

Tug of War

The Heterogeneous Effects of Outbidding between Terrorist Groups

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What is outbidding?

Explanation for violence in civil conflict with multiple anti-gov groups

- ① Groups use violence to increase public support at other groups' expense
- ② Terrorism signals resolve to a population that is uncertain about which group is strongest

Unique story in literature: enemy tangentially related to interaction

- Kydd & Walter 2006: "it provides a potential explanation for terrorist attacks that continue even when they seem unable to produce any real results"

Popularity entails instrumental benefits: recruitment, financing, information

- Crenshaw 1981; Acosta 2014; Fortna 2015

The outbidding mechanism

Group competition should generally exasperate violence (Bloom 2004)

- Eg: Rise of Hamas encouraging Fatah to use violence

Conrad and Greene (2015) summarize logic:

Since competition directly and indirectly threatens the resource base necessary to sustain the organization and ensure its effectiveness, it follows that terrorist organizations should make tactical choices in an effort to increase their share of resources within a competitive environment

Mixed empirical evidence

- No effects: Findley & Young 2015
- Mixed effects: Nemeth 2014; Conrad & Greene 2015
- Positive effects: Chenoweth 2010; Wood & Kathman 2015

How to study outbidding?

Mixed results are not surprising: Outbidding is consistent with both encouragement *but also deterrence*

- If a side becomes more effective, others may fight harder to keep up
- If a side becomes more effective, others may give up, creating a feedback loop

If deterrence exists, previous (reduced form) approaches are ambiguous:

- Focus on overall average effects, but countervailing/asymmetric effects may exist
- As such, *any* reduced form output is consistent with outbidding (non-falsifiable)

Measurement challenges

- Most proxy competitive incentives with number of groups
- Hard to measure popularity with cross-sectional data

Our approach

In this paper, we . . .

- ① Model outbidding as a dynamic contest between two groups
 - Competitive incentives: value of popularity, effectiveness of terrorism at boosting public opinion, cost of attacking
- ② Measure monthly relative popularity of Fatah and Hamas, 1994-2018
- ③ Fit model to data on popularity and terrorist attacks (GTD)
- ④ Use the fitted model to quantify how much competition affects violence

Competition has a double edge

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Competition from Hamas drives Fatah's violence, not vice versa

- Competition from Hamas led to a 34% increase in violence from Fatah during this 1994-2018 period
- For Hamas we find heterogeneous effects. During the Oslo era, competition *decreases* Hamas's propensity for violence, but encourages after.

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Asymmetric effects of competitive incentives:

- Increasing Hamas's incentives, increases violence by both groups
- Increasing Fatah's incentives, decreases violence by both groups

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Results emerge from asymmetric contests with a Fatah advantage

- Fatah is more effective at shifting public opinion, although Hamas has smaller attack costs

Contributions

Outbidding literature

- Even in the canonical example, competitive effects have heterogeneous and opposing effects on violence.
- Emerge without needing to consider the government or free riding.

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Policy

- Many recent presidential administrations provided support to build Fatah's capabilities
- Our model suggests countervailing effects: Lowers the costs of violence for Fatah (pro-peace), but may make Fatah value public support less (anti-peace)
- More work required to determine how U.S. and Israeli actions directly affect specific competitive incentives

Setup

Actors: Fatah and Hamas ($i = F, H$)

Time: $t = 1, 2, \dots$

Relative popularity of the groups: $s^t \in \mathcal{S} = \{s_1, \dots, s_K\} \subset \mathbb{R}$

- Public state variable observed in period t
- $s_k < s_l$ denotes Hamas is more popular relative to Fatah in state s_k than s_l
- equally spaced: $s_k - s_{k-1} = 2d$

Private State Variable: $\varepsilon_i^t \in \mathbb{R}^2$

- Private information, action-specific shocks
- $\varepsilon_i^t(a_i)$ indexes shock to action a_i for player i in period t

Strategic interaction

Timing in period t :

- 1 i observes relative popularity level s^t and private information ε_i^t
- 2 Simultaneously choose to commit terrorism ($a_i^t = 1$) or not ($a_i^t = 0$)
- 3 Payoffs are accrued
- 4 Transition to period $t + 1$

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Payoffs: $\underbrace{u_i(a_i^t, s^t; \theta)}_{\text{systematic}} + \underbrace{\varepsilon_i^t(a_i^t)}_{\text{idiosyncratic}}$

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- β_i is value of public support and κ_i is cost of terrorism
- Hypothesis: $\beta_H < 0$, $\beta_F > 0$, and $\kappa_i < 0$

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- Private information ε_i^{t+1} is drawn i.i.d. T1EV

Popularity follows AR-1 process

Tomorrow's s^{t+1} drawn from a discretized normal with mean:

$$\mu[a^t, s^t; \gamma] = \gamma_0 + \gamma_1 \cdot s^t + \underbrace{(\gamma_{H,1} + \gamma_{H,2} \cdot s^t)}_{\text{effect of } H\text{'s attacks}} \cdot a_H^t + \underbrace{(\gamma_{F,1} + \gamma_{F,2} \cdot s^t)}_{\text{effect of } F\text{'s attacks}} \cdot a_F^t$$

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- $\gamma_{i,1} + \gamma_{i,2} \cdot s^t$ is effectiveness at using terrorism to gain support
- Hypothesis: $\gamma_{H,1} < 0$ and $\gamma_{F,1} > 0$
- No expectations over how states affect effectiveness, $\gamma_{i,2}$

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Transition probabilities take the form:

$$f(s'; a^t, s^t, \gamma) = \begin{cases} \Phi\left(\frac{s_1 + d - \mu[a^t, s^t; \gamma]}{\sigma}\right) & s' = s_1 \\ \Phi\left(\frac{s' + d - \mu[a^t, s^t; \gamma]}{\sigma}\right) - \Phi\left(\frac{s' - d - \mu[a^t, s^t; \gamma]}{\sigma}\right) & s' \notin \{s_1, s_K\} \\ 1 - \Phi\left(\frac{s_K - d - \mu[a^t, s^t; \gamma]}{\sigma}\right) & s' = s_K. \end{cases}$$

Equilibria

Maximize expected utility with common discount factor δ

- δ not uniquely identified (Abbring and Daljord 2020)
- fix $\delta = 0.999$ after comparing likelihood for several choices of δ .

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Aguirregabiria and Mira (2007) characterize Markov Perfect equilibria

- Net-of-shock expected utility: $v_i(a_i, s)$
- v is an equilibrium if and only if $\mathcal{V}(v) = v$
- Eq. Choice probabilities: $P(a_i, s; v) = \frac{\exp\{v_i(a_i, s)\}}{\exp\{v_i(0, s)\} + \exp\{v_i(1, s)\}}$
- \mathcal{V} depends on $\beta, \kappa, \gamma, \sigma$, estimated below

Data

Treat a period as a month and need measures of actions and states

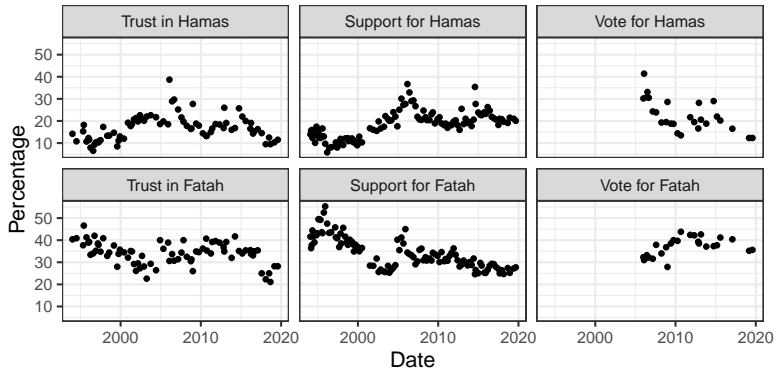
Monthly indicator of attacks using Global Terrorist Database

