indonesia, Mauritania, Myanmar), military regimes arising directly in the wake of regional rebellion are relatively rare in our data, and while their spell is comparatively long (32, 26, and 47 years

respectively), the model might struggle to produce clear results from so few positive observations. Second, a number of regimes are double-coded as having both regional and center-seeking rebellions prior to their onset (see Table A3). This is, amongst others, the case for Myanmar – one of the most long-lasting military dictatorships. As a result, differentiating the effect of center-seeking and regional rebellions at the origin of military regimes is complicated because of tied data.³ Conversely, given that these tied data bias the results against our hypotheses more generally, the fact that in all our models *Regional decay* has been highly significant makes us more confident in the validity of our findings.

Table A3: Regimes with both types of rebellion

<i>Previously experienced regional AND center-seeking rebellion</i>	<i>Directly following regional AND center-seeking rebellion</i>
Comoros (3)	Myanmar (46)
D. R. Congo (26, 3)	
Ethiopia (15, 2)	
Indonesia (32)	
Myanmar (45)	
Niger (1)	
Nigeria (12, 14)	
Pakistan (8)	

Note: Duration of military regime spell in brackets.

³ In additional tests, we experimented with recoding all tied origin variables as regional rebellions only. This removes the positive significant effect of the regional origin variable but still does not show a negative significant effect on democratization. Results are available upon request.

Table A4: Regional rebellions and military regime democratization

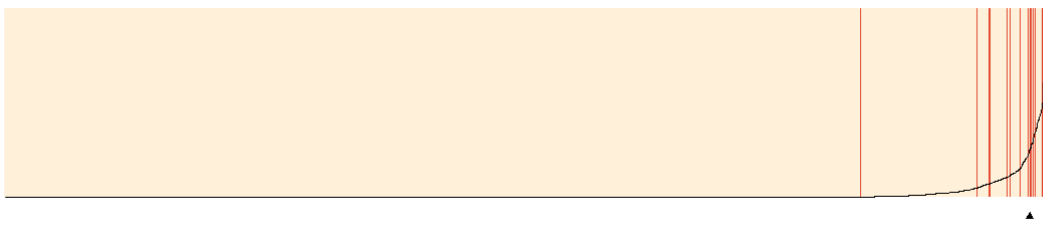
	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.759 (1.190)			2.161 (1.816)		
Previously experienced center-seeking rebellion t-1	0.844 (0.861)			-0.018 (1.378)		
Directly following regional rebellion t-1		0.650 (0.879)			2.212 (0.965)**	
Directly following center-seeking rebellion t-1		0.029 (1.483)			-0.775 (1.343)	
Regional rebellion (decay) t-1			-2.613 (1.496)*	-3.884 (1.858)**	-1.808 (0.975)*	-24.995 (6.222)***
Center-seeking rebellion (decay) t-1			1.713 (1.006)*	1.930 (1.513)	1.625 (1.131)	-1.636 (2.287)
GDP p.c. (log) t-1	0.088 (0.807)	0.497 (1.146)	0.471 (0.973)	0.165 (0.765)	0.558 (0.964)	-1.891 (8.406)
Population size (log) t-1	-1.501 (1.380)	-0.628 (1.180)	-0.881 (1.247)	-1.361 (1.211)	-0.382 (1.002)	9.899 (6.658)
Polity t-1	0.814 (0.099)***	0.786 (0.110)***	0.829 (0.119)***	0.854 (0.134)***	0.802 (0.103)***	1.528 (0.678)**
Mil. capabilities (cube root) t-1	34.619 (29.913)	14.964 (23.793)	22.660 (26.368)	30.100 (25.568)	9.588 (21.430)	365.337 (347.806)
Unrest (log) t-1	-0.184 (0.595)	-0.684 (0.661)	-0.187 (0.586)	-0.139 (0.552)	-0.678 (0.554)	-3.729 (3.063)
War (dummy) t-1	-1.963 (1.387)	-2.434 (1.410)*	-2.892 (1.486)*	-2.906 (1.578)*	-3.342 (1.470)**	-3.161 (2.580)
Rel. fractionalization t-1	4.988 (2.925)*	5.747 (2.132)***	5.961 (2.871)**	5.741 (2.865)**	6.710 (2.426)***	
Number of previous regime failures	0.697 (0.385)*	0.805 (0.419)*	0.671 (0.328)**	0.767 (0.459)*	0.911 (0.370)**	-2.812 (1.452)*
<i>N</i>	1,166	1,023	1,166	1,166	1,023	471
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$.

Separation plots to assess model fit

The two below figures show separation plots for our two baseline models. Separation plots arrange the fitted values in ascending order of probability with each line representing one observation (Greenhill, Ward, and Sacks 2011). Red lines are observations in which the outcome of interest actually occurs. The added line represents the predicted probabilities estimated for each observation. The small triangle beneath the plot indicates the expected number of total events predicted. A perfectly fitted model would thus have red and yellow lines perfectly separated at the position of the triangle, with a sharp corresponding rise in the predicted probability line. A very bad model would see the occurrence of actual outcomes at even very low levels of predicted probability and a slow gradual increase of the probability line. Bearing this in mind, the graphs below demonstrate a good model fit with high levels of separation and a visible spike in the predicted probability at the triangle separation point.

Baseline model: Military regime emergence



Note: Figure based on column 4, Table 1 in the main article.

Baseline model: Military regime democratization



Note: Figure based on column 3, Table A4 in the Online Appendix.

General note on robustness tests

The below section includes the following tests: additional/alternative variables, diffusion effects, and alternative time specifications. The tests are carried out for our military regime emergence and democratization models. Wherever possible, these tests are also carried out for our multinomial logit model of regime militarization.

Mundlak-Chamberlain models

As an alternative to our conditional logistic fixed effects regressions, we estimate the effect of rebellions on the emergence and democratization of military regimes using the Mundlak-Chamberlain (Chamberlain 1982; Mundlak 1978) approach. The model takes the following functional form:

$$\Pr(Y_{i,t} = 1 | Y_{i,t-1} = 0) = \text{Logit}(\alpha_0 + \beta_1(R_{i,t-1} - \bar{R}_l) + \beta_2(X_{i,t-1} - \bar{X}_l) + \beta_3\bar{R}_l + \beta_4\bar{X}_l + \beta_5\bar{Y}_l + \delta_3 + \mu_{i,t}), \quad (1),$$

where α designates the constant, R our measure of regional rebellions, X a vector of covariates (including our event count variable), and δ regional fixed effects. By adding the country means and the yearly deviation from the country means, the model enables us to separate cross-country from within-country effects. Note that in Table A6 we refrain from adding the yearly deviation for the variables measuring the rebellion prior or at regime onset (*Previously experienced regional/center-seeking rebellion; directly following regional/center-seeking rebellion*) as these are highly collinear with our standard measure of *Regional/Center-seeking rebellion*. Additionally, we add the country mean of our dependent variable Y , which allows us to account for unit heterogeneity by, effectively, adding a varying intercept (Wright, Frantz, and Geddes 2015). We also add μ , a vector of cubic time polynomials to

account for duration dependence (Carter and Signorino 2010). As before, all variables except for regional fixed effects and the event counter are lagged by one year to mitigate endogeneity concerns and ensure the correct order of causal dependence.

The results for military regime emergence are shown in Table A5. In line with our baseline model, the model finds that within-country changes of regional rebellions significantly increase the likelihood of military regime emergence. The effect is illustrated in the marginal effects plot in Figure A5. By contrast, cross-country differences in the average level of regional rebellions are not associated with a higher likelihood of military regime emergence, suggesting that our finding is driven by within-country variation of regional rebellions over time.

