

Public Reactions to International Legal  
Institutions: The ICC in a Developing  
Democracy, A Survey Experiment From  
Kyrgyzstan: Which is a Very Cool Place, But  
Unfortunately Had Recent Internal Violence...  
Ok, at This Point I'm Just Trolling Renee  
Because She Doesn't Like Long Titles

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*University of Texas at Austin*

Stephen Chaudoin

*University of Illinois at Urbana-Champaign*

# Overarching Question

*Under what conditions are international organizations (IOs) most effective?*

*→ What explains public reactions to IOs?*

My Framework:

- ▶ IO actions → contestation, which affect reactions to IOs.
- ▶ Show evidence from a survey experiment in Krygyzstan.

# Existing Theories

- ▶ IOs affect member state behavior by mobilizing subnational enforcement.
  - ▶ Human rights: Simmons 2009
  - ▶ Trade: Mansfield et al 2002; Chaudoin 2014
- ▶ Supported by survey evidence:
  - ▶ Bearce and Cook 2015, Chaudoin 2014, Chilton 2014, 2015, Kreps 2014, Tingley and Tomz 2014, Wallace 2013.

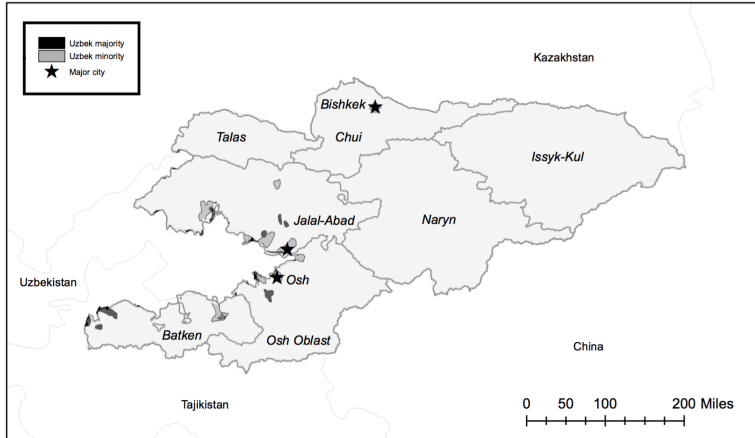
# My Theoretical Framework

- ▶ Subnational actors have heterogeneous preferences.
  - ▶ Variation across groups, pro- and anti-compliance.
  - ▶ Variation over time.
- ▶ IOs trigger competing efforts between groups.
  - ▶ Eg lobbying or protests.
  - ▶ Contest itself has tangible effects.

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  - ▶ Contest itself has tangible effects.
- ▶ This, in turn, affects how citizens react to IOs.

# Why Kyrgyzstan?



- ▶ Relatively recent, localized violence.
- ▶ Minimal exposure to the ICC.

# Hypotheses

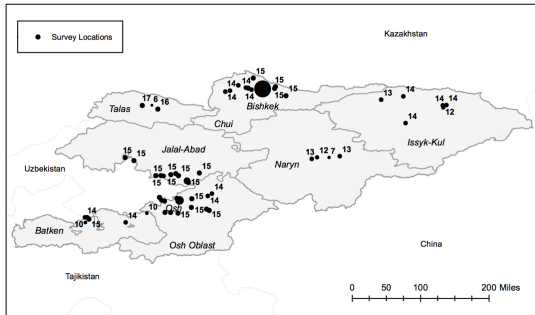
*How does contestation matter in this setting?*

- ▶ Proximity:
  - ▶ IO actions create the prospect of contestation.
  - ▶ → those in close proximity to the violence are less receptive to possible IO actions.

*(Not presented today)*

- ▶ Awareness of court *does not* blunt negative reactions.
- ▶ Alignment with the government *magnifies* negative reactions.

# Survey



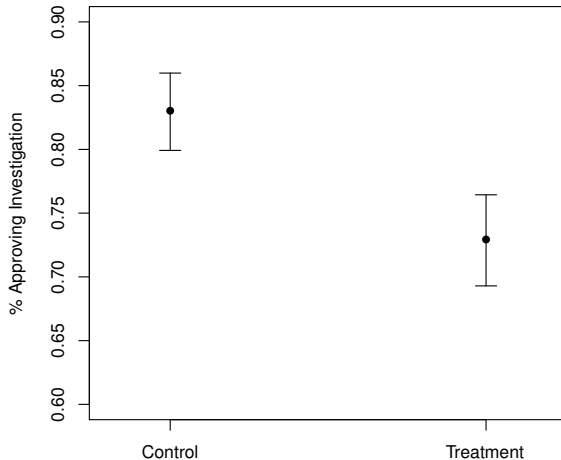
- ▶ 1,000 respondents.
- ▶ National, proportionate to region, urban/rural pop.
- ▶ Face-to-face Kyrgyz or Russian.
- ▶ November 2015.
- ▶ In-depth pilots.