- Hamas: Islamic Resistance Movement, *Al-Qassam* Brigades
- Fatah: *Al-Fatah*, Palestinian Liberation Organization

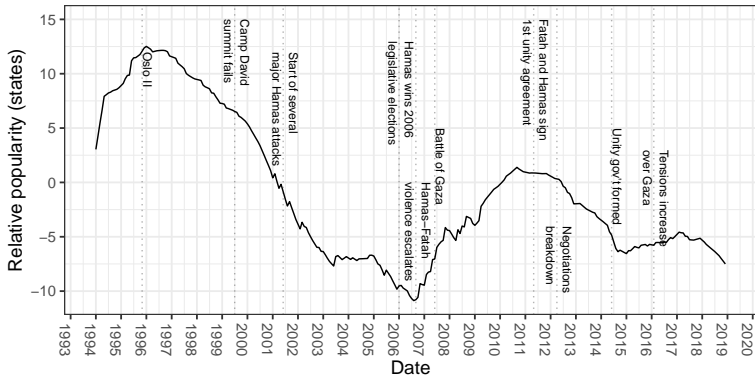
Public opinion polls from JMCC and PSR

- Rolling cross sections; access to polling totals, but not individual data
- Face-to-face interviews, representative sample of West Bank and Gaza Strip
- Appear in Bloom 2004; Clauset et al. 2010; Jaeger et al. 2012
- 1994–2018 gives us 300 months

Six public opinion indicators



Our measure of relative popularity



- Dynamic factor analysis backs out a continuous state variable \tilde{s}^t
- All indicators load onto \tilde{s}^t in the expected direction
- Timeline suggests that \tilde{s}^t is likely measuring relative popularity

Two-step estimation

Step 1: Estimate effectiveness ($\gamma_{i,1}$) and transition probs

- AR-1 models with continuous \tilde{s}^{t+1} as the dependent variable:

$$\tilde{s}^{t+1} = \gamma_0 + \gamma_1 \tilde{s}^t + \gamma_{H1} a_H^t + \gamma_{H2} (\tilde{s}^t \times a_H^t) + \gamma_{F1} a_F^t + \gamma_{F2} (\tilde{s}^t \times a_F^t)$$

- Unit root problems: point estimates are super-consistent but SEs are incorrect
- Use error correction methods for inference/robustness

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Step 2: Estimate value of public support and cost of terrorism (β_i, κ_i) and equilibrium expected utilities v

- Discretize state space and use $\hat{\gamma}$ and $\hat{\sigma}$ from step 1
- Use CMLE a la Su and Judd (2012)
- Murphy and Topel (1985) SE correction

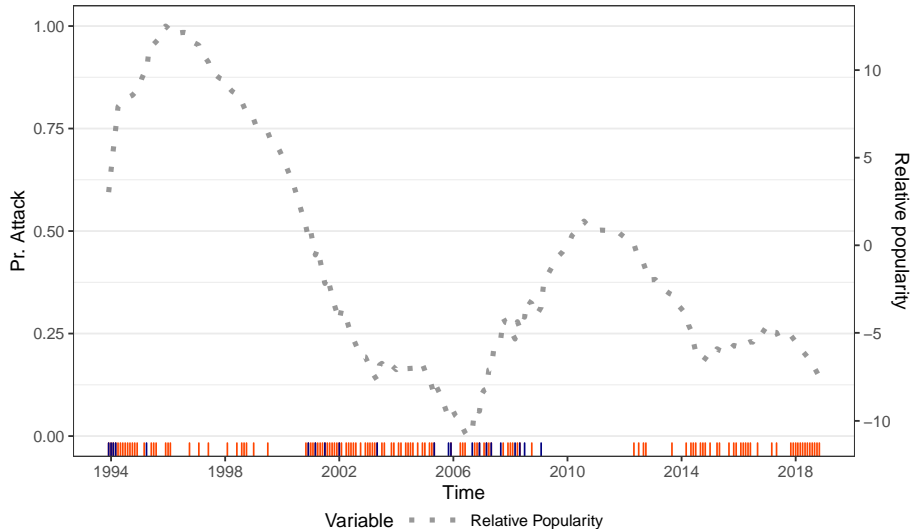
Effectiveness

<i>Model:</i>	AR(1)	ECM
<i>Dependent variable:</i>	ξ^{t+1}	$\Delta \xi^{t+1}$
Hamas attacks ($\gamma_{H,1}$)	-0.21	-0.21 (0.04)
Fatah attacks ($\gamma_{F,1}$)	1.12	1.03 (0.05)
Hamas attacks \times Lag states ($\gamma_{H,2}$)	0.01	0.002 (0.01)
Fatah attacks \times Lag states ($\gamma_{F,2}$)	0.03	0.01 (0.01)
Lag state (γ_1)	1.00	
Δ Lag state		0.33 (0.04)
Constant (γ_0)	-0.02	-0.01 (0.02)
σ	0.22	0.18

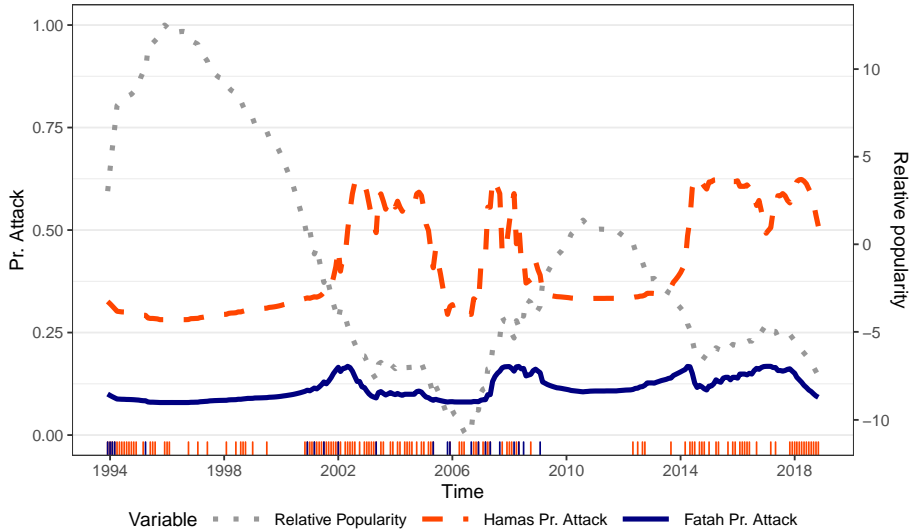
Values and costs

		Estimate	Std. Error
Value of public support , β_i	Hamas	-0.007	0.005
	Fatah	0.0005	0.0004
Attack costs, κ_i	Hamas	-0.95	0.28
	Fatah	-2.45	0.39
Log-Likelihood		-278.09	

Estimated EQ choice probabilities



Estimated EQ choice probabilities



Quantifying Effects of Competition

Single-agent comparison: How do i 's equilibrium attack probabilities compare to i 's single agent problem?

