

How Much Should We Trust Estimates from Multiplicative Interaction Models? Simple Tools to Improve Empirical Practice

Jens Hainmueller Jonathan Mummolo Yiqing Xu*

Abstract

Multiplicative interaction models are widely used in social science to examine whether the relationship between an outcome and an independent variable changes with a moderating variable. Current empirical practice tends to overlook two important problems. First, these models assume a linear interaction effect that changes at a constant rate with the moderator. Second, estimates of the conditional effects of the independent variable can be misleading if there is a lack of common support of the moderator. Replicating 46 interaction effects from 22 recent publications in five top political science journals, we find that these core assumptions often fail in practice, suggesting that a large portion of findings across all political science subfields based on interaction models are modeling artifacts or are at best highly model dependent. We propose a checklist of simple diagnostics to assess the validity of these assumptions and offer flexible estimation strategies that allow for nonlinear interaction effects and safeguard against excessive extrapolation. These statistical routines are available in both R and STATA.

*Jens Hainmueller is a Professor of Political Science, Stanford University, jhain@stanford.edu. Jonathan Mummolo is an Assistant Professor of Politics and Public Affairs, Princeton University, jmummolo@princeton.edu. Yiqing Xu is an Assistant Professor of Political Science, University of California, San Diego, yiqingxu@ucsd.edu. We thank Licheng Liu for excellent research assistance. We thank David Broockman, Daniel Carpenter, James Fowler, Justin Grimmer, Seth Hill, Macartan Humphreys, Kosuke Imai, Gabe Lenz, Neil Malhotra, John Marshall, Marc Ratkovic, Jas Sekhon, Sean Westwood and participants at the PolMeth, APSA and MPSA annual meetings and at methods workshops at Massachusetts Institute of Technology, Harvard University, Princeton University, Columbia University and University of California, San Diego for helpful feedback. We also thank the authors of the studies we replicate for generously sharing code and data. Replication data available in [Hainmueller, Mummolo and Xu \(2018\)](#).

B Supplementary Information

In this online appendix we present detailed results for the interaction effects we identified in the current literature for use in our replication analysis. For each case, we present: the excerpt(s) from the replicated study that contains the relevant substantive claim, the replication results of the original model, our diagnostic plots, and the estimates from the binning and kernel estimator that relax the linear interaction effect assumption. GAM plots are also included when both interacted variables are continuous.

Appendix: Table of Contents

.1	Adams et al. (2006) AJPS	B-3
.2	Aklin and Urpelainen (2013) AJPS	B-5
.3	Banks and Valentino (2012) AJPS	B-9
.4	Bodea and Hicks (2015 <i>a</i>) JOP	B-15
.5	Bodea and Hicks (2015 <i>b</i>) IO	B-19
.6	Carpenter and Moore (2014) APSR	B-27
.7	Chapman (2009) IO	B-29
.8	Clark and Golder (2006) CPS	B-31
.9	Clark and Leiter (2014) CPS	B-39
.10	Hellwig and Samuels (2007) CPS	B-41
.11	Hicken and Simmons (2008) AJPS	B-45
.12	Huddy, Mason and Aarøe (2015) APSR	B-47
.13	Kim and LeVeck (2013) APSR	B-51
.14	Malesky, Schuler and Tran (2012) APSR	B-57
.15	Neblo et al. (2010) APSR	B-65
.16	Pelc (2011) IO	B-67
.17	Petersen and Aarøe (2013) APSR	B-71
.18	Somer-Topcu (2009) JOP	B-75
.19	Tavits (2008) CPS	B-77
.20	Truex (2014) APSR	B-79
.21	Vernby (2013) AJPS	B-87
.22	Williams (2011) CPS	B-91
.23	Additional Results from Diagnostic Measures	B-95

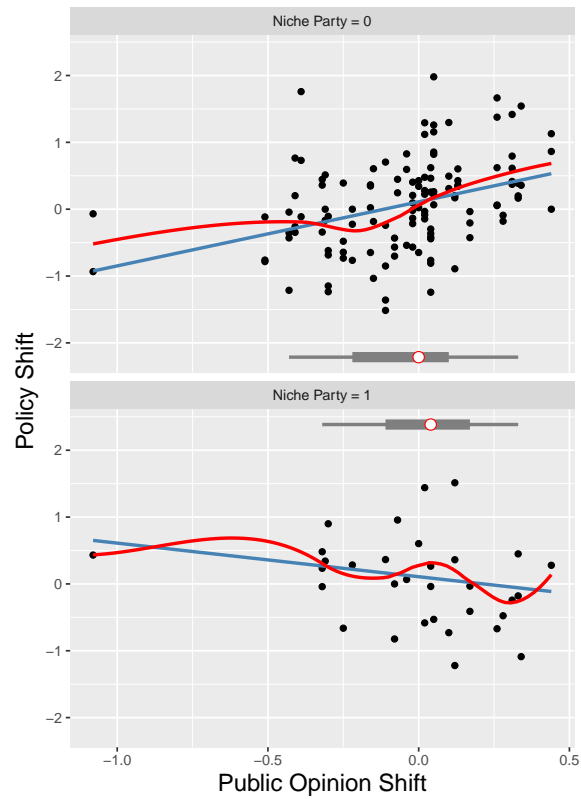
.1 [Adams et al. \(2006\)](#) AJPS

Claim on conditionality (Table 1 in manuscript): *“We find no evidence that niche parties responded to shifts in public opinion, while mainstream parties displayed consistent tendencies to respond to public opinion shifts”* (Abstract).

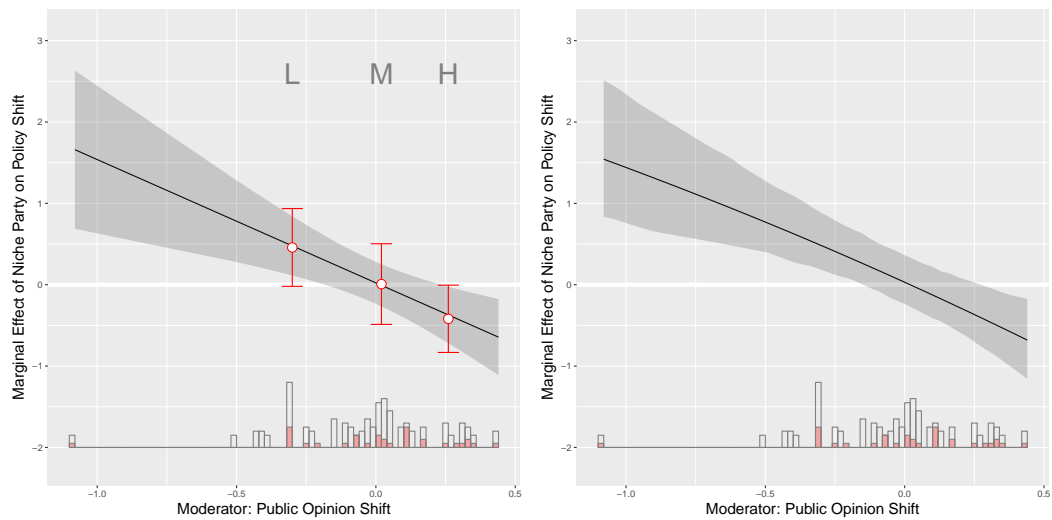
Key variables for the conditional relationship: Outcome Y: “party policy shift” (`pshift2`); treatment D: “public opinion shift” (`vshift`); moderator X: “niche parties” (`idparty`).

Note: In this replication, we treat “niche parties” as D and “public opinion shift” as X because the former is dichotomous. The interpretation of the interaction effect remains unchanged.

FIGURE B1. RESULTS FROM ADAMS ET AL. (2006)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

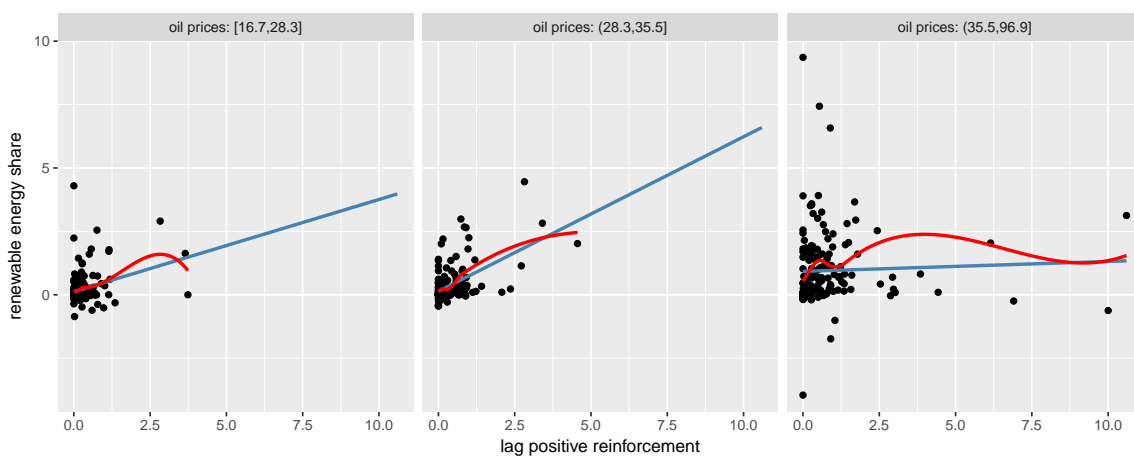
.2 Aklin and Urpelainen (2013) AJPS

First interaction:

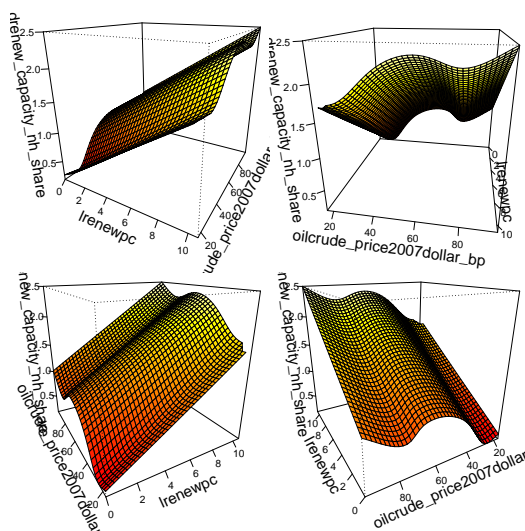
Claim on conditionality (Figure 1, left panel in manuscript): *“We examine formally how exogenous shocks, such as changes in international energy prices, interact with positive reinforcement factors, such as the growing strength of the renewables advocacy coalition. We find that political competition modifies the effect of path dependence on policy and outcomes. ... The effect of positive reinforcement also decreases with international energy prices.”* (Abstract).

Key variables for the conditional relationship: Outcome Y: “renewable share” (first differenced) (`drenew_capacity_nh_share`); treatment D: “oil prices” (`oilcrude_price2007dollar_bp`); moderator X: “lag positive reinforcement” (`lrenewpc`)

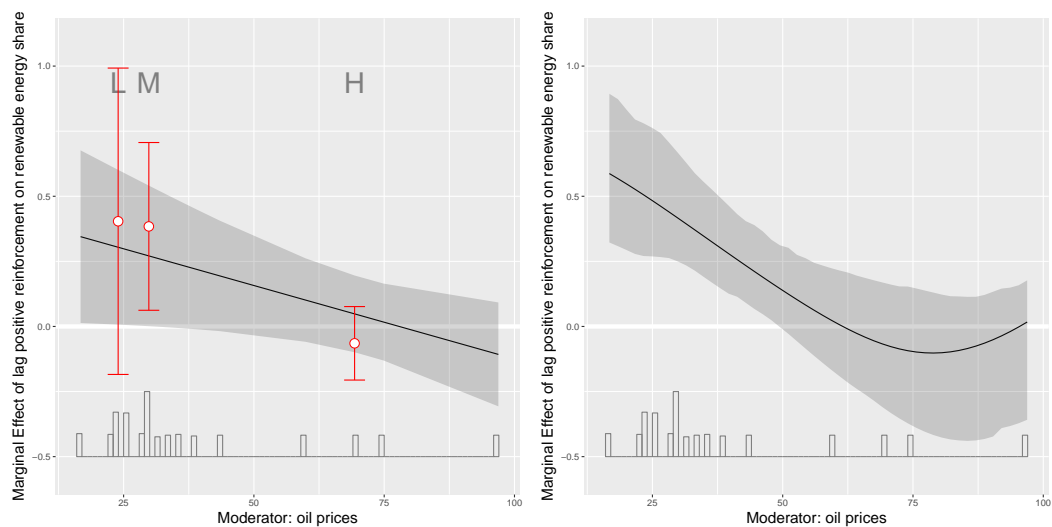
FIGURE B2. RESULTS FROM AKLIN AND URPELAINEN (2013)



(a) Raw data



(b) GAM Plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

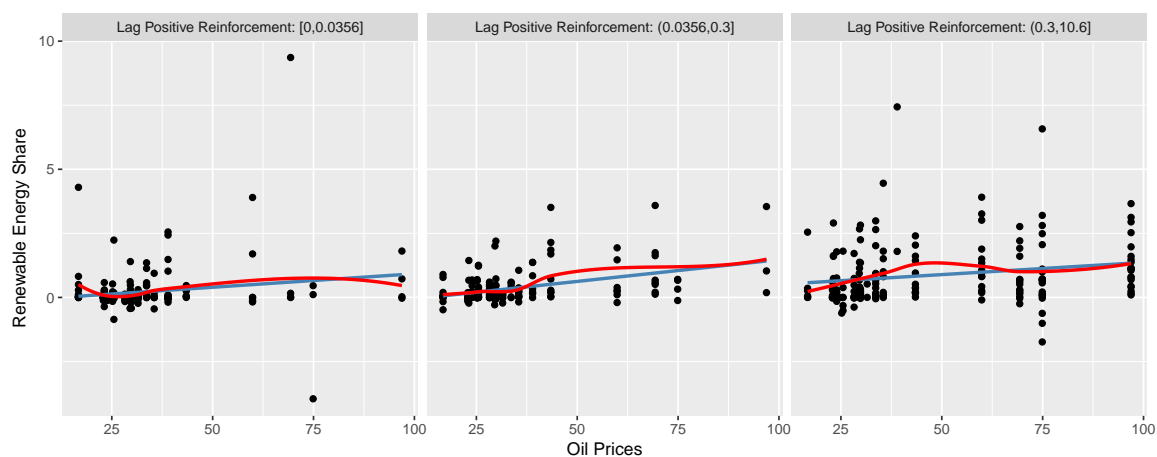
Second interaction:

Claim on conditionality (Figure 1, right panel in manuscript): *“We examine formally how exogenous shocks, such as changes in international energy prices, interact with positive reinforcement factors, such as the growing strength of the renewables advocacy coalition. We find that political competition modifies the effect of path dependence on policy and outcomes. ... The effect of positive reinforcement also decreases with international energy prices.”* (Abstract).

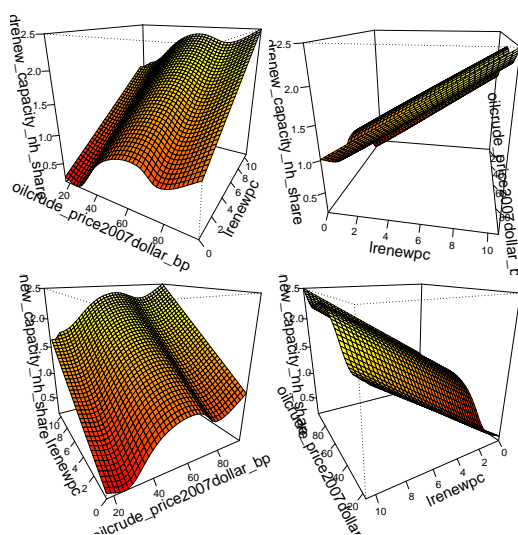
Key variables for the conditional relationship: Outcome Y: “renewable share” (first differenced) (`drenew_capacity_nh_share`); treatment D: “lag positive reinforcement” (`lrenewpc`); moderator X: “oil prices” (`oilcrude_price2007dollar_bp`).

Note: The dashed vertical line indicates the truncated interval of the moderator shown in the original marginal effect plot.

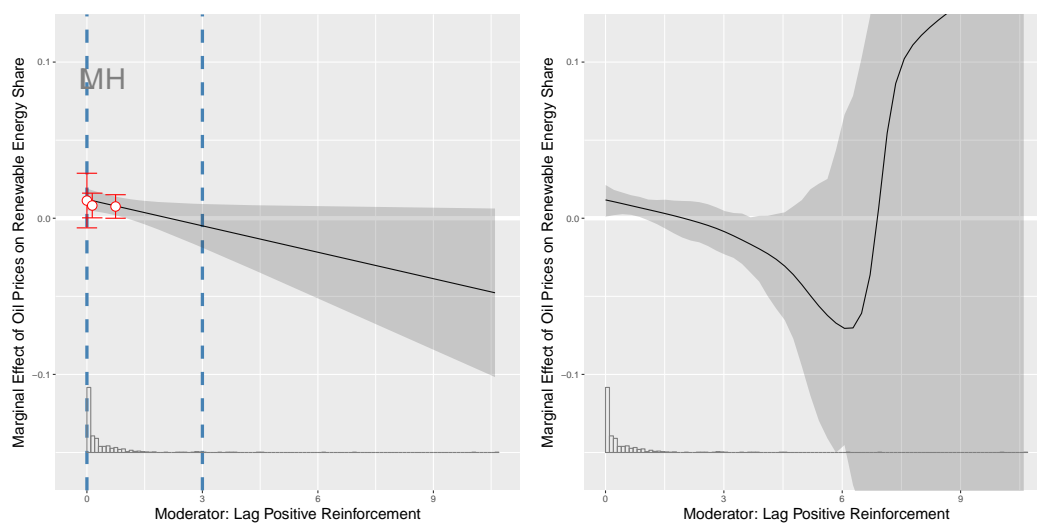
FIGURE B3. RESULTS FROM AKLIN AND URPELAINEN (2013)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

.3 Banks and Valentino (2012) AJPS

First interaction:

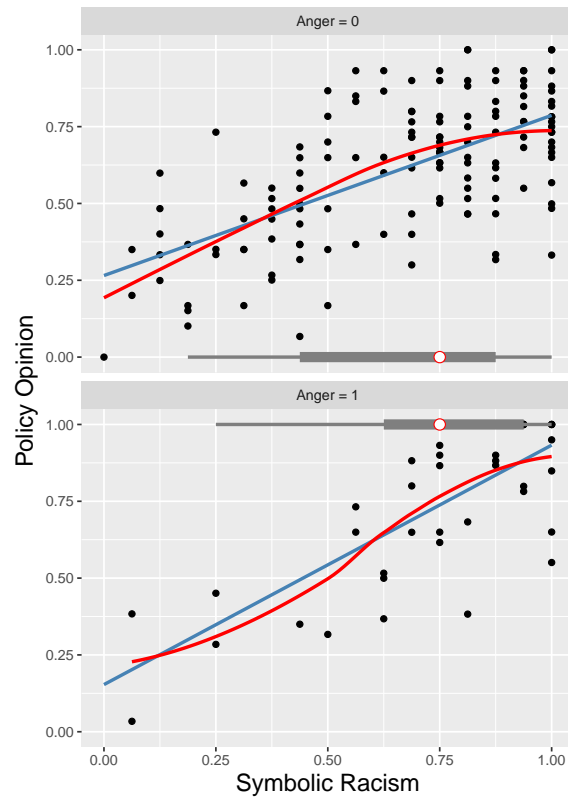
Claim on conditionality (Figure 1 in manuscript): *“One explanation for this is that a new racial belief system referred to as symbolic racism or racial resentment has replaced ‘old-fashioned racism.’ ... as a result, anger now serves as the primary emotional trigger of whites’ negative racial attitudes”* (Abstract).

“Figure 1 illustrates the marginal effect of each emotion on racial policy opinions across levels of symbolic racism (SR) ... As we predict, as SR increases, anger increasingly boosts opposition to racial policies such as affirmative action” (p. 292).

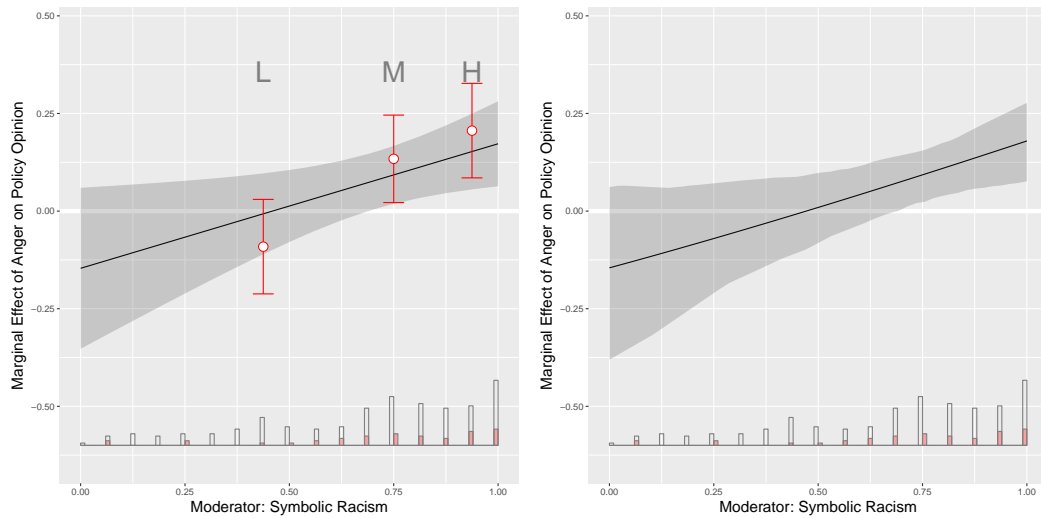
Key variables for conditional relationship: Outcome Y: “policy opinion” (`racpolicy`); treatment D: “anger” (`anger`); moderator X: “symbolic racism” (`racresent1`).

Note: Due to coding errors in the authors’ original analysis, the marginal effects plots we present here feature different intercepts than those in the published paper. We corrected the errors before applying our diagnostic functions.

FIGURE B4. RESULTS FROM [BANKS AND VALENTINO \(2012\)](#)



(a) Raw data



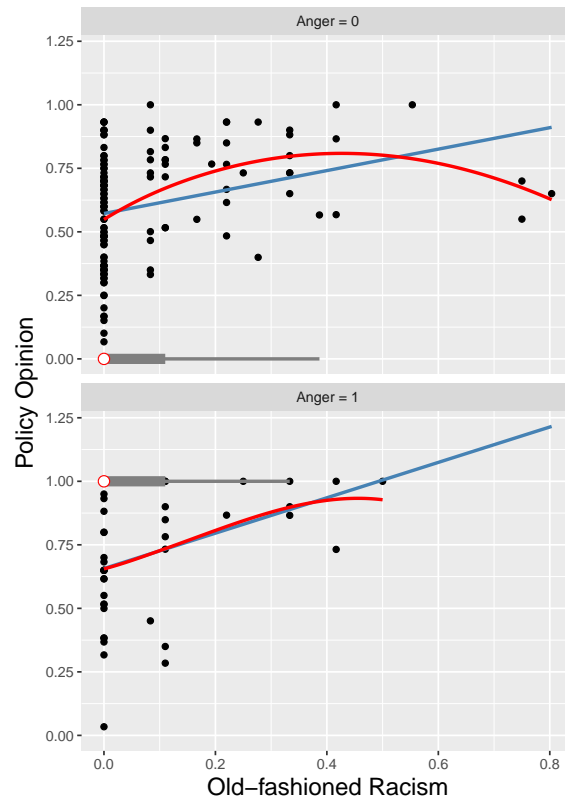
(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

Second interaction:

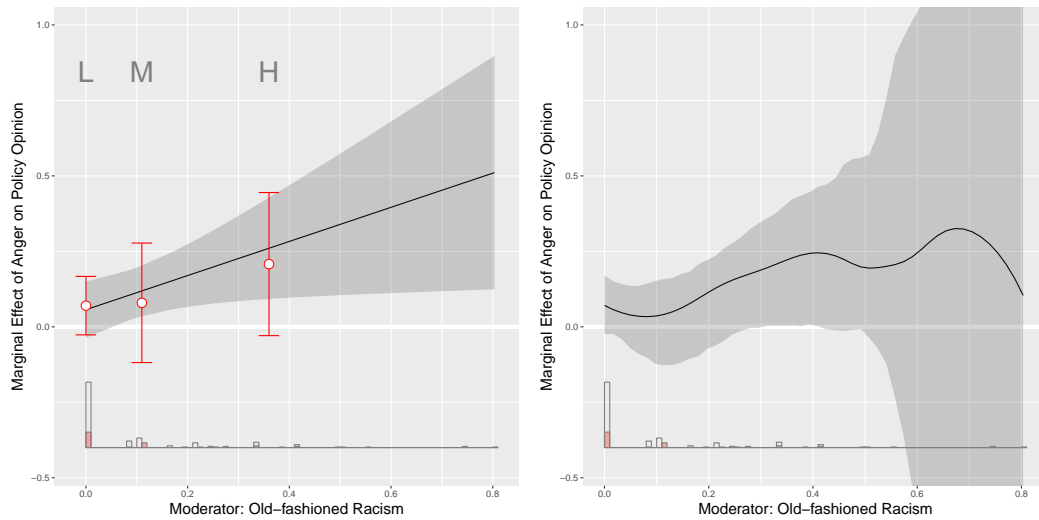
Claim on conditionality (Figure 2 in manuscript): *“Figure 2 displays these interactions visually and shows that the effects of anger and disgust are larger than that of fear, but these differences are not as large or statistically distinct. However, as OFR increases, both anger and disgust boost opposition to racial policies. At very high levels of OFR, both anger and disgust boost opposition significantly more than that in the (relaxed) control group.”* (p. 292)

Key variables for conditional relationship: Outcome Y: “policy opinion” (`racpolicy`); treatment D: “anger” (`anger`); moderator X: “old-fashioned racism” (`jimcrow13`).