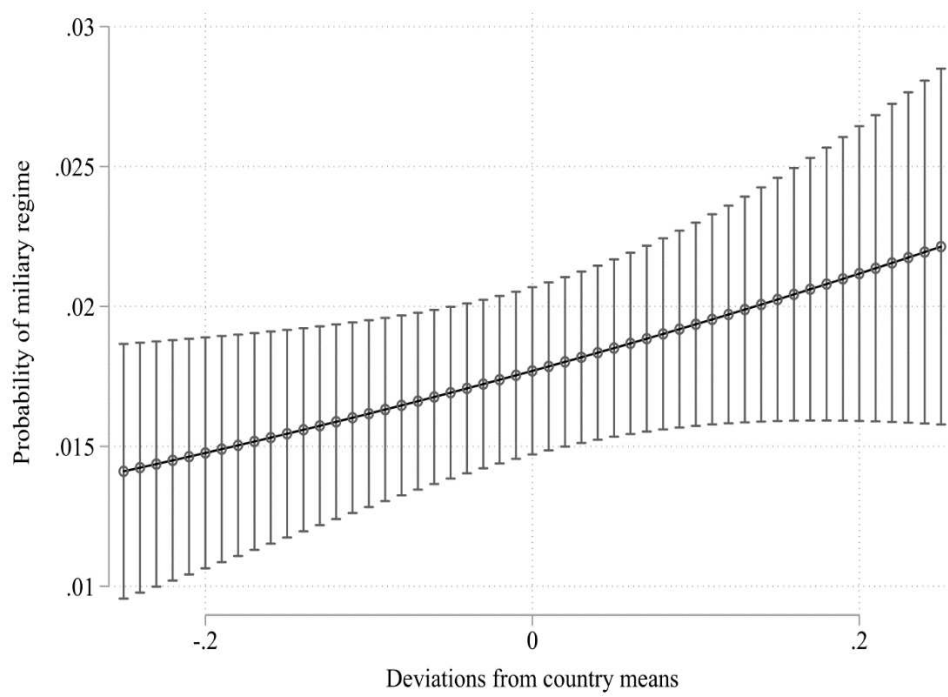
The results for military regime democratization are shown in Table A6. None of the variables for regional rebellions is significant. As we explain in the article, this is most probably due to tied data for regimes that experience both a regional and a center-seeking rebellion prior to regime onset.

Table A5: Mundlak-Chamberlain estimation of military regime emergence

	(1)	(2)
Regional rebellion (decay, mean)	-0.558 (0.979)	-0.682 (1.569)
Regional rebellion t-1 (decay, dev)	1.607 (0.762)**	1.308 (0.755)*
Center-seeking rebellion (decay, mean)	1.476 (0.671)**	0.496 (0.649)
Center-seeking rebellion t-1 (decay, dev)	0.391 (0.540)	0.077 (0.607)
GDP p.c. (log, mean)	-0.042 (0.286)	-0.218 (0.261)
GDP p.c. t-1 (log, dev)	-0.036 (0.533)	-0.071 (0.553)
Population size (log, mean)	0.201 (0.292)	0.292 (0.300)
Population size t-1 (log, dev)	2.661 (0.849)***	3.659 (0.876)***
Polity (mean)	0.043 (0.050)	0.162 (0.078)**
Polity t-1 (dev)	-0.265 (0.050)***	-0.365 (0.064)***
Mil. capabilities (cube root, mean)	-7.251 (8.429)	-5.096 (7.096)
Mil. capabilities t-1 (cube root, dev)	-16.044 (18.689)	-18.931 (19.885)
Unrest (log, mean)	0.881 (0.769)	1.008 (0.671)
Unrest t-1 (log, dev)	0.215 (0.203)	-0.012 (0.226)
War (dummy, mean)	4.432 (3.496)	3.892 (2.916)
War t-1 (dummy, dev)	0.746 (0.741)	-0.027 (0.886)
Rel. fractionalization (mean)	-0.793 (0.841)	-0.211 (0.725)
Previous failures (mean)	0.147 (0.029)***	-0.011 (0.043)
Previous failures (dev)	-0.191 (0.031)***	-0.128 (0.022)***
Y (mean)		8.353 (1.720)***
<i>NxT</i>	4,630	4,630
Time polynomials	Yes	Yes
Regional dummies	Yes	Yes

Mundlak-Chamberlain logistic regression with cluster-robust standard errors in parentheses. *Mean* designates the country averages, *dev* designates the annual deviation from country averages. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Figure A5: Effect of within-country changes in regional rebellions



Note: Marginal effects derived from Table A5, column 1. All other covariates set at their observed values (Hanmer and Ozan Kalkan 2013).

Table A6: Mundlak-Chamberlain estimation of military regime democratization

	(1)	(2)	(3)	(4)	(5)
Previously experienced regional rebellion (mean)	2.777 (3.403)			10.831 (9.747)	
Previously experienced center-seeking rebellion (mean)	-1.022 (1.256)			-5.283 (3.448)	
Directly following regional rebellion (mean)		0.901 (4.173)			-18.083 (32.346)
Directly following center-seeking rebellion (mean)		-1.847 (2.227)			-3.842 (5.472)
Regional rebellion (decay, mean)			-4.167 (8.185)	-18.157 (14.272)	8.392 (21.409)
Regional rebellion t-1 (decay, dev)			-30.648 (43.700)	-6.200 (13.656)	-71.368 (54.773)
Center-seeking rebellion (decay, mean)			11.036 (7.938)	20.118 (10.891)*	19.620 (9.216)**
Center-seeking rebellion t-1 (decay, dev)			2.886 (1.757)	3.987 (2.431)	2.456 (1.736)
Y (mean)	44.420 (40.531)	51.410 (47.585)	-30.814 (38.599)	-41.172 (48.098)	-93.795 (64.942)
<i>NxT</i>	1,166	1,151	1,166	1,166	1,151
Standard controls	Yes	Yes	Yes	Yes	Yes
Time polynomials	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes

Mundlak-Chamberlain logistic regression with cluster-robust standard errors in parentheses. *Mean* designates the country averages, *dev* designates the annual deviation from country averages. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Dummy instead of decaying variables for rebellion

To ascertain that our findings are not purely an artefact of our decaying measure of rebellions, we re-run our baseline models for military regime emergence and democratization using a simple dummy indicator of rebellion. The variable measures if a regional or center-seeking rebellion occurred in the previous year. The results are shown in Table A7 and Table A8.

As expected, constraining the effect of rebellions to one period weakens the association between regional rebellions and military regime emergence and democratization. That said, the pattern we have seen when using decaying variables reemerges when using simple dummies: In Table A7, only columns 4 and 5 are slightly insignificant, but with $p=0.105$ and 0.125 respectively reasonably close given that we are using a much coarser measure of regional rebellions. And in Table A8, we find a similar pattern to our baseline democratization model, suggesting that regional rebellions occurring during a military regime spell significantly decrease the likelihood of democratization. This is particularly the case when focusing on within-country variation only (Table A8, column 6), which is, as we know from Table A5, the main driver of our finding.

Table A7: Regional rebellions (dummy) and military regime emergence

	(1)	(2)	(3)	(4)	(5)
Regional rebellion (dummy) t-1	0.849 (0.457)*	0.859 (0.455)*	0.750 (0.455)*	0.709 (0.438)	1.421 (0.920)
Center-seeking rebellion (dummy) t-1	0.422 (0.336)	0.422 (0.336)	0.320 (0.317)	0.281 (0.324)	0.918 (0.610)
GDP p.c. (log) t-1	-0.166 (0.240)	-0.167 (0.239)	-0.179 (0.235)	0.001 (0.267)	-0.952 (0.852)
Population size (log) t-1	0.057 (0.126)	0.058 (0.123)	0.006 (0.122)	0.379 (0.213)*	2.455 (2.106)
Polity t-1	-0.125 (0.031)***	-0.126 (0.031)***	-0.140 (0.032)***	-0.152 (0.035)***	-0.408 (0.098)***
Previous failures	0.181 (0.211)	0.182 (0.212)	0.156 (0.235)	0.114 (0.249)	-0.926 (0.945)
War (dummy) t-1		-0.114 (0.766)	0.054 (0.746)	0.385 (0.734)	0.383 (1.053)
Unrest (log) t-1			0.483 (0.150)***	0.519 (0.153)***	-0.102 (0.306)
Rel. fractionalization t-1			-1.562 (0.692)**	-1.678 (0.746)**	
Mil. capabilities (cube root) t-1				-10.067 (5.041)**	-12.692 (32.218)
<i>NxT</i>	4,704	4,704	4,613	4,612	1,301
Time polynomials	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-4. Conditional logit model in column 5. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; ***