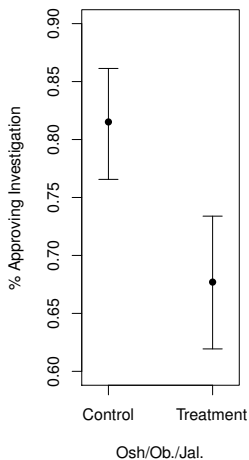
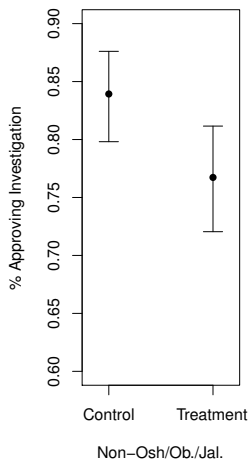
# Survey Instrument

- ▶ Brief introduction to the ICC.
- ▶ Are you aware of the ICC?
- ▶ Control: *Some people have suggested that the International Criminal Court should investigate the violence that occurred in other countries.*
- ▶ Treatment: *Some people have suggested that the International Criminal Court should investigate the violence that occurred in the Southern part of Kyrgyzstan in 2010.*
- ▶ *Do you think that these investigations would be a good or bad thing?*

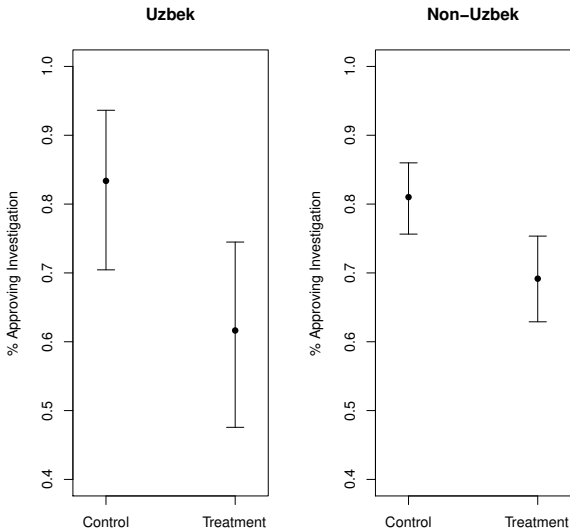
# Overall Treatment Effects



# Proximity Results: Region Effects



# Uzbeks vs Non-Uzbeks in Osh



# Enumerators

*Respondents from Osh city and Osh oblast perceived the [investigation question] in the [forms with the treatment] extremely negatively. These respondents said that **this question had to be raised in 2010, and now there is no necessity to raise this question up, since it was difficult to improve the situation but the stability has been finally restored.***

- SIAR Technical Report 2015

- ▶ The prospect of contestation made respondents less receptive to the ICC.

# Kyrgyzstan Take-Aways

- ▶ Proximity can magnify negative reactions.
  - ▶ IO's increase contestation, which makes public reactions more negative.
- ▶ Future work: Philippine war on drugs.
  - ▶ Goal: fine grained measures of contestation and public opinion.
  - ▶ Scrape all tweets about Duterte, WOD.
  - ▶ Media report generated events data.
  - ▶ On Thursday, the ICC opened up a preliminary investigation into the Philippines.

# Public Reactions to International Legal Institutions: The ICC in a Developing Democracy

Stephen Chaudoin *University of Illinois*

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► Model Setup

► Effect on Contest

► Value to IO

## ► Kyrgyzstan:

► Bayes

► Balance

► Full Sur. Instr.

► Regions

► Sum Stats

► Protest Tmt.

► ICC Intro.

► Tmt. Simple

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► Deaths

► Aut. Events

► Twitter Workflow

► Goldstein

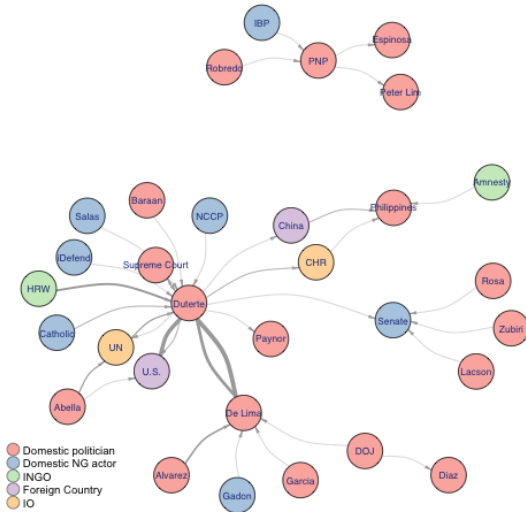
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# Philippines Background

- ▶ June 2016: Duterte elected, war on drugs begins.
- ▶ June 2017: 8,000 people dead.
- ▶ The war on drugs conducted publicly.
- ▶ The war on drugs likely violates international law.
- ▶ A chance to observe patterns of contestation before and after international actions.
- ▶ Objectives:
  1. A fine-grained dataset of who does what, when.
  2. Tighter link to changes in public opinion.