- Quantifies the total effect of competition on violence

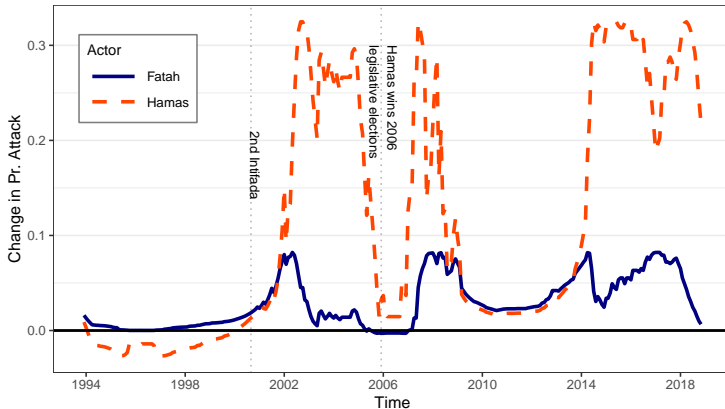
Comparative statics: How do equilibrium attack probabilities change with different competitive incentives? (Deferred for time)

Direction of effects are not obvious even with model

Effects of competition

Subtract single-agent attack probabilities from estimated EQ ones

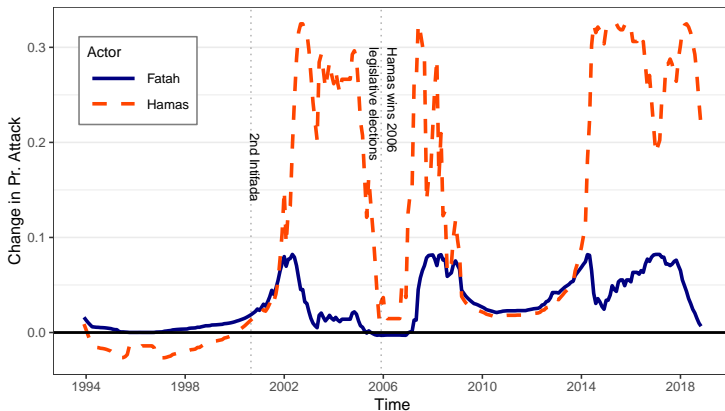
- EQ probs capture violence *with* competition
- Single-agent probs capture violence *without* competition



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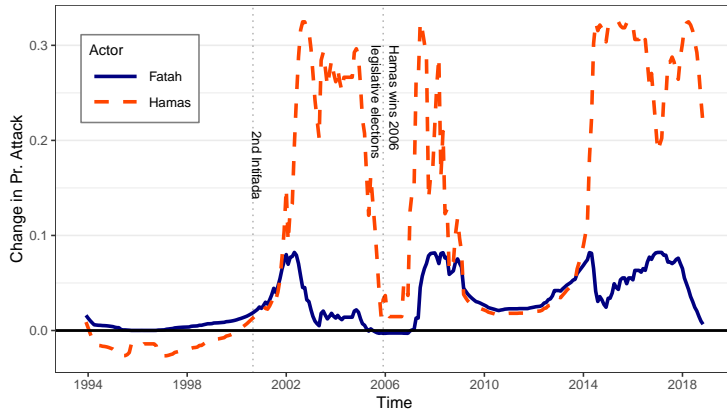


Takeaway 1: Competition from Hamas encourages Fatah to use more violence

Effects of competition

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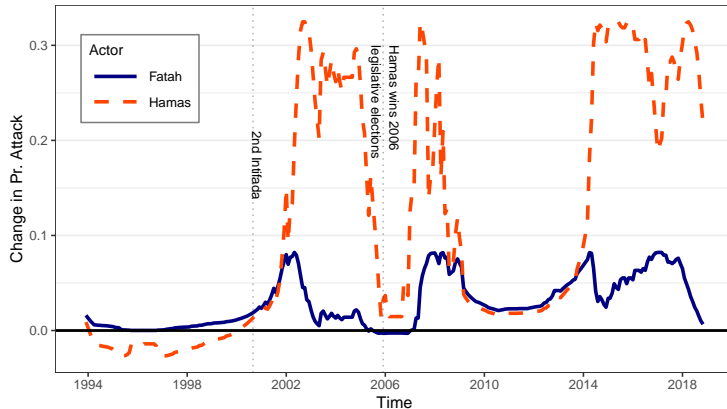


Takeaway 2: Averaging over time, Fatah uses 34% more violence than it would without Hamas attacks (expected)

Effects of competition

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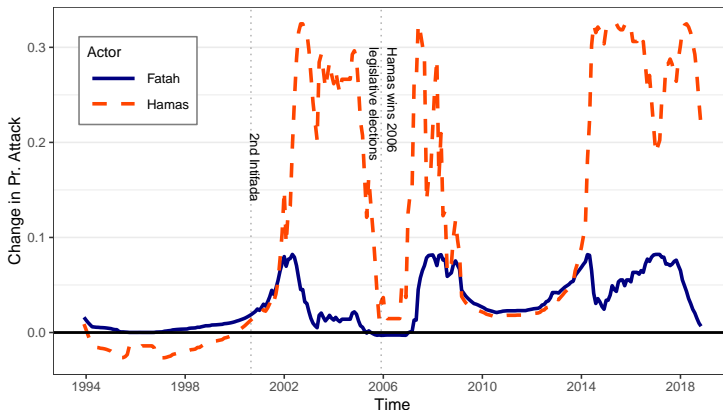


Takeaway 3: Averaging over time (post-Oslo), Hamas would use 37% more violence than it would without Fatah attacks (expected)

Effects of competition

Subtract single-agent attack probabilities from estimated EQ ones

- EQ probs capture violence *with* competition
- Single-agent probs capture violence *without* competition



Takeaway 4: Average masks heterogeneity: when Fatah is popular (during Oslo), competition depresses Hamas attacks by about 4% (not expected)

Effects of competition: By era

Subtract single-agent attack probabilities from estimated EQ ones

- EQ probs capture violence *with* competition
- Single-agent probs capture violence *without* competition

	Oslo era	2nd Intifada	post-2006 election
Hamas	−0.014 (0.001)	0.177 (0.010)	0.148 (0.007)
Fatah	0.005 (0.0003)	0.027 (0.002)	0.043 (0.001)

Effects of competition: By era

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Takeaway 1: Heterogeneous effects over time *and* across groups

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Takeaway 2: Competition from Hamas is making Fatah more violent over time

Effects of competition: By era

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Takeaway 3: Competition from Fatah was only deterring during the Oslo era

Additional results

Robustness checks

- 1 Vary the end of the time frame: 2000, 2006, 2011, 2015
- 2 Grid search for best δ
- 3 Specification of measurement model
- 4 Specification of first-step model

Comparative statics

- 1 Changes to the effectiveness of terrorism γ
- 2 Changes to the value of support β
- 3 Changes to the cost of terrorism κ

Model fit

- 1 Choice probabilities and attack decisions
- 2 Comparison to “no competition” model

Concluding implications

For the literature: group competition has countervailing effects on Hamas and Fatah's use of violence

Stronger incentives to compete can decrease violence

- Fatah is effective at shifting support, so when its incentives to compete increase, it deters Hamas from attacking, creating a feedback loop
- Explanation does not require free riding or government behavior

Even in canonical outbidding example, competition has a complicated relationship with violence

We help to explain why it is difficult to find evidence of outbidding

Concluding implications

For policy: US often supporting Fatah to counterbalance Hamas

- Kalman 2006

Could be effective depending on dominant effect:

- ① Peace enhancing if foreign support increases Fatah's effectiveness or decreases its cost
- ② Peace diminishing if foreign support decreases Fatah's value of local support

Uniformly raising costs of terrorism is not necessarily best

- Focus on increasing Hamas's costs *and* decreasing Fatah's.

Thank you!

Appendix

▸ Equilibrium characterization

▸ CMLE

▸ Numerical example

▸ Descriptive statistics on attacks

▸ Example: JMCC

▸ Example: PSR

▸ Correlation of public opinion indicators

▸ Measurement Model

▸ Model fit

▸ Robustness: First stage

▸ Robustness: Time frame

▸ Robustness: Discount factor

▸ Homotopy method

▸ Changing effectiveness

▸ Changing values

▸ Changing costs

Equilibrium characterization

i 's net-of-shock expected utility: $v_i(a_i, s)$

- i chooses a_i in state s if and only if

$$a_i \in \operatorname{argmax}_{a_i \in \{0,1\}} \{v_i(a_i, s) + \varepsilon_i(a_i)\}.$$

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v is an EQ if and only if $v_i(a_i, s) = \mathcal{V}_i^v(a_i, s)$ for all (i, a_i, s) ,
where:

$$\mathcal{V}_i^v(a_i, s) = u_i(a_i, s; \theta) + \sum_{a_j} P_j^v(a_j, s) \left[\delta \sum_{s'} f_l(s'; a_i, a_j, s, \gamma) V_i^v(s') \right]$$

CMLE: Su and Judd 2012

Data: $Y = \{(a^t, s^t)\}_{t=1}^T$.