$p < 0.01$

Table A8: Regional rebellions (dummy) and military regime democratization

	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.759 (1.190)			1.575 (1.421)		
Previously experienced center-seeking rebellion t-1	0.844 (0.861)			0.611 (1.053)		
Directly following regional rebellion t-1		0.650 (0.879)			1.686 (0.983)*	
Directly following center-seeking rebellion t-1		0.029 (1.483)			-0.374 (1.582)	
Regional rebellion (dummy) t-1			-2.286 (1.491)	-3.422 (1.765)*	-1.233 (1.023)	-16.375 (2.434)***
Center-seeking rebellion (dummy) t-1			1.160 (1.116)	1.047 (1.364)	0.822 (1.338)	-3.039 (2.106)
GDP p.c. (log) t-1	0.088 (0.807)	0.497 (1.146)	0.388 (0.942)	0.209 (0.754)	0.490 (1.034)	-4.731 (7.567)
Population size (log) t-1	-1.501 (1.380)	-0.628 (1.180)	-0.949 (1.215)	-1.368 (1.268)	-0.523 (1.032)	7.758 (10.625)
Polity t-1	0.814 (0.099)***	0.786 (0.110)***	0.815 (0.109)***	0.821 (0.105)***	0.790 (0.101)***	1.989 (0.750)***
Mil. capabilities (cube root) t-1	34.619 (29.913)	14.964 (23.793)	23.762 (25.888)	30.289 (27.057)	12.450 (21.284)	481.046 (291.278)*
Unrest (log) t-1	-0.184 (0.595)	-0.684 (0.661)	-0.170 (0.599)	-0.131 (0.584)	-0.657 (0.607)	-4.846 (2.408)**
War (dummy) t-1	-1.963 (1.387)	-2.434 (1.410)*	-2.583 (1.567)*	-2.744 (1.582)*	-2.903 (1.568)*	-4.279 (2.346)*
Rel. fractionalization t-1	4.988 (2.925)*	5.747 (2.132)***	5.707 (2.746)**	5.148 (2.855)*	6.195 (2.242)***	
Number of previous regime failures	0.697 (0.385)*	0.805 (0.419)*	0.638 (0.318)**	0.763 (0.367)**	0.852 (0.412)**	-2.662 (1.451)*
<i>NxT</i>	1,166	1,023	1,166	1,166	1,023	471

Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Using coups as dependent variable

As we argue in the paper, the concept of militarization overlaps with but is not identical to military coups. To test the degree of overlap empirically, we rerun our baseline logit model using a binary indicator for coups instead of militarized regimes as our dependent variable. We test coup data from the Powell and Thyne (2011) (column 1) and Marshall and Marshall (2014) (column 2). We would expect the direction of the sign and strength of the sign to be similar. This is precisely what Table A9 shows. Like in our baseline model, the coefficient of *Regional rebellion* is positive; the association is slightly weaker when using the Powell and Thyne dataset with $p=0.15$.

Table A9: Regional rebellions and military coups

	(1)	(2)
Regional rebellion (decay) t-1	0.384 (0.264)	0.514 (0.260)**
Center-seeking rebellion (decay) t-1	0.238 (0.153)	0.297 (0.138)**
GDP p.c. (log) t-1	-0.358 (0.100)***	-0.397 (0.090)***
Population size (log) t-1	0.006 (0.109)	-0.069 (0.098)
Polity t-1	-0.007 (0.012)	-0.012 (0.011)
War (dummy) t-1	-0.099 (0.410)	0.169 (0.332)
Unrest (log) t-1	0.303 (0.106)***	0.336 (0.093)***
Rel. fractionalization t-1	-0.278 (0.299)	-0.071 (0.288)
Mil. capabilities (cube root) t-1	-1.573 (2.660)	-2.121 (2.621)
Number of previous coups	0.002 (0.019)	0.014 (0.021)
<i>NxT</i>	6,263	6,260
Time polynomials	Yes	Yes
Regional dummies	Yes	Yes

Pooled logit model with cluster-robust standard errors in parentheses. Column 1 uses the Powell and Thyne data; column 2 uses the Marshall and Marshall data * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Testing the effect of rebellion distance to capital

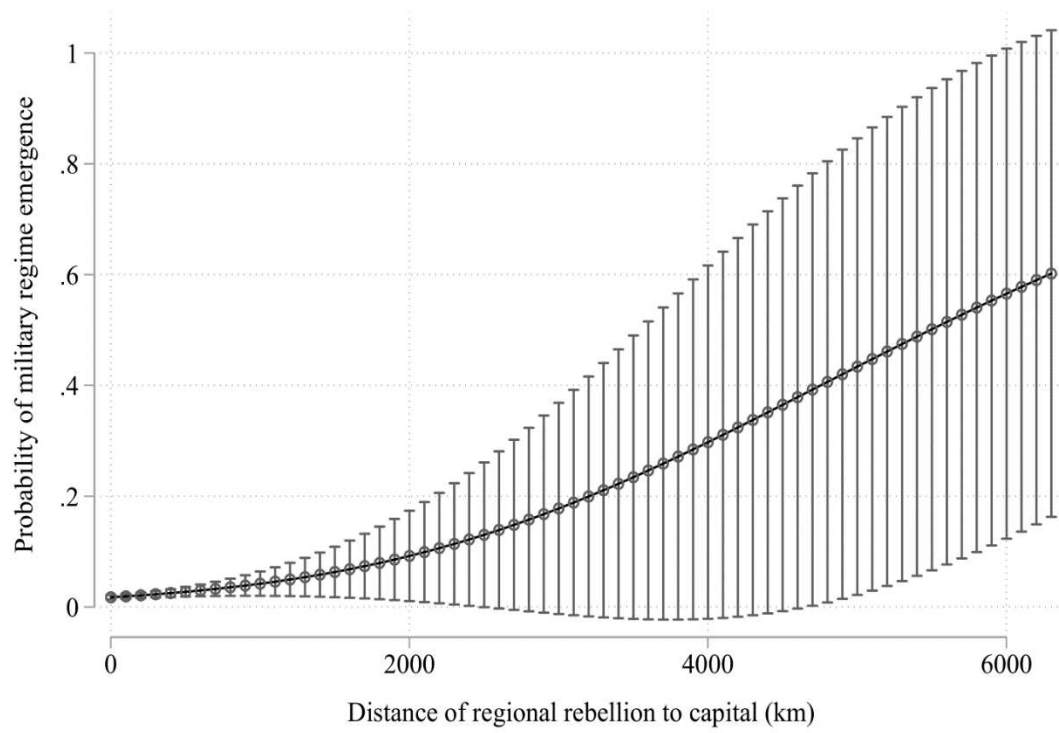
The proposed disalignment mechanism implies strong information asymmetries between civilian and military elites as a result of the distance between the regional rebellion and urban centers, especially the capital. An observable implication of this mechanism is that the effect of regional rebellions on the likelihood of military regime emergence should increase the farther the rebellion is away from the capital. We test this corollary using the geocoded version of the PRIO conflict dataset (1946-2005; Allansson, Melander, and Themnér 2017; Gleditsch et al. 2002). The new distance-weighted variables measure the distance of a rebellion from the capital city, with 0 indicating no rebellion, 1 a rebellion in the capital city, and further integers signifying larger distances from the capital. The results are shown in Table A10 below and strongly confirm our prior. The coefficient of *Regional rebellion* is strongly positively significant, meaning the farther a rebel a regional rebellion is away from the capital, the more likely the emergence of a military regime becomes. This effect is illustrated graphically in below. Interestingly, while the effect is only weakly significant, even center-seeking rebellion begin to exhibit a similar pattern once their distance to the capital city is taken into account and, as our theory predicts, information asymmetries become more salient.

Table A10: Rebellion distance and military regime emergence

	(1)
Regional rebellion (distance-weighted) t-1	0.001 (0.000)***
Center-seeking rebellion (distance-weighted) t-1	0.004 (0.002)*
GDP p.c. (log) t-1	-0.009 (0.280)
Population size (log) t-1	0.402 (0.211)*
Polity t-1	-0.160 (0.038)***
War (dummy) t-1	0.042 (0.733)
Unrest (log) t-1	0.488 (0.167)***
Rel. fractionalization t-1	-1.756 (0.786)**
Mil. capabilities (cube root) t-1	-10.661 (5.053)**
Previous failures	0.144 (0.234)
<i>NxT</i>	4,612
Time polynomials	Yes
Regional dummies	Yes

Pooled logit model with cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Figure A6: Capital distance of rebellions and militarized regimes



Note: Marginal effects derived from Table A10, column 1. All other covariates set at their observed values (Hanmer and Ozan Kalkan 2013).

