Appendix: Trump and Public Support for Democracy Abroad

This appendix provides more detail about the multilevel model in the manuscript and association between the Trump administration and democratic support in Asia. Please note that the World Values survey does not permit disseminating copies of the data in replication archives. We used version 1.6 of the World Values survey longitudinal data, which can be downloaded from: https://www.worldvaluessurvey.org/WVSDocumentationWVL.jsp.

1 Model Specification

The varying slopes model lets the relationship between U.S. performance and other year-level factors shift across regions, rather than assuming a constant effect. The binomial model of the proportion of high democratic support given the number of respondents within each state-year group can be expressed as follows, with the probability of high support θ for each country-year observation partially pooled such that $\theta_i \sim N(\mu_i, \sigma)$:

$$y \sim Binomial(Num.Resp, \theta_i)$$
 (1)

We model the expected value of high democratic support μ in each state-year observation as follows:

$$\mu = \alpha + \alpha_{\text{region}} + \alpha_{\text{year}} + \mathbf{X}\beta + \mathbf{Z}\gamma + \mathbf{G}\lambda_{\text{region}}$$
 (2)

There are three sets of variables here. The year level variables matrix G has varying slopes by region. The state level variable coefficients tied to Z and X do not change across regions.

The specification subsumes regional intercepts and year varying slopes expressed in G into a multivariate normal prior with covariance matrix Σ and an LKJ(2) prior, further optimized with a Choleskey factorization:

$$\alpha_{\text{region}}, \lambda_{\text{region}} \sim \text{multivariate normal}(\mu_{\text{region}}, \mu_{\lambda}, \Sigma)$$
 (3)

Table 1 summarizes the other prior distributions in the multilevel model. All priors are weakly informative relative to the scale of the data, and the regression parameters are robust to unusually large effect sizes. We employ a non-centered parameterization of the year intercepts in fitting the model.

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p(\alpha) \sim N(0, 1)
p(\alpha_{year}) \sim N(G\lambda, \sigma^{yr})
p(\sigma^{yr}) \sim \text{half-}N(0, 1)
p(\alpha_{state}) \sim N(\mu_{state}, \sigma^{st})
p(\mu_{state}) \sim N(0, 1)
p(\sigma^{st}) \sim \text{half-}N(0, .5)
p(\beta) \sim \text{student-t}(5, 0, 1)
p(\gamma) \sim \text{student-t}(5, 0, 1)
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Table 1: Summary of Priors in Multilevel Model

2 Hamiltonian Monte Carlo Diagnostics

We fit these models with STAN (Carpenter et al., 2016). There were no divergent iterations in either sample running 4 chains for 2,000 iterations with 1,000 warm up iterations in any of the imputed datasets. The split \hat{R} statistic is less than 1.1 for all parameters as well. Energy diagnostics and the effective sample size for the posterior medians and tails are also satisfactory. All of these estimates suggest that the Hamiltonian Monte Carlo adequately explored the posterior distribution.

3 Variables

This section summarizes the key variables in the model with tabulations. We start with the year-level variables that measure U.S. success and failure. Table 2 summarizes these variables and their distribution. Table 3 summarizes the state-level variables across all imputed datasets. Last, Table 4 shows summary statistics for the imputed individual level variables in the analysis, which we averaged at the state year. It also includes the distribution of high democratic support sums and the number of respondents, which are key quantities in the binomial proportions.

Table 2: Year Level Variables

	Unique (#)	Missing (%)	Mean	SD	Min	Median	Max
US GDP Growth	30	0	2.59	1.63	-2.54	2.77	4.75
US Democracy	24	0	0.80	0.03	0.73	0.80	0.86
US Human Rights	30	0	0.72	0.67	-0.17	0.45	1.88
US Protests	30	0	0.69	0.17	0.42	0.69	1.05
US GINI	23	0	36.79	1.93	31.60	37.45	38.70
Clinton	2	0	0.20	0.41	0.00	0.00	1.00
W.Bush	2	0	0.27	0.45	0.00	0.00	1.00
Obama	2	0	0.23	0.43	0.00	0.00	1.00
Trump Pres	2	0	0.10	0.31	0.00	0.00	1.00
Chinese Growth	30	0	8.94	2.45	3.91	9.18	14.23
US Intervention	3	0	-0.10	0.55	-1.00	0.00	1.00