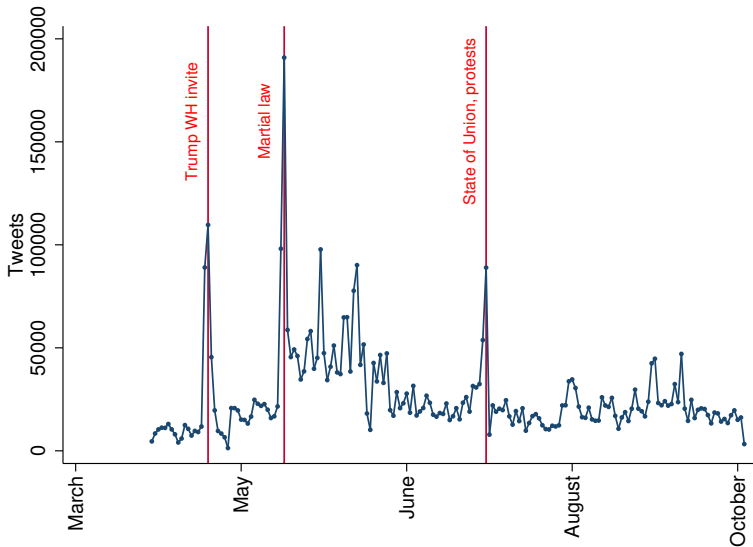
# Contestation Network



# Link Actions with Public Opinion

- ▶ Traditional polls:
  - ▶ Quarterly, nationally representative
  - ▶ Relevant, stable set of questions
- ▶ Social media data:
  - ▶ 66% of Filipinos are social media users.
  - ▶ Since Jan. 2017, I've scraped every tweet about Duterte, WOD, ICC.
  - ▶ Currently: 6M tweets, 28K added daily.

# Twitter and Events, 2017



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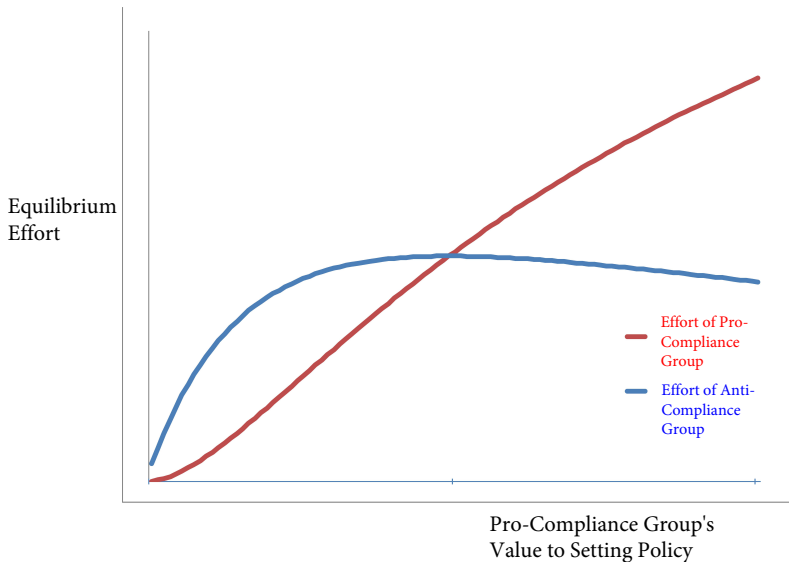
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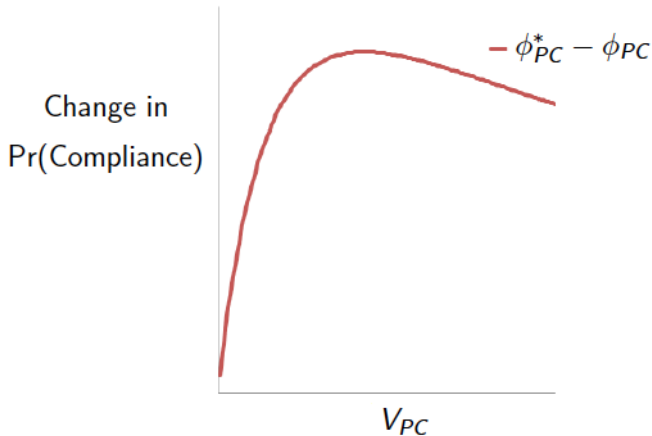
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- ▶ Objective:  $\Pi_i = \phi_i(e_i, e_j)V_i - e_i c_i$
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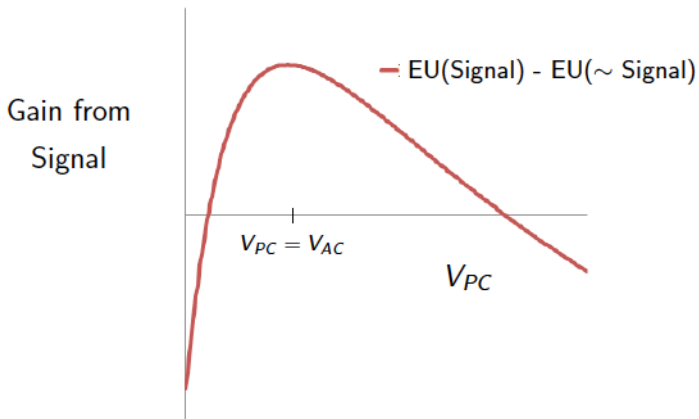
# Equilibrium Effort