Instrumental variable approach

A potential source of bias is endogeneity and reversed causality. Military repression, for example, might aggravate regional grievances, thus leading to regional rebellions rather than vice-versa. Indeed, Fjelde (2010) finds that military regimes are more likely to become involved in civil wars. Moreover, both regional rebellions and military regimes might be jointly associated with an uncontrolled omitted confounder. Depending on the strength of endogeneity and/or reverse causality, both scenarios could considerably bias our findings. The standard econometric approach to this problem has been to use instrumental variables. Finding instruments that are both sufficiently predictive and only influence the outcome Y through its effect on X (exclusion restriction) is often challenging, however (Bollen 2012; Sovey and Green 2011). In our case, this challenge is aggravated by the fact that we need to instrument for two endogenous predictors as both regional and center-seeking rebellions can plausibly be considered endogenous and/or in a reverse-causal relationship with military regimes.

We believe to have identified two instruments that fulfil the stringent conditions for instrumental variables: First, taking into account the emphasis the civil war literature has placed on the geographical conditions conducive to civil wars (e.g. Fearon and Laitin 2003), we turn to terrain characteristics as an instrument. Both center-seeking and separatist rebels will have to hide at times from government forces, which can be greatly facilitated or complicated by the shape of the terrain. The varying availability of hideouts is thus likely to play some role in the decision of whether or not to mount a rebellion. We therefore use *Rugged terrain in (thousands of) ha (log) t-1* taken from Nunn and Puga (2012) as our first instrument. Second, building on the insights from diffusion arguments of macro-social outcomes (Gleditsch and Ward 2006; Miller 2015), we use the number of *Battle-related deaths in neighboring countries t-1* as an instrument for regional and center-seeking

rebellion. Our logic relies on an implicit deterrence mechanism from large-scale human losses in neighboring countries on potential rebels. Neighborhood is defined as all contiguous states not separated by more than 24 miles of water.⁴ We use the Correlates of War Direct Contiguity dataset (Stinnett et al. 2002, v3.2) to construct these variables.

Could it be that that terrain characteristics and battle-related deaths in the neighborhood have a direct effect on military regime emergence or breakdown that is not mediated by domestic rebellion? Existing theory does not suggest this. Geography can plausibly be considered as exogenous. As for the second instrument, violence in neighboring countries could certainly increase the readiness of domestic military units in order to prevent spill-over effects; yet the mechanisms that would lead from, say, increased border patrols to a military takeover seem to be rather obscure.

Technically, we conduct our instrumental variable regression using a limited-information maximum likelihood estimator (LIML) as such estimators have been shown to be more robust to weak instrumental variable bias (Stock and Yogo 2005).⁵ Fortunately, all our F statistics meet the Stock-Yogo (2005) critical value on instrument strength. This suggests a rather strong predictive power of our instruments, so that any potential weak instrument bias will be limited. The results of the IV regressions are shown in Table A11. In line with our baseline model, regional rebellions remain a strong predictor for the emergence of military regimes. We do not recover, however, our previous finding regarding military regime democratization: while the sign of regional rebellion remains negative, the coefficient is far from statistical significance. This emphasizes the merely indicative nature of our finding on democratization at this stage.

⁴ This is one of the possible operationalizations of contiguity proposed in the dataset. Results are similar for alternative definitions of contiguity and are available upon request.

⁵ We use the *ivreg2* command in Stata to run these regressions.

Table A11: Regional conflict and military regime emergence/democratization (IV approach)

	(1)	(2)
<i>Second stage outcome</i>	Military regime <i>emergence</i>	Military regime <i>democratization</i>
Regional rebellion (decay) t-1	0.232** (0.112)	-0.0434 (0.180)
Center-seeking rebellion (decay) t-1	-0.201 (0.151)	0.0274 (0.0787)
<i>First stage outcome</i>	Regional rebellion decay t-1	Regional rebellion decay t-1
Battle-related deaths in neighboring countries t-1	-0.004*** (0.0015)	-0.0084** (0.004)
Rugged terrain in ha (log) t-1	0.0054 (0.0033)	0.0085 (0.0069)
<i>First stage outcome</i>	Center-seeking rebellion decay t-1	Center-seeking rebellion decay t-1
Battle-related deaths in neighboring countries t-1	0.0015 (0.0022)	0.0020 (0.0058)
Rugged terrain in ha (log) t-1	0.0065** (0.0026)	0.0441*** (0.0159)
<i>N x T</i> second stage	4,889	1,093
Time polynomials	Yes	Yes
Regional dummies	Yes	Yes
Standard controls	Yes	Yes
Cragg-Donald Wald F statistic	7.73	11.62
Stock-Yogo critical value	7.03	7.03

Two-stage instrumental variable regression with limited-information maximum likelihood estimator (LIML) and cluster-robust standard errors in parentheses. The Cragg-Donald Wald F statistic and the Stock-Yogo critical values are taken from a model assuming conditional homoscedasticity as the Stock-Yogo test is only valid in the presence of i.i.d. errors. * p<0.1; ** p<0.05; *** p<0.01

Additional control variables

As it is standard practice, we test the robustness of our baseline models to the addition of further control variables. Specifically, we add *Oil rents (log)* (Ross 2013), *Population density* (UNDP 2015), *Trade/GDP* (World Bank 2017), a dummy for the post-Cold War period (*Post Cold War*), and a measure of the prior regime type (*Prior regime*) (Svolik 2012) to the regression. The latter is a categorical variable measuring if a regime is autocratic, democratic, or there was no authority. These additional variables aim to take into account that (i) military regimes might be more likely to emerge in resource-abundant countries; (ii) civilian involvement in counter-insurgency might be more likely in more densely populated areas, decreasing the likelihood of military regime emergence; (iii) more open economies might exhibit more volatile business cycles, ensuing instability and potentially military take-overs or regime breakdowns; (iv) military regimes might be less likely to emerge in the post-Cold War era; and (v) certain regime types might be more likely to transition into military rule. We run all these regressions using *Ethnic fractionalization* instead of religious fractionalization as control variable to make sure our findings are robust to ethnic diversity (Alesina et al. 2003).

In the light of the results displayed in Table A12-Table A14, our main findings appear practically unchanged by these additional variables. Regional rebellions remain a highly significant predictor of military take-overs; and regional rebellions occurring during a regime spell tend to decrease the likelihood of military regime democratization. The latter finding seems to be particularly driven by within-country variation (see the conditional logit model in Table A14, column 6). As before, our expectation that regimes emerging out of or in contexts of regional rebellions is not supported by the model, as shown by the positive significant effect for *Previously experienced regional rebellion* and *Directly following regional rebellion*. As we discussed in the article, while we are puzzled by these findings in

the light of strong descriptive patterns, the absence of a clear finding might have to do with the effect of ties in our data.

As for the multinomial logit model on regime militarization (Table A13), the effect of rebellion intensity remains highly significant in the unrestricted sample (column 1). In the restricted sample (column 2) for military regimes only, the coefficient of *Duration of regional rebellion* becomes slightly insignificant ($p=0.155$). At the same, we lose about 150 observations in this regression compared to our baseline model, which is likely to be driving the loss of significance.

Table A12: Regional rebellions and military regime emergence (additional controls)

	(1)	(2)	(3)	(4)	(5)	(6)
Regional rebellion (decay) t-1	0.923 (0.420)**	0.950 (0.416)**	1.159 (0.478)**	1.129 (0.488)**	0.913 (0.536)*	2.041 (1.022)**
Center-seeking rebellion (decay) t-1	0.171 (0.353)	0.200 (0.354)	0.256 (0.379)	0.293 (0.380)	0.147 (0.428)	1.366 (0.943)
Ethnic fractionalization t-1	1.175 (0.738)	0.736 (0.742)	0.665 (0.772)	0.698 (0.773)	1.099 (0.732)	
Oil rents (log) t-1	-0.082 (0.061)	-0.088 (0.062)	-0.043 (0.071)	-0.041 (0.070)	-0.064 (0.068)	0.372 (0.396)
Pop. density t-1		-0.003 (0.002)	-0.003 (0.002)	-0.003 (0.002)	-0.002 (0.002)	-0.039 (0.053)
Trade/GDP t-1			-0.009 (0.008)	-0.007 (0.008)	-0.006 (0.008)	0.010 (0.013)
Post Cold War				-0.408 (0.355)	-0.524 (0.365)	-0.540 (1.062)
Prior regime: Autocracy t-1 (<i>baseline cat.: democracy</i>)					-1.434 (0.713)**	0.256 (0.953)
Prior regime: No authority t-1 (<i>baseline cat.: democracy</i>)					-0.819 (0.997)	-0.516 (1.615)
<i>NxT</i>	4,605	4,605	4,092	4,092	4,092	1,099
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A13: Regional rebellions and regime militarization (additional controls)