FIGURE B5. RESULTS FROM [BANKS AND VALENTINO \(2012\)](#)



(a) Raw data



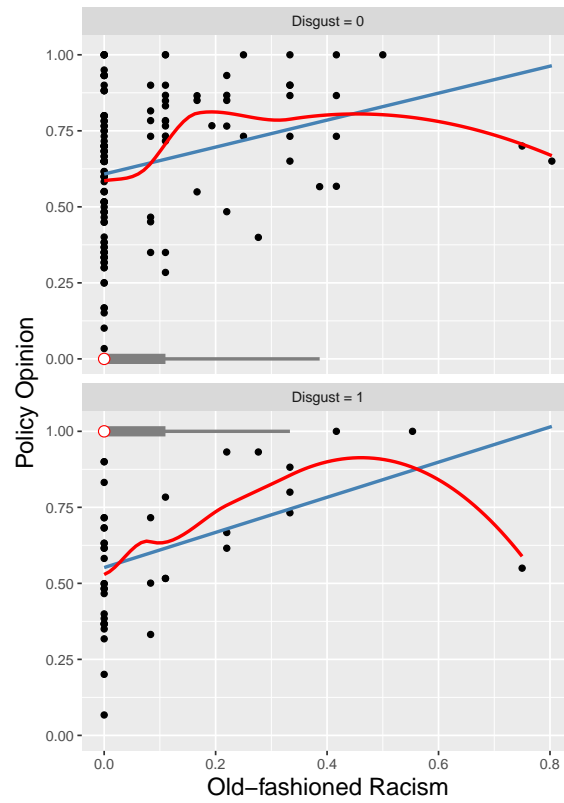
(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

Third interaction:

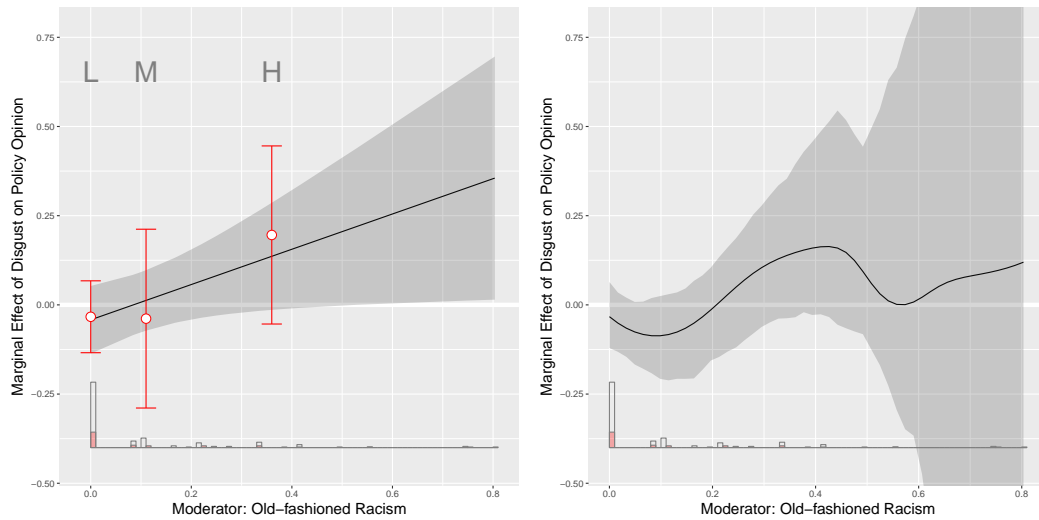
Claim on conditionality (Figure 2 in manuscript): *“Figure 2 displays these interactions visually and shows that the effects of anger and disgust are larger than that of fear, but these differences are not as large or statistically distinct. However, as OFR increases, both anger and disgust boost opposition to racial policies. At very high levels of OFR, both anger and disgust boost opposition significantly more than that in the (relaxed) control group.”* (p. 292)

Key variables for conditional relationship: Outcome Y: “policy opinion” (`racpolicy`); treatment D: “disgust” (`disgust`); moderator X: “old-fashioned racism” (`jimcrow13`).

FIGURE B6. RESULTS FROM [BANKS AND VALENTINO \(2012\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

.4 Bodea and Hicks (2015a) JOP

First interaction:

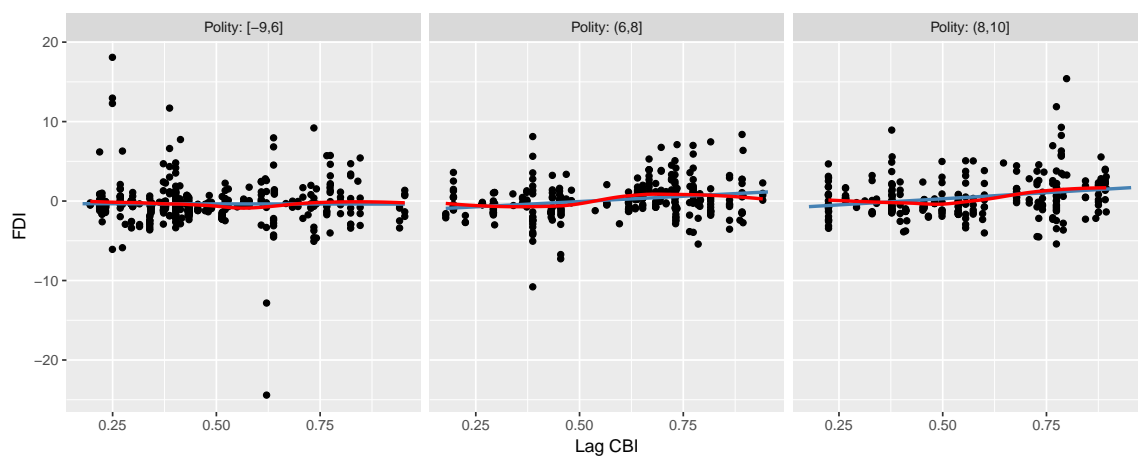
Claim on conditionality (Figure 1a in manuscript): *“In addition, we show that CBI affects the flow and cost of capital in non-OECD countries ... where political institutions allow the central bank to de facto be credible.”* (Abstract).

“We plot the marginal effect of CBI as democracy increases in non-OECD countries in Figure 1(a). The marginal effect is significant only at high levels of Polity, supporting Hypothesis 2.1” (p. 278).

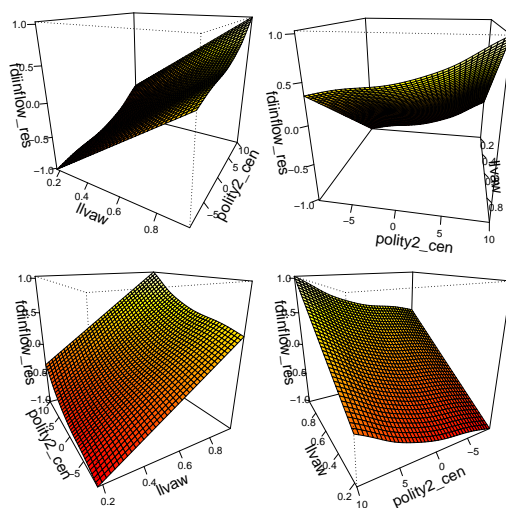
Key variables for the conditional relationship: Outcome Y: “FDI” (fdiinflw, demeaned); treatment D: “lag CBI” (llvaw); moderator X: “Polity” (polity2_cen).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

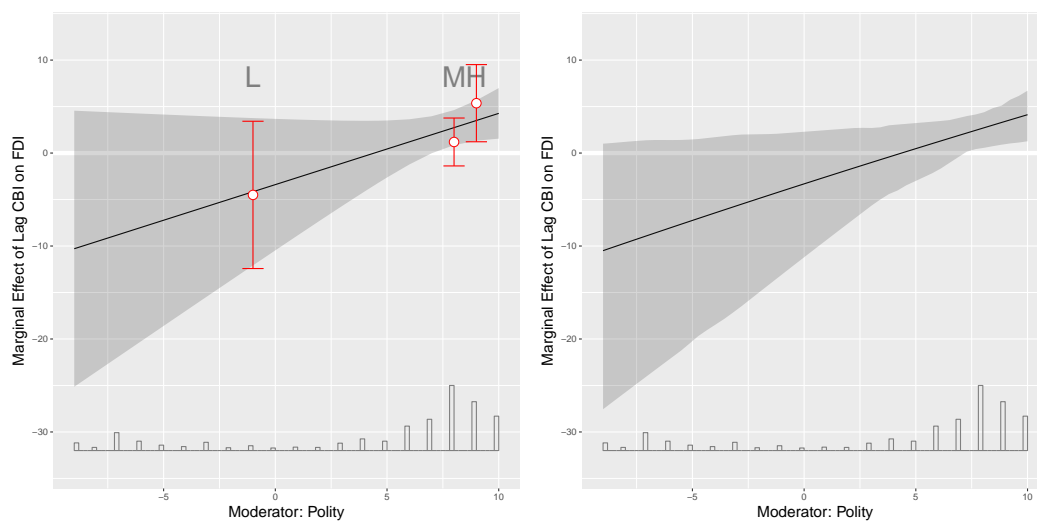
FIGURE B7. RESULTS FROM BODEA AND HICKS (2015a)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

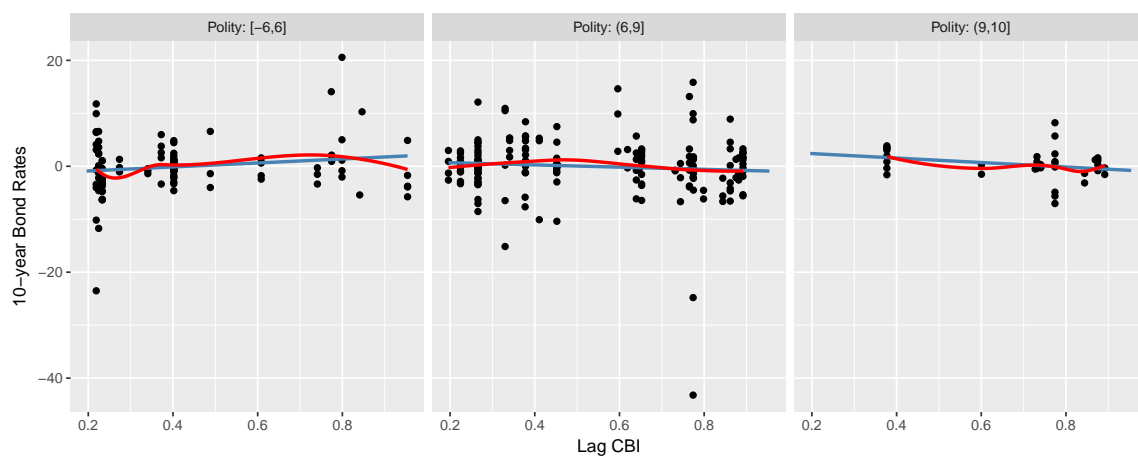
Second interaction:

Claim on conditionality: *Figure 1(c) shows that the marginal effect of CBI is downward sloping and negative at higher levels of democracy. For the 10-year bonds, more independent central banks reduce borrowing costs for democratic governments but have no effect in nondemocracies.* (p. 279)

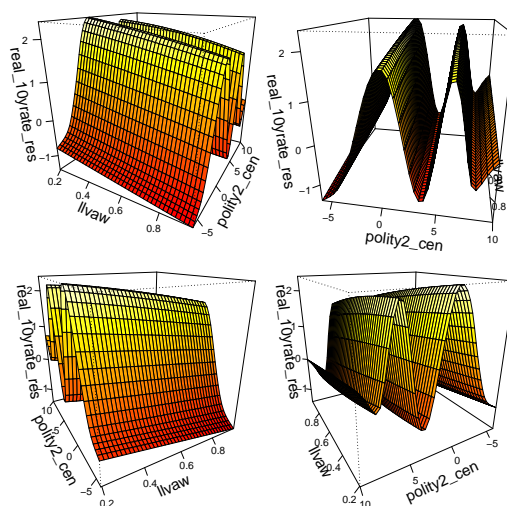
Key variables for the conditional relationship (Figure 1c): Outcome Y: “10-year bond rates ” (`real_10yrate`, demeaned); treatment D: “lag CBI” (`11vaw`); moderator X: “Polity” (`polity2_cen`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

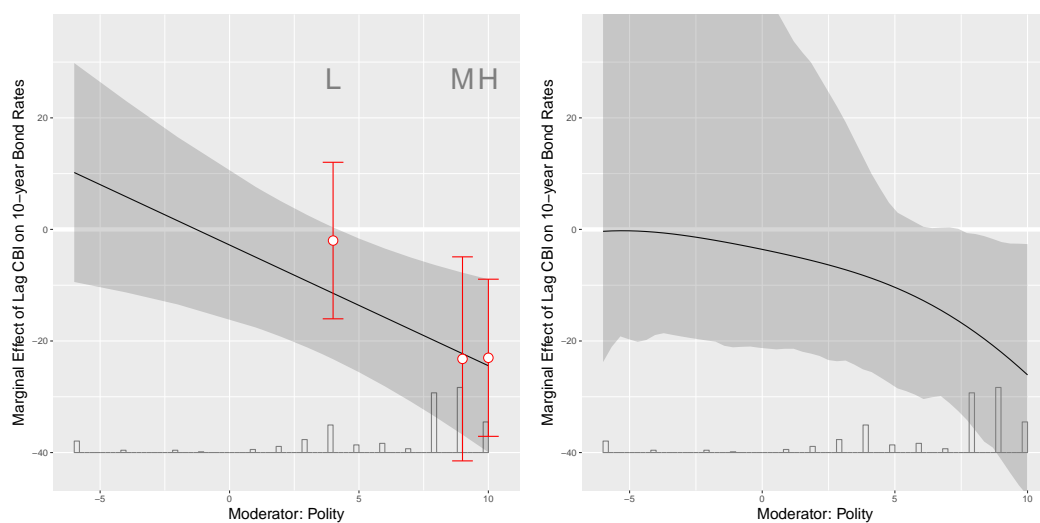
FIGURE B8. RESULTS FROM [BODEA AND HICKS \(2015a\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.5 Bodea and Hicks (2015b) IO

First Interaction:

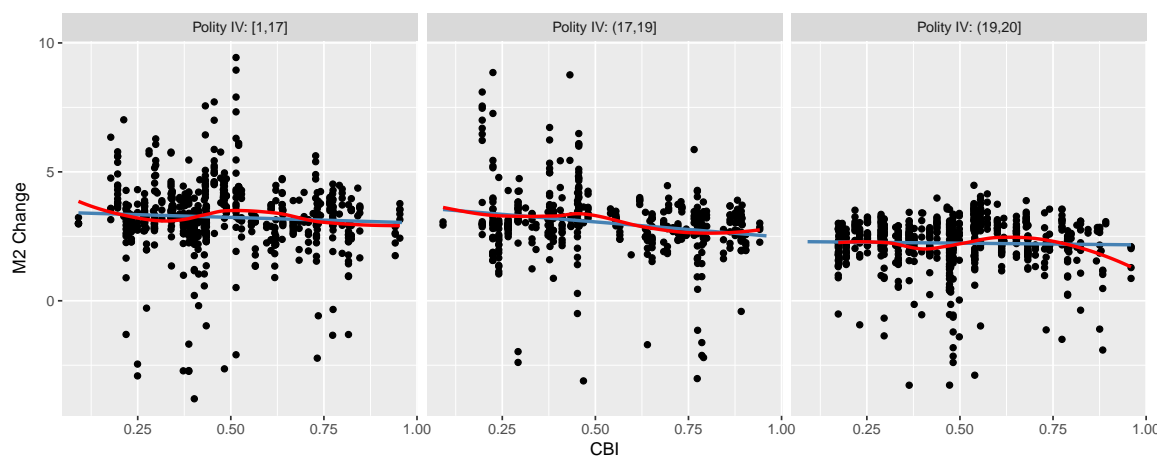
Claim on conditionality (Figure 2 in manuscript): *“Despite mixed empirical evidence, in the past two decades central bank independence (CBI) has been on the rise under the assumption that it ensures price stability. ... Empirical results are robust and support a discipline effect conditioned by political institutions, as well as a credibility effect.”* (Abstract)

“Figure 2 shows the marginal effect of CBI as POLITY and FREEDOM HOUSE democracy vary. The graphs confirm our expectations; the marginal effect of CBI is downward sloping but it is negative and statistically significant at high levels of democracy only (POLITY scores above 16). At low levels of POLITY, the marginal effect of CBI is positive but statistically insignificant. Similarly, the marginal effect of CBI is negative and significant only when the FREEDOM HOUSE score is greater than about 5.” (p. 49)

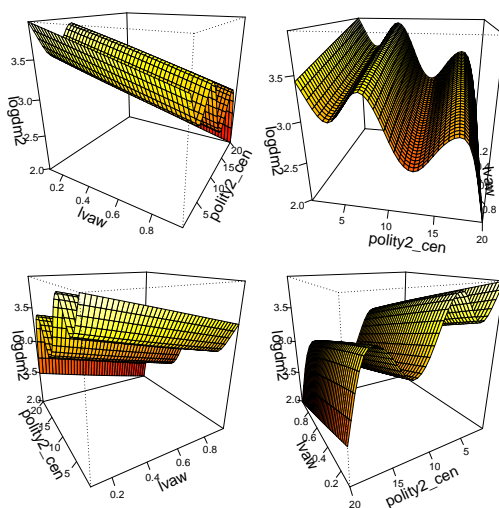
Key variables for conditional relationship 1: Outcome Y: “M2 growth” (logdm2); treatment D: “central bank independence” (CBI); moderator X: “Polity IV score” (polity2_cen).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

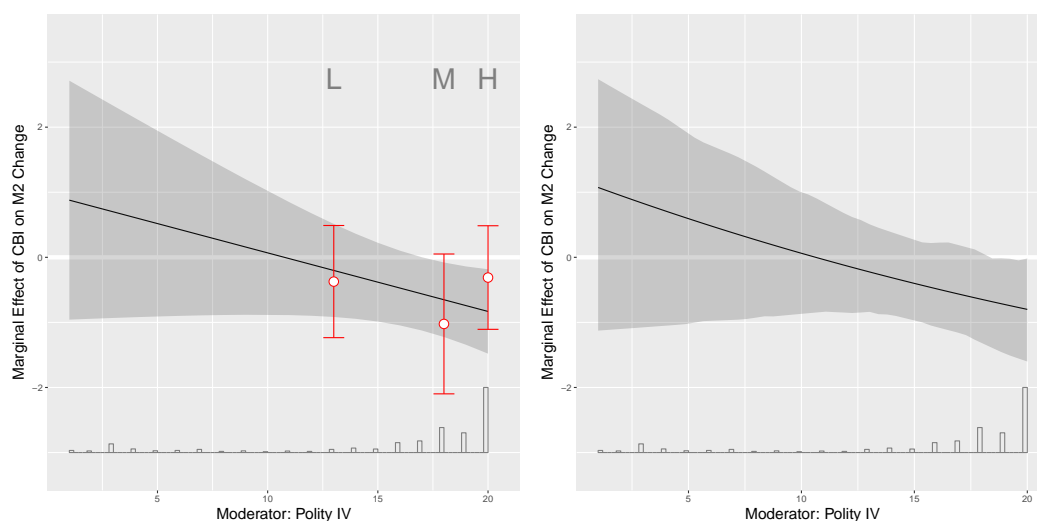
FIGURE B9. RESULTS FROM [BODEA AND HICKS \(2015b\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

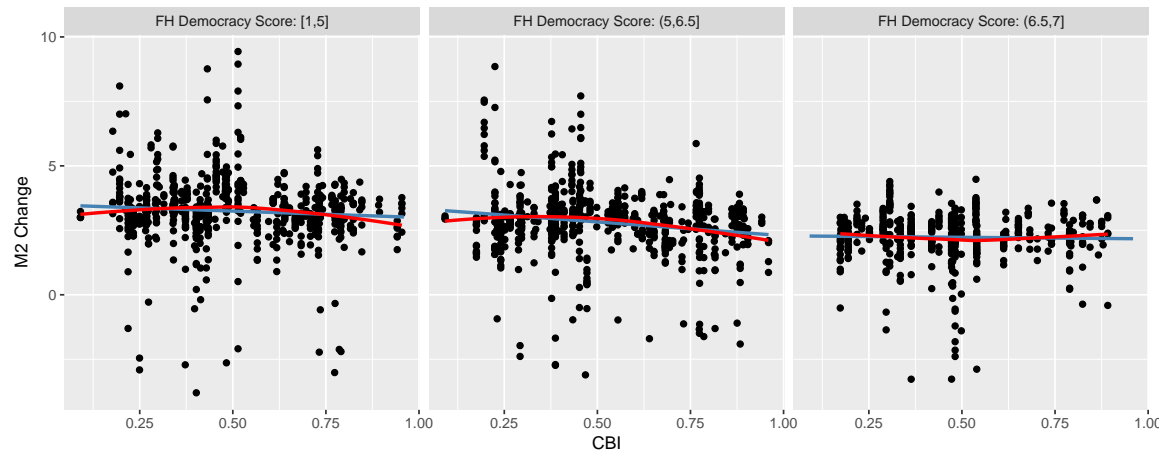
Second Interaction:

Claim on conditionality (Figure 2 in manuscript): *“Figure 2 shows the marginal effect of CBI as POLITY and FREEDOM HOUSE democracy vary. The graphs confirm our expectations; the marginal effect of CBI is downward sloping but it is negative and statistically significant at high levels of democracy only (POLITY scores above 16). At low levels of POLITY, the marginal effect of CBI is positive but statistically insignificant. Similarly, the marginal effect of CBI is negative and significant only when the FREEDOM HOUSE score is greater than about 5.”* (p. 49)

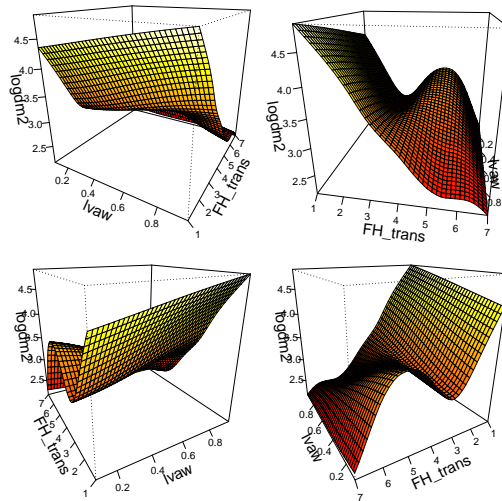
Key variables for conditional relationship 2 : Outcome Y: “M2 growth” (`logdm2`); treatment D: “central bank independence” (CBI); moderator X: “Freedom House democracy score” (`FH_trans`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

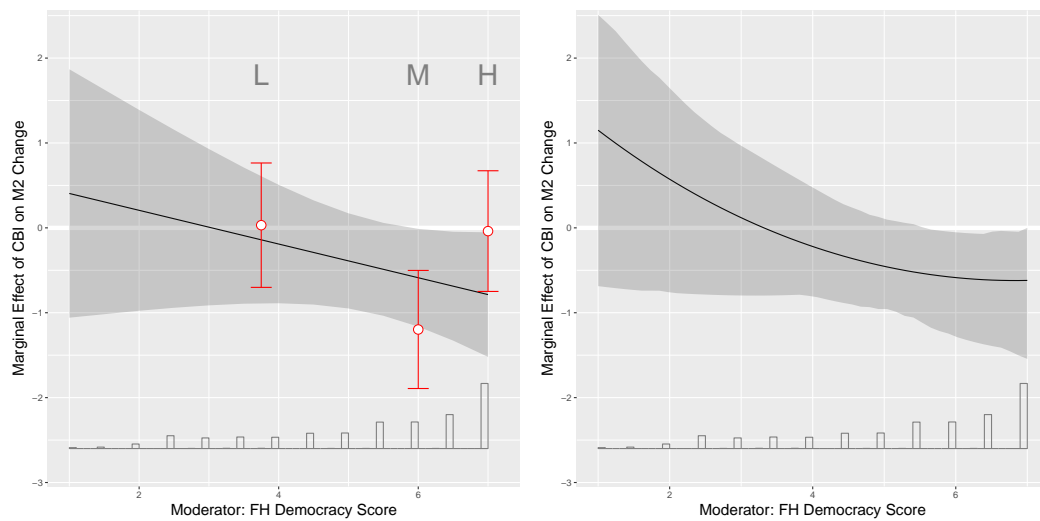
FIGURE B10. RESULTS FROM [BODEA AND HICKS \(2015b\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

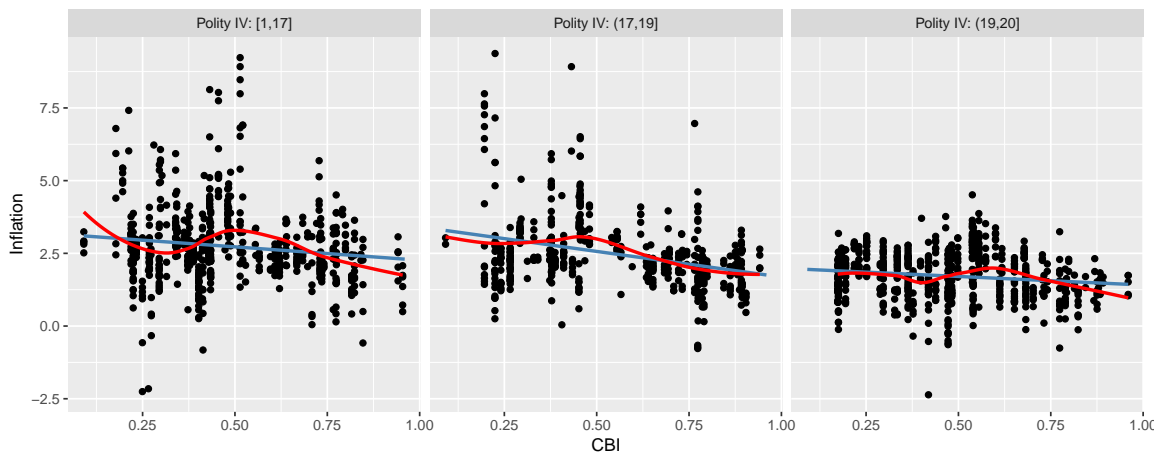
Third Interaction:

Claim on conditionality (Figure 4 in manuscript): *“Figure 4 shows, however, that the marginal effect of CBI is significant only at high levels of Polity and FREEDOM HOUSE. The marginal effect line is downward sloping, suggesting that only for Polity scores greater than about 14 (FREEDOM HOUSE scores greater than about 4.5) does CBI significantly reduce inflation.”* (p. 52)

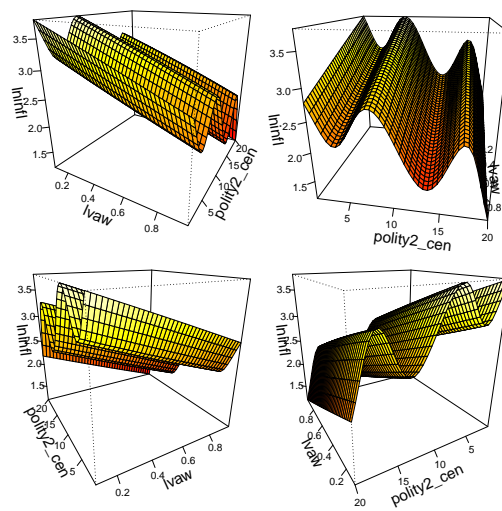
Key variables for conditional relationship: Outcome Y: “inflation” (`lninfl`); treatment D: “central bank independence” (CBI); moderator X: “Polity IV score” (`polity2_cen`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

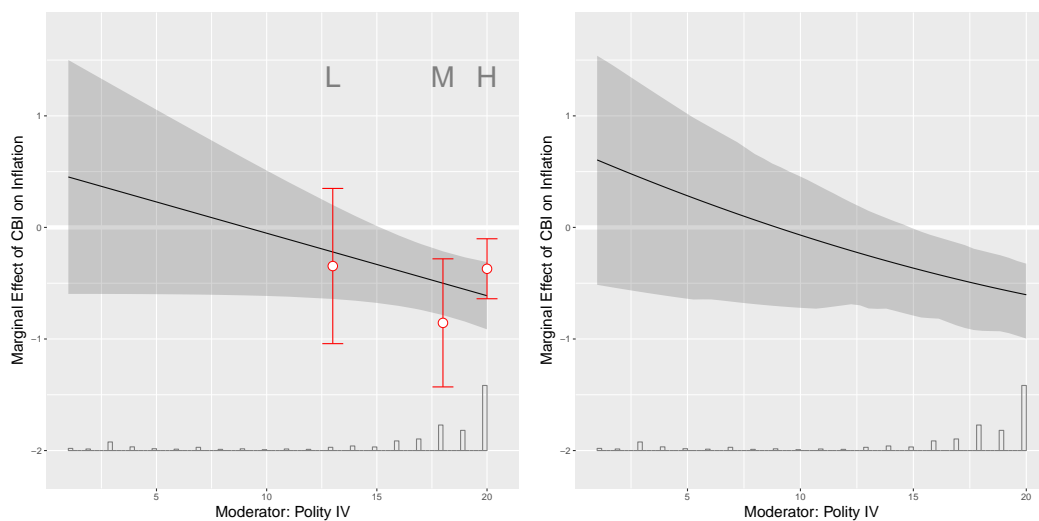
FIGURE B11. RESULTS FROM [BODEA AND HICKS \(2015b\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
 (d) Marginal Effects from Kernel Estimator