# Effect of Institutional on Contest



# Value to Institution





# Treatment Effect: Bayesian

- ▶ Quantities:
  - ▶  $\theta_t$ : proportion approving investigation, treatment regime  $t \in \{\text{Tmt}, \text{Ctrl}\}$ .
  - ▶  $n_t$ : number in regime  $t$
  - ▶  $a_t$ : number approving in regime  $t$
- ▶ Priors:  $\theta_t^o \sim \beta(0.5, 0.5)$
- ▶ Posteriors:  $\theta_t^p \sim \beta(a_t + 0.5, n_t - a_t + 0.5)$
- ▶ Sample 5,000 draws from  $\theta_t^p$

# Balance in Treatment Across Regions

Variable	Control	Treatment	Adj. Diff.	z Score
Region 1	0.17800	0.17600	-0.00200	-0.08281
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Region 8	0.07600	0.07800	0.00200	0.11856
Region 9	0.05000	0.05000	0.00000	0.00000

Overall  $\chi^2$ : 0.0843

p-value = 1.000

Hansen and Bowers (2008)

# Balance in Treatment by Covariate

Variable	Control	Treatment	Adj. Diff.	z Score
Kyrgyz Lang.	0.668	0.662	-0.006	-0.201
Under 50	0.674	0.638	-0.036	-1.20
Male	0.318	0.482	0.164	5.29***
Any PS Educ.	0.358	0.384	0.026	0.851
Employed	0.244	0.284	0.04	1.43
Income Ab. Av.	0.788	0.766	-0.022	-0.835

Overall  $\chi^2$ : 30.9

p-value = 2.61e-05

Hansen and Bowers (2008)

- ▶ Treatment effect not significantly different for males.

# Regional Populations

Province	Urban population			Rural population		
	Pop.	%	No. of int.	Pop.	%	No. of int.
<i>Batken</i>	66,813	1.95%	20	195,933	5.71%	57
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<b>TOTAL</b>	<b>1,263,428</b>	<b>36.81%</b>	<b>368</b>	<b>2,168,978</b>	<b>63.19%</b>	<b>632</b>

# Full Survey Outline

- ▶ Demographics: Age, Gender, Education level, employment, income, region
- ▶ Government opinion: Satisfied with government, things getting better
- ▶ Protest: first, second most important issues
- ▶ Protest Experiment
- ▶ ICC Intro, have you heard of the ICC?
- ▶ ICC Experiment
- ▶ Outcome measures

# ICC Intro in Survey

*As you may or may not know, Kyrgyzstan has taken steps to join an international organization called the International Criminal Court. The International Criminal Court is located in The Hague, in the Netherlands but addresses issues in many countries. The court tries to investigate and prosecute individuals who are accused of serious crimes like genocide, crimes against humanity or crimes committed during wartime.*

- ▶ “Have you heard of the ICC? [Yes/No]”
  - ▶ 25% Yes
  - ▶ Osh/Osh Obl.: 22% Yes
  
- ▶ ABA US Polls: 32% “heard [at least] a little bit about the ICC.”

# Summary Statistics

	Full mean	Osh/Ob./Jal. mean	Non-Osh mean	Osh city mean	Osh oblast mean	Jalal-Abad mean
Treatment	0.50	0.50	0.50	0.50	0.50	0.50
App. Inv.	0.63	0.63	0.62	0.68	0.70	0.53
App. Inv. Num.	2.98	3.03	2.92	2.76	2.97	2.92
App. ICC	0.54	0.50	0.60	0.68	0.73	0.44
App. ICC Num.	2.96	2.92	3.01	2.82	3.08	2.98
DK/RTA	0.19	0.22	0.16	0.00	0.06	0.32
Heard of ICC	0.24	0.29	0.17	0.42	0.17	0.10
Government Approval	2.63	2.59	2.68	2.48	2.77	2.65
Uzbek	0.12	0.05	0.20	0.18	0.23	0.18
Age Under 50	0.66	0.66	0.64	0.84	0.65	0.59
Male	0.40	0.40	0.40	0.32	0.40	0.44
Post Sec. Ed.	0.37	0.42	0.30	0.76	0.22	0.26
Employed	0.26	0.28	0.25	0.40	0.28	0.17
Inc. Aver.	0.78	0.73	0.84	0.94	0.85	0.79
N	1,000	579	421	50	192	179

# Treatment Effects: Simple

<i>Investigation Approval (Binary)</i>					
Treatment Group	N	% Approv.	Diff.	S.E.	t-stat
Control	396	83.1			
Treatment	410	72.9	-10.2	0.03	-3.50

<i>Investigation Approval (Categorical)</i>				
Treatment Group	Very bad (1)	Somewhat bad (2)	Somewhat good (3)	Very good (4)
Control	22 5.6%	45 11.4%	208 52.5%	121 30.6%
Treatment	51 12.4%	60 14.6%	183 44.6%	116 28.3%