		(1)	(2)
<i>Indirect Military</i>	Duration of regional rebellion t-1	0.041 (0.043)	
	Duration of center-seeking rebellion t-1	0.018 (0.029)	
	Ethnic fractionalization t-1	-0.027 (0.632)	
	Oil rents (log) t-1	-0.003 (0.057)	
	Pop. density t-1	-0.000 (0.001)	
	Trade/GDP t-1	0.003 (0.004)	
	Post Cold War	0.191 (0.269)	
	Prior regime: Autocracy t-1 (baseline cat.: democracy)	0.165 (0.355)	
	Prior regime: No authority t-1 (baseline cat.: democracy)	-1.234 (0.588)**	
<i>Direct Military</i>	Duration of regional rebellion t-1	0.117 (0.028)***	0.073 (0.051)
	Duration of center-seeking rebellion t-1	-0.108 (0.034)***	-0.125 (0.049)**
	Ethnic fractionalization t-1	-0.430 (1.029)	-0.324 (1.303)
	Oil rents (log) t-1	-0.131 (0.088)	-0.141 (0.117)
	Pop. density t-1	-0.004 (0.003)	-0.005 (0.004)
	Trade/GDP t-1	-0.003 (0.005)	-0.006 (0.006)
	Post Cold War	-0.906 (0.443)**	-1.181 (0.538)**
	Prior regime: Autocracy t-1 (baseline cat.: democracy)	-0.042 (0.386)	-0.669 (0.456)
	Prior regime: No authority t-1 (baseline cat.: democracy)	-0.817 (0.704)	-0.103 (0.650)
<i>NxT</i>	5,355	1,057	
Time polynomials	Yes	Yes	
Regional dummies	No	No	
Mil. regime t-1=1	No	Yes	
Standard controls	Yes	Yes	

Multinomial logit regression with cluster-robust standard errors. Column 1 unrestricted sample, base category: non-military regimes. Column 2 restricted sample on military regimes only, base category: indirect military rule. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A14: Regional rebellions and military regime democratization (additional variables)

	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	4.324 (2.125)**			7.108 (2.027)***		
Previously experienced center-seeking rebellion t-1	0.368 (0.721)			-1.079 (0.972)		
Directly following regional rebellion t-1		9.346 (4.266)**			6.067 (1.875)***	
Directly following center-seeking rebellion t-1		3.213 (1.896)*			3.682 (2.057)*	
Regional rebellion (decay) t-1			-0.451 (1.296)	-3.883 (1.544)**	5.356 (5.679)	-97.412 (44.847)**
Center-seeking rebellion (decay) t-1			1.997 (1.578)	2.601 (2.153)	-1.628 (1.804)	-9.563 (7.867)
Ethnic fractionalization t-1	5.538 (3.747)	7.816 (4.158)*	4.679 (3.292)	4.846 (3.489)	9.420 (4.908)*	
Oil rents (log) t-1	-0.805 (0.324)**	-1.778 (0.786)**	-0.596 (0.291)**	-0.988 (0.352)***	-2.048 (0.916)**	-2.511 (1.526)*
Pop. density t-1	0.004 (0.009)	-0.002 (0.019)	0.002 (0.009)	0.004 (0.010)	-0.001 (0.018)	1.316 (0.722)*
Trade/GDP t-1	-0.022 (0.017)	-0.028 (0.010)***	-0.006 (0.012)	-0.030 (0.015)**	-0.045 (0.019)**	
Post Cold War	-1.354 (0.994)	-3.513 (2.303)	-2.148 (1.050)**	-0.858 (0.764)	-3.959 (2.762)	12.621 (7.103)*
<i>NxT</i>	1,029	900	1,029	1,029	900	471
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Reginal dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; ***

$p < 0.01$

Controlling for diffusion effects

There is ample evidence that democratization is driven, in part, by regional diffusion effects (e.g. Elkins and Simmons 2005; Gleditsch and Ward 2006; Patel, Bunce, and Wolchik 2014). Similar dynamics can be imagined for the emergence of specific authoritarian regime types, such as military regimes. To ensure that our findings are not driven by diffusion, we control for the *percentage of military regimes in neighborhood* in our emergence regressions, and the *percentage of democratic regimes in neighborhood* for our democratization models. Neighborhood is defined as all contiguous states not separated by more than 24 miles of water. We use the Correlates of War Direct Contiguity dataset to construct these variables (Stinnett et al. 2002, v3.2). The results of this robustness test, shown in Table A15-Table A17, are substantively identical to those from our baseline models.

Table A15: Regional rebellions and military regime emergence (diffusion effects)

	(1)	(2)
Regional rebellion (decay) t-1	0.861 (0.445)*	2.241 (0.986)**
Center-seeking rebellion (decay) t-1	-0.034 (0.328)	0.851 (0.702)
% mil. regimes in neighborhood (contiguous) t-1	1.487 (0.407)***	4.243 (1.209)***
GDP p.c. (log) t-1	-0.116 (0.298)	-0.781 (0.745)
Population size (log) t-1	0.062 (0.284)	3.500 (1.921)*
Polity t-1	-0.156 (0.036)***	-0.412 (0.110)***
War (dummy) t-1	0.324 (0.742)	0.375 (1.049)
Unrest (log) t-1	0.477 (0.170)***	-0.031 (0.285)
Rel. fractionalization t-1	-1.470 (0.739)**	
Mil. capabilities (cube root) t-1	-0.031 (8.033)	-13.701 (33.551)
Previous failures	0.084 (0.253)	-1.731 (0.822)**
<i>NxT</i>	4,292	1,205
Time polynomials	Yes	Yes
Regional dummies	Yes	Yes

Pooled logit model in column 1. Conditional logit model in column 2. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A16: Regional rebellions and regime militarization (diffusion effects)

		(1)	(2)
<i>Indirect Military</i>	Duration of regional rebellion t-1	0.008 (0.041)	
	Duration of center-seeking rebellion t-1	-0.011 (0.031)	
	% mil. regimes in neighborhood (contiguous) t-1	1.143 (0.431)***	
<i>Direct Military</i>	Duration of regional rebellion t-1	0.116 (0.027)***	0.116 (0.051)**
	Duration of center-seeking rebellion t-1	-0.142 (0.030)***	-0.143 (0.052)***
	% mil. regimes in neighborhood (contiguous) t-1	1.257 (0.476)***	0.053 (0.593)
<i>NxT</i>		5,355	1,057
Time polynomials		Yes	Yes
Regional dummies		No	No
Mil. regime t-1=1		No	Yes
Standard controls		Yes	Yes

Multinomial logit regression with cluster-robust standard errors. Column 1 unrestricted sample, base category: non-military regimes. Column 2 restricted sample on military regimes only, base category: indirect military rule. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A17: Regional rebellions and military regime democratization (diffusion effects)

	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.809 (2.069)			2.208 (2.460)		
Previously experienced center-seeking rebellion t-1	1.373 (0.999)			0.624 (1.595)		
Directly following regional rebellion t-1		0.240 (0.977)			1.854 (1.367)	
Directly following center-seeking rebellion t-1		0.115 (1.577)			-0.667 (1.461)	
Regional rebellion (decay) t-1			-2.640 (1.494)*	-4.084 (2.212)*	-1.583 (1.195)	-21.809 (3.192)***
Center-seeking rebellion (decay) t-1			1.832 (1.060)*	1.759 (1.488)	1.511 (1.423)	-1.682 (1.988)
% dem. regimes in neighborhood (contiguous) t-1	-1.074 (1.919)	1.307 (2.009)	-0.773 (1.839)	-1.037 (1.877)	0.134 (2.455)	10.784 (3.794)***
<i>NxT</i>	1,146	1,003	1,146	1,146	1,003	457
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; ***

$p < 0.01$

Alternative times specifications

We test the sensitivity of our baseline models to alternative time specifications in Table A18-Table A20 below. Specifically, we rerun the models using cubic splines, the log of time, and decade dummies with time polynomials. Our main findings remain substantively unchanged.