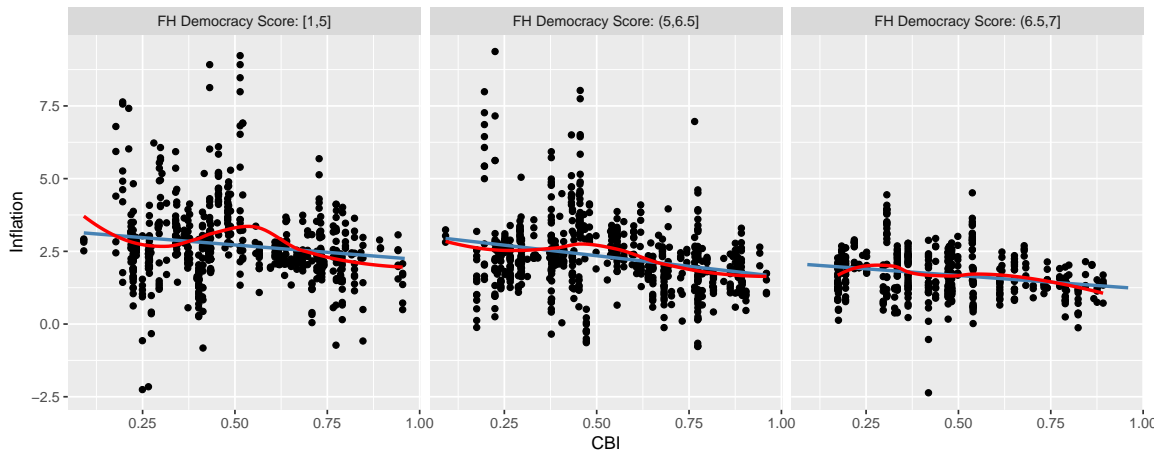
Fourth Interaction:

Claim on conditionality (Figure 4 in manuscript): *“Figure 4 shows, however, that the marginal effect of CBI is significant only at high levels of Polity and FREEDOM HOUSE. The marginal effect line is downward sloping, suggesting that only for Polity scores greater than about 14 (FREEDOM HOUSE scores greater than about 4.5) does CBI significantly reduce inflation.”* (p. 52)

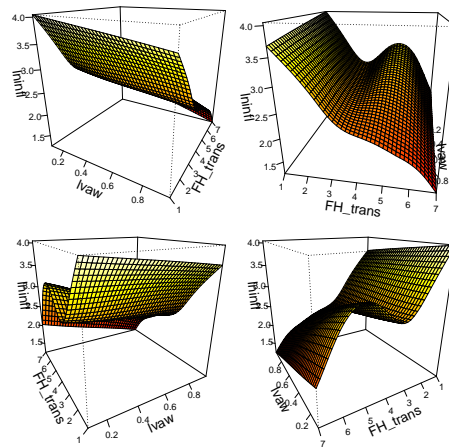
Key variables for conditional relationship: Outcome Y: “inflation” (`lninfl`); treatment D: “central bank independence” (CBI); moderator X: “Freedom House democracy score” (`FH_trans`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

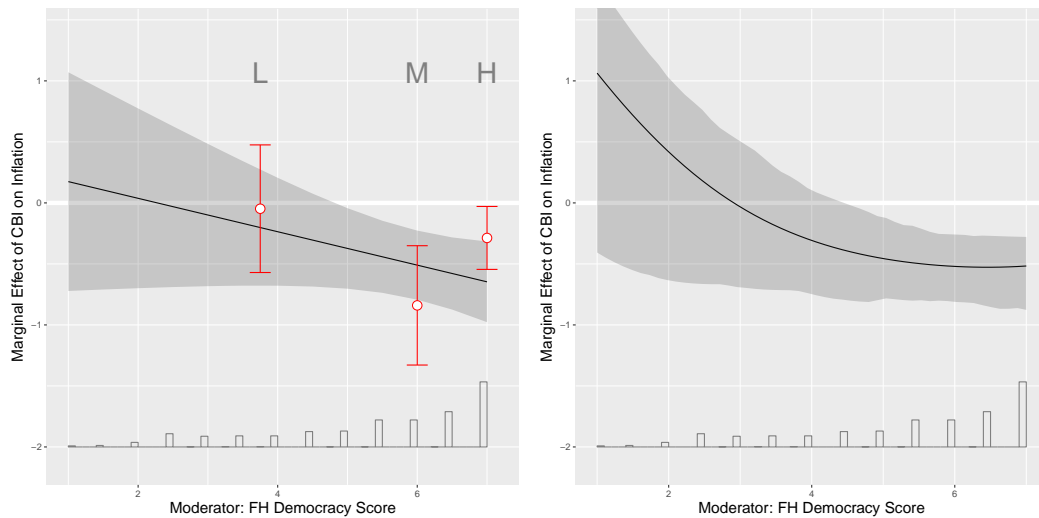
FIGURE B12. RESULTS FROM [BODEA AND HICKS \(2015b\)](#)



(a) Raw data



(b) GAM



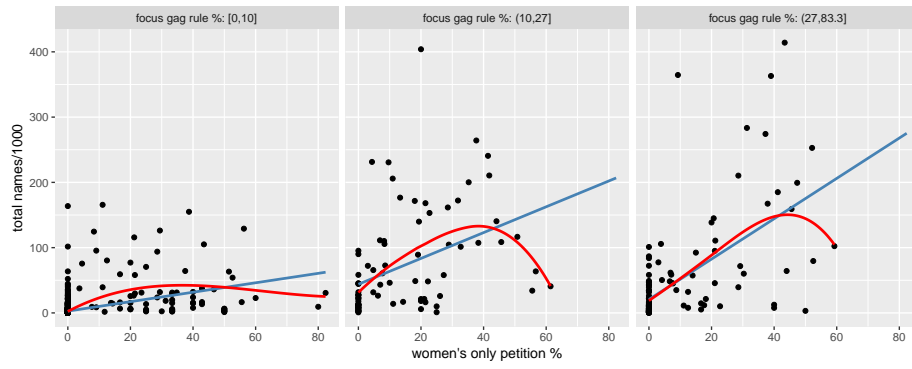
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
 (d) Marginal Effects from Kernel Estimator

.6 Carpenter and Moore (2014) APSR

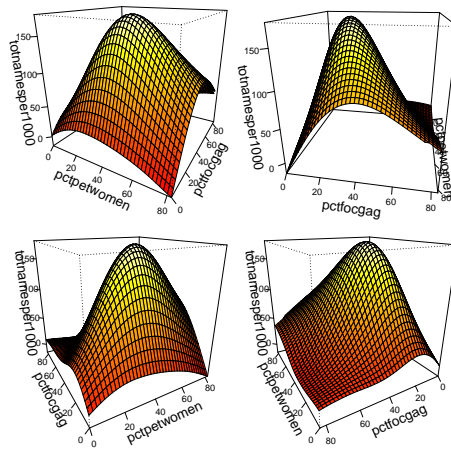
Claim on conditionality (Figure 4 in manuscript): “[W]e note that women’s canvassing was far more efficacious when the prayer of the petition contained a protest against the gag rule. Figure 4 presents the marginal effect of the percentage of county petitions canvassed by women (in terms of additional signatures per 1,000 county population) as a function of the percentage of county petitions whose prayer focuses on the gag rule. ...The marginal-effects plot demonstrates that the effect of women’s canvassing is positive and statistically differentiable from zero for all values of the gag-rule focus variable.” (pp. 490-91).

Key variables for the conditional relationship: Outcome Y: “total names per 1000” (totnamesper1000); treatment D: “percent women’s only petition” (pctpetwomen); moderator X: “percent focus gag rule” (pctfocgag).

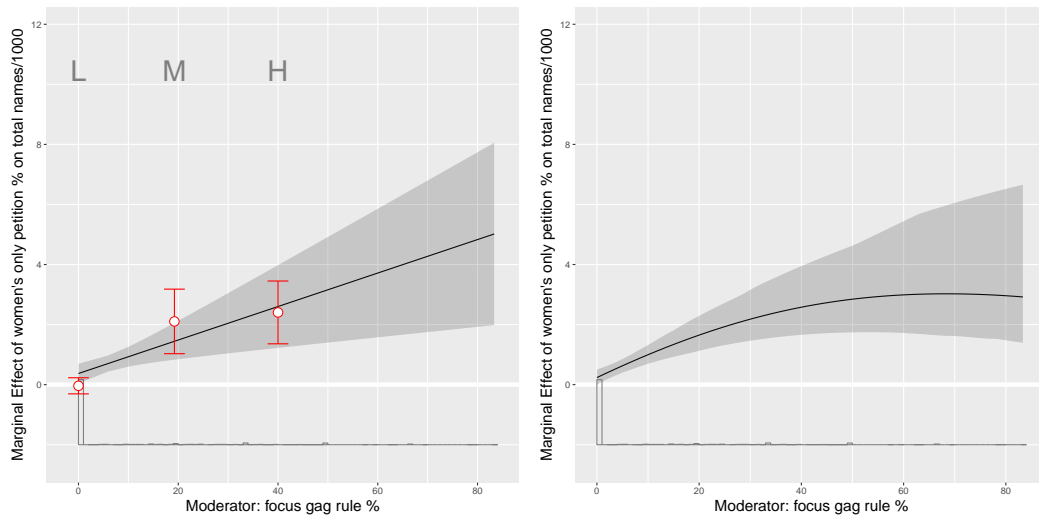
FIGURE B13. RESULTS FROM CARPENTER AND MOORE (2014)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.7 Chapman (2009) IO

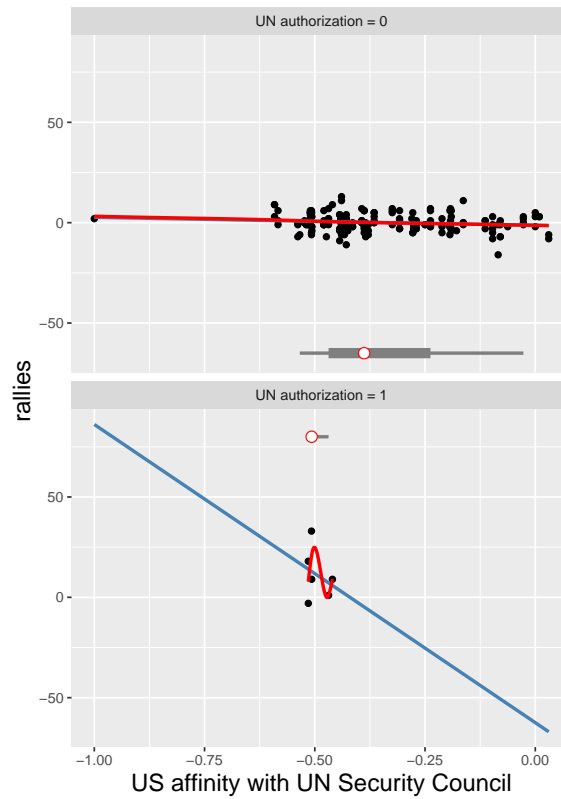
Claim on conditionality: *“This article tests this conditional relationship in the context of changes in presidential approval surrounding military disputes, using a measure of preference distance between the United States and veto-wielding members of the UN Security Council. Findings indicate that short-term changes in presidential approval surrounding the onset of military disputes in the United States between 1946 and 2001 have been significantly large when accompanied by a positive resolution for a Security Council that is more distant in terms of foreign policy preferences”* (Abstract).

“Rallies with UN authorization are only larger than average when the pivotal member is ideologically distant from the United States... Clearly, the effect of authorization on rallies decreases as similarity increases: foreign policy actions that receive authorization from a less conservative institution receive similar rallies to those that do not receive authorization from an IO” (p. 756)

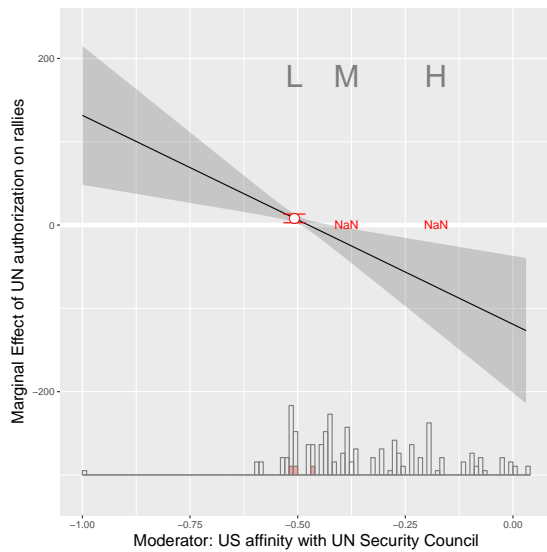
Key variables for the conditional relationship: Outcome Y: “rallies” (**rally**); treatment D: “UN authorization” (**unauth**); moderator X: “US affinity with UN Security Council ” (**S**).

Note: Among 196 observations, there are only 6 positive cases (**unauth=1**).

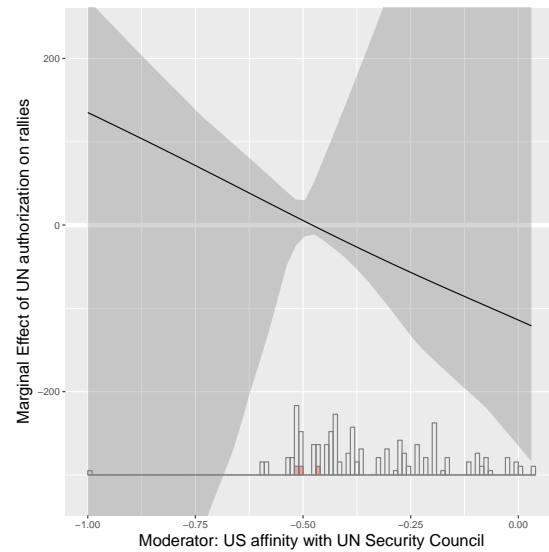
FIGURE B14. RESULTS FROM [CHAPMAN \(2009\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)



(c) Marginal Effects from Kernel Estimator

.8 Clark and Golder (2006) CPS

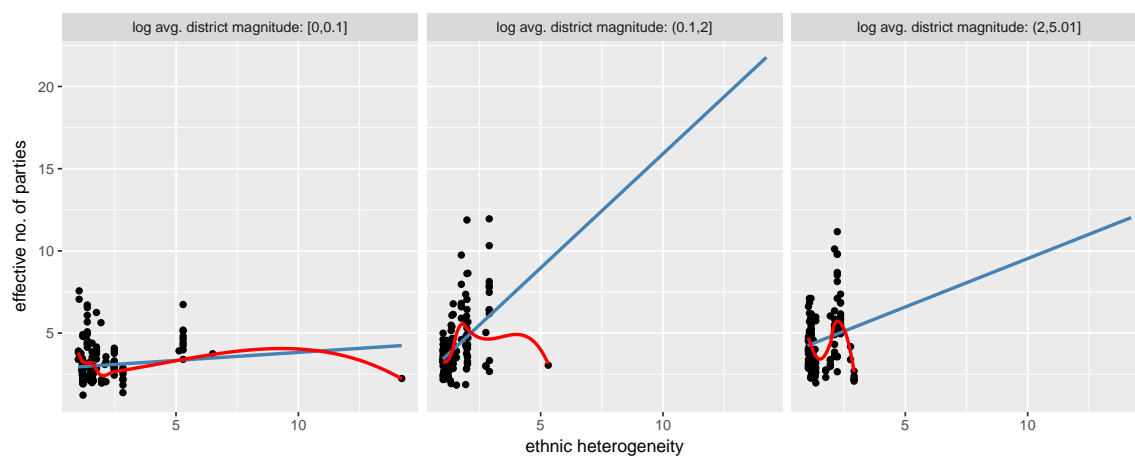
First Interaction:

Claims on conditionality (Figure 1 in manuscript): *“All three figures (in Figure 1) clearly illustrate that in established democracies, ethnic heterogeneity significantly increases the number of parties once the electoral system is sufficiently permissive. This is exactly what Duverger’s theory predicts”* (p. 700).

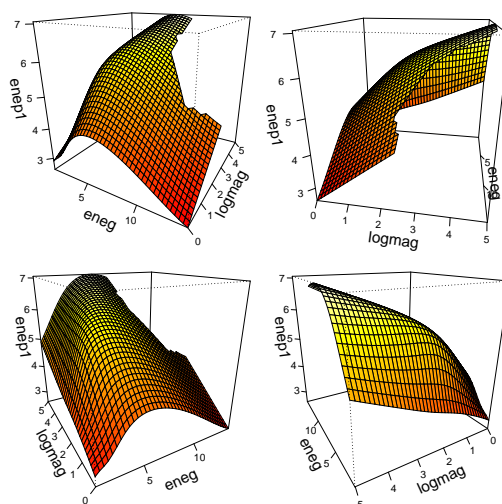
Key variables for the first conditional relationship: Outcome Y: “effective number of parties” (`enep1`); treatment D: “ethnic heterogeneity” (`eneg`); moderator X: “log average district magnitude” (`logmag`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

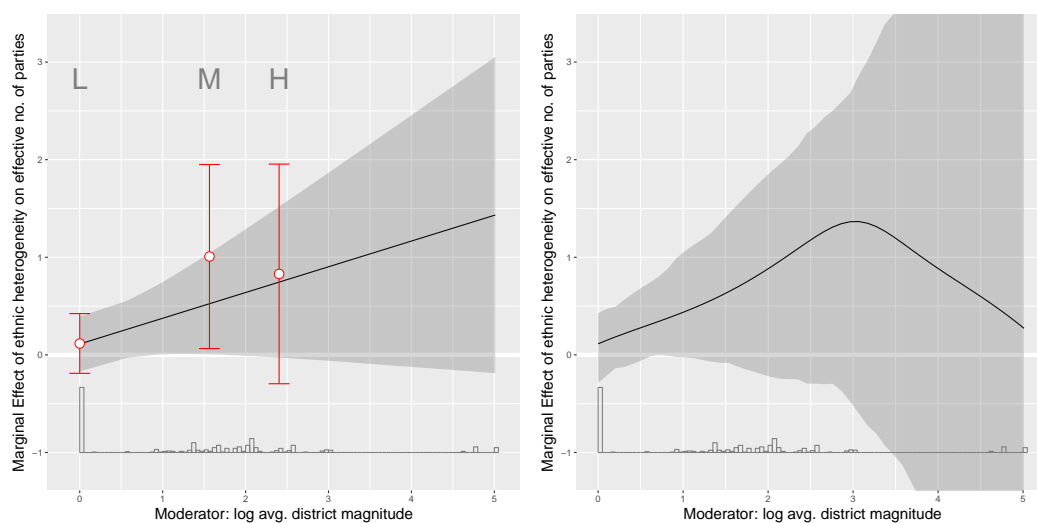
FIGURE B15. RESULTS FROM CLARK AND GOLDER (2006)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

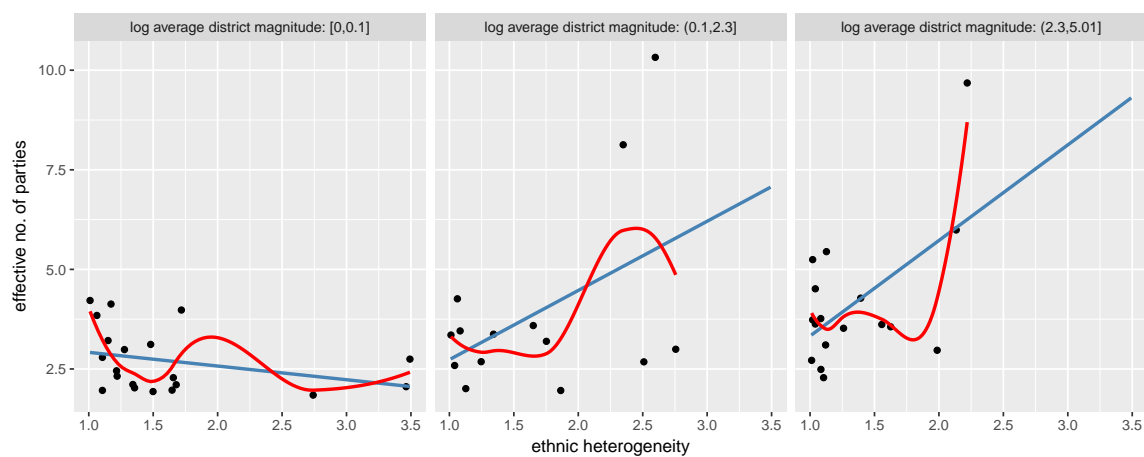
Second Interaction:

Claims on conditionality (Figure 1 in manuscript, middle panel): *“All three figures (in Figure 1) clearly illustrate that in established democracies, ethnic heterogeneity significantly increases the number of parties once the electoral system is sufficiently permissive. This is exactly what Duverger’s theory predicts”* (p. 700).

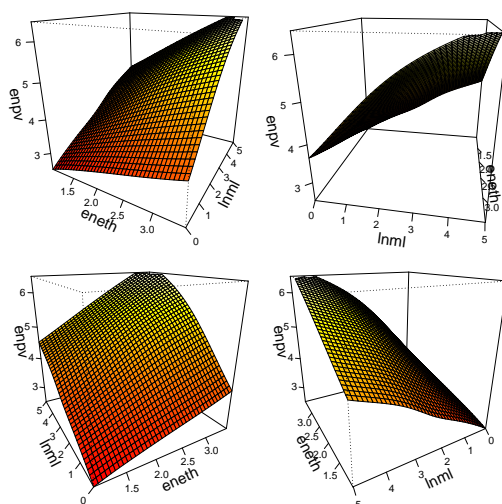
Key variables for the first conditional relationship: Outcome Y: “effective number of parties” (**enpv**); treatment D: “ethnic heterogeneity” (**eneth**); moderator X: “log average district magnitude” (**lnml**).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

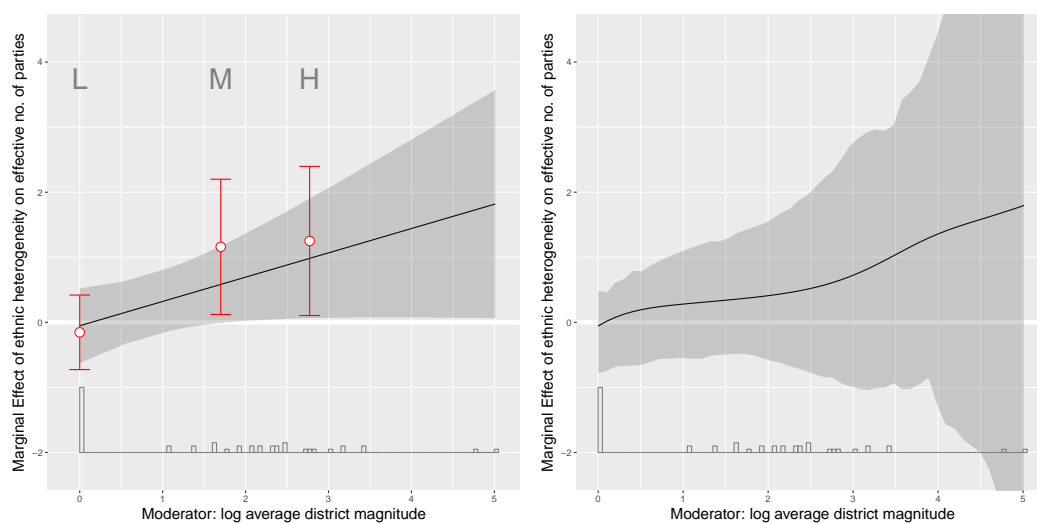
FIGURE B16. RESULTS FROM CLARK AND GOLDER (2006)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)

(d) Marginal Effects from Kernel Estimator

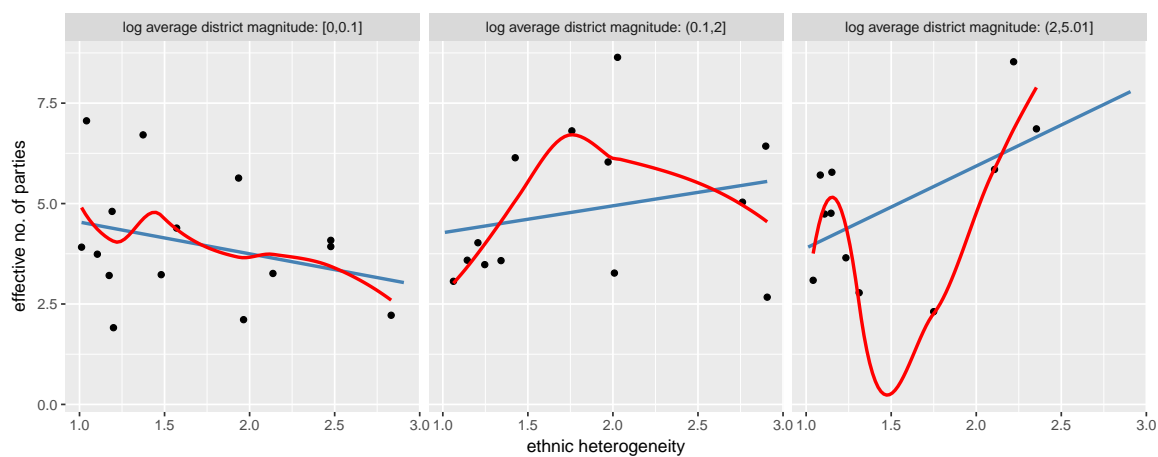
Third Interaction:

Claims on conditionality (Figure 1 in manuscript, bottom panel): *“All three figures (in Figure 1) clearly illustrate that in established democracies, ethnic heterogeneity significantly increases the number of parties once the electoral system is sufficiently permissive. This is exactly what Duverger’s theory predicts”* (p. 700).

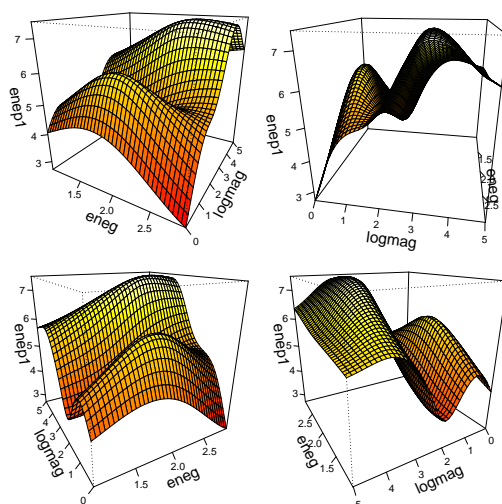
Key variables for the first conditional relationship: Outcome Y: “effective number of parties” (`enep1`); treatment D: “ethnic heterogeneity” (`eneg`); moderator X: “log average district magnitude” (`logmag`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

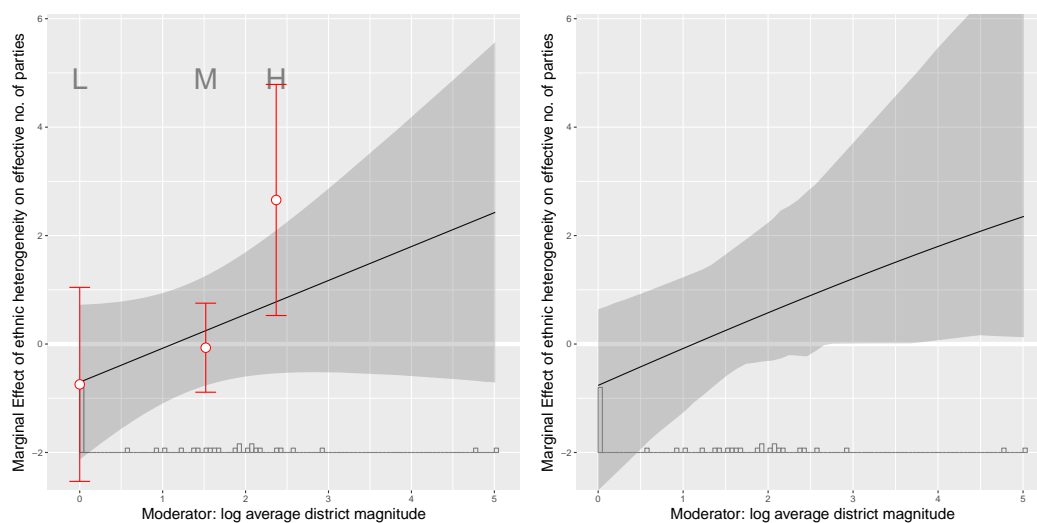
FIGURE B17. RESULTS FROM CLARK AND GOLDER (2006)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)

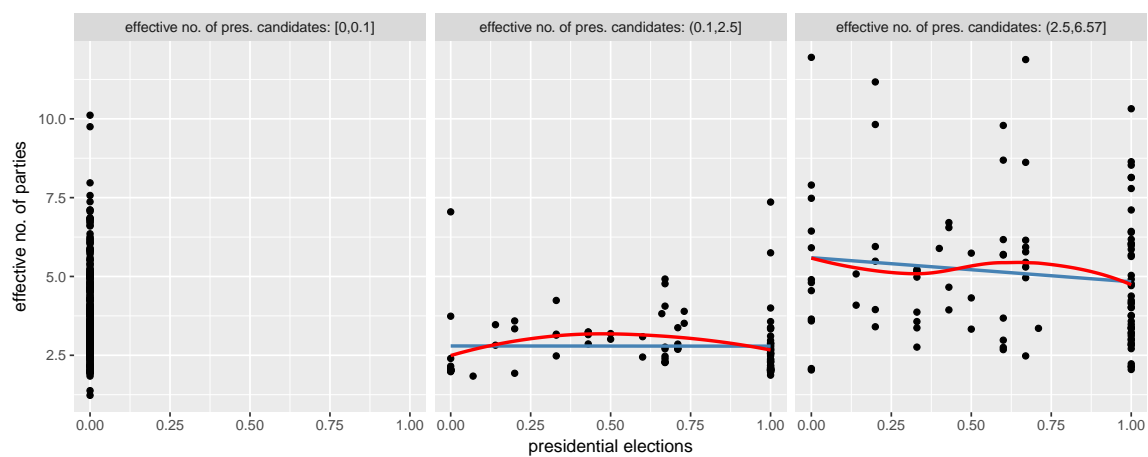
Fourth Interaction:

Claims on conditionality (Figure 2 in manuscript): *“Figure 2 plots the marginal effect of temporally proximate presidential elections. ... It should be clear that temporally proximate presidential elections have a strong reductive effect on the number of parties when there are few presidential candidates”* (p. 702).