# Protest Treatment

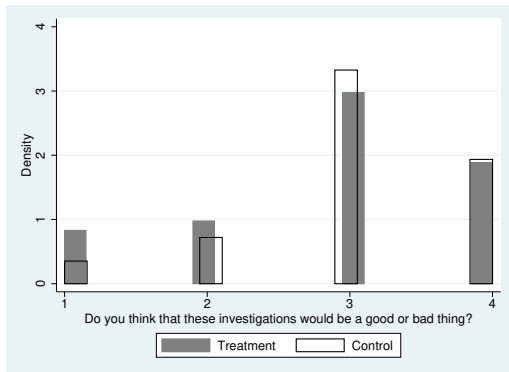
	Region Cluster	Region FE	Exclude "DK/RTA"	with Controls	
	(1)	(2)	(3)	(4)	(5)
Protest Tmt.	-.120	-.120	-.132	-.067	-.099
	(0.131)	(0.149)	(0.133)	(0.17)	(0.135)
Kyrgyz Gov. Approv.					0.142
					(0.1)
Kyrgyz Lang.					-.070
					(0.149)
Under 50					0.167
					(0.143)
Male					0.018
					(0.141)
Any PS Educ.					0.332
					(0.149)**
Employed					-.021
					(0.164)
Income Ab. Av.					0.03
					(0.163)
N	1,000	1,000	1,000	806	960

# Treatment Effects: Logits

	Logit (1)	Region FE (2)	w/ controls (3)	Region FE w/ controls (4)
Treatment	-.600 (0.134)***	-.606 (0.175)***	-.574 (0.139)***	-.578 (0.18)***
Uzbek			-.145 (0.12)	-.118 (0.284)
Under 50			-.039 (0.167)	-.035 (0.19)
Male			-.151 (0.205)	-.175 (0.187)
Post Sec. Ed.			0.16 (0.211)	0.269 (0.197)
Employed			-.080 (0.255)	-.017 (0.208)
Income Ab. Av.			-.176 (0.24)	-.177 (0.224)
Const.	1.591 (0.1)***		1.786 (0.251)***	
Obs.	806	806	806	806

# Treatment Effects: Inv. Appr., Categorical

Figure: Distribution of Inv. Approval, Categorical



1: Very bad idea, 4: Very good idea

# Multilevel Model

- ▶ Let treatment effect vary by region.
- ▶ Include region-specific intercepts and controls.

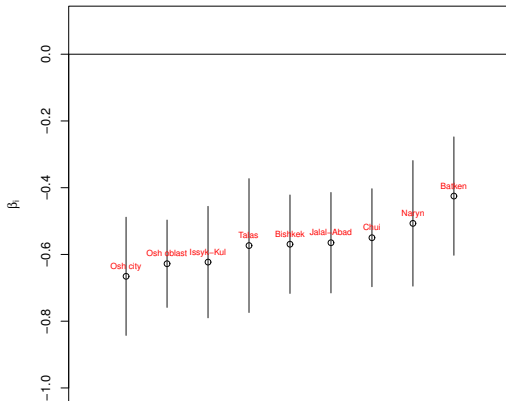
$$y_i = \beta_0 + \beta_i \textit{Treatment}_i + X_i \Gamma + \epsilon_i$$

$$\beta_0 = u_{1j}$$

$$\beta_i = u_{2j}$$

(1)

# Multilevel Model: Results

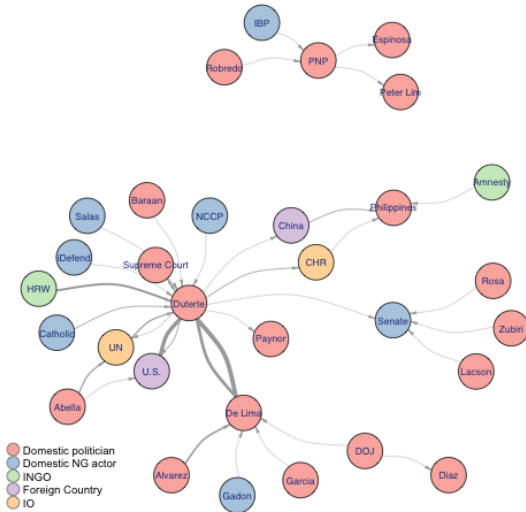


Region  $i$ , ranked by Estimated Treatment Effect  $\beta_i$

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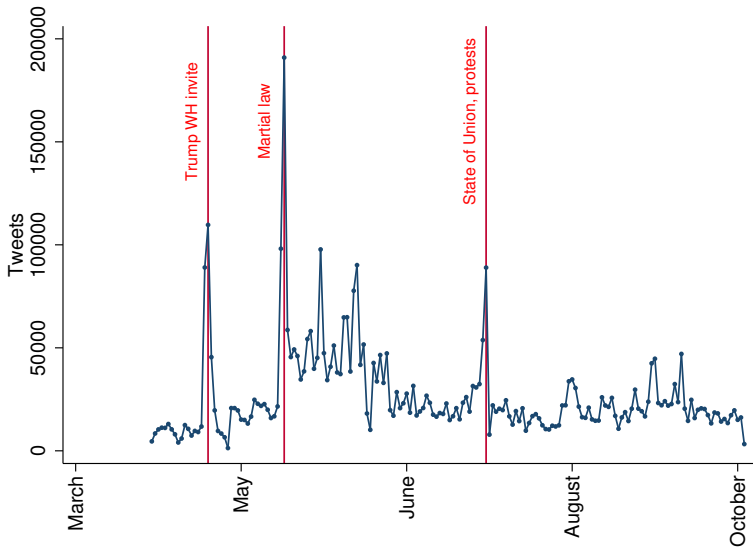


# Link Actions with Public Opinion

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  - ▶ Currently: 5M tweets, 28K added daily.