- Month observations between 1994 and 2018: $T = 300$
- Actions from GTD. s^t is discretized version of our estimated relative popularity

Maximize logit log-likelihood (v, θ) :

$$\mathcal{L}(\mathbf{v} \mid Y) = \sum_{t=1}^T [\log P_H^v(a_H^t, s^t) + \log P_F^v(a_F^t, s^t)]$$

Subject to the constraint:

$$\mathcal{V}_i^v(a_i, s; \theta, \hat{\gamma}) = v_i(a_i, s), \text{ for all } i, a_i, s$$

This produces estimates of $\theta = (\beta_H, \beta_F, \kappa_H, \kappa_F)$.

- Note: $\hat{\gamma}$ estimated in step 1 via AR-1 model

Symmetric Setup

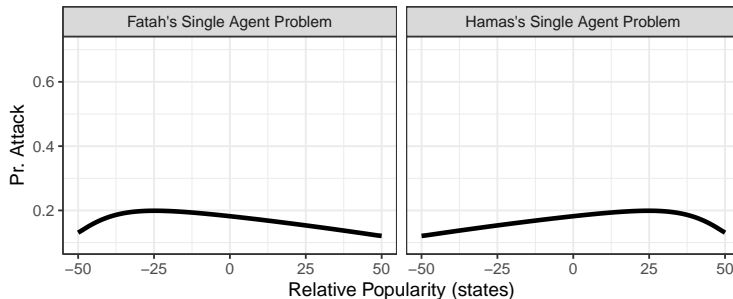
- States: $\mathcal{S} = \{-50, -49, \dots, 49, 50\}$
- Value: $\beta_F = -\beta_H = \frac{1}{500}$
- Cost: $\kappa_i = -2$
- Effectiveness: $\gamma_{F,1} = -\gamma_{H,1} = 1$ and $\gamma_{i,2} = 0$
- Mean of s^{t+1} : $\mu[a^t, s^t; \gamma] = s^t - a_H^t + a_F^t$
- Standard deviation of s^{t+1} : $\sigma = 2$

Single-Agent Problem

What is the probability that i attacks at relative popularity level s when expecting $-i$ to never attack?

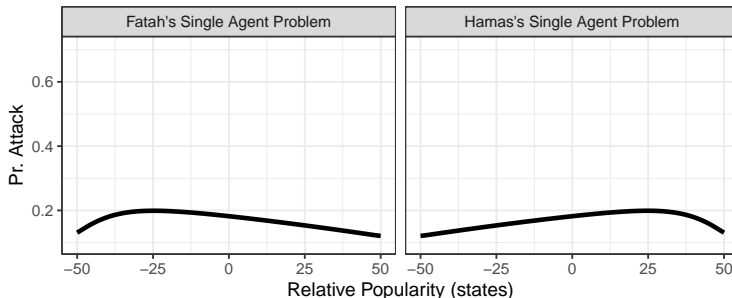
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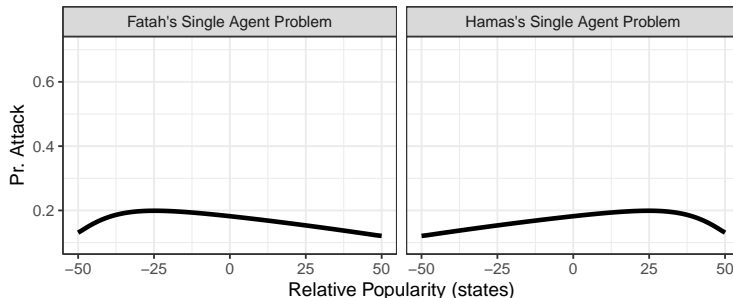
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Takeaway 1: i attacks in least favorable states (with end point effects)

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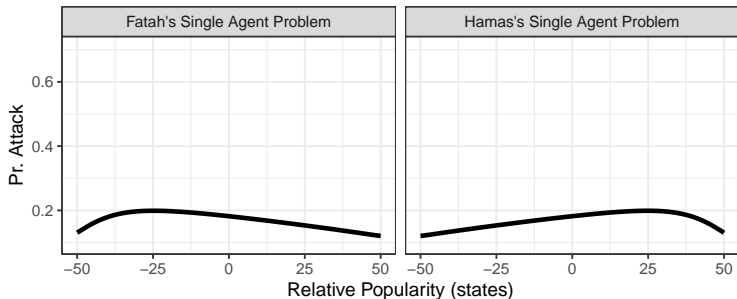
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Takeaway 2: Attack probabilities are small, from 0.1 to 0.2

Single-Agent Problem

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Takeaway 3: Attack probabilities are negatively correlated across states

Equilibria of the game

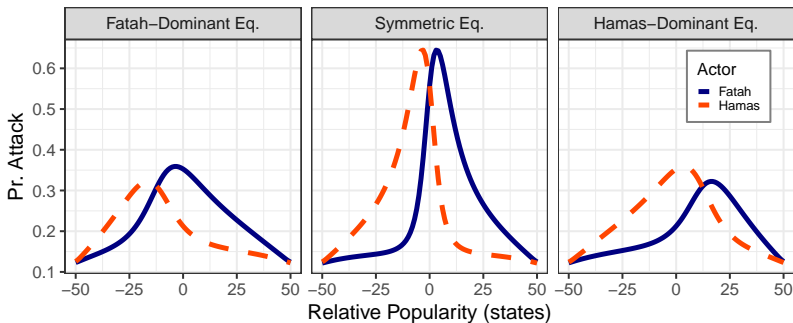
What is the equilibrium probability that i attacks in state s ?

► [Appendix](#)

► [back to talk](#)

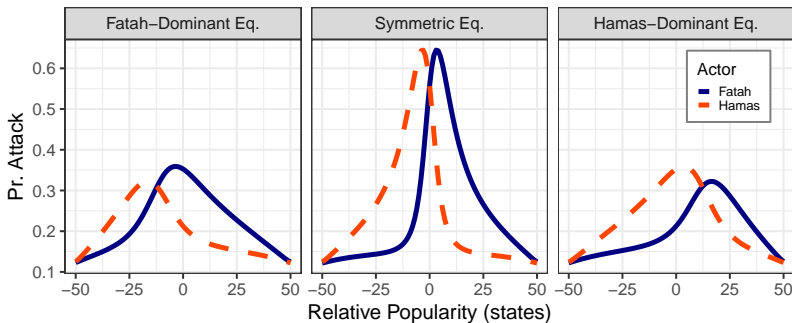
Equilibria of the game

What is the equilibrium probability that i attacks in state s ?



Equilibria of the game

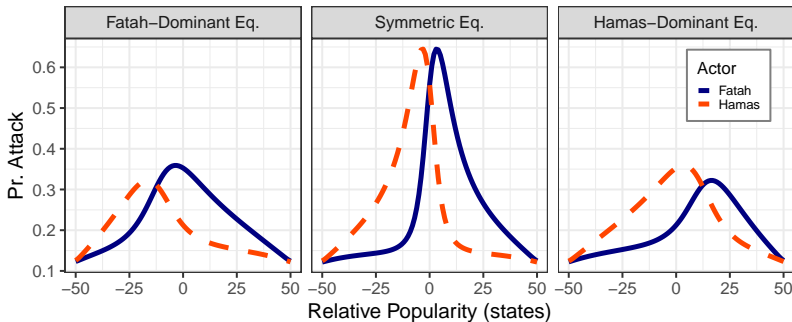
What is the equilibrium probability that i attacks in state s ?