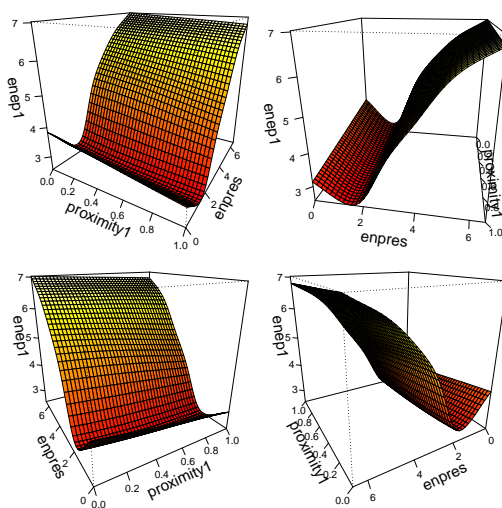
Key variables for the second conditional relationship: Outcome Y: “effective number of parties” (`enep1`); treatment D: “proximate presidential elections” (`proximity1`); moderator X: “effective number of pres. candidates” (`enpres`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

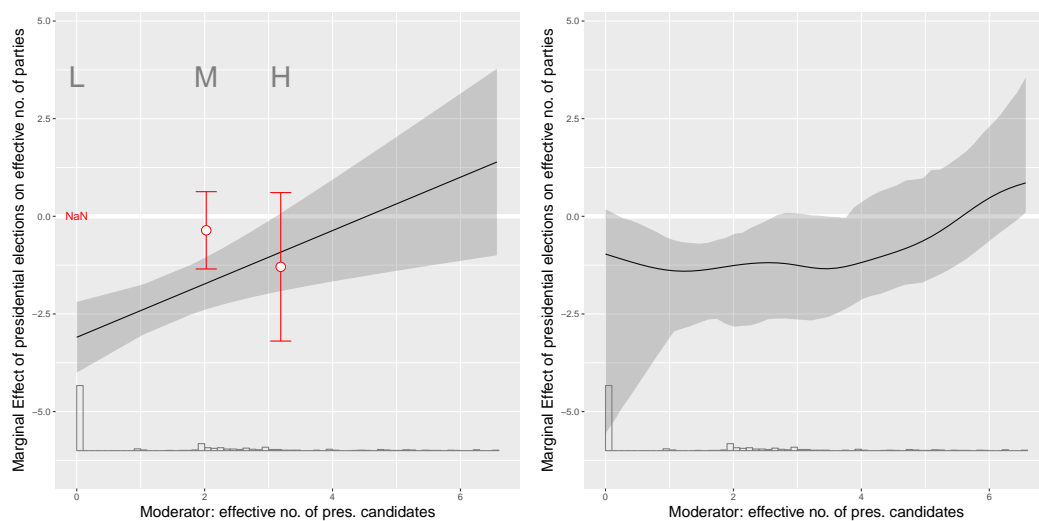
FIGURE B18. RESULTS FROM CLARK AND GOLDER (2006)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

.9 Clark and Leiter (2014) CPS

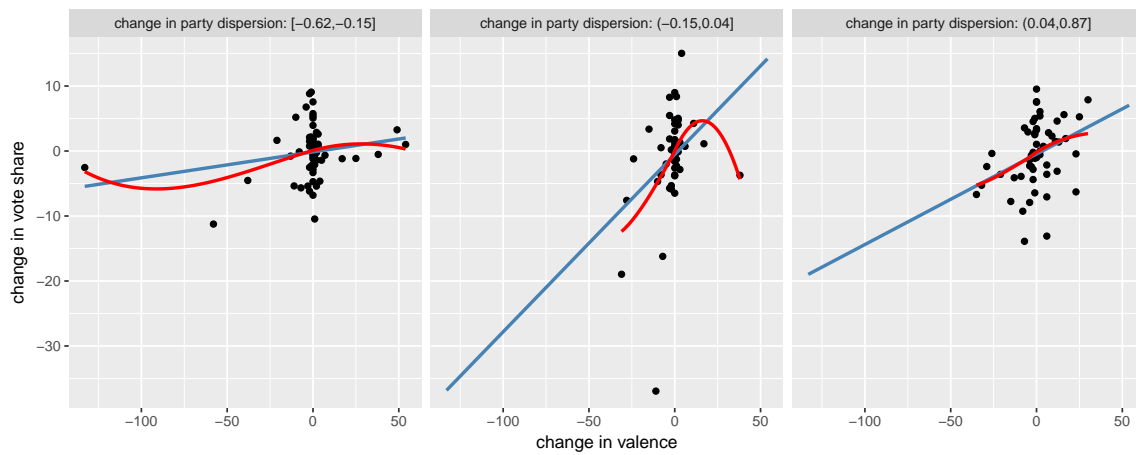
Claim on conditionality (Figure 2 in manuscript): *“We perform empirical analyses . . . and test the hypothesis that when parties are more ideologically proximate to the mean voter position, character-based valence attributes will be of greater significance in determining parties’ electoral fortunes. Surprisingly, we find no support for this hypothesis. Instead, our analyses suggest that the more ideologically dispersed parties are, the more likely it is that character-based valence attributes will affect parties’ vote shares.”* (Abstract).

“This relationship indicates that, as parties become more dispersed on the left-right spectrum, voters weigh changes in parties’ character-based valence attributes more heavily in their voting decisions.” (p. 185).

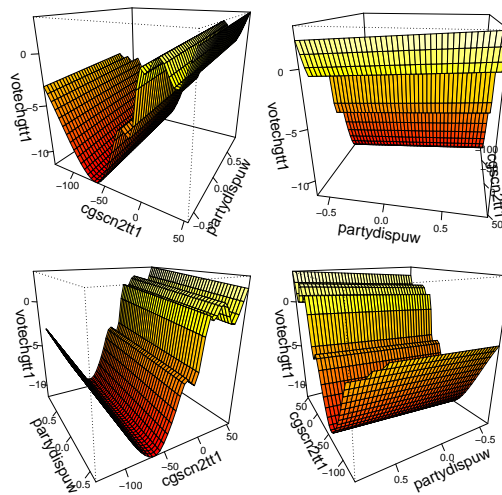
Key variables for the conditional relationship: Outcome Y: “change in vote share” (`res_votechgtt1`); treatment D: “change in valence” (`cgsn2tt1`); moderator X: “change in party dispersion” (`partydispw`).

Note: The dashed vertical line indicates the truncated interval of the moderator shown in the original marginal effect plot.

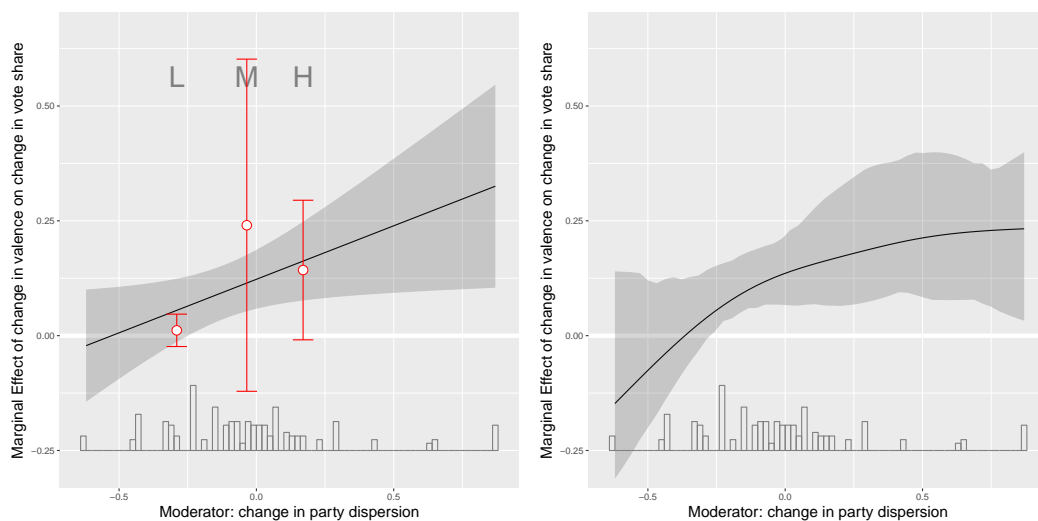
FIGURE B19. RESULTS FROM CLARK AND LEITER (2014)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

.10 Hellwig and Samuels (2007) CPS

First Interaction:

Claims on conditionality (Figure 1 in manuscript): *“Results support a government constraint hypothesis: Exposure to the world economy weakens connections between economic performance and support for political incumbents”* (Abstract).

“Figure 1 shows that trade openness reduces the positive relationship between economic performance and vote share for the incumbent.” (p. 294).

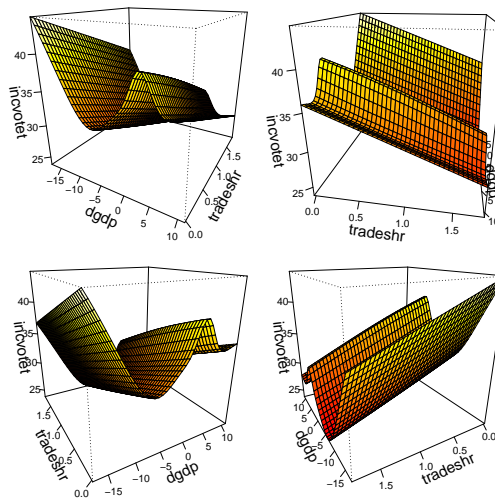
Key variables for the conditional relationships: Outcome Y: “election” (`incvotet`); treatment D: “economy” (`dgdg`); moderator X: “trade as share of GDP ” (`tradeshr`, in Figure 1).

Note: The dashed vertical line indicates the truncated interval of the moderator shown in the original marginal effect plot.

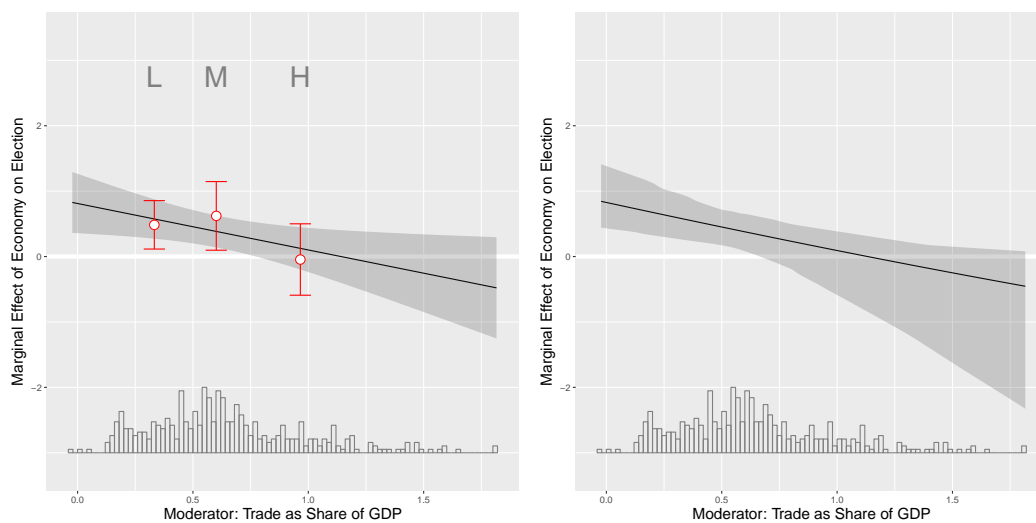
FIGURE B20. RESULTS FROM [HELLWIG AND SAMUELS \(2007\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

Second Interaction:

Claims on conditionality (Figure 2 in manuscript): *“Figure 2 shows that the exposure to international capital flows also reduces the relationship between economic performance and election outcomes”* (p. 294).

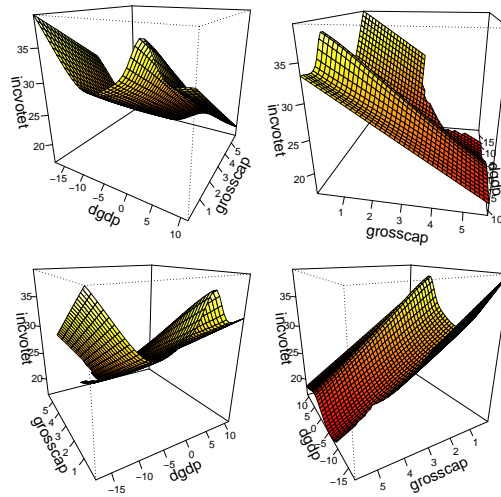
Key variables for the conditional relationships: Outcome Y: “election” (incvotet); treatment D: “economy” (dgdp); moderator X: “capital flows as share of GDP” (grosscap) .

Note: The dashed vertical line indicates the truncated interval of the moderator shown in the original marginal effect plot.

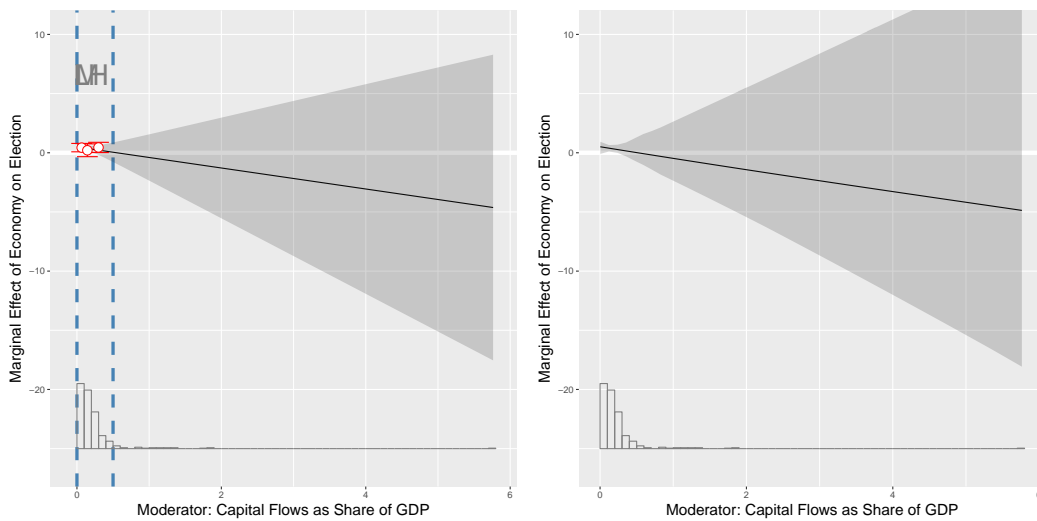
FIGURE B21. RESULTS FROM [HELLWIG AND SAMUELS \(2007\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

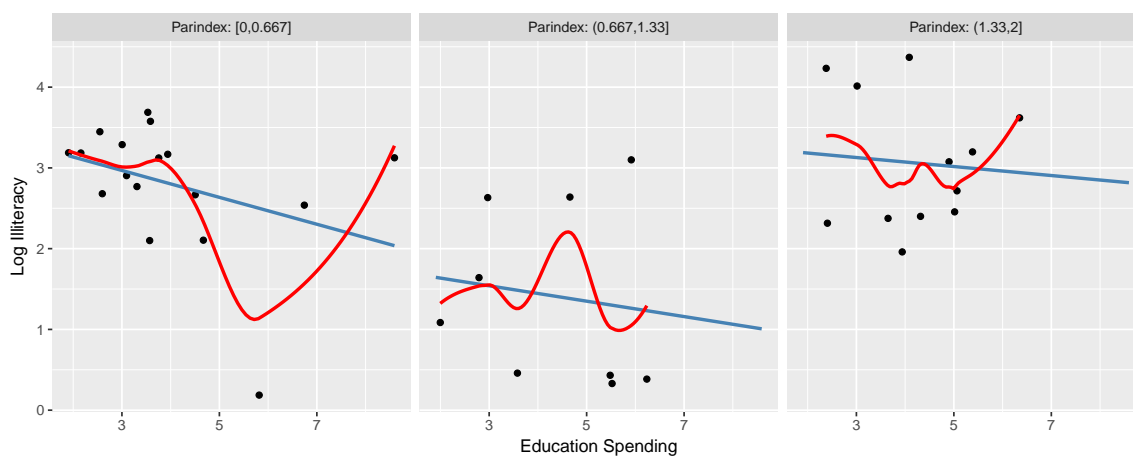
.11 Hicken and Simmons (2008) AJPS

Claim on conditionality (Figure 1, top left panel in manuscript): *“We find that though personal vote systems spend just as much on education as party vote systems, particularism in personal vote systems dampens the marginal effect of increased education spending on illiteracy and at its highest levels, incentives to cultivate a personal vote completely undermine the positive effects of increased education spending on literacy”* (Abstract).

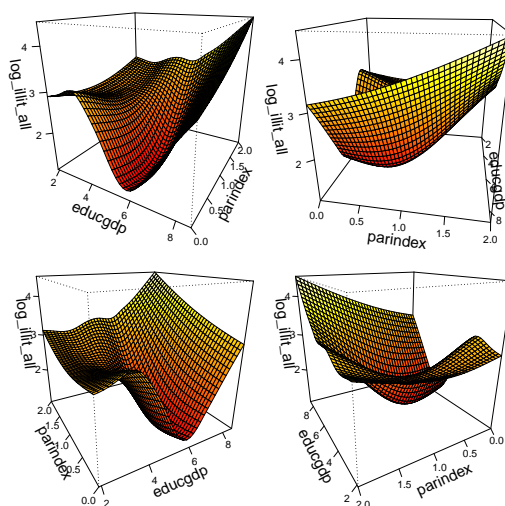
Key variables for the conditional relationship: Outcome Y: “log illiteracy” (`log_illit_all`); treatment D: “education spending” (`educgdp`); moderator X: “parindex” (`parindex`), which is the average of three institutional variables: “Ballot”, “Pool”, and “Vote”. We also use the three variables as moderators and investigate the conditional relationships separately.

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

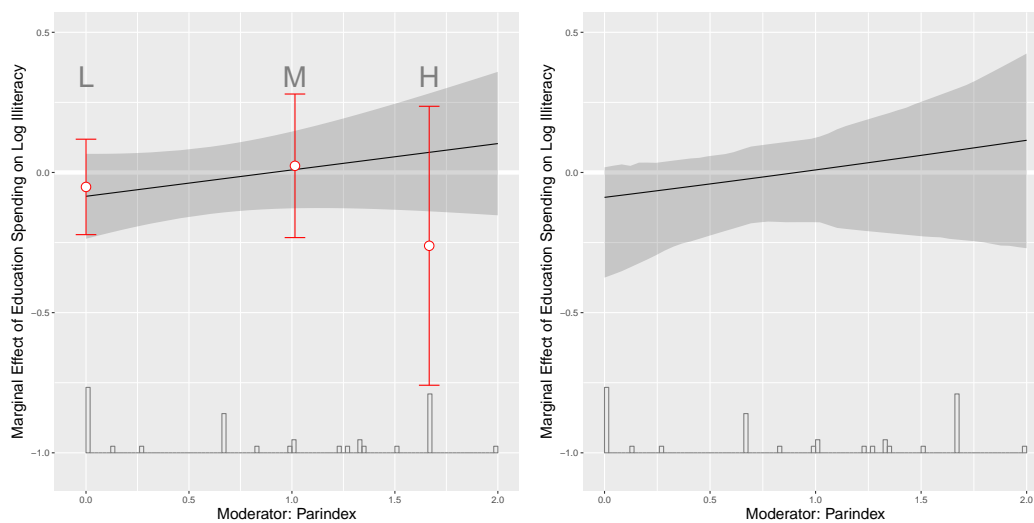
FIGURE B22. RESULTS FROM [HICKEN AND SIMMONS \(2008\)](#)



(a) Raw data



(b) GAM



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

.12 Huddy, Mason and Aarøe (2015) APSR

First Interaction:

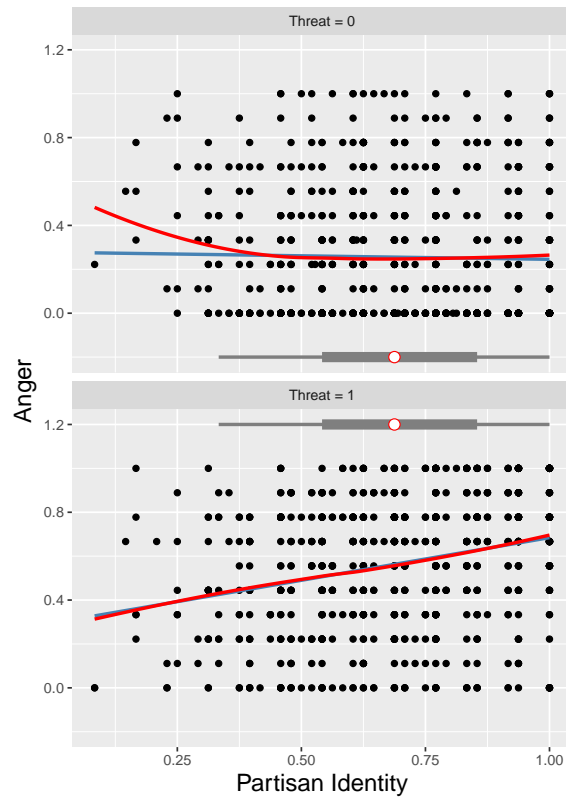
Claim on conditionality (Figure 2, top left panel in manuscript): *“A series of experiments underscore the power of partisan identity to generate action-oriented emotions that drive campaign activity. Strongly identified partisans feel angrier than weaker partisans when threatened with electoral loss and more positive when reassured of victory”* (Abstract).

“The figure (Figure 2) shows clearly that threat and reassurance arouse the most powerful emotion among the strongest partisan identifiers in the blog study.” (p. 12).

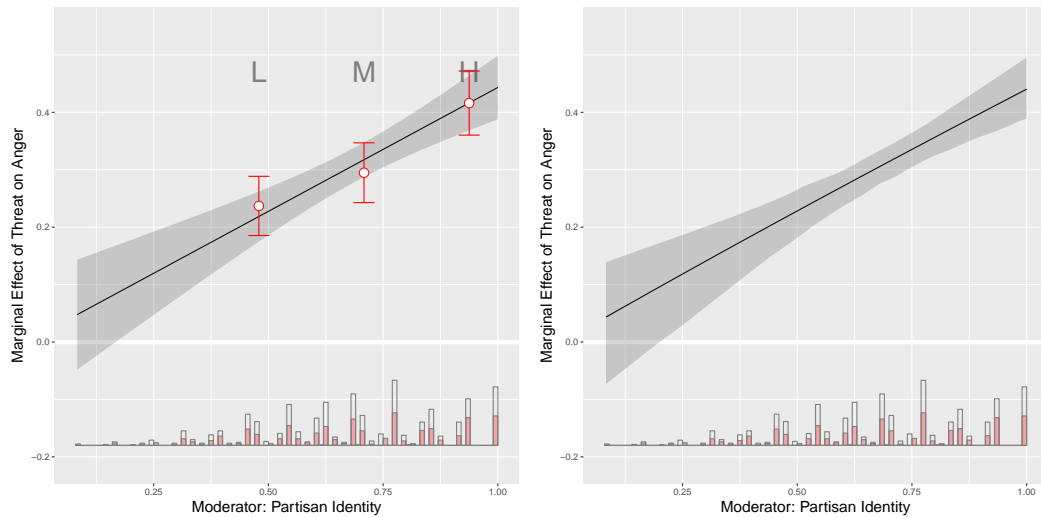
Key variables for the conditional relationship: Outcome Y: “anger” (`totangry`); treatment D: “threat” (`threat`); moderator X: “partisan identity ” (`pidentity`).

Key variables for the conditional relationship 2 (Figure 2B): Outcome Y: “enthusiasm” (`totpos`); treatment D: “support” (`support`); moderator X: “partisan identity ” (`pidentity`).

FIGURE B23. RESULTS FROM HUDDY, MASON AND AARØE (2015)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(c) Marginal Effects from Kernel Estimator

Second Interaction:

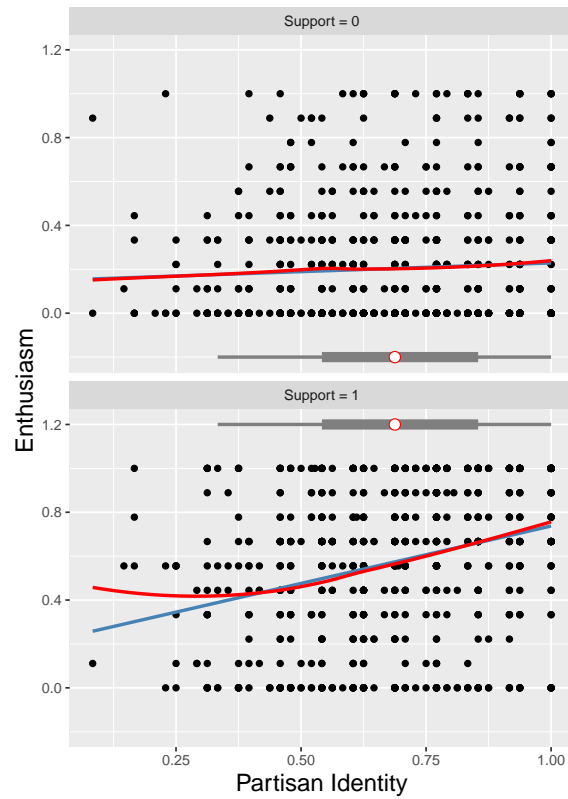
Claim on conditionality (Figure 2, top right panel in manuscript): “A series of experiments underscore the power of partisan identity to generate action-oriented emotions that drive campaign activity. Strongly identified partisans feel angrier than weaker partisans when threatened with electoral loss and more positive when reassured of victory” (Abstract).