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# Philippines Data and Contestation

- ▶ As international institutions intervene:
  - ▶ Does this trigger blanket mobilization by pro-IO groups?
  - ▶ Or contestation from both types of groups?
- ▶ As contestation unfolds:
  - ▶ Is there a uniform increase in opposition to the war?
  - ▶ Or increased polarization as groups compete?

# Conclusions

- ▶ Preference heterogeneity and contestation explains:
  - ▶ When IOs are effective.
  - ▶ How citizens react.
- ▶ Raises normative questions about what IO actions are most desirable.

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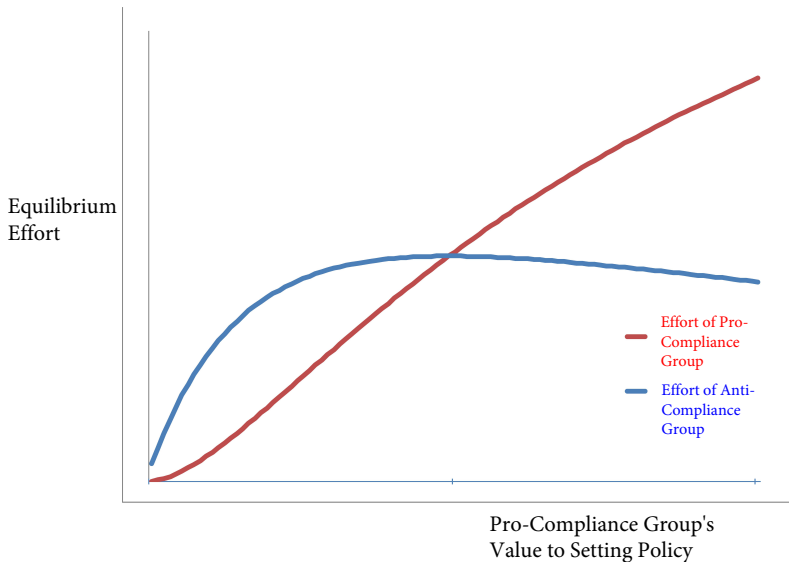
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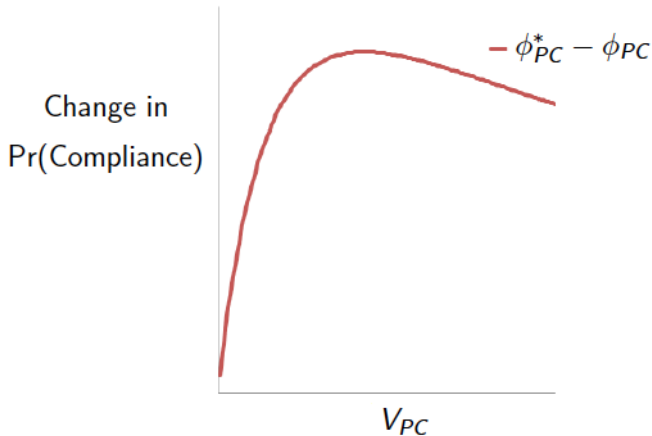
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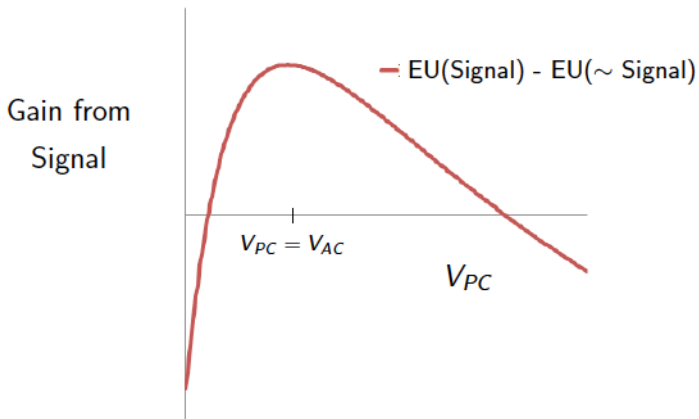
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- ▶ ABA US Polls: 32% “heard [at least] a little bit about the ICC.”