Table A18: Regional rebellions and military regime emergence (alternative time specifications)

	(1)	(2)	(3)	(4)	(5)	(6)
	<i>Logit</i>	<i>Clogit</i>	<i>Logit</i>	<i>Clogit</i>	<i>Logit</i>	<i>Clogit</i>
Regional rebellion (decay) t-1	0.940 (0.428)**	2.309 (1.004)**	0.931 (0.424)**	2.358 (0.910)***	0.913 (0.424)**	2.303 (1.197)*
Center-seeking rebellion (decay) t-1	0.324 (0.346)	1.147 (0.749)	0.205 (0.349)	1.049 (0.674)	0.392 (0.349)	0.916 (0.846)
GDP p.c. (log) t-1	-0.005 (0.269)	-0.876 (0.836)	0.016 (0.257)	-0.244 (0.750)	-0.012 (0.273)	-1.017 (0.953)
Population size (log) t-1	0.362 (0.212)*	2.820 (2.020)	0.380 (0.211)*	4.827 (1.977)**	0.371 (0.223)*	1.776 (2.269)
Polity t-1	-0.152 (0.036)***	-0.420 (0.096)***	-0.161 (0.035)***	-0.379 (0.083)***	-0.135 (0.037)***	-0.390 (0.104)***
War (dummy) t-1	0.379 (0.728)	0.368 (1.084)	0.319 (0.753)	0.363 (0.994)	0.306 (0.745)	0.197 (0.911)
Unrest (log) t-1	0.544 (0.154)***	-0.071 (0.308)	0.533 (0.153)***	-0.064 (0.274)	0.507 (0.155)***	-0.154 (0.299)
Rel. fractionalization t-1	-1.691 (0.747)**		-1.689 (0.751)**		-1.450 (0.752)*	
Mil. capabilities (cube root) t-1	-10.215 (5.011)**	-19.317 (29.699)	-10.535 (4.923)**	-32.703 (33.536)	-11.705 (5.220)**	-22.140 (37.913)
Previous failures	0.165 (0.242)	-1.001 (1.002)	0.104 (0.262)	-1.863 (0.938)**	0.463 (0.293)	-1.542 (0.942)
<i>NxT</i>	4,630	1,301	4,630	1,301	4,630	1,301
Cubic splines	Yes	Yes	No	No	No	No
Log of time	No	No	Yes	Yes	No	No
Decade dummies	No	No	No	No	Yes	Yes
Time polynomials	No	No	No	No	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1, 3, and 5. Conditional logit model in column 2, 4, and 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A19: Regional rebellions and regime militarization (alternative time specifications)

		(1)	(2)	(3)	(4)	(5)	(6)
<i>Indirect Military</i>	Duration of regional rebellion t-1	0.019 (0.046)	0.015 (0.042)	0.021 (0.036)			
	Duration of center-seeking rebellion t-1	-0.020 (0.033)	-0.019 (0.031)	-0.018 (0.029)			
<i>Direct Military</i>	Duration of regional rebellion t-1	0.119 (0.032)***	0.120 (0.025)***	0.112 (0.028)***	0.116 (0.050)**	0.116 (0.050)**	0.098 (0.037)***
	Duration of center-seeking rebellion t-1	-0.142 (0.029)***	-0.148 (0.029)***	-0.150 (0.027)***	-0.144 (0.051)***	-0.144 (0.050)***	-0.142 (0.041)***
<i>NxT</i>		6,052	6,052	6,052	1,200	1,200	1,200
Cubic splines		Yes	No	No	Yes	No	No
Log of time		No	Yes	No	No	Yes	No
Decade dummies		No	No	Yes	No	No	Yes
Time polynomials		No	No	Yes	No	No	Yes
Regional dummies		No	No	No	No	No	No
Mil. regime t-1=1		No	No	No	Yes	Yes	Yes
Standard controls		Yes	Yes	Yes	Yes	Yes	Yes

Multinomial logit regression with cluster-robust standard errors. Columns 1-3 unrestricted sample, base category: non-military regimes. Columns 4-6 restricted sample on military regimes only, base category: indirect military rule. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A20: Regional rebellions and military regime democratization (alternative time specifications)

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	<i>Logit</i>	<i>Logit</i>	<i>Clogit</i>	<i>Logit</i>	<i>Logit</i>	<i>Clogit</i>	<i>Logit</i>	<i>Logit</i>	<i>Clogit</i>
Prev. exp. regional rebellion t-1	2.485 (1.657)			1.486 (1.587)			2.701 (1.996)		
Prev. exp. center-seeking rebellion t-1	-0.106 (1.302)			-0.029 (1.446)			0.124 (0.984)		
Dir. following regional rebellion t-1		2.125 (0.946)**			2.184 (1.210)*			2.588 (1.401)*	
Dir. following center-seeking rebellion t-1		-0.354 (1.149)			-0.420 (1.205)			0.284 (1.381)	
Regional rebellion (decay) t-1	-3.963 (1.710)**	-1.387 (0.948)	-21.754 (3.244)***	-3.437 (1.872)*	-1.524 (1.071)	-25.277 (232.768)	-5.335 (2.854)*	-3.632 (1.726)**	-24.155 (8.595)***
Center-seeking rebellion (decay) t-1	1.924 (1.490)	1.305 (1.128)	-0.142 (1.363)	1.866 (1.546)	1.672 (1.137)	1.378 (1.157)	2.609 (1.664)	2.192 (1.312)*	2.146 (2.299)
<i>NxT</i>	1,166	1,023	471	1,166	1,023	471	1,166	1,023	478
Cubic splines	Yes	Yes	Yes	No	No	No	No	No	No
Log of time	No	No	No	Yes	Yes	Yes	No	No	No
Decade dummies	No	No	No	No	No	No	Yes	Yes	Yes
Time polynomials	No	No	No	No	No	No	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-2, 4-5, and 7-9. Conditional logit model in column 3, 6, and 9. Cluster-robust standard errors in parentheses. *

$p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Alternative measure for military regimes

The robustness tests below replicate our baseline models using an indicator of military regimes based on Geddes et al. (2014). Importantly, we depart from her schema in two ways. First, when a regime is classified by her as a hybrid regime, we recode the regime according to Lai and Slater (2006), who classify authoritarian regimes as one of four types: Strongman, Junta, Bossism, or Machine. We treat the first two types as instances of military regimes and the latter two types as party regimes. We make this departure from Geddes because her typology measures variation in regime type solely by how “access to office and the fruits of office” are allocated; that is, by military committee, party committee, or a single individual (Geddes 1999b, 121–122). Yet, authoritarian regimes should be differentiated by their executory powers and not just by their decision-making procedures. Doing so substantially reduces hybridization in the classification of all regime types.

Second, and by similar reasoning, we recode regimes classified as *purely personalist* by Geddes as *military* regimes, since almost all cases of personalist regimes rely on the military to maintain power and execute policy. Where Geddes sees a strong party apparatus accompanying personal rule, she classifies it as a party-personalist hybrid. This lends added confidence that Geddes’ purely personalist regimes lack a ruling party that rivals the power of the military.

The results of this robustness test are displayed in Table A21 and Table A22. Regarding military regime emergence, the findings using Geddes et al.’s data are nearly identical to our baseline model, both in the pooled and in the fixed effects regression. As for democratization, the results are much weaker using the alternative measure. While the coefficient of *Regional rebellion (decay)* is negative throughout, it only reaches levels of statistical significance in Table A22, column 4. This

underlines the fact that the results for military regime democratization should be viewed as suggestive, rather than conclusive.