Takeaway 1: Attack probabilities are larger, from 0.1 to 0.4 or 0.6

Equilibria of the game

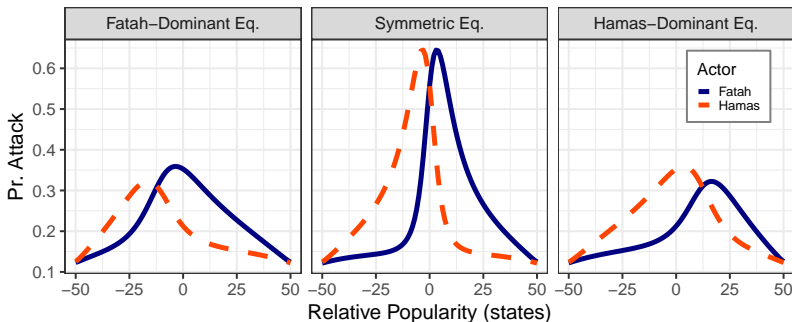
What is the equilibrium probability that i attacks in state s ?



Takeaway 2: Attack probabilities are positively correlated across states

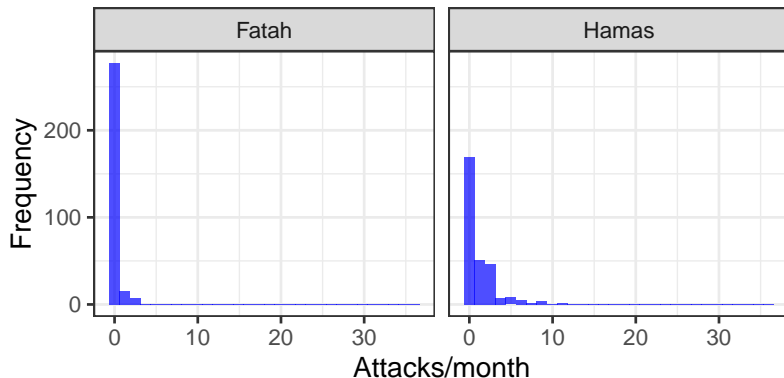
Equilibria of the game

What is the equilibrium probability that i attacks in state s ?



Takeaway 3: i is most likely to attack the most when improving i 's popularity decreases — i 's attack probability the most

Observed terrorism



Hamas Mean: 1.5 attacks/month. Median: 0. Range: 0-36.

Fatah Mean: < 1 attacks/month. Median: 0. Range: 0-15.

JMCC Example

JMCC: “Which Political or religious faction do you trust the most”

- Open ended question. Sample sizes around 1200.
- We record percent answering Fatah, percent answering Hamas.

Q31- Which Political or religious faction do you trust the most?

	Total	West Bank	Gaza
	n=1199	n= 749	n=450
Fatah	35.4	34.8	36.2
Hamas	20.0	18.6	22.4
PFLP	3.2	3.1	3.3
Other Islamic factions	3.1	1.9	5.1
Others	3.5	4.0	2.7
I don't trust anyone	29.2	31.1	26.0
No answer	5.6	6.5	4.3

*This was an open-ended question no options were read to the interviewee

- August 2015 poll. 52.5% from West Bank, 10.0% from Jerusalem, 37.5% from the Gaza Strip. 50.5% were male. Average age 37.

PSR Example

PSR: “Which of the following political parties do you support?”

- Sample sizes around 1200. We record percent answering Fatah, percent answering Hamas.

Which of the following political parties do you support?

1) PPP	1.4	1.6	0.9
2) PFLP	3.1	3.2	3.0
3) Fateh	32.3	31.7	33.3
4) Hamas	29.3	25.4	36.0
5) DFLP	1.0	1.4	0.2
6) Islamic Jihad	1.6	1.8	1.1
7) Fida	0.2		0.5
8) National Initiative (Mubadara)	1.0	1.2	0.7
9) Independent Islamists	3.3	3.9	2.3
10) Independent Nationalists	4.3	4.2	4.6
11) None of the above	22.4	25.3	17.4
12) Other, specify	0.2	0.4	0.0

- December 2006 poll. 1270 adults: 830 in the West Bank and 440 in the Gaza Strip. Columns are total, West Bank, Gaza Strip.

Correlation of public opinion indicators

	Trust in Hamas	Trust in Fatah	Support for Hamas	Support for Fatah	Vote for Hamas	Vote for Fatah
Trust in Hamas	1.00	−0.19	0.77	−0.44	0.98	−0.72
Trust in Fatah	−0.19	1.00	−0.26	0.69	−0.54	0.92
Support for Hamas	0.77	−0.26	1.00	−0.57	0.89	−0.83
Support for Fatah	−0.44	0.69	−0.57	1.00	0.56	−0.32
Vote for Hamas	0.98	−0.54	0.89	0.56	1.00	−0.73
Vote for Fatah	−0.72	0.92	−0.83	−0.32	−0.73	1.00

- each variable correlates as expected with at least three of the five other indicators suggesting that they can be collapsed onto one dimension.

Measurement Model: Setup

Recall our 6 public opinion measures:

$$y^t = \begin{bmatrix} \% \text{ of Population that trusts Fatah} \\ \% \text{ of Population that trusts Hamas} \\ \% \text{ of Population that supports Fatah} \\ \% \text{ of Population that supports Hamas} \\ \% \text{ of Population that plans to vote for Fatah} \\ \% \text{ of Population that plans to vote for Hamas} \end{bmatrix}^t,$$

Define z^t as $z_j^t = \frac{y_j^t - \bar{y}_j}{\sqrt{\text{Var}(y_j)}}$, $j = 1, \dots, 6$.

Our **dynamic factor analysis** is comprised of :

$$z^t = \tilde{s}^t \cdot \omega + \xi^t$$

$$\tilde{s}^t = \rho \tilde{s}^{t-1} + \alpha_0 + a_H^{t-1} \cdot \alpha_H + a_F^{t-1} \cdot \alpha_F + \eta^t$$

$$\xi^t \sim N(0, \mathbf{1})$$

$$\eta^t \sim N(0, 1)$$

Measurement Model: Results

Parameters (ω, α) estimated using MLE and EM using the MARSS package for R

Equation	Variable	Estimate
Factor Weights (ω)	Trusts Hamas	-0.09
	Trust Fatah	0.06
	Supports Hamas	-0.11
	Supports Fatah	0.10
	Votes Hamas	-0.07
	Votes Fatah	0.07
AR(1) term (ρ)	Lagged DV	0.99
Additional inputs (α)	Constant	-0.01
	Hamas attack	-0.28
	Fatah attack	1.05

Measurement Model: Robustness

- 1 Main specification (described above)
- 2 Fixed AR(1) term for \tilde{s}^t .
- 3 Estimated and homoskedastic variance in ξ^t .
- 4 Estimated and heteroskedastic variance in ξ^t .
- 5 A model with no α s (setting $\alpha = 0$)
- 6 Remove the “plans to vote for” surveys

Measurement Model: Robustness

We correlate \tilde{S}^t across models:

	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6
Model 1	1.00	1.00	0.99	0.87	0.99	0.99
Model 2	1.00	1.00	1.00	0.88	1.00	0.99
Model 3	0.99	1.00	1.00	0.88	1.00	0.98
Model 4	0.87	0.88	0.88	1.00	0.89	0.91
Model 5	0.99	1.00	1.00	0.89	1.00	0.99
Model 6	0.99	0.99	0.98	0.91	0.99	1.00

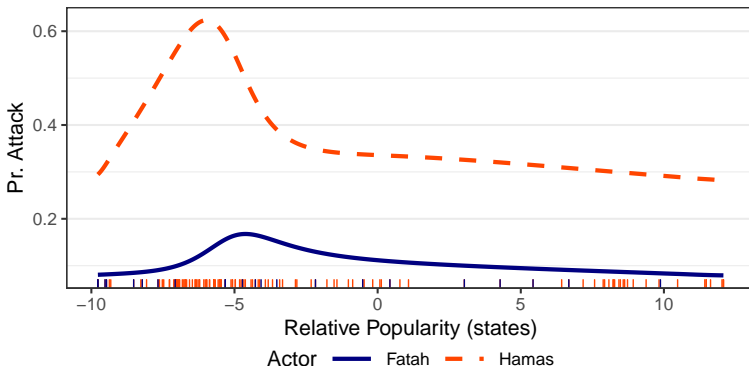
[► Appendix](#)[► back to talk](#)[► Additional results](#)

Equilibrium choice probabilities

What is the equilibrium probability that i attacks in state s ?