# Summary Statistics

	Full mean	Osh/Ob./Jal. mean	Non-Osh mean	Osh city mean	Osh oblast mean	Jalal-Abad mean
Treatment	0.50	0.50	0.50	0.50	0.50	0.50
App. Inv.	0.63	0.63	0.62	0.68	0.70	0.53
App. Inv. Num.	2.98	3.03	2.92	2.76	2.97	2.92
App. ICC	0.54	0.50	0.60	0.68	0.73	0.44
App. ICC Num.	2.96	2.92	3.01	2.82	3.08	2.98
DK/RTA	0.19	0.22	0.16	0.00	0.06	0.32
Heard of ICC	0.24	0.29	0.17	0.42	0.17	0.10
Government Approval	2.63	2.59	2.68	2.48	2.77	2.65
Uzbek	0.12	0.05	0.20	0.18	0.23	0.18
Age Under 50	0.66	0.66	0.64	0.84	0.65	0.59
Male	0.40	0.40	0.40	0.32	0.40	0.44
Post Sec. Ed.	0.37	0.42	0.30	0.76	0.22	0.26
Employed	0.26	0.28	0.25	0.40	0.28	0.17
Inc. Aver.	0.78	0.73	0.84	0.94	0.85	0.79
N	1,000	579	421	50	192	179

# Treatment Effects: Simple

<i>Investigation Approval (Binary)</i>					
Treatment Group	N	% Approv.	Diff.	S.E.	t-stat
Control	396	83.1			
Treatment	410	72.9	-10.2	0.03	-3.50

<i>Investigation Approval (Categorical)</i>				
Treatment Group	Very bad (1)	Somewhat bad (2)	Somewhat good (3)	Very good (4)
Control	22 5.6%	45 11.4%	208 52.5%	121 30.6%
Treatment	51 12.4%	60 14.6%	183 44.6%	116 28.3%



# Protest Treatment

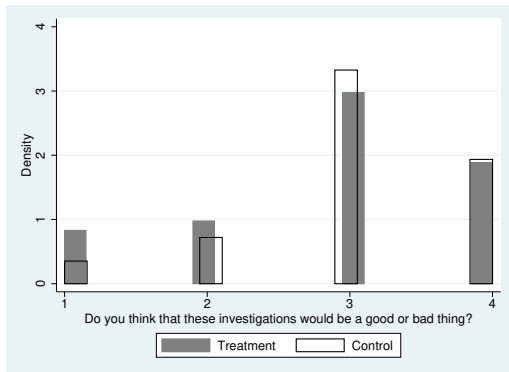
	Region Cluster	Region FE	Exclude "DK/RTA"	with Controls	
	(1)	(2)	(3)	(4)	(5)
Protest Tmt.	-.120	-.120	-.132	-.067	-.099
	(0.131)	(0.149)	(0.133)	(0.17)	(0.135)
Kyrgyz Gov. Approv.					0.142
					(0.1)
Kyrgyz Lang.					-.070
					(0.149)
Under 50					0.167
					(0.143)
Male					0.018
					(0.141)
Any PS Educ.					0.332
					(0.149)**
Employed					-.021
					(0.164)
Income Ab. Av.					0.03
					(0.163)
N	1,000	1,000	1,000	806	960

# Treatment Effects: Logits

	Logit (1)	Region FE (2)	w/ controls (3)	Region FE w/ controls (4)
Treatment	-.600 (0.134)***	-.606 (0.175)***	-.574 (0.139)***	-.578 (0.18)***
Uzbek			-.145 (0.12)	-.118 (0.284)
Under 50			-.039 (0.167)	-.035 (0.19)
Male			-.151 (0.205)	-.175 (0.187)
Post Sec. Ed.			0.16 (0.211)	0.269 (0.197)
Employed			-.080 (0.255)	-.017 (0.208)
Income Ab. Av.			-.176 (0.24)	-.177 (0.224)
Const.	1.591 (0.1)***		1.786 (0.251)***	
Obs.	806	806	806	806

# Treatment Effects: Inv. Appr., Categorical

Figure: Distribution of Inv. Approval, Categorical



1: Very bad idea, 4: Very good idea

# Multilevel Model

- ▶ Let treatment effect vary by region.
- ▶ Include region-specific intercepts and controls.

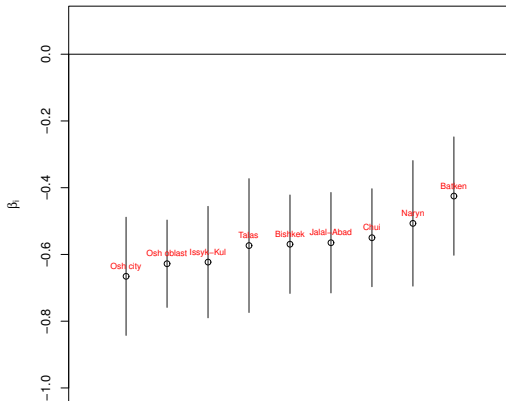
$$y_i = \beta_0 + \beta_i \textit{Treatment}_i + X_i \Gamma + \epsilon_i$$

$$\beta_0 = u_{1j}$$

$$\beta_i = u_{2j}$$

(2)