Table A21: Regional rebellions and military regime emergence (GWF data)

	(1)	(2)	(3)	(4)	(5)
Regional rebellion (decay) t-1	1.029 (0.430)**	1.023 (0.429)**	0.838 (0.450)*	0.937 (0.444)**	1.944 (0.757)**
Center-seeking rebellion (decay) t-1	0.701 (0.316)**	0.699 (0.317)**	0.651 (0.325)**	0.710 (0.323)**	0.857 (0.567)
GDP p.c. (log) t-1	-0.431 (0.201)**	-0.435 (0.200)**	-0.606 (0.237)**	-0.451 (0.285)	-2.674 (1.331)**
Population size (log) t-1	0.043 (0.134)	0.050 (0.134)	-0.108 (0.148)	0.187 (0.330)	2.928 (1.800)
Polity t-1	-0.131 (0.026)***	-0.131 (0.026)***	-0.147 (0.028)***	-0.146 (0.029)***	-0.316 (0.050)***
Previous failures	0.008 (0.014)	0.008 (0.014)	0.007 (0.016)	-0.004 (0.017)	-0.157 (0.078)**
War (dummy) t-1		-0.339 (0.755)	-0.265 (0.743)	-0.128 (0.753)	0.444 (1.056)
Unrest (log) t-1			0.780 (0.193)***	0.806 (0.184)***	0.602 (0.253)**
Rel. fractionalization t-1			-0.633 (0.804)	-0.506 (0.831)	
Mil. capabilities (cube root) t-1				-7.806 (7.602)	-9.579 (41.881)
<i>NxT</i>	4,382	4,382	4,293	4,103	1,183
Time polynomials	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-4. Conditional logit model in column 5. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; ***

$p < 0.01$

Table A22: Regional rebellions and military regime democratization (GWF data)

	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.333 (0.789)			0.723 (0.803)		
Previously experienced center-seeking rebellion t-1	-0.356 (0.489)			-0.021 (0.437)		
Directly following regional rebellion t-1		-1.606 (1.631)			-1.701 (1.706)	
Directly following center-seeking rebellion t-1		-0.838 (0.607)			-0.502 (0.556)	
Regional rebellion (decay) t-1			-0.569 (0.541)	-0.953 (0.578)*	-0.926 (0.771)	-0.435 (0.929)
Center-seeking rebellion (decay) t-1			-0.980 (0.636)	-0.992 (0.622)	-0.658 (0.597)	-0.826 (1.649)
<i>NxT</i>	1,337	951	1,337	1,337	951	548
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Alternative half-lives for our decaying variables

To recall, our main variables measuring exposure to rebellions – regional and center-seeking – are decaying variables with a half-life of one year, meaning that half of the effect decays after one year. To see if our findings are driven by this particular choice of half-life, we rerun our baseline models with alternative half-lives of two, three, and five years respectively. The results, displayed in Table A23-Table A28, are substantively identical to our baseline models, albeit slightly weaker in the case of military regime democratization with a half-life of five years.

Table A23: Regional rebellions and military regime emergence (half-life 2)

	(1)	(2)	(3)	(4)	(5)
Regional rebellion (HL2, decay) t-1	1.004 (0.436)**	1.014 (0.432)**	0.898 (0.423)**	0.876 (0.409)**	2.283 (1.027)**
Center-seeking rebellion (HL2, decay) t-1	0.372 (0.371)	0.374 (0.371)	0.293 (0.356)	0.261 (0.360)	1.348 (0.794)*
GDP p.c. (log) t-1	-0.151 (0.238)	-0.152 (0.237)	-0.166 (0.233)	0.017 (0.264)	-0.774 (0.890)
Population size (log) t-1	0.038 (0.129)	0.041 (0.126)	-0.011 (0.125)	0.367 (0.216)*	2.776 (2.113)
Polity t-1	-0.127 (0.031)***	-0.128 (0.031)***	-0.142 (0.032)***	-0.154 (0.035)***	-0.408 (0.096)***
Previous failures	0.195 (0.215)	0.196 (0.216)	0.165 (0.238)	0.126 (0.251)	-1.136 (1.003)
War (dummy) t-1		-0.143 (0.770)	0.039 (0.749)	0.370 (0.742)	0.337 (1.066)
Unrest (log) t-1			0.482 (0.151)***	0.516 (0.155)***	-0.098 (0.310)
Rel. fractionalization t-1			-1.572 (0.694)**	-1.685 (0.748)**	
Mil. capabilities (cube root) t-1				-10.185 (5.012)**	-21.722 (33.329)
<i>NxT</i>	4,726	4,726	4,631	4,630	1,301
Time polynomials	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-4. Conditional logit model in column 5. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A24: Regional rebellions and military regime democratization (half-life 2)

	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.759 (1.190)			2.662 (2.030)		
Previously experienced center-seeking rebellion t-1	0.844 (0.861)			-0.482 (1.809)		
Directly following regional rebellion t-1	0.650 (0.879)				1.847 (2.149)	
Directly following center-seeking rebellion t-1	0.029 (1.483)				-0.923 (1.293)	
Regional rebellion (HL2, decay) t-1		-2.758 (1.524)*	-4.340 (1.951)**		-1.436 (2.053)	-13.441 (4.607)***
Center-seeking rebellion (HL2, decay) t-1		1.866 (0.989)*	2.460 (1.785)		1.955 (1.086)*	1.090 (2.333)
<i>NxT</i>	1,166	1,023	1,166	1,166	1,023	471
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A25: Regional rebellions and military regime emergence (half-life 3)

	(1)	(2)	(3)	(4)	(5)
Regional rebellion (HL3, decay) t-1	0.974 (0.430)**	0.983 (0.425)**	0.869 (0.415)**	0.859 (0.403)**	2.350 (1.063)**
Center-seeking rebellion (HL3, decay) t-1	0.363 (0.375)	0.366 (0.375)	0.293 (0.362)	0.259 (0.365)	1.474 (0.833)*
GDP p.c. (log) t-1	-0.146 (0.238)	-0.147 (0.237)	-0.161 (0.233)	0.023 (0.265)	-0.723 (0.914)
Population size (log) t-1	0.038 (0.130)	0.040 (0.127)	-0.012 (0.126)	0.368 (0.216)*	2.832 (2.138)
Polity t-1	-0.128 (0.031)***	-0.128 (0.031)***	-0.143 (0.032)***	-0.154 (0.035)***	-0.409 (0.096)***
Previous failures	0.190 (0.217)	0.190 (0.217)	0.160 (0.239)	0.121 (0.252)	-1.177 (1.010)
War (dummy) t-1		-0.137 (0.763)	0.038 (0.746)	0.369 (0.740)	0.266 (1.079)
Unrest (log) t-1			0.485 (0.152)***	0.518 (0.155)***	-0.099 (0.315)
Rel. fractionalization t-1			-1.568 (0.695)**	-1.680 (0.750)**	
Mil. capabilities (cube root) t-1				-10.285 (5.034)**	-23.936 (34.249)
<i>NxT</i>	4,726	4,726	4,631	4,630	1,301
Time polynomials	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-4. Conditional logit model in column 5. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A26: Regional rebellions and military regime democratization (half-life 3)

	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.759 (1.190)			3.107 (2.096)		
Previously experienced center-seeking rebellion t-1	0.844 (0.861)			-0.769 (2.207)		
Directly following regional rebellion t-1		0.650 (0.879)			0.931 (2.894)	
Directly following center-seeking rebellion t-1		0.029 (1.483)			-0.961 (1.290)	
Regional rebellion (HL3, decay) t-1			-2.791 (1.565)*	-4.838 (2.045)**	-0.529 (2.733)	-14.690 (15.613)
Center-seeking rebellion (HL3, decay) t-1			1.872 (0.992)*	2.758 (2.116)	2.090 (1.124)*	2.548 (2.387)
<i>NxT</i>	1,166	1,023	1,166	1,166	1,023	471
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A27: Regional rebellions and military regime emergence (half-life 5)

	(1)	(2)	(3)	(4)	(5)
Regional rebellion (HL5, decay) t-1	0.930 (0.419)**	0.938 (0.414)**	0.826 (0.404)**	0.837 (0.395)**	2.381 (1.119)**
Center-seeking rebellion (HL5, decay) t-1	0.338 (0.382)	0.341 (0.382)	0.281 (0.372)	0.246 (0.375)	1.612 (0.894)*
GDP p.c. (log) t-1	-0.141 (0.237)	-0.142 (0.237)	-0.157 (0.233)	0.031 (0.265)	-0.659 (0.945)
Population size (log) t-1	0.039 (0.130)	0.041 (0.128)	-0.013 (0.127)	0.373 (0.217)*	2.862 (2.172)
Polity t-1	-0.128 (0.031)***	-0.128 (0.031)***	-0.143 (0.032)***	-0.155 (0.035)***	-0.410 (0.096)***
Previous failures	0.182 (0.220)	0.183 (0.220)	0.151 (0.242)	0.114 (0.255)	-1.214 (1.011)
War (dummy) t-1		-0.130 (0.751)	0.034 (0.741)	0.367 (0.736)	0.156 (1.087)
Unrest (log) t-1			0.488 (0.152)***	0.520 (0.155)***	-0.098 (0.322)
Rel. fractionalization t-1			-1.559 (0.696)**	-1.671 (0.754)**	
Mil. capabilities (cube root) t-1				-10.482 (5.076)**	-26.702 (35.579)
<i>NxT</i>	4,726	4,726	4,631	4,630	1,301
Time polynomials	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-4. Conditional logit model in column 5. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A28: Regional rebellions and military regime democratization (half-life 5)