“The figure (Figure 2) shows clearly that threat and reassurance arouse the most powerful emotion among the strongest partisan identifiers in the blog study.” (p. 12).

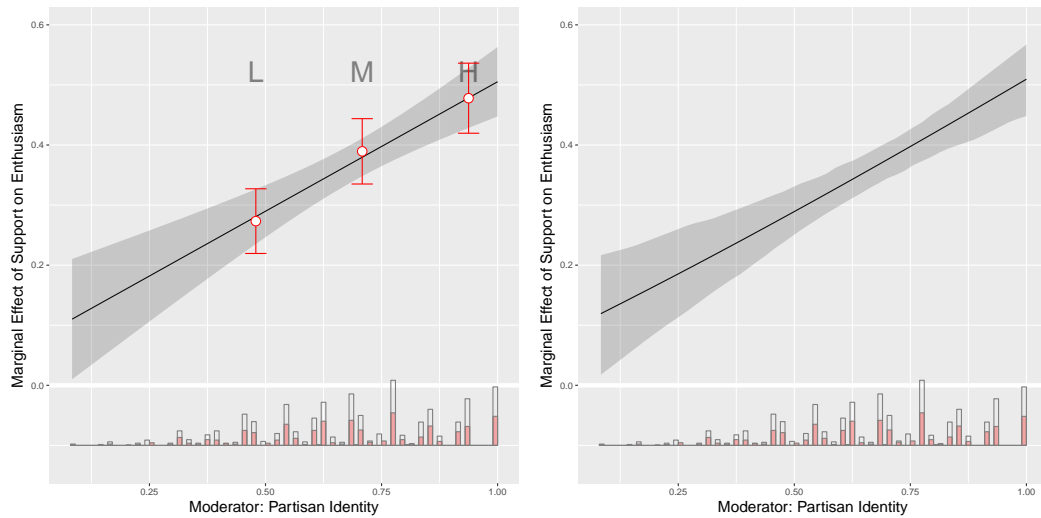
Key variables for the conditional relationship: Outcome Y: “anger” (totangry); treatment D: “threat” (threat); moderator X: “partisan identity ” (pidentity).

Key variables for the conditional relationship: Outcome Y: “enthusiasm” (totpos); treatment D: “support” (support); moderator X: “partisan identity ” (pidentity).

FIGURE B24. RESULTS FROM HUDDY, MASON AND AARØE (2015)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

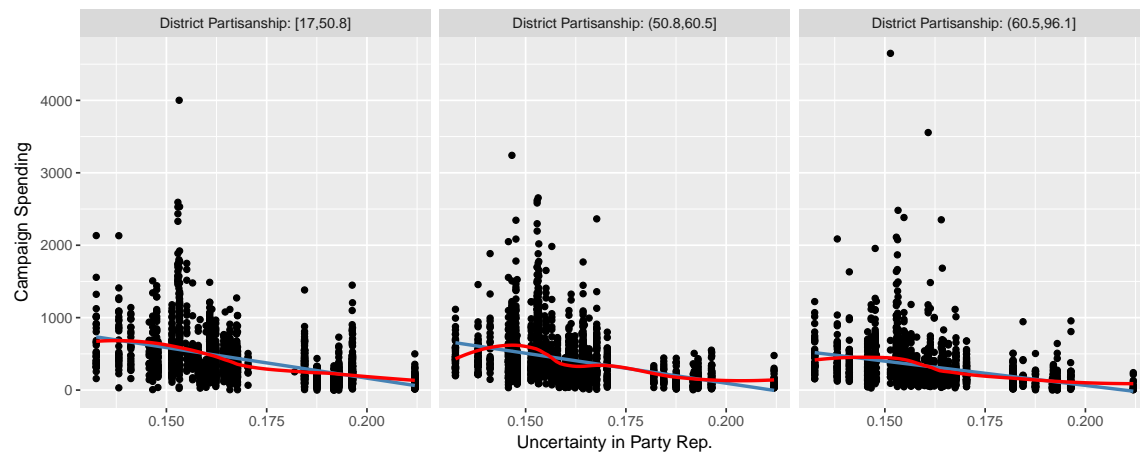
.13 Kim and LeVeck (2013) APSR

First interaction:

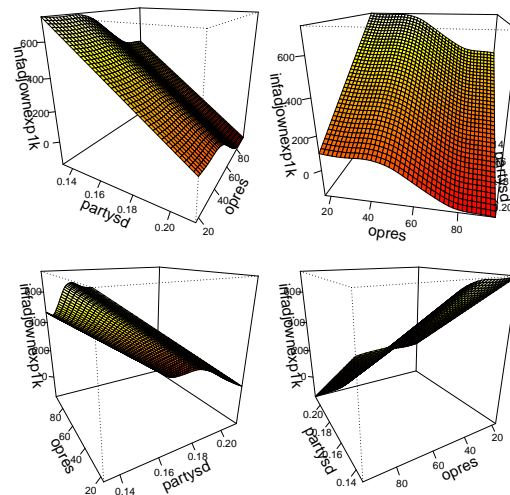
Claim on conditionality (Figure 2 in manuscript): *“Figure 2 shows that the effect of the greater uncertainty in the incumbent party’s reputation on campaign spending attenuates as districts become less marginal. Indeed, greater uncertainty in the incumbent’s party reputation seems to actually increase spending in the least marginal districts.”* (499).

Key variables for the conditional relationship: Outcome Y: “Campaign Spending” (`infadjownexp1k`); treatment D: “uncertainty in incumbent party’s position” (`partysd`); moderator X: “district partisanship” (`opres`).

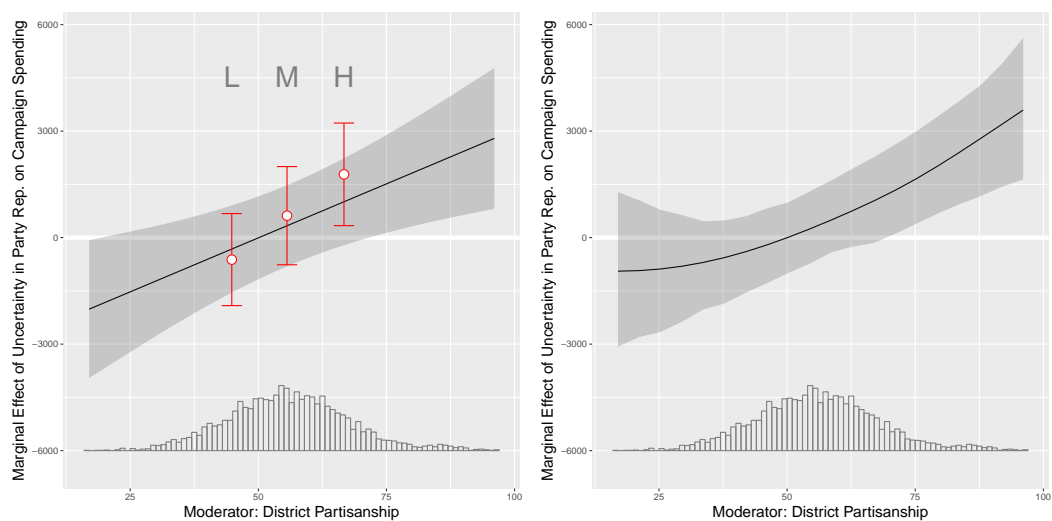
FIGURE B25. RESULTS FROM [KIM AND LeVECK \(2013\)](#)



(a) Raw data



(b) GAM plot



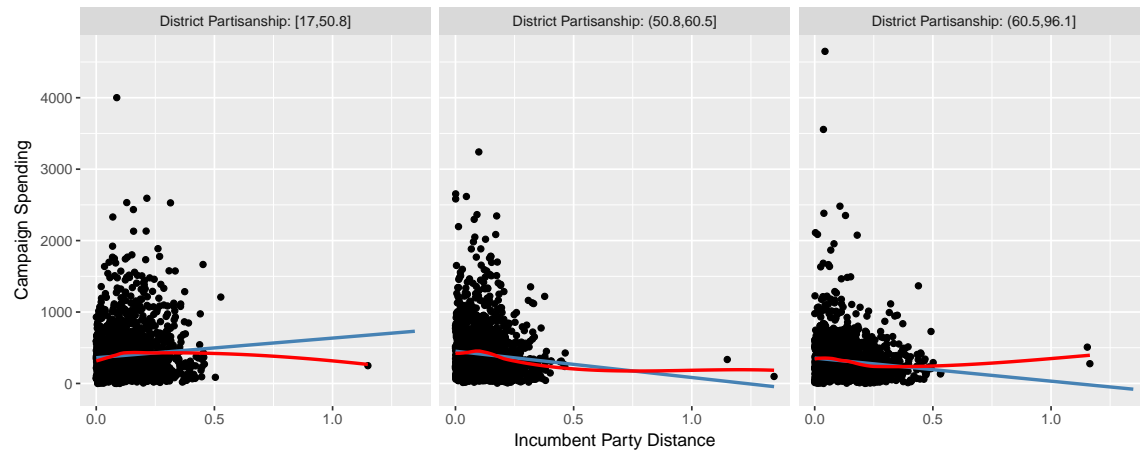
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)

Second interaction:

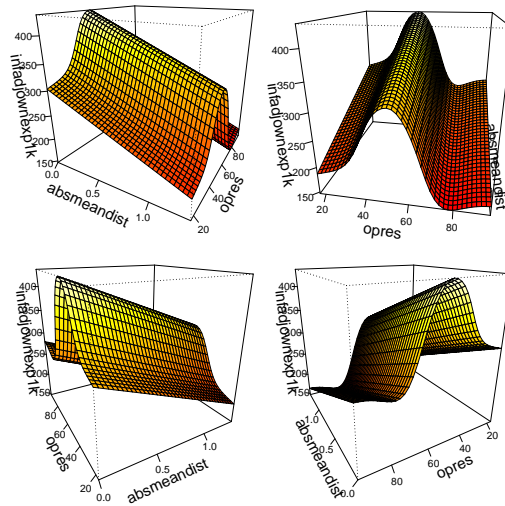
Claim on conditionality (Figure 3 in manuscript): *“On the other hand, it is less helpful for the incumbent to be far from his party in less marginal districts. Accordingly, the positive coefficient on Incumbent-Party Distance \times District Partisanship shows that distance from the party decreases incumbent spending less as a district becomes less marginal.”* (500).

Key variables for the conditional relationship: Outcome Y: “Campaign Spending” (`infadjownexp1k`); treatment D: “incumbent party distance” (`absmeandist`); moderator X: “district partisanship” (`opres`).

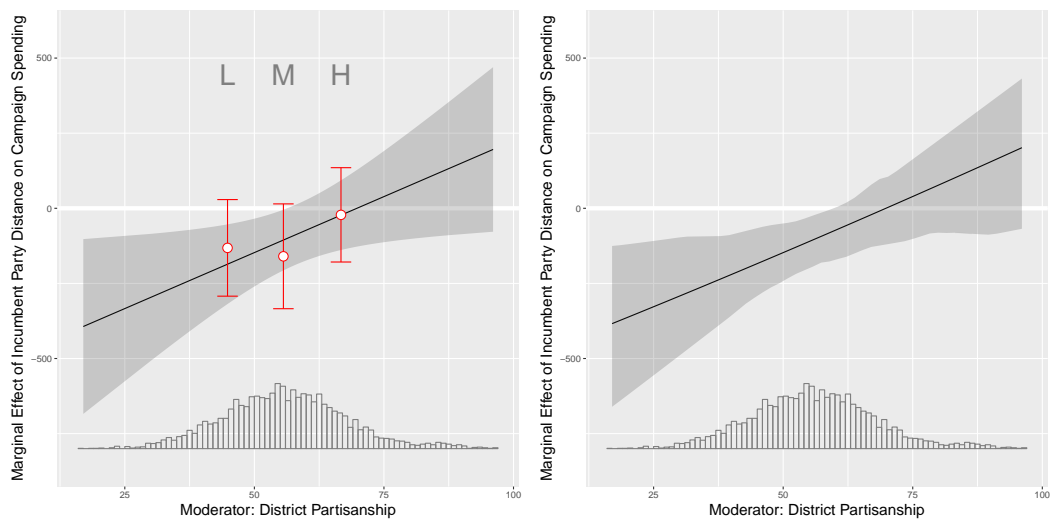
FIGURE B26. RESULTS FROM [KIM AND LeVECK \(2013\)](#)



(a) Raw data



(b) GAM plot



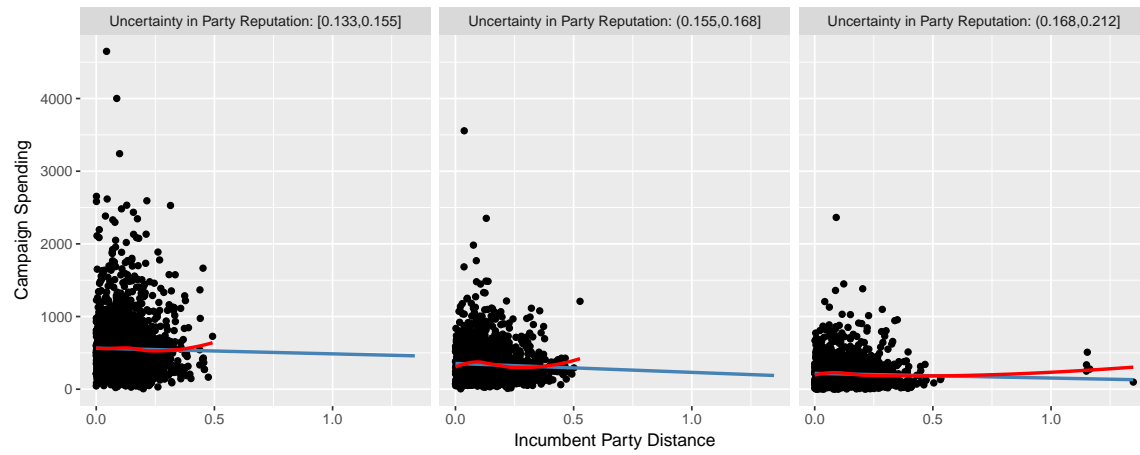
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

Third interaction:

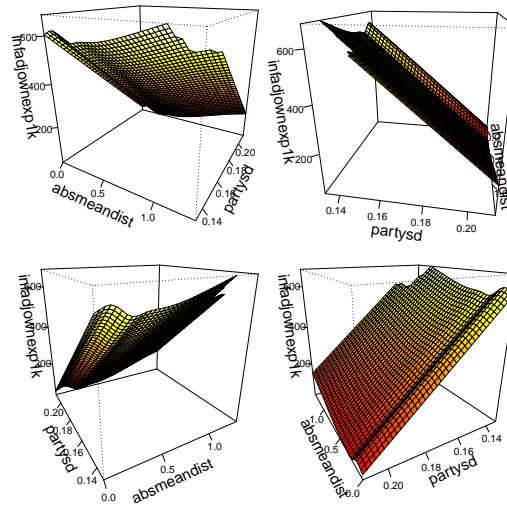
Claim on conditionality (Figure 4 in manuscript): *“Figure 4 indicates that in 1972, deviating from the party did not decrease an incumbent’s spending. Distancing oneself from the party only reduced incumbents’ spending once the parties become sufficiently unified in their voting behavior.”* (500).

Key variables for the conditional relationship: Outcome Y: “Campaign Spending” (`infadjownexp1k`); treatment D: “incumbent party distance” (`absmeandist`); moderator X: “uncertainty in party reputation” (`partysd`).

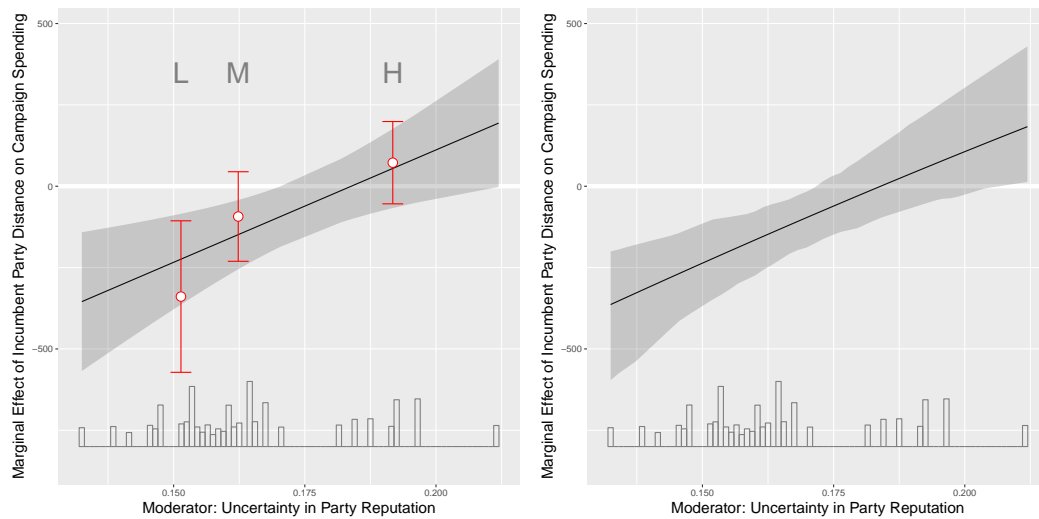
FIGURE B27. RESULTS FROM [KIM AND LeVECK \(2013\)](#)



(a) Raw data



(b) GAM plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.14 Malesky, Schuler and Tran (2012) APSR

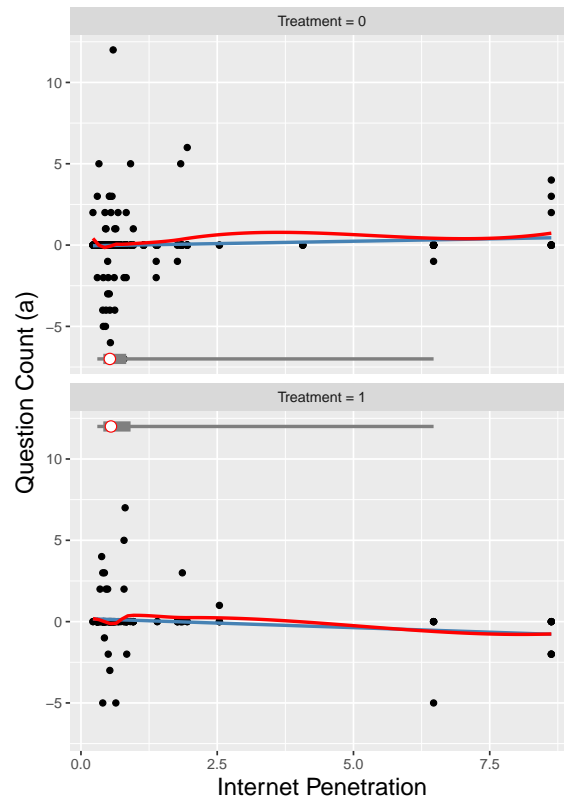
First interaction:

Claim on conditionality (Figure 1, top left panel in manuscript): *“We find no evidence of a direct effect of the transparency treatment on delegate performance; however, further analysis reveals that delegates subjected to high treatment intensity demonstrate robust evidence of curtailed participation and damaged reelection prospects. These results make us cautious about the export of transparency without electoral sanctioning.”* (Abstract).

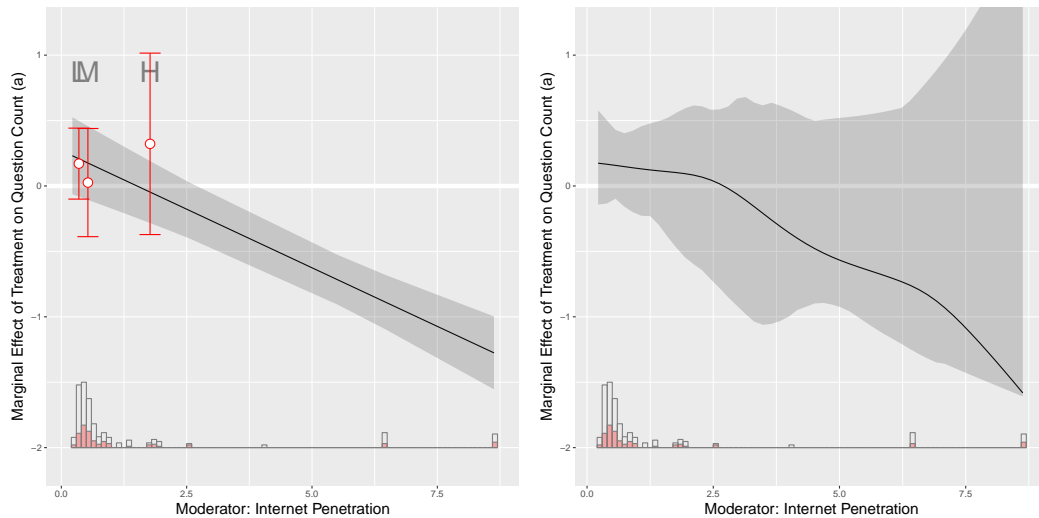
Key variables for the conditional relationship: Outcome Y: “change in questions asked” (`d.question_count`); treatment D: “sunshine” (`t2`); moderator X: “internet penetration” (`internet_users100`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

FIGURE B28. RESULTS FROM MALESKY, SCHULER AND TRAN (2012)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator

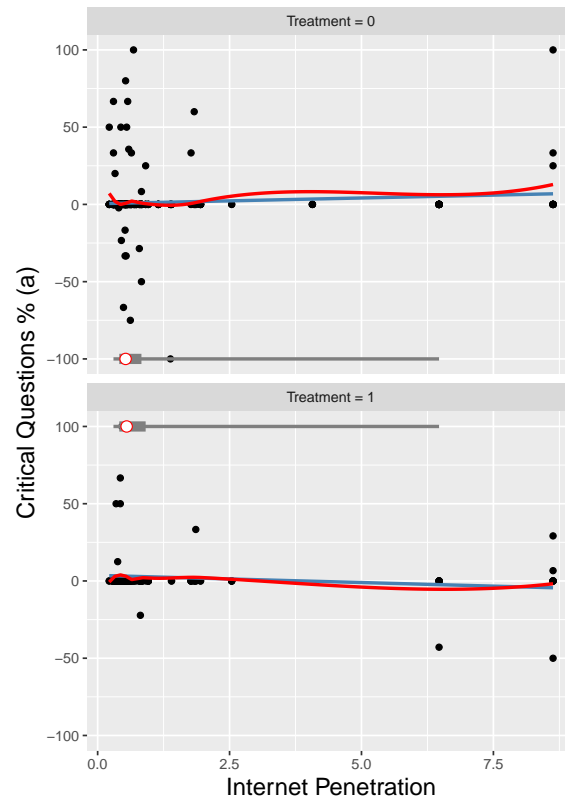
Second interaction:

Claim on conditionality (Figure 1, bottom left panel in manuscript): *“We find no evidence of a direct effect of the transparency treatment on delegate performance; however, further analysis reveals that delegates subjected to high treatment intensity demonstrate robust evidence of curtailed participation and damaged reelection prospects. These results make us cautious about the export of transparency without electoral sanctioning.”* (Abstract).

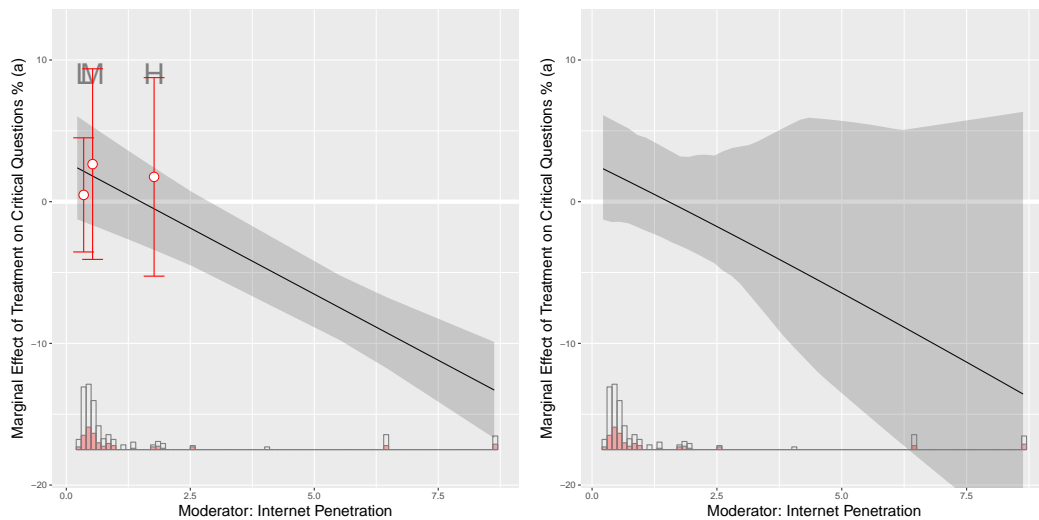
Key variables for the conditional relationship: Outcome Y: “change in critical questions (%)” (`d.criticize_total_per`); treatment D: “sunshine” (`t2`); moderator X: “internet penetration” (`internet_users100`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

FIGURE B29. RESULTS FROM MALESKY, SCHULER AND TRAN (2012)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

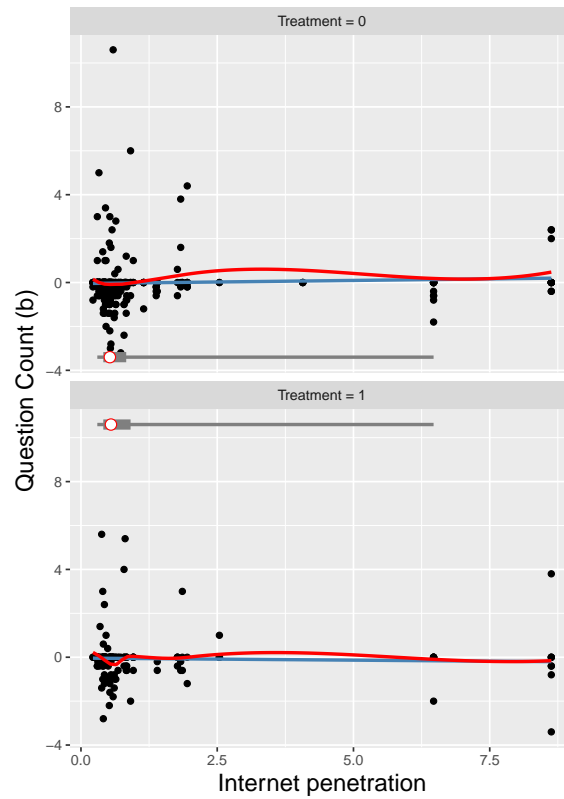
Third interaction:

Claim on conditionality (Figure 1, top right panel in manuscript): *“We find no evidence of a direct effect of the transparency treatment on delegate performance; however, further analysis reveals that delegates subjected to high treatment intensity demonstrate robust evidence of curtailed participation and damaged reelection prospects. These results make us cautious about the export of transparency without electoral sanctioning.”* (Abstract).