Table 3: State Level Variables

	Unique (#)	Missing (%)	Mean	SD	Min	Median	Max
GDP per Capita	250	0	14696.05	17188.09	273.49	7673.12	91565.73
GDP Growth	248	0	4.33	5.42	-24.00	4.10	54.16
Information Flow	233	0	64.97	19.26	13.28	69.05	94.82
Social Globalization	233	0	60.00	18.75	12.00	61.71	90.81
Bank Crisis	2	0	0.09	0.28	0.00	0.00	1.00
Liberal Democracy	219	0	0.49	0.27	0.03	0.51	0.89
Conflict Battle Deaths	29	0	445.33	3125.12	0.00	0.00	35071.00
Human Rights	236	0	0.16	1.54	-2.57	-0.00	3.97
Inequality	157	0	37.51	9.00	17.50	36.90	63.20
U.S. Aid	150	0	170.18	774.22	0.00	0.32	10149.89

Table 4: Individual Level Variables

	Unique (#)	Missing (%)	Mean	SD	Min	Median	Max
Sum: High Democ. Support	1219	0	701.78	367.22	43.00	661.00	2665.00
Avg: Political Interest	2269	0	2.68	0.31	1.27	2.72	3.38
Avg: Country Aim	2977	0	1.74	0.25	1.23	1.73	2.62
Avg: Left or Right	4500	0	5.83	0.62	2.77	5.81	8.95
Avg: Government Confidence	3137	0	2.66	0.39	1.25	2.72	3.45
Avg: Rate Political System	4744	0	5.19	0.78	2.35	5.28	8.61
Avg: Nationalism	3822	0	0.19	0.10	0.00	0.17	0.45
Avg: Financial Satisfaction	3334	0	5.72	0.98	3.06	5.94	8.21
Avg: Respect Authority	2557	0	0.27	0.18	0.02	0.23	0.89
Number of Respondents	172	0	1445.89	570.84	240.00	1212.00	4078.00

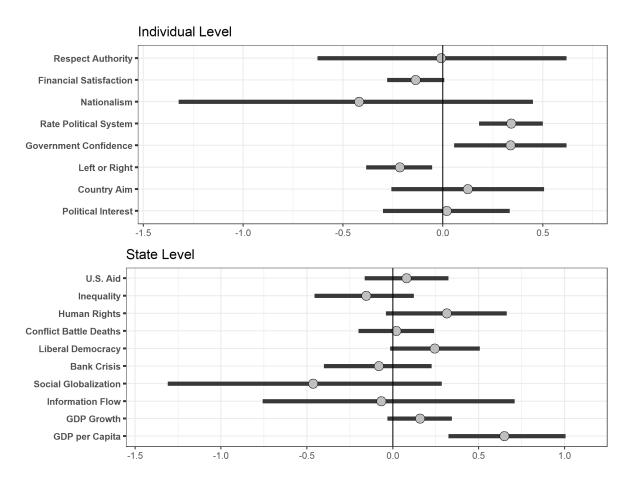


Figure 1: Estimated association between individual and state level variables and the probability of high democratic support. Point estimates mark the posterior median, and error bars summarize the 90% credible interval.

4 Other Estimates

This section of the appendix summarizes state and individual level variable estimates from the multilevel model in the manuscript. Some individual and state-level variables also impact support for democracy outside the United States, as Figure 1 shows. Because we average the individual variables at the country-year level to facilitate binomial model fitting, the individual coefficient estimates compare averages across observations. Higher average ratings of the political system and confidence in the government in a country increase the likelihood of high democratic support. Countries where individuals are more right-wing on average express lower support for democracy. The nationalism and financial satisfaction averages are also negative, but are more uncertain.

At the state level, GDP per capita is positively associated with high democratic support. The GDP growth, human rights and liberal democracy coefficient estimates are largely positive, but consistent with slight negative relationships. The income inequality and social globalization coefficients are mostly negative with substantial uncertainty.

5 Trump and Democratic Support in Asia

The models in the manuscript find a negative impact of the Trump administration on support for democracy in Asia. In this section of the appendix, we explore that finding further. To do so, we fit separate models for each Asian state with WVS data before and after Trump took office in early 2017.

Fitting separate models for each state is a simple way to assess whether the impact of the Trump administration and other factors varies by country. It also allows us to draw on the full range of the democratic support variable, instead of simplifying it to a dummy, as in the manuscript. Because democratic support takes on relatively few integer values, we fit Bayesian ordinal logit models with flexible outcome thresholds using BRMS (Bürkner, 2017). The lack of cross-sectional variation and limited number of years in these analyses meant we could only include two state and year level variables without introducing perfect multicollinearity into the model. We therefore included a national economic growth variable and the Trump administration dummy, in addition to a series of controls at the individual level. For each state, we fit models to five imputed datasets, and then combined them into a single posterior, following standard practice.

We find that the Trump presidency reduced support for democracy throughout Asia. Figure 2 plots the estimated coefficient on the Trump administration dummy variable from each model of democratic support over time in Asian states. Most of the coefficient estimates are negative and substantively large as well.