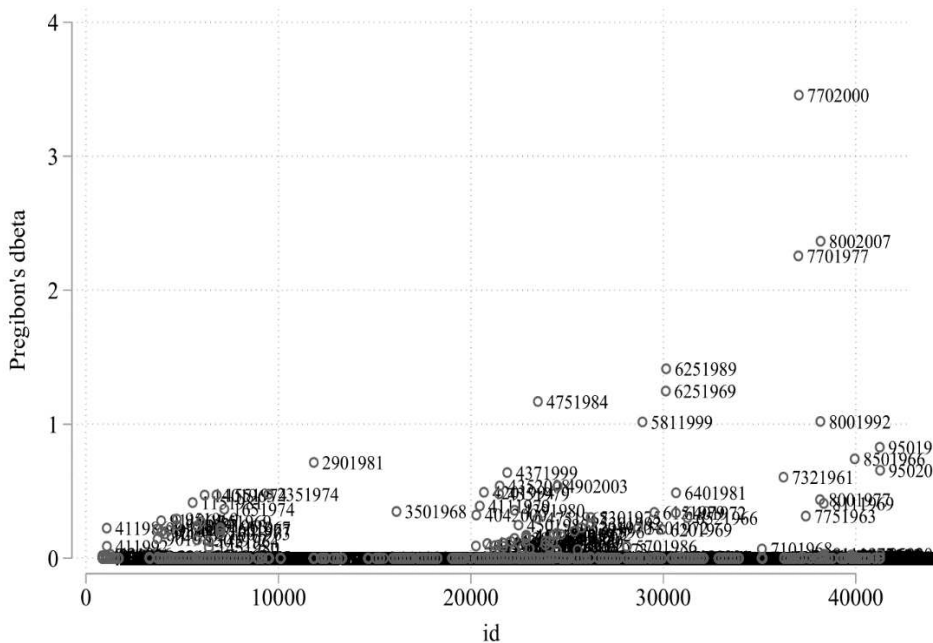
	(1)	(2)	(3)	(4)	(5)	(6)
Previously experienced regional rebellion t-1	0.759 (1.190)			3.951 (2.284)*		
Previously experienced center-seeking rebellion t-1	0.844 (0.861)			-1.167 (2.944)		
Directly following regional rebellion t-1		0.650 (0.879)			0.011 (3.018)	
Directly following center-seeking rebellion t-1		0.029 (1.483)			-0.948 (1.296)	
Regional rebellion (HL5, decay) t-1			-2.673 (1.632)	-5.794 (2.366)**	0.341 (2.805)	-35.328 (36.325)
Center-seeking rebellion (HL5, decay) t-1			1.767 (0.985)*	3.134 (2.832)	2.185 (1.198)*	4.311 (3.421)
<i>NxT</i>	1,166	1,023	1,166	1,166	1,023	471
Time polynomials	Yes	Yes	Yes	Yes	Yes	Yes
Regional dummies	Yes	Yes	Yes	Yes	Yes	Yes
Standard controls	Yes	Yes	Yes	Yes	Yes	Yes

Pooled logit model in columns 1-5. Conditional logit model in column 6. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Influential observations

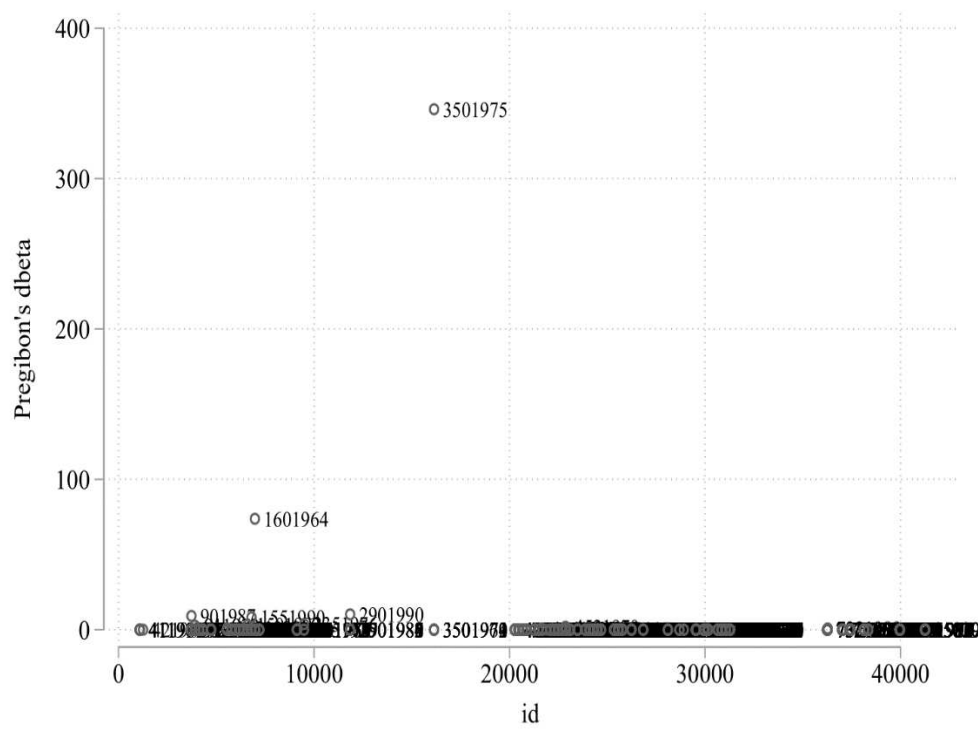
To ensure that our results are not solely driven by a few influential observations, we rerun our regressions while dropping observations with particularly high leverage. To identify these observations, we are using Pregibon's $dbeta$ which provides information on the influence on parameter estimates of each individual observation. There is no critical value for this measure and influential observations have to be identified using eyeballing. In light of Figure A7, Pakistan (cowcode770) and Thailand (cowcode 800) seem to contain influential observations. Figure A8 highlights Argentina (cowcode 160) and Greece (cowcode 350) as influential observations. We therefore remove these countries from the regression and rerun our baseline models. The results shown in Table A29 and Table A30 are substantively identical to our core findings, albeit slightly weaker in both cases. We thus conclude that our results are not a pure artefact of a few high-leverage observations.

Figure A7: Influential observations for military regime emergence



Note: Labels represent cowcode-years.

Figure A8: Influential observations for military regime democratization



Note: Labels represent cowcode-years.

Table A29: Regional rebellions and military regime emergence (w/o influential observations)

	(1)
Regional rebellion (decay) t-1	0.793 (0.451)*
Center-seeking rebellion (decay) t-1	0.444 (0.363)
GDP p.c. (log) t-1	-0.169 (0.249)
Population size (log) t-1	0.330 (0.224)
Polity t-1	-0.144 (0.034)***
War (dummy) t-1	0.163 (0.883)
Unrest (log) t-1	0.562 (0.162)***
Rel. fractionalization t-1	-1.585 (0.755)**
Mil. capabilities (cube root) t-1	-10.092 (5.103)**
Previous failures	-0.089 (0.250)
<i>NxT</i>	4,368
Time polynomials	Yes
Regional dummies	Yes

Pooled logit model replicating column 4, Table 1 in main article, without Pakistan and Thailand. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table A30: Regional rebellions and military regime democratization (w/o influential observations)

	(1)
Previously experienced regional rebellion t-1	1.717 (2.193)
Previously experienced center-seeking rebellion t-1	-1.823 (1.117)
Regional rebellion (decay) t-1	-4.259 (2.371)*
Center-seeking rebellion (decay) t-1	2.350 (1.484)
GDP p.c. (log) t-1	0.011 (0.854)
Population size (log) t-1	-0.075 (1.059)
Polity t-1	1.034 (0.218)***
Mil. capabilities (cube root) t-1	-4.079 (20.354)
Unrest (log) t-1	0.156 (0.668)
War (dummy) t-1	-3.257 (1.705)*
Rel. fractionalization t-1	12.127 (3.967)***
Number of previous regime failures	0.979 (0.521)*
<i>NxT</i>	1,144
Time polynomials	Yes
Regional dummies	Yes

Pooled logit model replicating column 4, Table 4 in main article, without Argentina and Greece. Cluster-robust standard errors in parentheses. * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

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