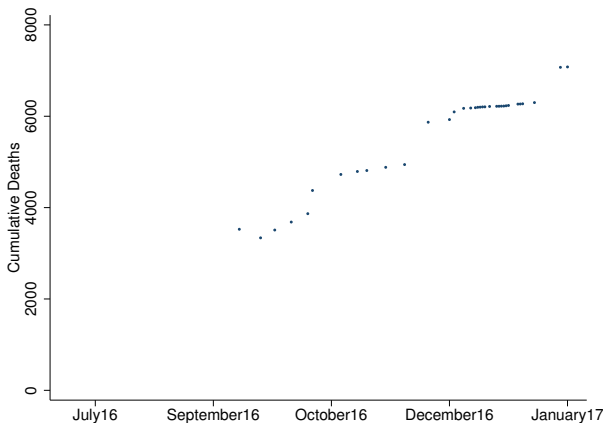
# Multilevel Model: Results



Region  $i$ , ranked by Estimated Treatment Effect  $\beta_i$

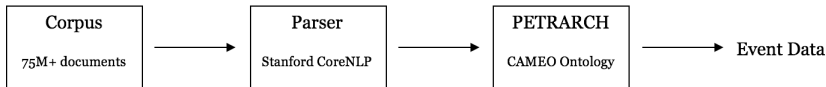
# Deaths from War on Drugs

## Cumulative Deaths through Jan. 2017



Source: Rappeler's "By the Numbers"

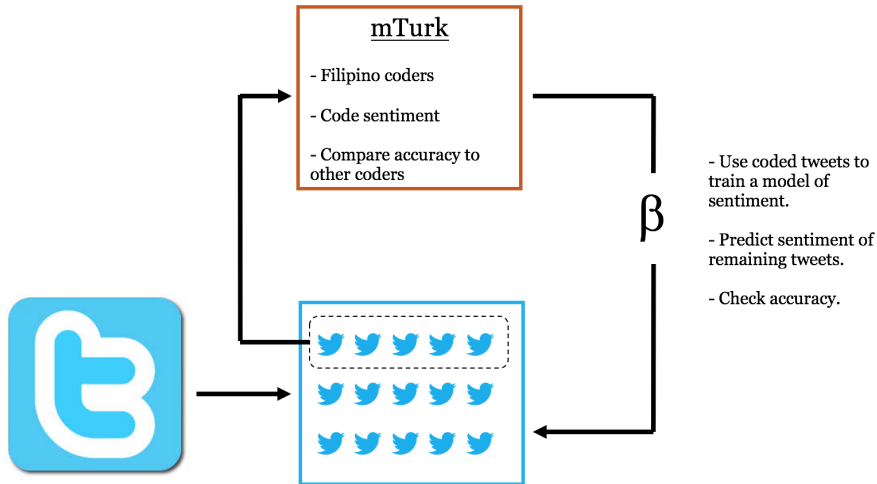
# Automated Event Coding



<u>Source</u>	<u>Target</u>	<u>Verbs</u>	<u>Action</u>
NGOHRI	PHLCOP	['has', 'said', 'accused']	11: Disapprove - 112: Accuse --- 1122: Accuse of human rights abuses

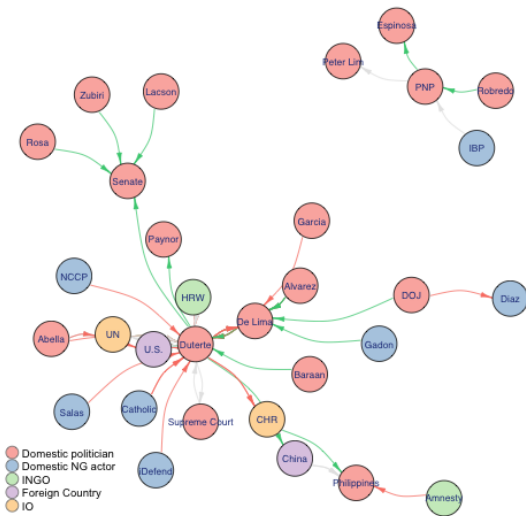
- ▶ Identify 14,510 events from 122,720 documents
- ▶ June 2016 - Sept 2017.

# Twitter Workflow





## Event Network: Goldstein Scale



# Types of Variation

- ▶ Intensive margin:
  - ▶ How many actions does  $i$  take against  $j$ ?
  - ▶ How many resources does  $i$  use against  $j$ ?
- ▶ Extensive margins:
  - ▶ How many actors does  $i$  target?
  - ▶ How many actors target  $j$ ?
- ▶ Qualitative margins:
  - ▶ What actions does  $i$  take against  $j$ ?
  - ▶ Qualities of media coverage of those actions?
- ▶ Temporal margins:
  - ▶ How do these change over time?
  - ▶ In response to shocks?

# Public Reactions to International Legal Institutions: The ICC in a Developing Democracy

Stephen Chaudoin *University of Illinois*

## ► Model:

► Model Setup   ► Effect on Contest   ► Value to IO

## ► Kyrgyzstan:

► Bayes   ► Balance   ► Full Sur. Instr.   ► Regions   ► Sum Stats   ► Protest Tmt.   ► ICC Intro.

► Tmt. Simple   ► Tmt. Logit   ► Categorical DV   ► Multilevel Model

## ► Philippines:

► Deaths   ► Aut. Events   ► Twitter Workflow   ► Goldstein   ► Margins