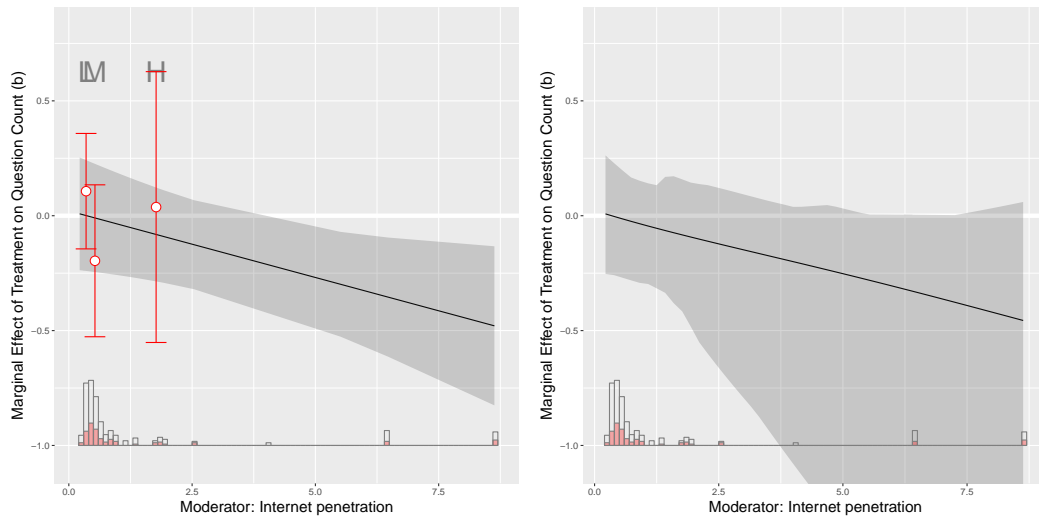
Key variables for the conditional relationship: Outcome Y: “change in questions asked” (`diff_quest`); treatment D: “sunshine” (`t2`); moderator X: “internet penetration” (`internet_users100`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

FIGURE B30. RESULTS FROM MALESKY, SCHULER AND TRAN (2012)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)

Fourth interaction:

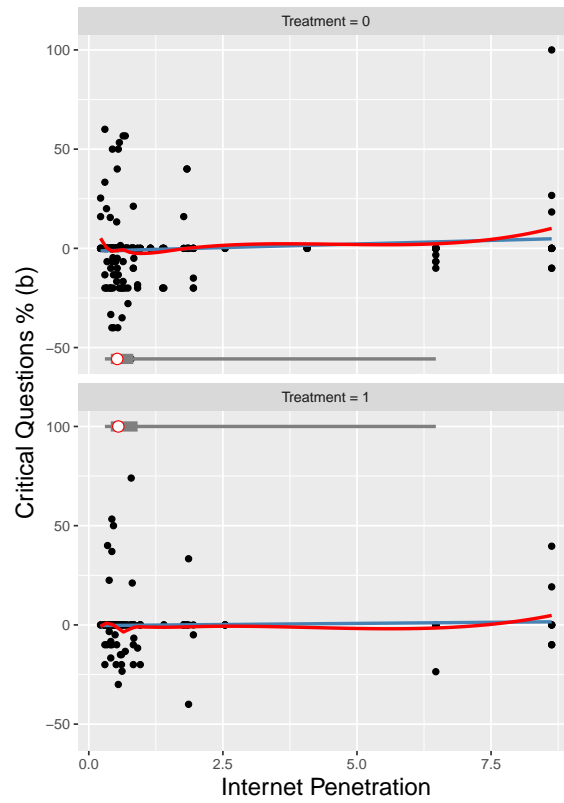
Claim on conditionality (Figure 1, bottom right panel in manuscript):

“We find no evidence of a direct effect of the transparency treatment on delegate performance; however, further analysis reveals that delegates subjected to high treatment intensity demonstrate robust evidence of curtailed participation and damaged reelection prospects. These results make us cautious about the export of transparency without electoral sanctioning.” (Abstract).

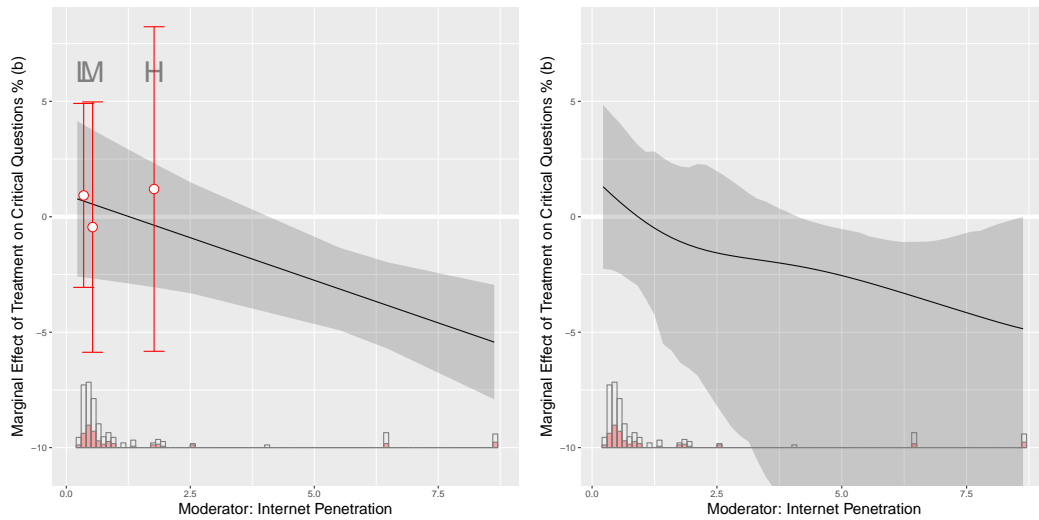
Key variables for the conditional relationship: Outcome Y: “change in critical questions (%)” (`diff_crit`); treatment D: “sunshine” (`t2`); moderator X: “internet penetration” (`internet_users100`).

Note: The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

FIGURE B31. RESULTS FROM MALESKY, SCHULER AND TRAN (2012)



(a) Raw data



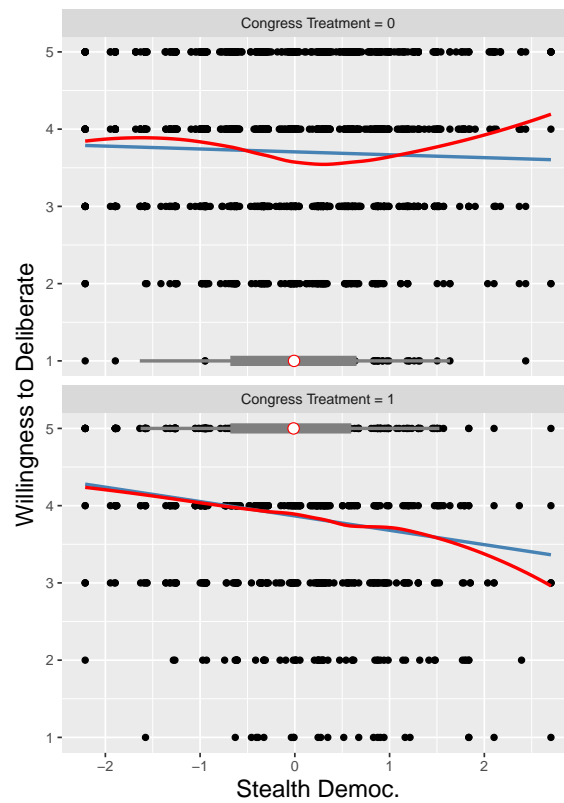
(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

.15 Neblo et al. (2010) APSR

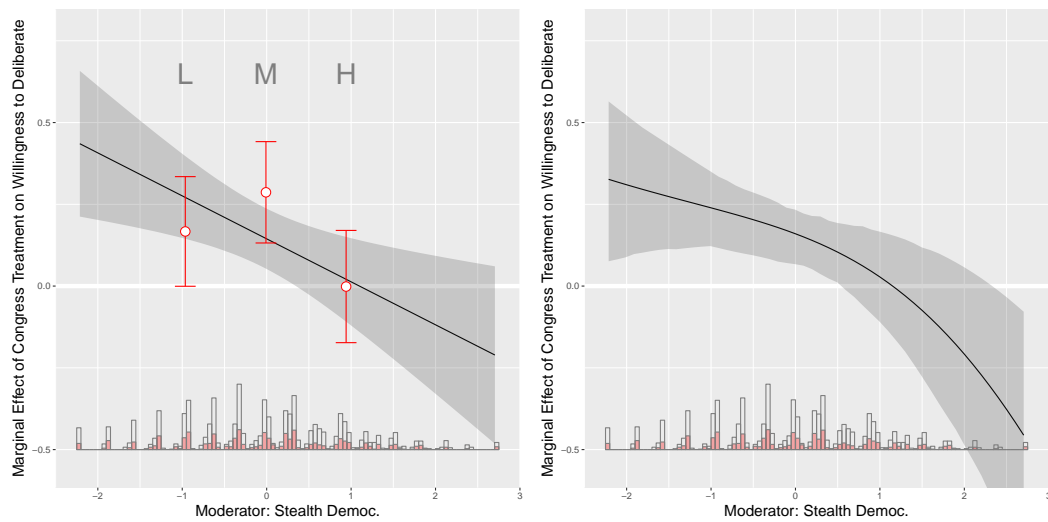
Claim on conditionality (Table 1 in manuscript): “...[T]he interaction between stealth and the experimental ‘Congress’ condition was negative and highly significant, indicating that, with the other variables controlled, people high on stealth were not as attracted as were others by the hypothetical prospect of talking with their (presumptively corrupt) members of Congress.” (574).

Key variables for the conditional relationship: Outcome Y: “willingness to deliberate” (**willing**); treatment D: “Congress treatment” (**treatcong2**); moderator X: “stealth democracy” (**stealth2.ct**).

FIGURE B32. RESULTS FROM [NEBLO ET AL. \(2010\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator

.16 Pelc (2011) IO

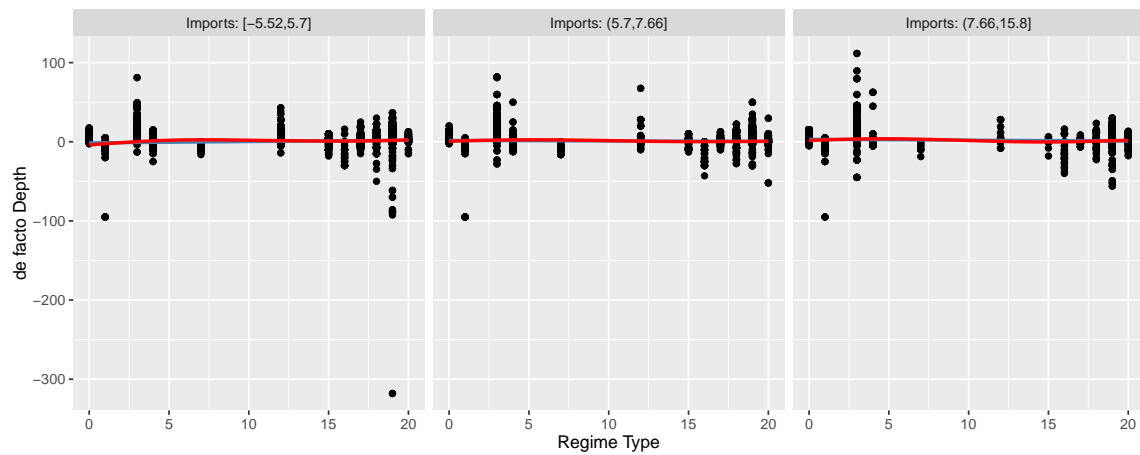
First interaction:

Claim on conditionality (Table 4, column 2 in manuscript): *“The interaction term between regime type and industry imports is substantively and significantly negative. . . . Democracies still display far greater de facto depth across all products looking at the regime coefficient, but those industries that are most valuable to members, and that have thus faced the greatest pressure during talks, exhibit considerable push-back in the form of hiked tariffs.”* (pp. 663-664).

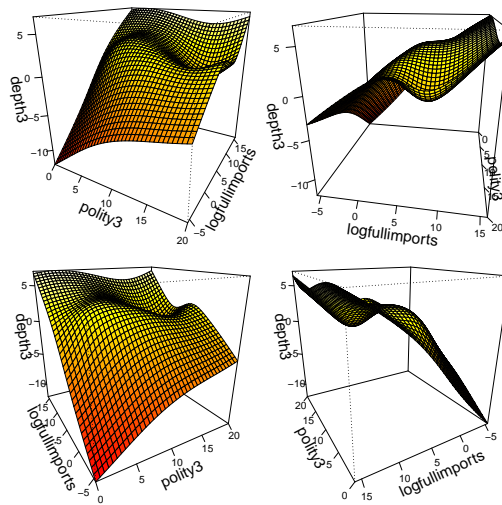
Key variables for the conditional relationship: Outcome Y: “de facto depth” (depth3); treatment D: “regime type” (polity3); moderator X: “imports” (logfullimports).

Note: The reason why the confidence intervals in the kernel plot are huge and highly asymmetric is because in the original analysis the standard errors are clustered on the reporter variable and there are only 17 clusters. Our replications use a block bootstrap to mimic this choice. We do correct for the potential problem of poor finite sample properties given the small number of clusters.

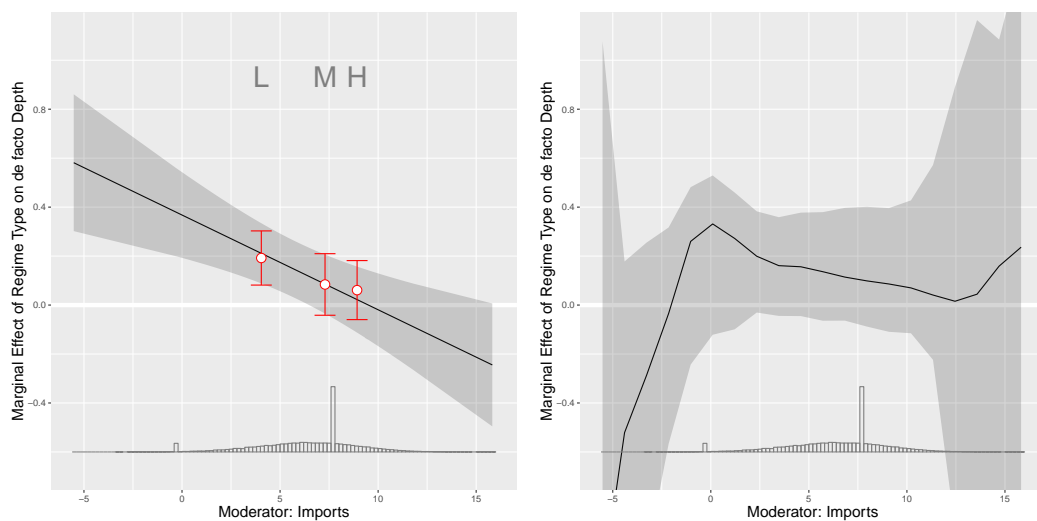
FIGURE B33. RESULTS FROM [PELC \(2011\)](#)



(a) Raw data



(b) GAM plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

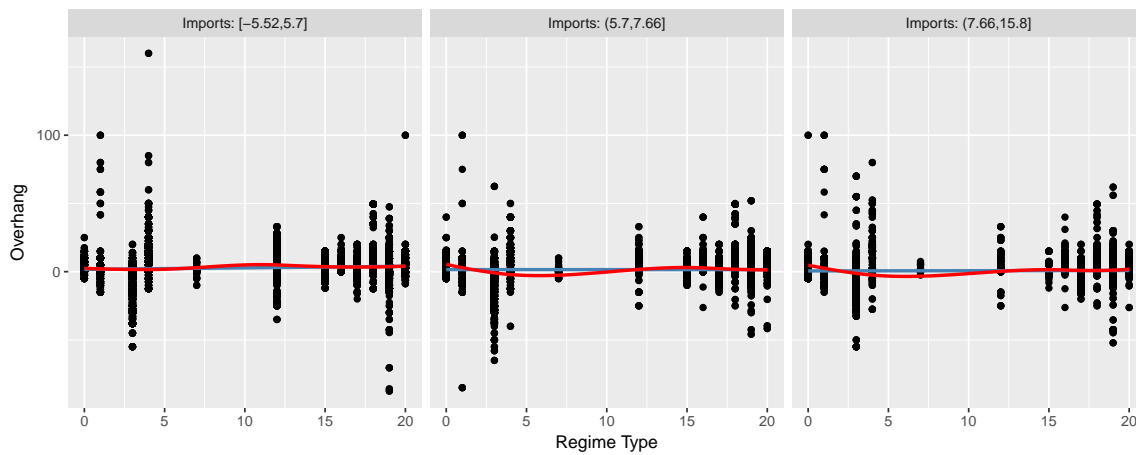
Second Interaction:

Claim on conditionality (Table 4, column 3 in manuscript): *‘Democratic countries, more vulnerable to interest group pressure, are observed “spending” their flexibility in these key industries. As a result, binding overhang is significantly decreased for these valuable democratic industries, as can be seen in the third column.’* (664).

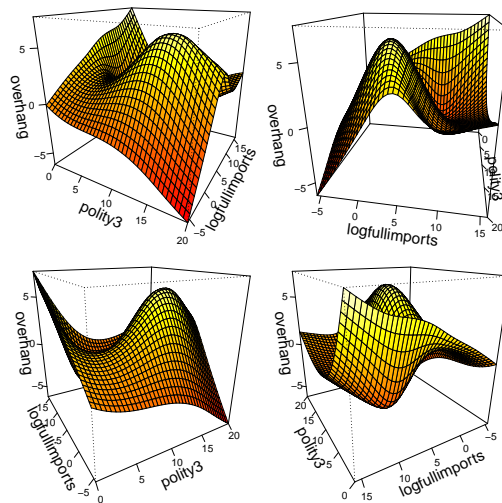
Key variables for the conditional relationship: Outcome Y: “overhang” (`overhang`); treatment D: “regime type” (`polity3`); moderator X: “imports” (`logfullimports`).

Note: The reason why the confidence intervals in the kernel plot are huge and highly asymmetric is because in the original analysis the standard errors are clustered on the reporter variable and there are only 17 clusters. Our replications use a block bootstrap to mimic this choice. We do correct for the potential problem of poor finite sample properties given the small number of clusters.

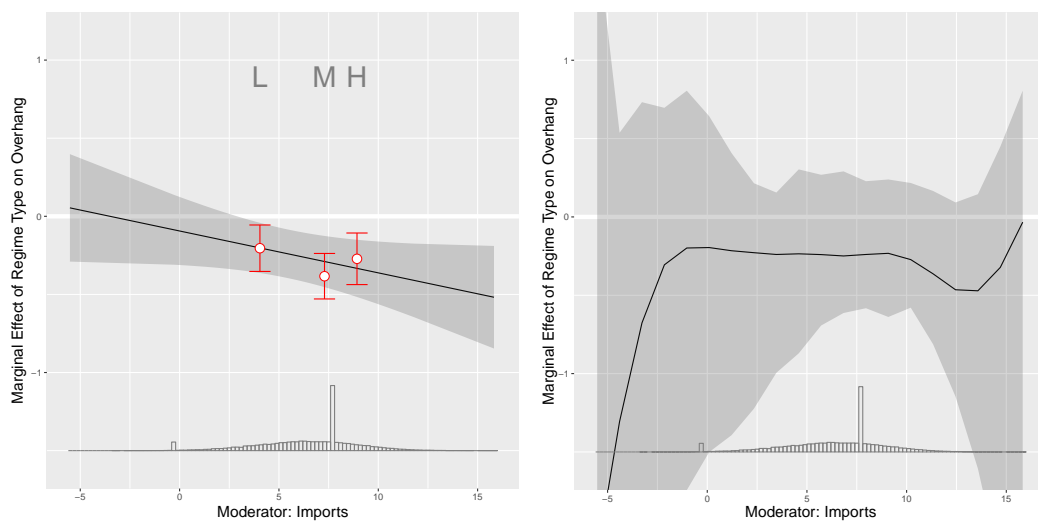
FIGURE B34. RESULTS FROM [PELC \(2011\)](#)



(a) Raw data



(b) GAM plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.17 Petersen and Aarøe (2013) APSR

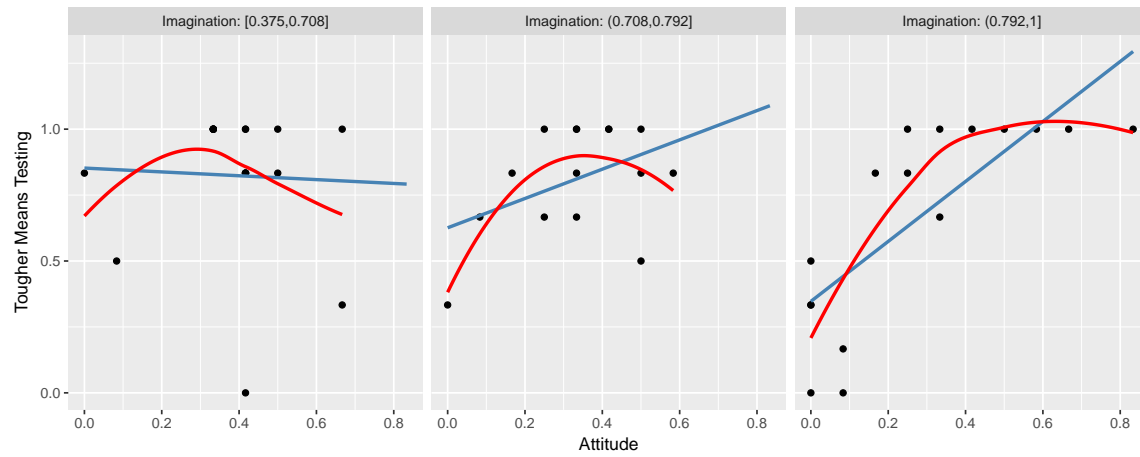
First interaction:

Claim on conditionality (Figure 1A in manuscript): *“As can be observed in the low-vividness condition (panel A), when vivid social cues were lacking, we found a strong tendency for imaginative people to filter in their own stereotypes to a greater extent than did unimaginative people. That is, as imagination increases, the predicted marginal effect of prior stereotypes on support for tougher means testing increases as well (as indicated by the positively sloped line), and as the associated confidence intervals cease to include zero, this increase becomes significant.”* (p. 286).

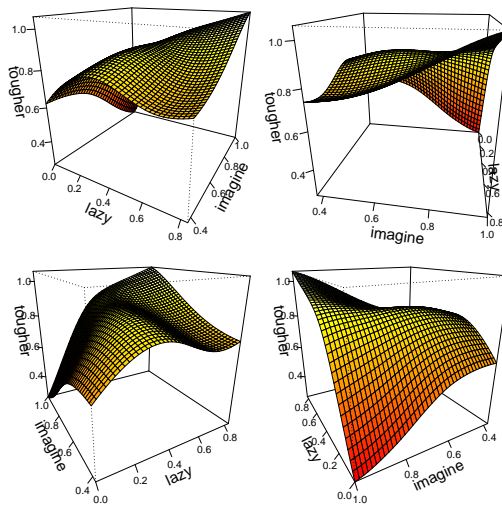
Key variables for the conditional relationship: Outcome Y: “support for means testing” (**tougher**); treatment D: “attitude (prior stereotypes)” (**lazy**); moderator X: “imagination” (**imagine**).

Note: The dashed vertical line indicates the truncated interval of the moderator shown in the original marginal effect plot.

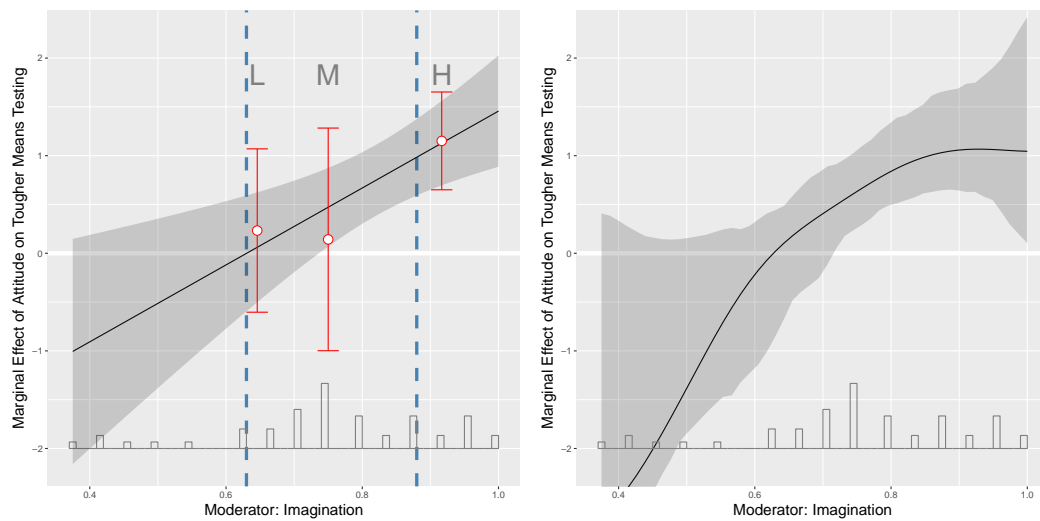
FIGURE B35. RESULTS FROM PETERSEN AND AARØE (2013)



(a) Raw data



(b) GAM plot



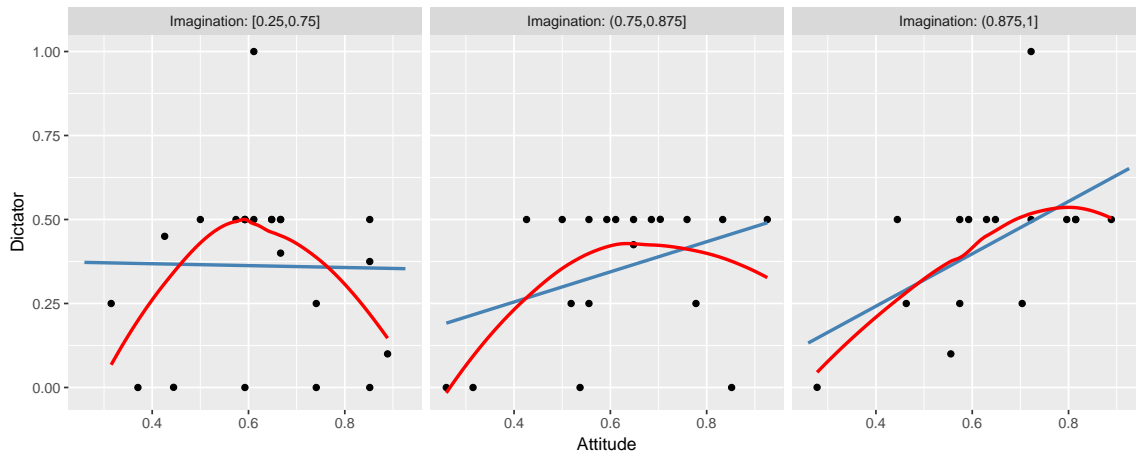
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

Second interaction:

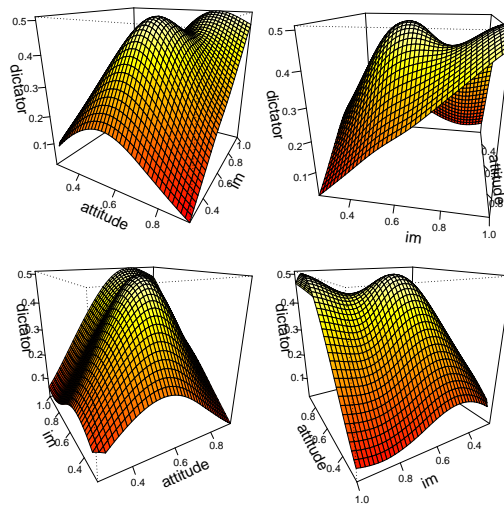
Claim on conditionality (Figure 2A in manuscript): *“As can be seen, imagination significantly increases the effect of people’s political principles on incentivized behavior such that the imaginative are more likely to stick to their principles (i.e., donate if they are supportive of welfare) in the face of short-term temptations to sacrifice their principles for money”* (p. 288).

Key variables for the conditional relationship: Outcome Y: “donation in dictator game” (**dictator**); treatment D: “attitude (support for welfare)” (**attitude**); moderator X: “imagination” (**im**).