The Trump effect is largest in Bangladesh, but this may reflect only three WVS waves in Bangladesh, which creates limited temporal variation in that model. The Kazakhstan coefficient has substantial uncertainty for similar reasons. Only in Australia and Japan is the Trump administration correlated with increased democratic support.

Indonesia, Thailand, Taiwan, the Philippines, South Korea, Pakistan, Malaysia and China all saw lower public support for democracy during the Trump administration. The common change in a group with such varied characteristics gives some credence to a common trend or shock. The mechanism likely differs across these countries, however. South Korea and the Philippines are U.S. allies and Taiwan has close ties to the United States. China's relationship with the Trump administration was often antagonistic. Pakistan has some ties to the U.S., along with close relations with China.

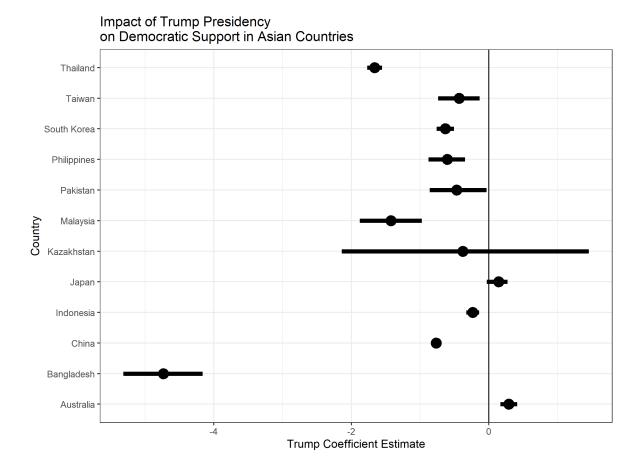


Figure 2: Estimated association between Donald Trump's presidency and public support for democracy in Asian countries, 1980 to 2019. Point estimates mark the posterior median, and error bars summarize the 90% credible interval. Coefficient estimates reflect pooled estimates from five Bayesian ordered logit models fit to five imputed datasets for each state.

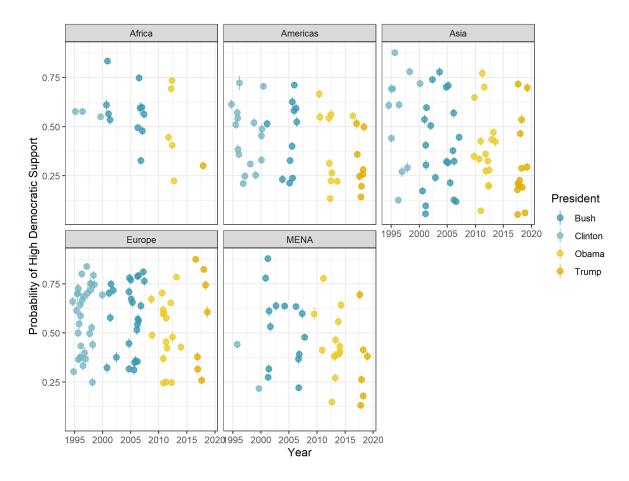


Figure 3: Estimated probability of high democratic support by region from multilevel model with regional varying slopes, 1989 to 2019. Colors differentiate between presidential administrations. MENA stands for Middle East and North Africa.

6 WVS Coverage by Region and Presidency

Here, we detail data coverage in the WVS across regions during our sample. The WVS proceeds country by country, and our sample lacks data on key control variables in 2020, so we lose out on data from later in the seventh wave.¹ Thus, as Figure 3 shows, we have more estimates for countries in Asia and consistent coverage of the Americas, with more sparse coverage in Africa and Europe.

The Trump finding may carry over into other regions, but data coverage limits our ability to assess this. A similar relationship is plausible in Africa and the Middle East. Although the Trump presidency is negatively associated with democratic support in the Americas, the association is weaker than in Asia.

The likely patterns in Europe are less clear. On the one hand, the Australian and Japanese estimates from our analysis of Asian countries suggests that the Trump administration was less

¹With completely missing data for most countries, imputation has less value.

likely to damage perceptions of democracy among wealthy consolidated democracies. On the other, democratic support fell in South Korea and Taiwan during the Trump administration. Thus, Trump may have done little damage to democratic support in Europe, or reduced support in some countries, or driven up support for democracy through some backlash mechanism.

Figure 3 further indicates that the largely negative association between the Clinton administration and support for democracy in the Middle East and North Africa is likely the result of sparse data. There are only two countries with WVS data during the Clinton administration. Larger samples from the region during other presidencies show substantial variation in support for democracy and higher average democratic support.

References

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