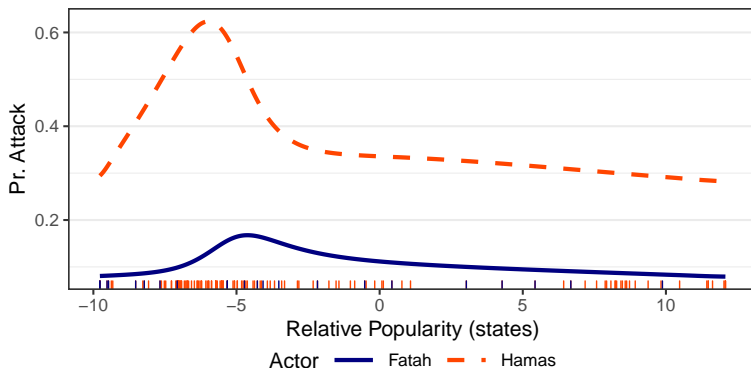
Equilibrium choice probabilities

What is the equilibrium probability that i attacks in state s ?



Equilibrium choice probabilities

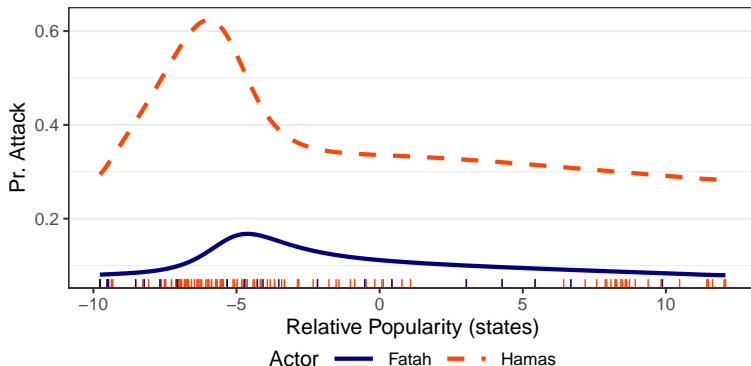
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Takeaway 1: Attack probabilities are positively correlated across states

Equilibrium choice probabilities

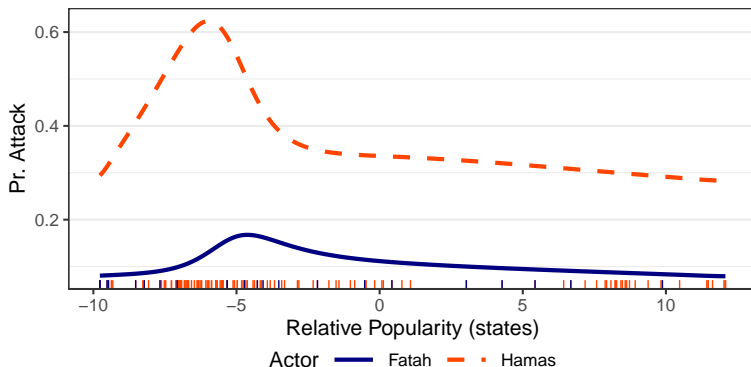
What is the equilibrium probability that i attacks in state s ?



Takeaway 2: i is most likely to attack the most when improving i 's popularity decreases — i 's attack probability the most

Equilibrium choice probabilities

What is the equilibrium probability that i attacks in state s ?



Takeaway 3: There is a set of attacks that are poorly explained by the model. Even with this, we find evidence of competition

Model comparison

Two model fit exercises to compare the fitted model to a “no competition” model where neither side can use violence to move public opinion

[▶ Appendix](#)[▶ back to talk](#)

Model comparison

Two model fit exercises to compare the fitted model to a “no competition” model where neither side can use violence to move public opinion

- Expected number of attacks correctly predicted

$$\text{ePCP}_i = \frac{1}{T} \sum_{t=1}^T P(a_i^t; s^t, \hat{v}_i),$$

- Likelihood ratio test to compare the models

Model comparison

Two model fit exercises to compare the fitted model to a “no competition” model where neither side can use violence to move public opinion

	Main model	No Competition
ePCP–Hamas	0.54	0.51
ePCP–Fatah	0.83	0.86
ePCP–Overall	0.46	0.44
LR test (d.f. = 6)	$\chi^2 = 275$ ($p < 0.01$)	

Robustness: First stage

<i>Controlling for</i>	Fatah	Hamas
Base Model	(0.94, 1.14)	(-0.23, -0.19)
No interaction	(1.00, 1.00)	(-0.21, -0.21)
Unemployment	(0.94, 1.21)	(-0.21, -0.18)
Support for violence	(0.96, 1.14)	(-0.24, -0.22)
2nd Intifada	(0.94, 1.16)	(-0.21, -0.23)
Time since last election	(0.95, 1.16)	(-0.21, -0.19)
Palestinians killed	(0.87, 1.44)	(-0.12, -0.60)

Values are the estimated marginal effects of a terrorist attack on relative popularity at the extremes $\tilde{s}_{t-1} = \{-10, 12\}$. All estimates are from an ECM model.

Robustness: Time frame

	Full Sample	2014 Agreement	2011 Agreement	2006 Elections	Second Intifada [†]
β_H	-0.007 (0.004)	-0.007 (0.005)	-0.007 (0.004)	-0.021 (0.013)	-0.007 (0.004)
β_F	0.0005 (0.0003)	0.0006 (0.0003)	0.0005 (0.0003)	0.0005 (0.0002)	0.0005 (0.0152)
κ_H	-0.95 (0.23)	-0.90 (0.23)	-0.69 (0.22)	-0.74 (0.26)	-0.89 (0.10)
κ_F	-2.45 (0.28)	-2.47 (0.31)	-2.31 (0.28)	-2.34 (0.26)	-2.89 (0.23)
T	300	243	208	144	80
LL	-278.09	-230.98	-206.84	-125.27	-71.82

Note: [†]CMLE did not converge, estimates from nested-pseudo-likelihood (NPL) estimator. Samples begin at Jan. 1994 and end in one month prior to the event listed.

Robustness: Discount factor

δ	Log-Likelihood
0	-284.18
0.9	-280.98
0.925	-281.54
0.95	-282.43
0.975	-283.67
0.99	-280.35
0.999	-278.09
0.9999*	-288.14

Note: * Model failed to converge.

Homotopy Algorithm

Algorithm 1: COMPARATIVE STATICS (CS) using a homotopy

Input: An initial coefficient vector θ_0 and equilibrium v_0 ; new values $\tilde{\theta}$; and a tuning parameter $n \in \mathbb{N}$. For
 NEWTON solver, a convergence tolerance $\varepsilon > 0$ and maximum number iterations $m \in \mathbb{N}$.