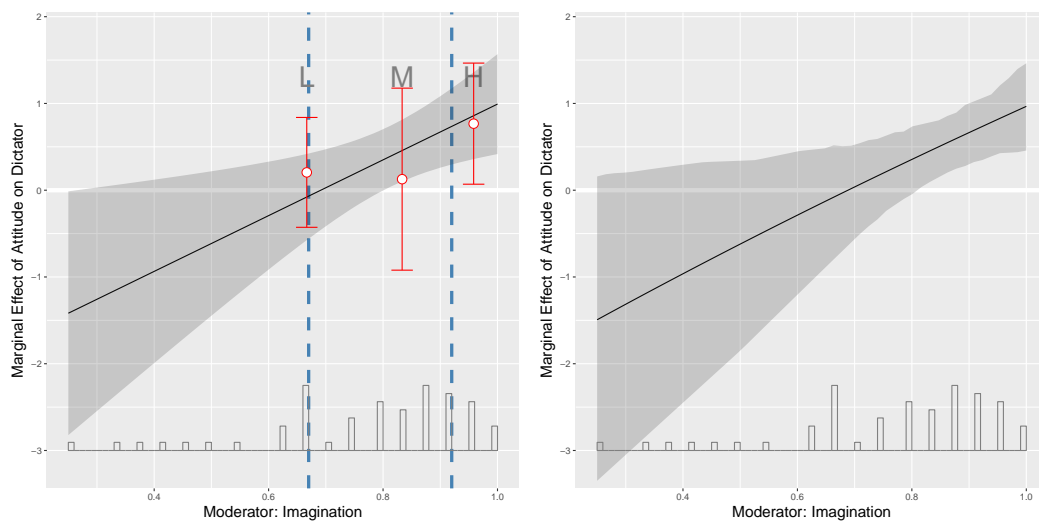
FIGURE B36. RESULTS FROM [PETERSEN AND AARØE \(2013\)](#)



(a) Raw data



(b) GAM plot



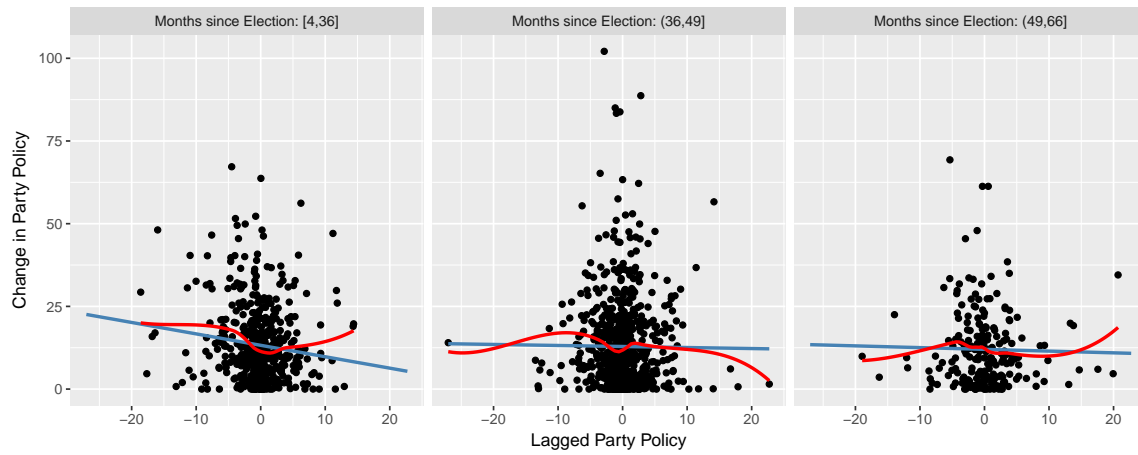
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.18 **Somer-Topcu (2009) JOP**

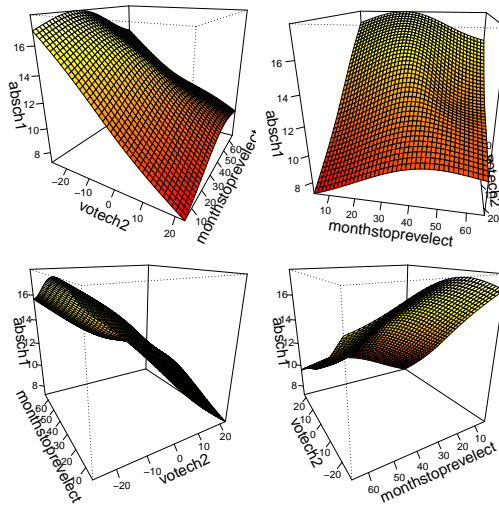
Claim on conditionality (Figure 2 in manuscript): *“As can be seen, parties change their positions if they lose votes, and the effect dissipates as time (x-axis) elapses.”* (244).

Key variables for the conditional relationship: Outcome Y: “change in party policy” (`absch1`); treatment D: “lagged party policy” (`votech2`); moderator X: “months since election” (`monthstopreselect`).

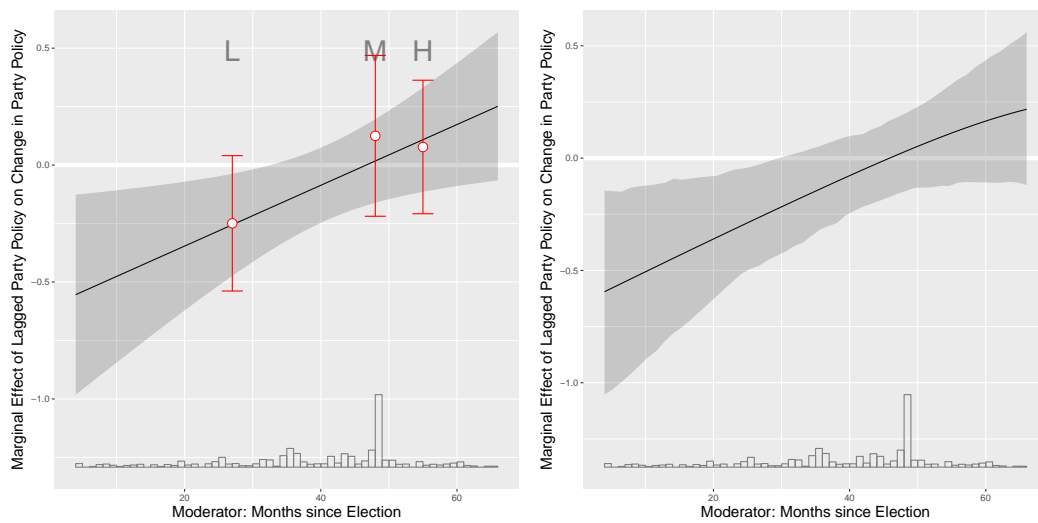
FIGURE B37. RESULTS FROM [SOMER-TOPCU \(2009\)](#)



(a) Raw data



(b) GAM plot



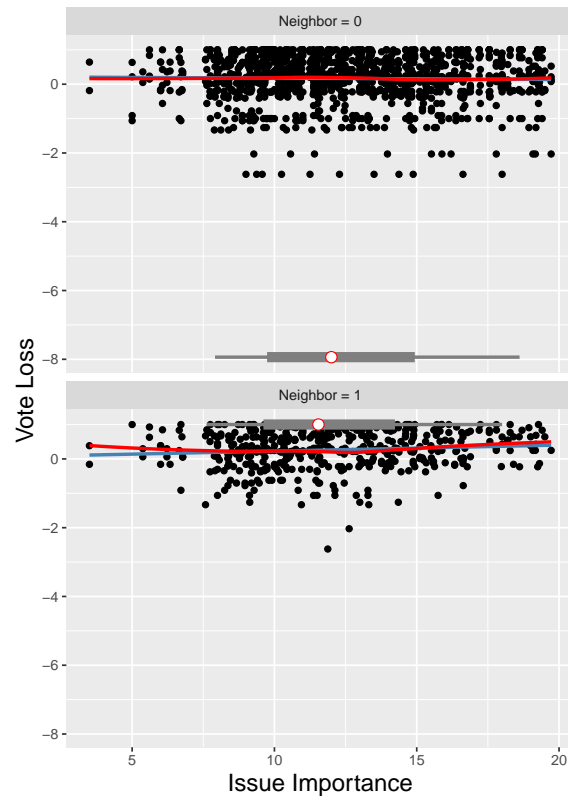
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.19 Tavits (2008) CPS

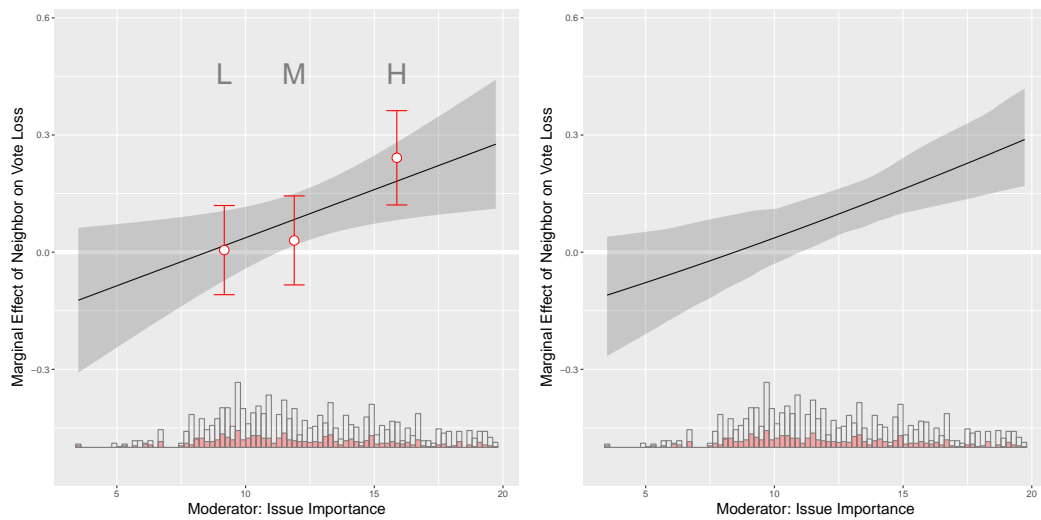
Claim on conditionality (Figure 1 in manuscript): *“As the graph shows, neighbors to the new party start to lose significantly more votes compared to other parties when the issue importance reaches 11, and the effect becomes stronger as issue importance increases. The vote loss of neighbors is the highest on the most important issue to the new party.”* (9).

Key variables for the conditional relationship: Outcome Y: “vote loss” (`voteslost`); treatment D: “neighbor” (`neighbor`); moderator X: “issue importance” (`importance`).

FIGURE B38. RESULTS FROM TAVITS (2008)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

.20 **Truex (2014)** APSR

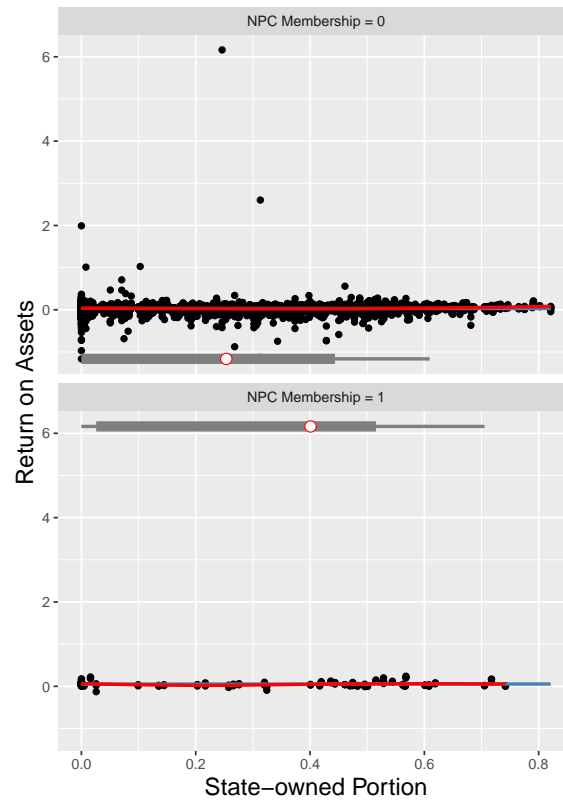
First Interaction:

Claim on conditionality (Figure 3, top left panel in manuscript): *“For a firm with no shares owned by the state, the marginal effect of NPC membership on ROA is about 2.4 percentage points, and 4.3 points for MARGIN. For firms with greater than 50% shares state owned, the effect appears negligible. We observe a similar conditional relationship for revenue, with the benefits of membership decreasing substantially with firm size. The ‘returns to office’ appear greatest for smaller, private firms.”* (243).

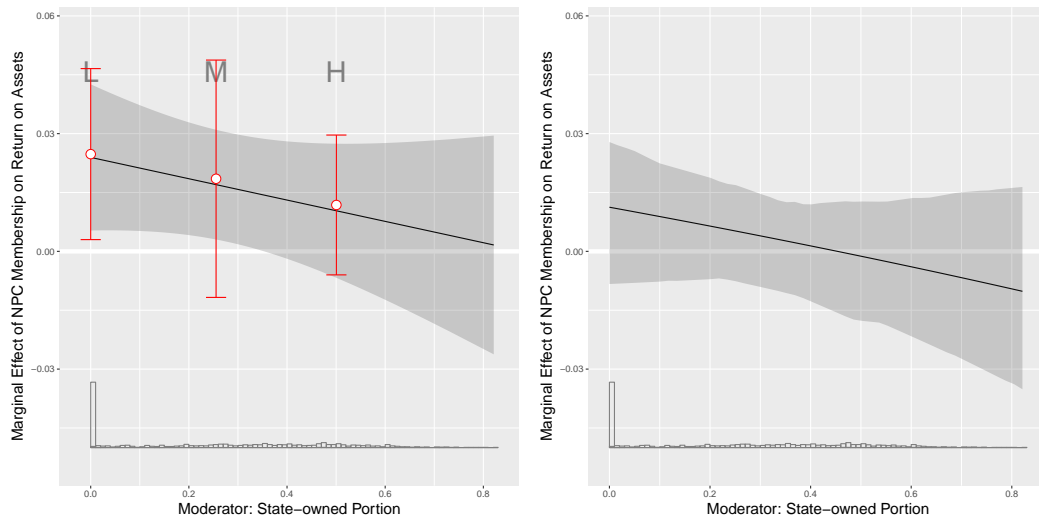
Key variables for the conditional relationship: Outcome Y: “return on assets” (roa); treatment D: “NPC membership” (npc); moderator X: “state-owned portion” (so_portion).

Note: We reweight the data as the author does.

FIGURE B39. RESULTS FROM [TRUEX \(2014\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator

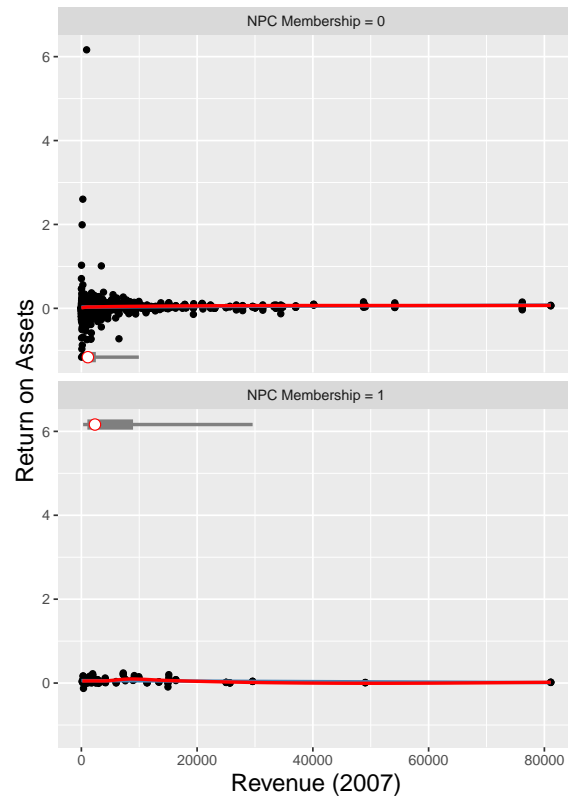
Second Interaction:

Claim on conditionality (Figure 3, top right panel in manuscript): *“For a firm with no shares owned by the state, the marginal effect of NPC membership on ROA is about 2.4 percentage points, and 4.3 points for MARGIN. For firms with greater than 50% shares state owned, the effect appears negligible. We observe a similar conditional relationship for revenue, with the benefits of membership decreasing substantially with firm size. The ‘returns to office’ appear greatest for smaller, private firms.”* (243).

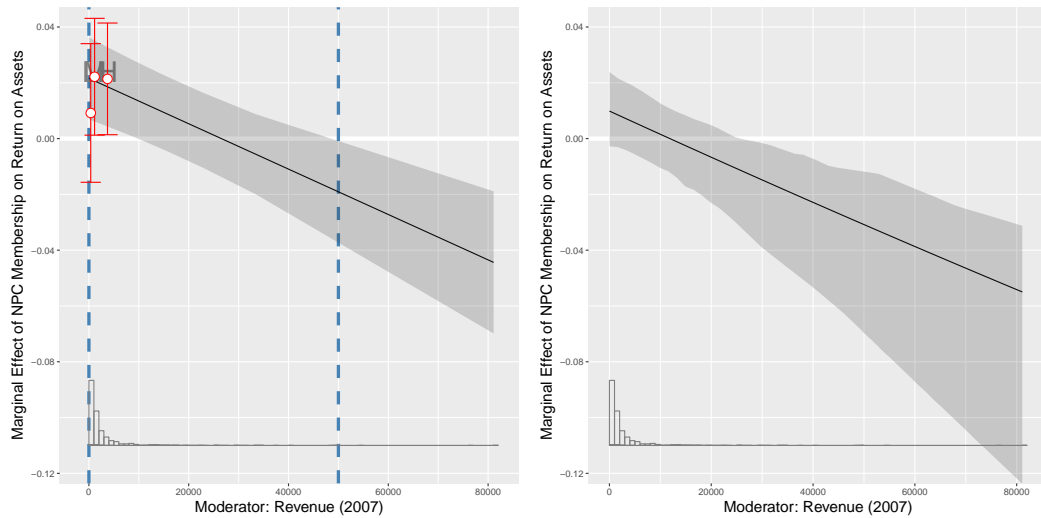
Key variables for the conditional relationship: Outcome Y: “return on assets” (`roa`); treatment D: “NPC membership” (`npc`); moderator X: “revenue (2007)” (`rev2007`).

Note: In the binning plot below, the dashed vertical lines indicate the range of the moderator displayed in the original manuscript. We reweight the data as the author does.

FIGURE B40. RESULTS FROM [TRUEX \(2014\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator

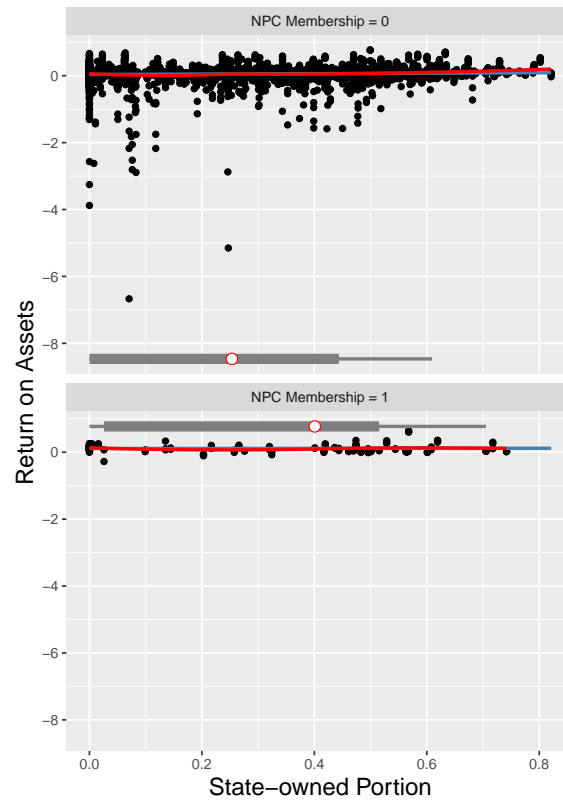
Third Interaction:

Claim on conditionality (Figure 3, bottom left panel in manuscript): *“For a firm with no shares owned by the state, the marginal effect of NPC membership on ROA is about 2.4 percentage points, and 4.3 points for MARGIN. For firms with greater than 50% shares state owned, the effect appears negligible. We observe a similar conditional relationship for revenue, with the benefits of membership decreasing substantially with firm size. The ‘returns to office’ appear greatest for smaller, private firms.”* (243).

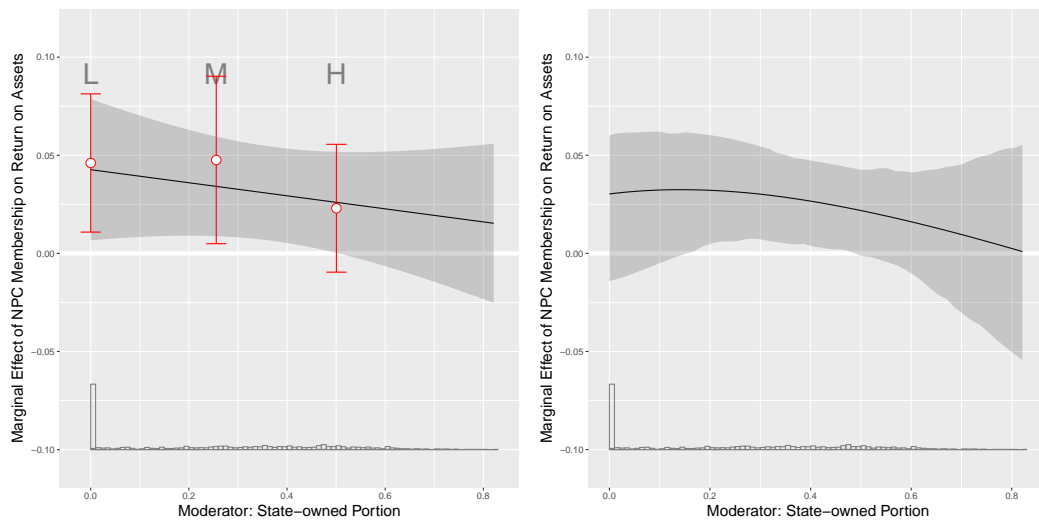
Key variables for the conditional relationship: Outcome Y: “profit margin” (`margin`); treatment D: “NPC membership” (`npc`); moderator X: “state-owned portion” (`so_portion`).

Note: The dashed vertical line indicates the truncated interval of the moderator shown in the original marginal effect plot. We reweight the data as the author does.

FIGURE B41. RESULTS FROM [TRUEX \(2014\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator (black line) and from Binning Estimator (white dots)

Fourth Interaction:

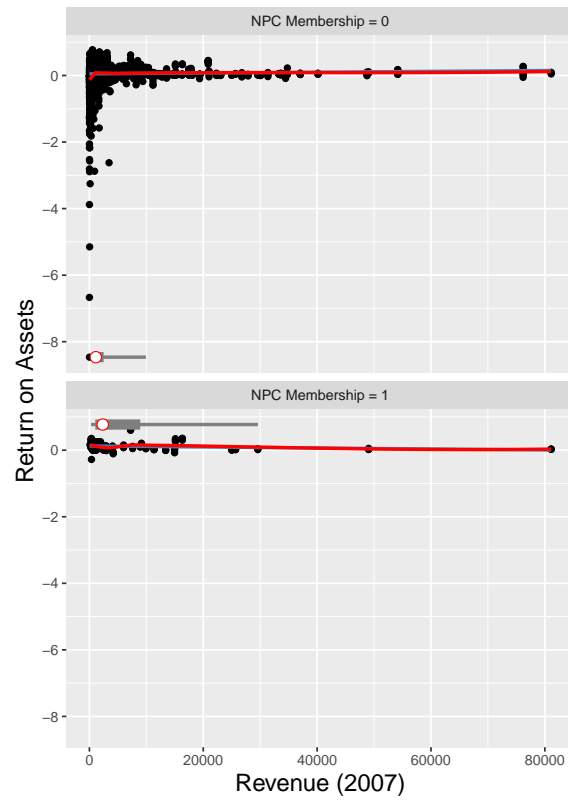
Claim on conditionality (Figure 3, bottom right panel in manuscript):

“For a firm with no shares owned by the state, the marginal effect of NPC membership on ROA is about 2.4 percentage points, and 4.3 points for MARGIN. For firms with greater than 50% shares state owned, the effect appears negligible. We observe a similar conditional relationship for revenue, with the benefits of membership decreasing substantially with firm size. The ‘returns to office’ appear greatest for smaller, private firms.” (243).

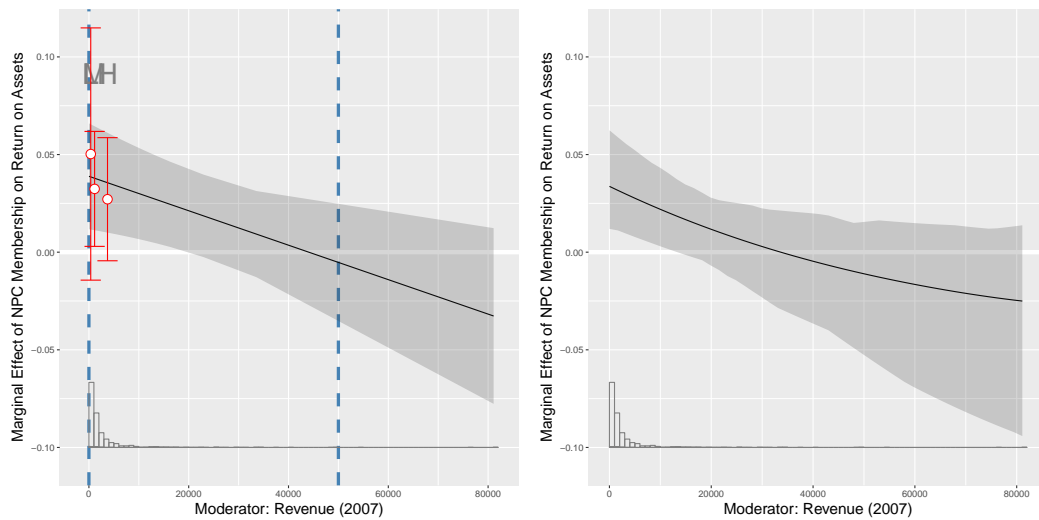
Key variables for the conditional relationship: Outcome Y: “profit margin” (`rev2007`); treatment D: “NPC membership” (`margin`); moderator X: “revenue (2007)” (`npc`).

Note: In the binning plot below, the dashed vertical lines indicate the range of the moderator displayed in the original manuscript. We reweight the data as the author does.

FIGURE B42. RESULTS FROM [TRUEX \(2014\)](#)



(a) Raw data



(b) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (c) Marginal Effects from Kernel Estimator

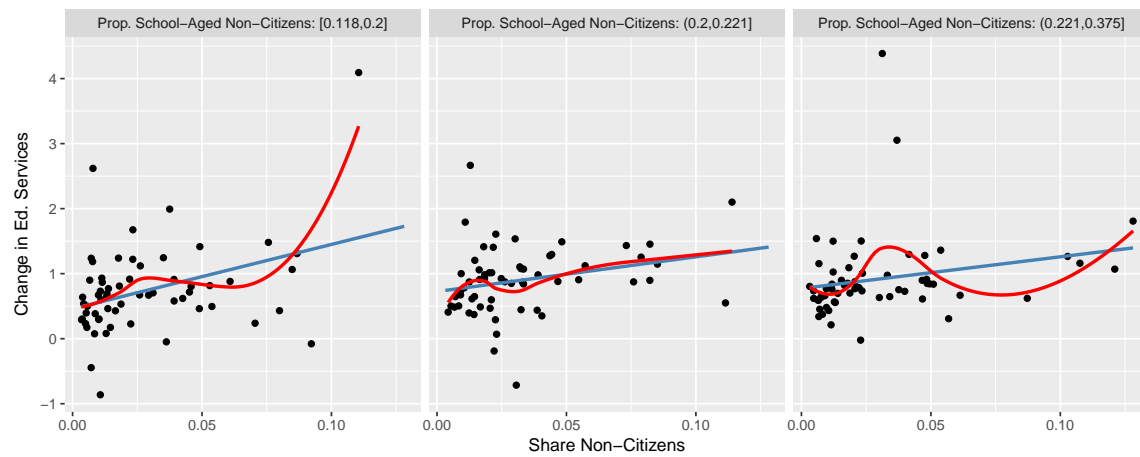
.21 Vernby (2013) AJPS

First interaction:

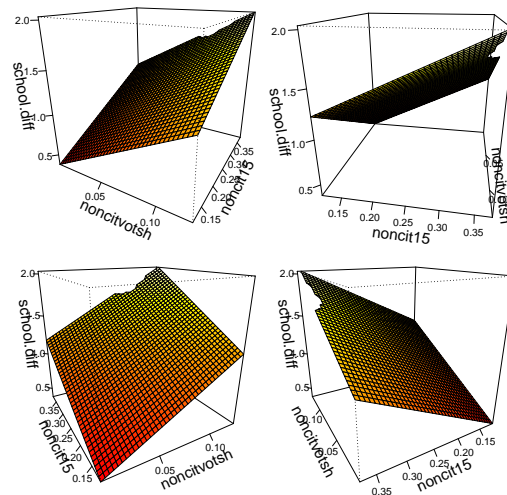
Claim on conditionality (Figure 2, left panel in manuscript): *“The impact of the reform on education services was larger where many noncitizens were school-aged, even if the interaction term is not statistically significant. . . . In the case of education services, the marginal effect increases more than tenfold as we go from a situation where 8% (the empirical minimum) of noncitizens are school-aged, to a situation where 38% (the empirical maximum) are school-aged. However, the 95% confidence interval is fairly wide and the marginal effect is only statistically significant when more than 18% of noncitizens are school-aged.”* (p. 23).

Key variables for the conditional relationship: Outcome Y: “change in ed. services” (`school.diff`); treatment D: “share noncitizens in electorate” (`noncitvotsh`); moderator X: “Proportion school aged noncitizens” (`noncit15`).

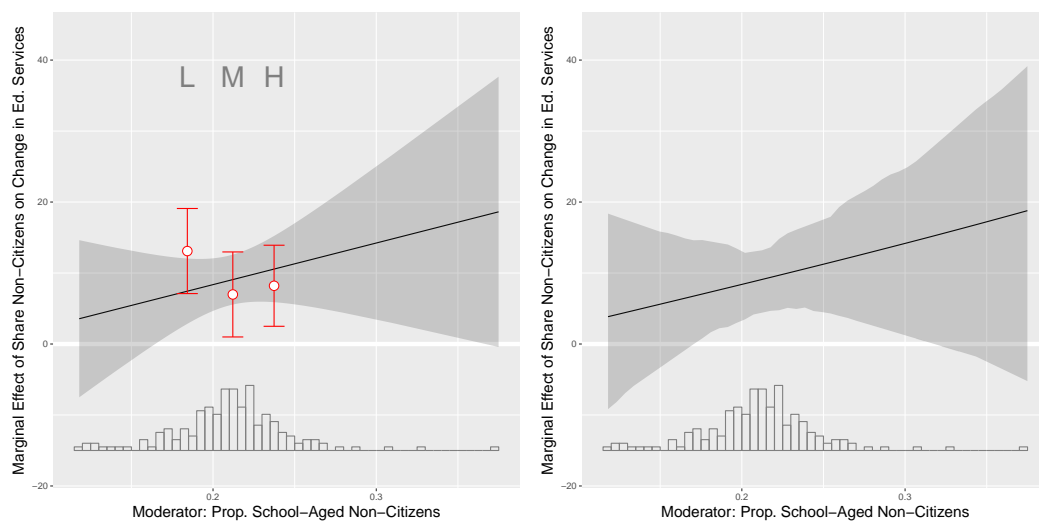
FIGURE B43. RESULTS FROM [VERNBY \(2013\)](#)



(a) Raw data



(b) GAM plot



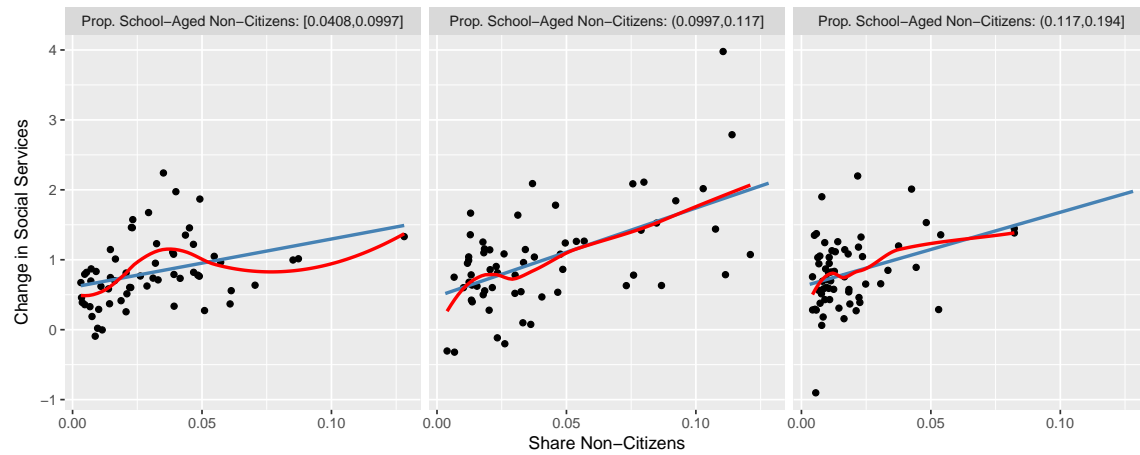
(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

Second interaction:

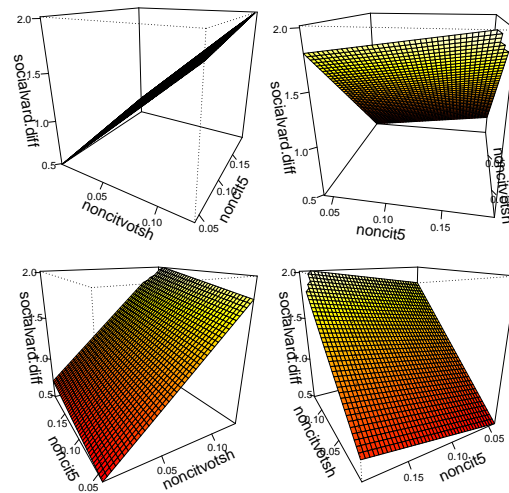
Claim on conditionality (Figure 2, right panel in manuscript): *“From the second column, it can be seen that the reform’s impact on social and family services was larger where a large share of noncitizens were preschool aged. . . . Turning to spending on social and family services, a move from a situation where 4% of noncitizens are school-aged, to a situation where 20% are, leads to an almost threefold increase in the marginal effect of the share of noncitizens in the electorate. Again, the 95% confidence interval is wide, and the marginal effect becomes statistically significant where the preschool-aged make up 6% or more of the municipal noncitizen population.”* (23).

Key variables for the conditional relationship: Outcome Y: “change in social services” (`socialvard.diff`); treatment D: “share noncitizens in electorate” (`noncitvotsh`); moderator X: “Proportion school aged noncitizens” (`noncit15`).

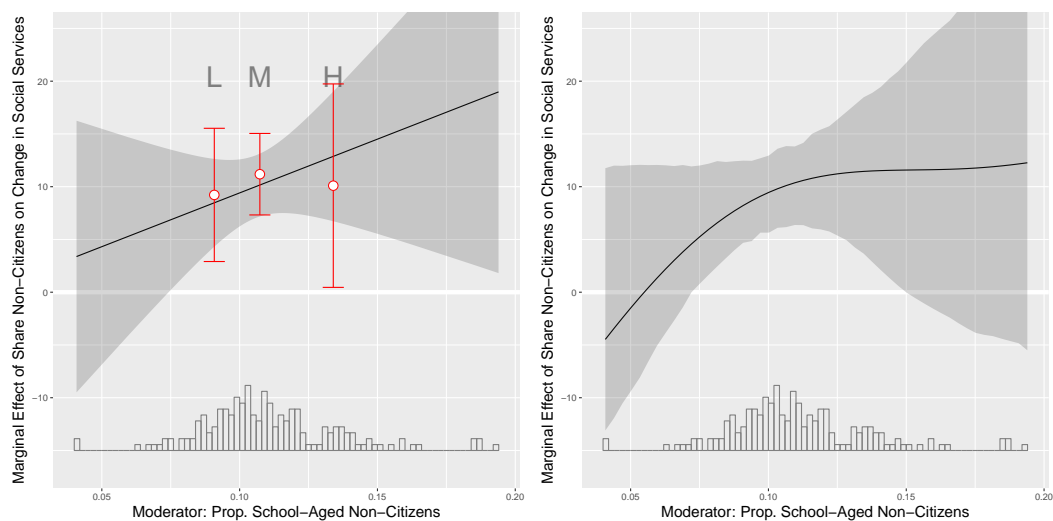
FIGURE B44. RESULTS FROM [VERNBY \(2013\)](#)



(a) Raw data



(b) GAM plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

.22 Williams (2011) CPS

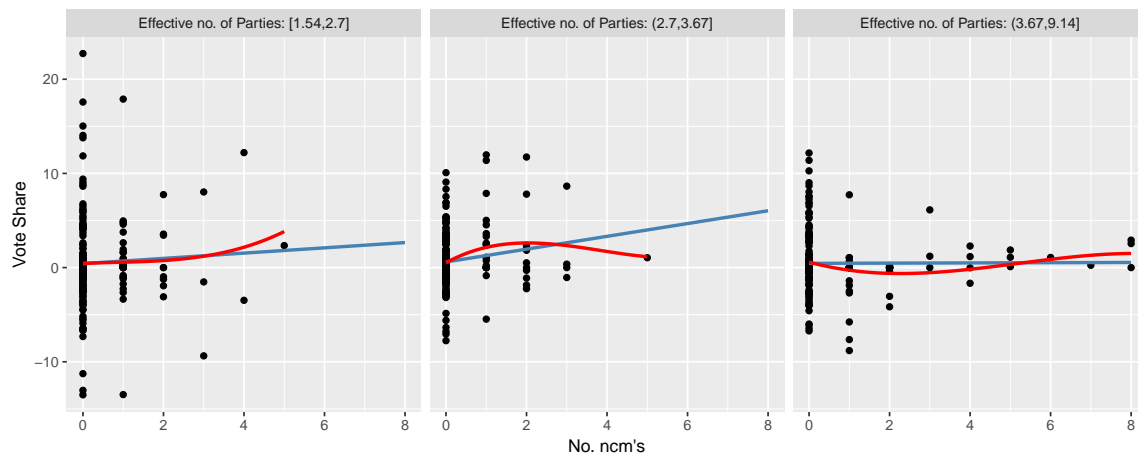
First Interaction:

Claim on conditionality (Figure 3 in manuscript): *“As the effective number of parties increases (to more than five effective parties), the beneficial electoral impacts of NCMs disappear.”* (19).

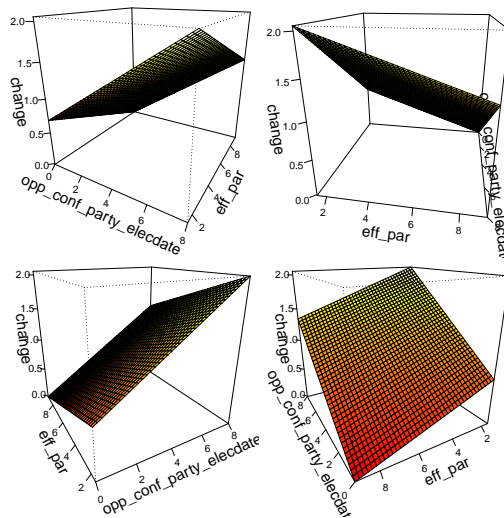
Key variables for the conditional relationship: Outcome Y: “vote change” (`change`); treatment D: “no confidence motion” (`opp_conf_party_elecdte`); moderator X: “effective no. of parties” (`eff_par`).

Note: In the binning plot below, the dashed vertical lines indicate the range of the moderator displayed in the original manuscript. The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

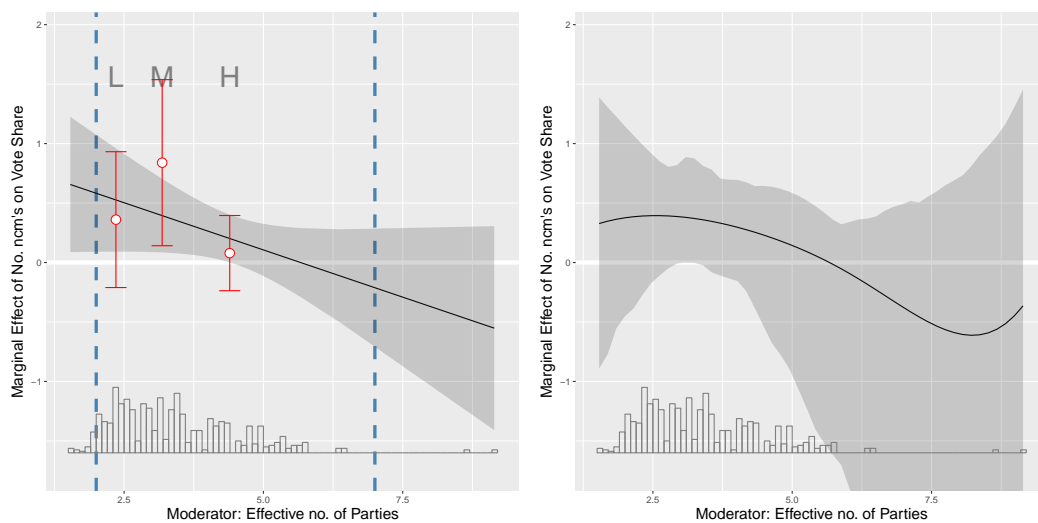
FIGURE B45. RESULTS FROM WILLIAMS (2011)



(a) Raw data



(b) GAM plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots) (d) Marginal Effects from Kernel Estimator

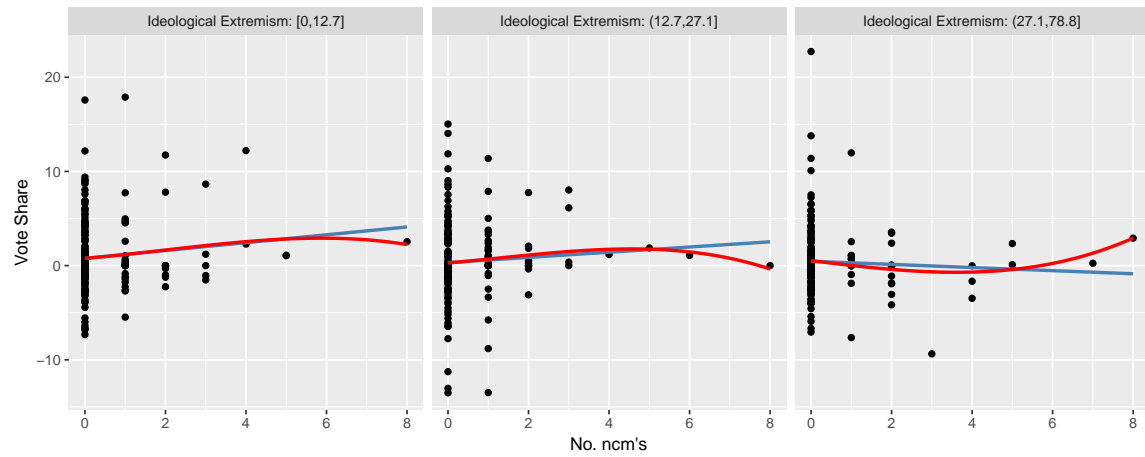
Second Interaction:

Claim on conditionality (Figure 4 in manuscript): *“For ideologically moderate parties (those with absolute ideology scores lower than 30), proposing NCMs increases their vote shares by as much as 0.75%. These marginal effects are statistically significant at the 90% confidence level. However, as the proposing party’s ideology becomes more extreme and farther from the median voter, the beneficial impacts of NCMs are eliminated because voters view these signals as ‘cheap talk.’ For parties located 30 or more points from the center, proposing NCMs have no significant effect on vote choice. This supports the notion that the capturable voter will change her or his vote to viable government alternatives only based on credible signals.”* (20).

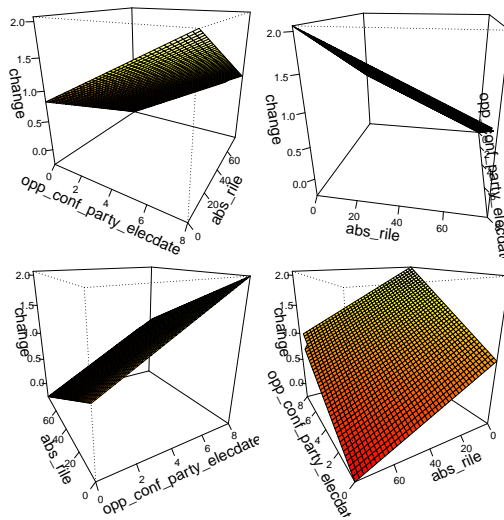
Key variables for the conditional relationship: Outcome Y: “vote change” (`change`); treatment D: “no confidence motion” (`opp_conf_party_elecdate`); moderator X: “ideological extremism” (`abs_rile`).

Note: In the binning plot below, the dashed vertical lines indicate the range of the moderator displayed in the original manuscript. The authors show 90% confidence intervals in the paper, while in both the binning plot and the kernel smoothing plot, we use 95% confidence intervals.

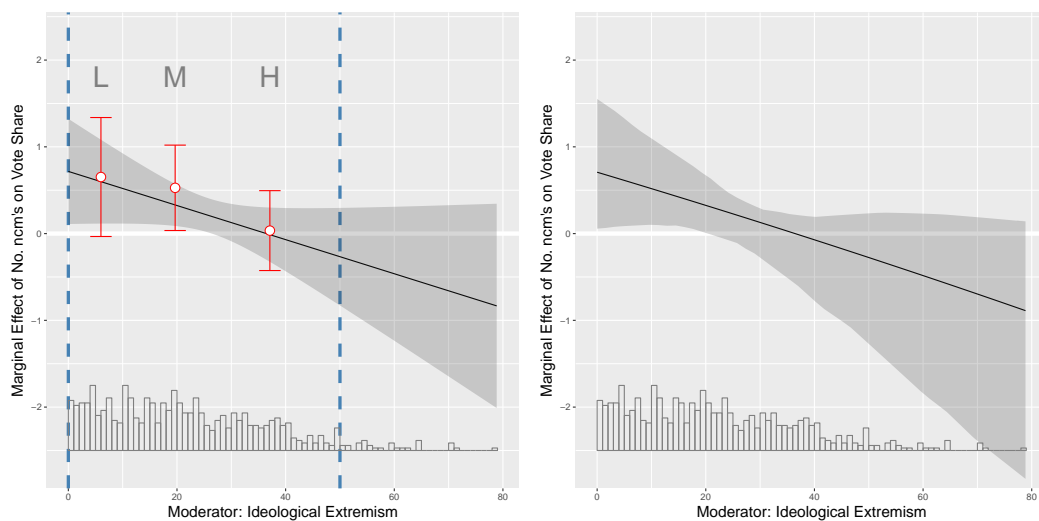
FIGURE B46. RESULTS FROM WILLIAMS (2011)



(a) Raw data



(b) GAM plot



(c) Marginal Effects from Replicated Model (black line) and from Binning Estimator (white dots)
(d) Marginal Effects from Kernel Estimator

.23 Additional Results from Diagnostic Measures

Study	Journal	L-kurtosis	Wald test <i>p</i> -value	Low vs. High <i>p</i> -value
Adams et al. (2006)	AJPS	0.161	0.220	0.007
Aklin and Urpelainen (2013)	AJPS	0.381	0.000	0.697
Aklin and Urpelainen (2013)	AJPS	0.187	0.000	0.121
Banks and Valentino (2012)	AJPS	0.057	0.220	0.001
Banks and Valentino (2012)	AJPS	0.242	0.150	0.287
Banks and Valentino (2012)	AJPS	0.242	0.280	0.094
Bodea and Hicks (2015 <i>a</i>)	JOP	0.109	0.020	0.027
Bodea and Hicks (2015 <i>a</i>)	JOP	0.184	0.010	0.000
Bodea and Hicks (2015 <i>b</i>)	IO	0.166	0.100	0.917
Bodea and Hicks (2015 <i>b</i>)	IO	0.024		0.891
Bodea and Hicks (2015 <i>b</i>)	IO	0.172		0.949
Bodea and Hicks (2015 <i>b</i>)	IO	0.028		0.416
Carpenter and Moore (2014)	APSR	0.467	0.000	0.000
Chapman (2009)	IO	0.077	n.a.*	n.a.*
Clark and Golder (2006)	CPS	0.047	0.000	0.226
Clark and Golder (2006)	CPS	-0.040	0.010	0.034
Clark and Golder (2006)	CPS	0.029	0.000	0.013
Clark and Golder (2006)	CPS	0.058	0.00	n.a.*
Clark and Leiter (2014)	CPS	0.210	0.780	0.264
Hellwig and Samuels (2007)	CPS	0.146	0.330	0.096
Hellwig and Samuels (2007)	CPS	0.393	0.840	0.953
Hicken and Simmons (2008)	AJPS	-0.112	0.080	0.416
Huddy, Mason and Aarøe (2015)	APSR	0.065	0.550	0.000
Huddy, Mason and Aarøe (2015)	APSR	0.065	0.230	0.000
Kim and LeVeck (2013)	APSR	0.143	0.030	0.001
Kim and LeVeck (2013)	APSR	0.143	0.020	0.207
Kim and LeVeck (2013)	APSR	0.100	0.000	0.002
Malesky, Schuler and Tran (2012)	APSR	0.426	0.040	0.692
Malesky, Schuler and Tran (2012)	APSR	0.426	0.030	0.762
Malesky, Schuler and Tran (2012)	APSR	0.426	0.220	0.834
Malesky, Schuler and Tran (2012)	APSR	0.426	0.030	0.947
Neblo et al. (2010)	APSR	0.113	0.010	0.169
Pelc (2011)	IO	0.161	0.000	0.001
Pelc (2011)	IO	0.161	0.000	0.346
Petersen and Aarøe (2013)	APSR	0.024	0.690	0.194
Petersen and Aarøe (2013)	APSR	0.127	0.500	0.236
Somer-Topcu (2009)	JOP	0.124	0.170	0.114
Tavits (2008)	CPS	0.071	0.730	0.005
Truex (2014)	APSR	-0.063	0.000	0.303
Truex (2014)	APSR	0.469	0.020	0.407
Truex (2014)	APSR	-0.063	0.000	0.290
Truex (2014)	APSR	0.469	0.060	0.507
Vernby (2013)	AJPS	0.243	0.100	0.223
Vernby (2013)	AJPS	0.199	0.790	0.879
Williams (2011)	CPS	0.101	0.550	0.368
Williams (2011)	CPS	0.083	0.000	0.147

Notes: * Unable to produce a test statistic because of insufficient variation in data.

References

- Adams, James, Michael Clark, Lawrence Ezrow and Garrett Glasgow. 2006. "Are Niche Parties Fundamentally Different from Mainstream Parties? The Causes and the Electoral Consequences of Western European Parties' Policy Shifts, 1976–1998." *American Journal of Political Science* 50(3):513–529.
- Aklin, Michaël and Johannes Urpelainen. 2013. "Political Competition, Path Dependence, and The Strategy of Sustainable Energy Transitions." *American Journal of Political Science* 57(3):643–658.
- Banks, Antoine J. and Nicholas A. Valentino. 2012. "Emotional Substrates of White Racial Attitudes." *American Journal of Political Science* 56(2):286–297.
- Bodea, Cristina and Raymond Hicks. 2015a. "International Finance and Central Bank Independence: Institutional Diffusion and the Flow and Cost of Capital." *The Journal of Politics* 77(1):268–284.
- Bodea, Cristina and Raymond Hicks. 2015b. "Price Stability and Central Bank Independence: Discipline, Credibility, and Democratic Institutions." *International Organization* 69(1):35–61.
- Carpenter, Daniel and Colin D. Moore. 2014. "When Canvassers Became Activists: Antislavery Petitioning and the Political Mobilization of American Women." *American Political Science Review* 108(3):479–498.
- Chapman, Terrence L. 2009. "Audience Beliefs and International Organization Legitimacy." *International Organization* 63(04):733–764.
- Clark, Michael and Debra Leiter. 2014. "Does the Ideological Dispersion of Parties Mediate the Electoral Impact of Valence? A Cross-national Study of Party Support in Nine Western European Democracies." *Comparative Political Studies* 47(2):171–202.
- Clark, William Roberts and Matt Golder. 2006. "Rehabilitating Duverger's Theory Testing the Mechanical and Strategic Modifying Effects of Electoral Laws." *Comparative Political Studies* 39(6):679–708.
- Hainmueller, Jens, Jonathan Mummolo and Yiqing Xu. 2018. "Replication Data for: How Much Should We Trust Estimates from Multiplicative Interaction Models?"

- Simple Tools to Improve Empirical Practice.” *Harvard Dataverse* . doi:[10.7910/DVN/Q1V00G](https://doi.org/10.7910/DVN/Q1V00G).
- Hellwig, Timothy and David Samuels. 2007. “Voting in Open Economies The Electoral Consequences of Globalization.” *Comparative Political Studies* 40(3):283–306.
- Hicken, Allen and Joel W. Simmons. 2008. “The Personal Vote and the Efficacy of Education Spending.” *American Journal of Political Science* 52(1):109–124.
- Huddy, Leonie, Lilliana Mason and Lene Aarøe. 2015. “Expressive Partisanship: Campaign Involvement, Political Emotion, and Partisan Identity.” *American Political Science Review* 109(01):1–17.
- Kim, Henry A. and Brad L. LeVeck. 2013. “Money, Reputation, and Incumbency in US House Elections, or Why Marginals Have Become More Expensive.” *American Political Science Review* 107(3):492–504.
- Malesky, Edmund, Paul Schuler and Anh Tran. 2012. “The Adverse Effects of Sunshine: A Field Experiment on Legislative Transparency in An Authoritarian Assembly.” *American Political Science Review* 106(04):762–786.
- Neblo, Michael A., Kevin M. Esterling, Ryan P. Kennedy, David M.J. Lazer and Anand E. Sokhey. 2010. “Who Wants to Deliberate—and Why?” *American Political Science Review* 104(3):566–583.
- Pelc, Krzysztof J. 2011. “Why do Some Countries Get Better WTO Accession Terms Than Others?” *International Organization* 65(4):639–672.
- Petersen, Michael Bang and Lene Aarøe. 2013. “Politics in the Mind’s Eye: Imagination as a Link between Social and Political Cognition.” *American Political Science Review* 107(2):275–293.
- Somer-Topcu, Zeynep. 2009. “Timely Decisions: The Effects of Past National Elections on Party Policy Change.” *The Journal of Politics* 71(1):238–248.
- Tavits, Margit. 2008. “Policy Positions, Issue Importance, and Party Competition in New Democracies.” *Comparative Political Studies* 41(1):48–72.
- Truex, Rory. 2014. “The Returns to Office in a ‘Rubber Stamp’ Parliament.” *American Political Science Review* 108(2):235–251.

- Vernby, Kåre. 2013. "Inclusion and Public Policy: Evidence from Sweden's Introduction of Noncitizen Suffrage." *American Journal of Political Science* 57(1):15–29.
- Williams, Laron K. 2011. "Unsuccessful Success? Failed No-confidence Motions, Competence Signals, and Electoral Support." *Comparative Political Studies* 44(11):1474–1499.