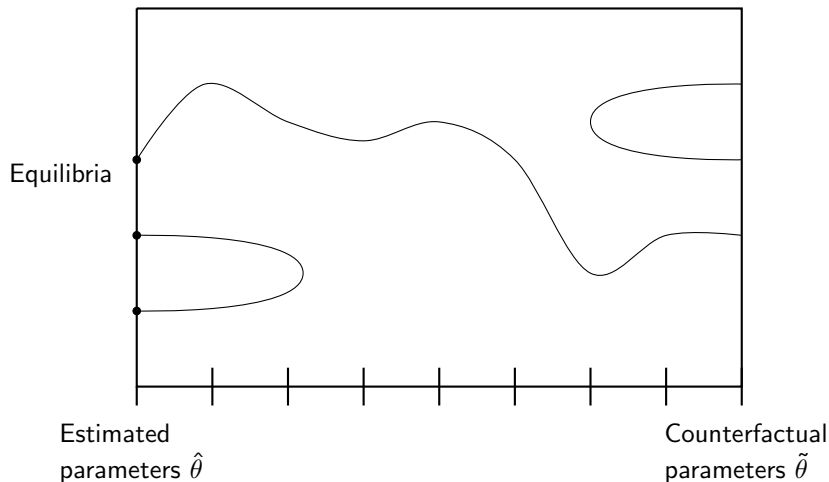
Output: An equilibrium \tilde{v} under new parameters $\tilde{\theta}$

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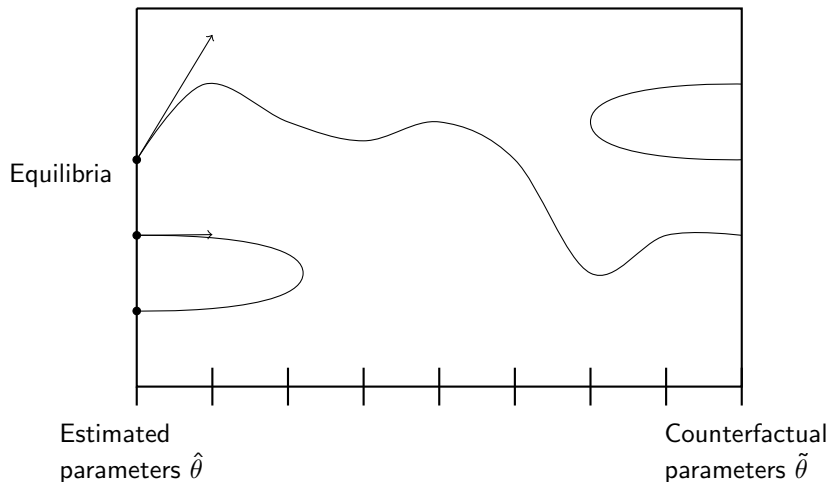
1  $\theta_{old} \leftarrow \theta_0$ 
2  $f_{old} \leftarrow f(\theta_0)$ 
3  $v_{old} \leftarrow v_0$ 
4 for  $i \leftarrow 1$  to  $n$  do
5    $\lambda \leftarrow \frac{i}{n}$ 
6    $\theta_{new} \leftarrow (1 - \lambda)\theta_0 + \lambda\tilde{\theta}$ 
7    $f_{new} \leftarrow f((1 - \lambda)\theta_0) + f(\lambda\tilde{\theta})$ 
8    $slope \leftarrow -(D_\theta \mathcal{V}(v_{old} \mid \theta_{old}, f_{old}))' (D_v \mathcal{V}(v_{old} \mid \theta_{old}, f_{old}))^{-1}$ 
9    $start \leftarrow v_{old} + [\theta_{new} - \theta_{old}] slope$ 
10   $(v_{new}, success) \leftarrow \text{NEWTON}(start, \mathcal{V}(v \mid \theta_{new}, f_{new}) - v, \varepsilon, m)$ 
11  if success then
12     $\theta_{old} \leftarrow \theta_{new}$ 
13     $f_{old} \leftarrow f(\theta_{new})$ 
14     $v_{old} \leftarrow v_{new}$ 
15  else
16     $\tilde{v} \leftarrow \text{"Warning: Convergence Problems."}$ 
17    break
18  $\tilde{v} \leftarrow v_{new}$ 
19 return  $\tilde{v}$ 

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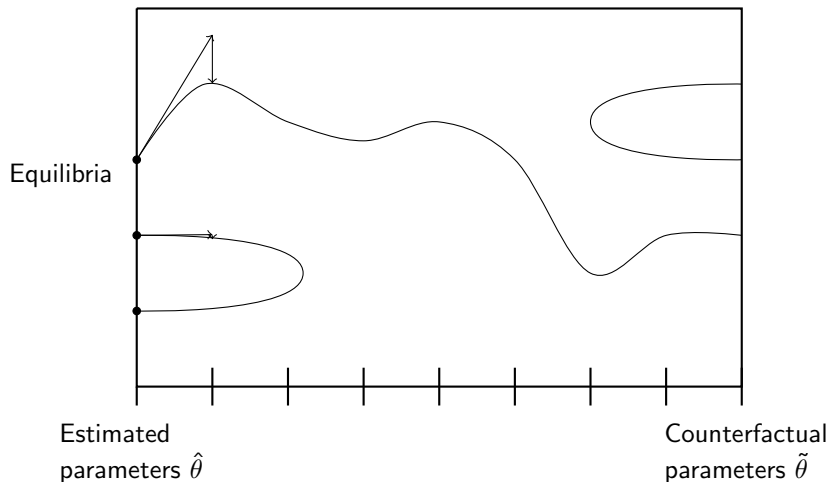
Homotopy method for comparative statics



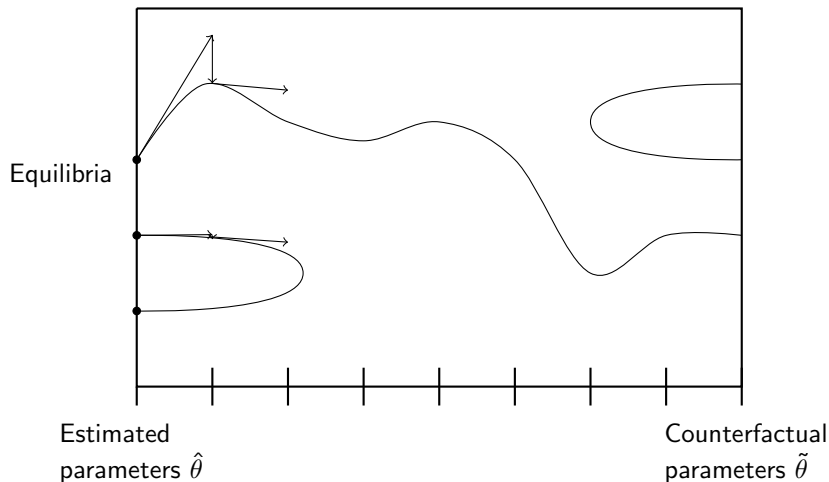
Homotopy method for comparative statics



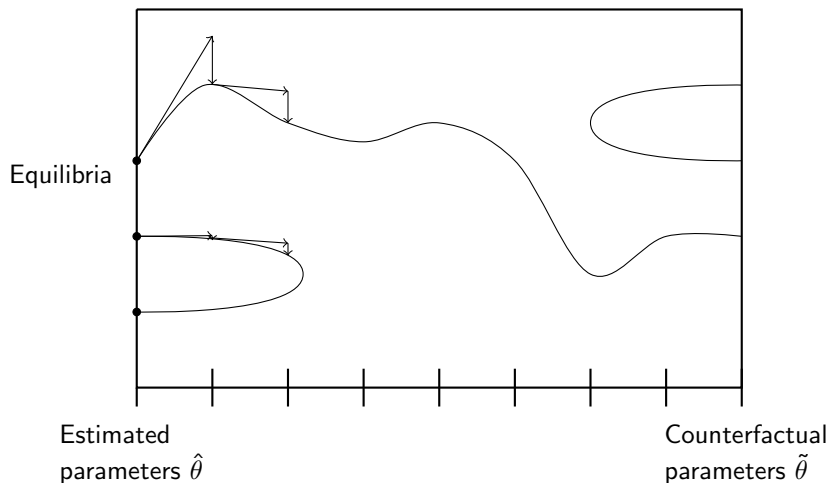
Homotopy method for comparative statics



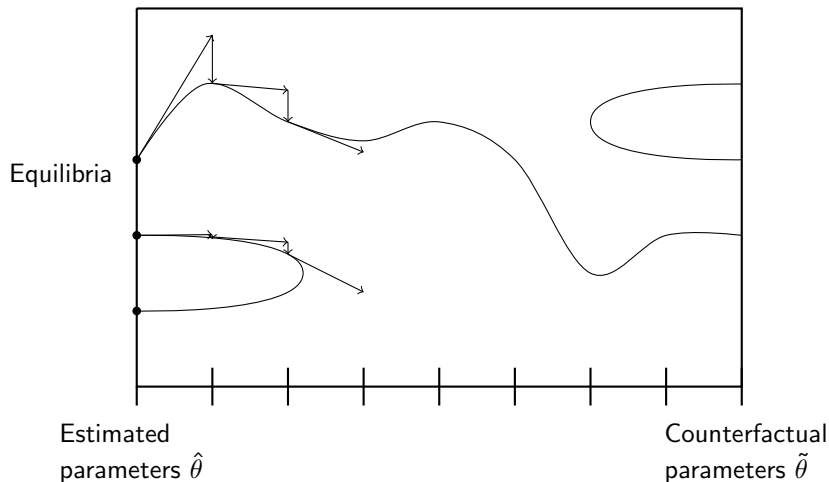
Homotopy method for comparative statics



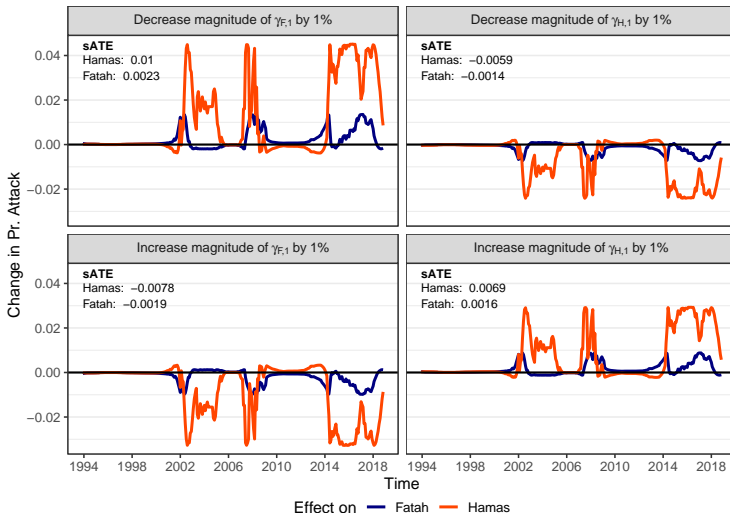
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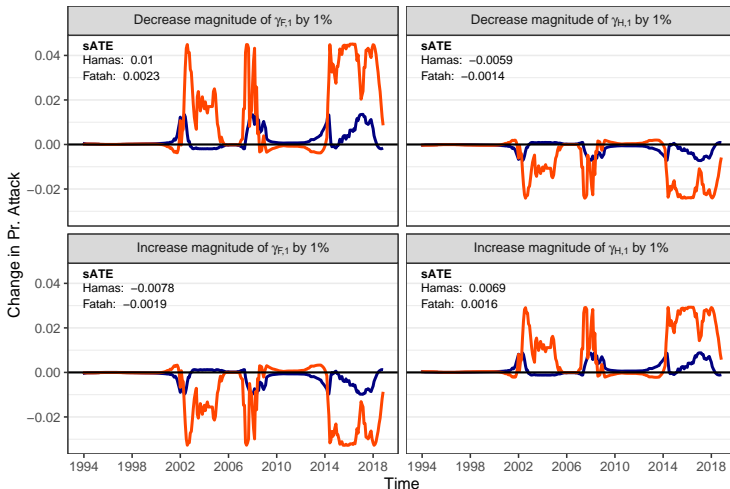
Homotopy method for comparative statics



Changing effectiveness



Changing effectiveness



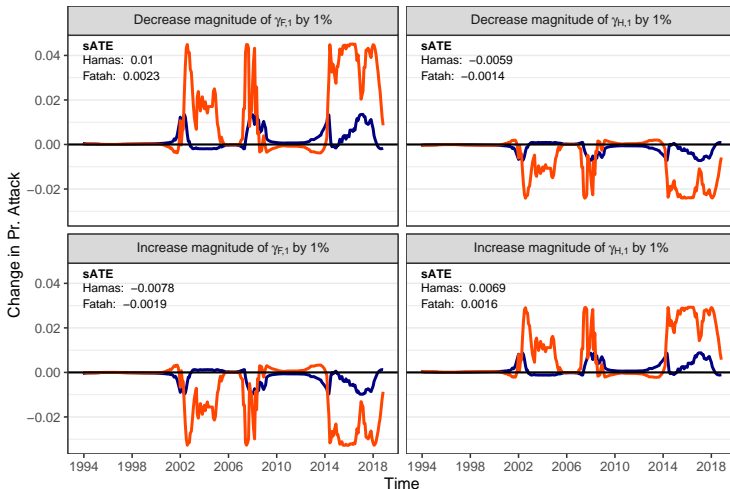
Takeaway 1: $\uparrow |\gamma_{H,1}|$ implies \uparrow violence (expected)

► Appendix

► Additional results

► Return to talk

Changing effectiveness



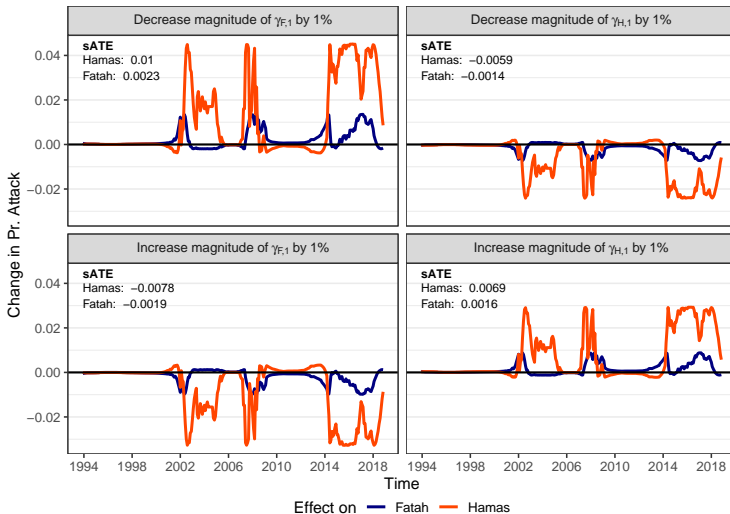
Takeaway 2: $\uparrow |\gamma_{F,1}|$ implies \downarrow violence (not expected)

► Appendix

► Additional results

► Return to talk

Changing effectiveness



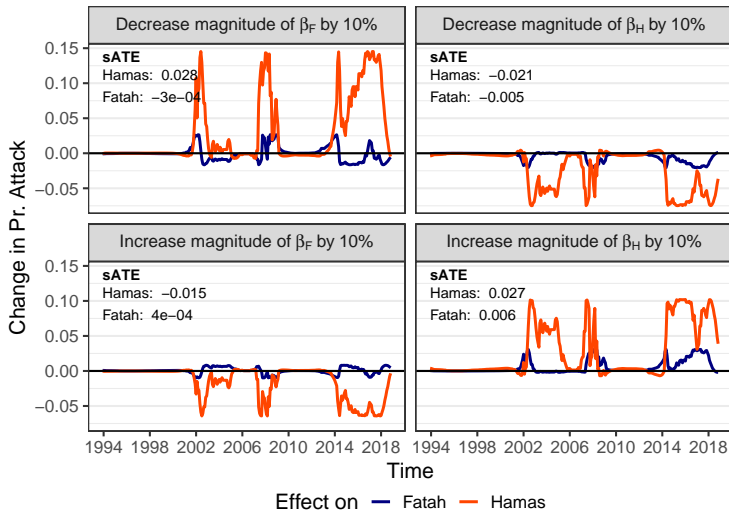
Takeaway 3: Same conclusions when changing values or costs

► Appendix

► Additional results

► Return to talk

Changing values



Changing attack costs

