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# Unequal Before the Law: Political Incentives and Selective Drug Enforcement in Colombia

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What explains variation in drug enforcement?



What explains variation in drug enforcement?

“For my friends, everything; for my enemies, the law.”

— Óscar R. Benavides

# A Tale of Two Coca-Growing Municipalities



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Similar geography  
Neighboring municipalities

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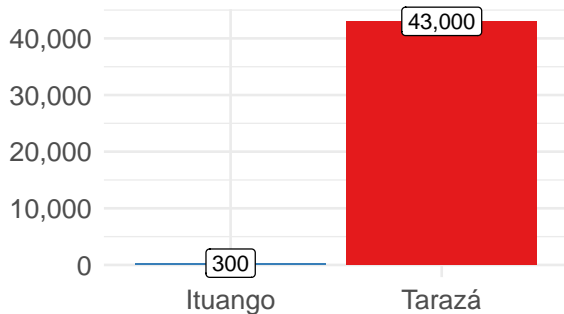
$\approx 6.5\times$  more cultivation in Tarazá



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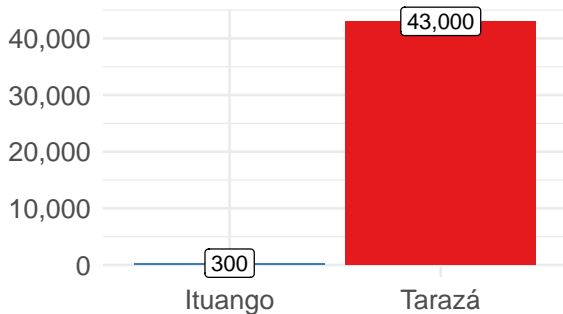
Aerial fumigation (hectares)



143× **more** eradication in Tarazá

# A Tale of Two Coca-Growing Municipalities

Aerial fumigation (hectares)



143× **more** eradication in Tarazá

- 1/4 of Tarazá's land area
- > 1/2 of NYC area,  $\approx$  1/5 of OC area
- 70× UC Irvine's campus area

# Ituango and Tarazá: Friends and Foes

Friendly criminal groups in Ituango, **enemy** criminal groups in Tarazá



# Preview of Results

**Intensification** of enforcement to criminal **foes**  
**Restraint** in enforcement to criminal **friends**

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Urbanicity

Ideology

Candidates

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# Explaining Variation in Drug Enforcement

Conventional wisdom: constraints

Enforcement varies due to

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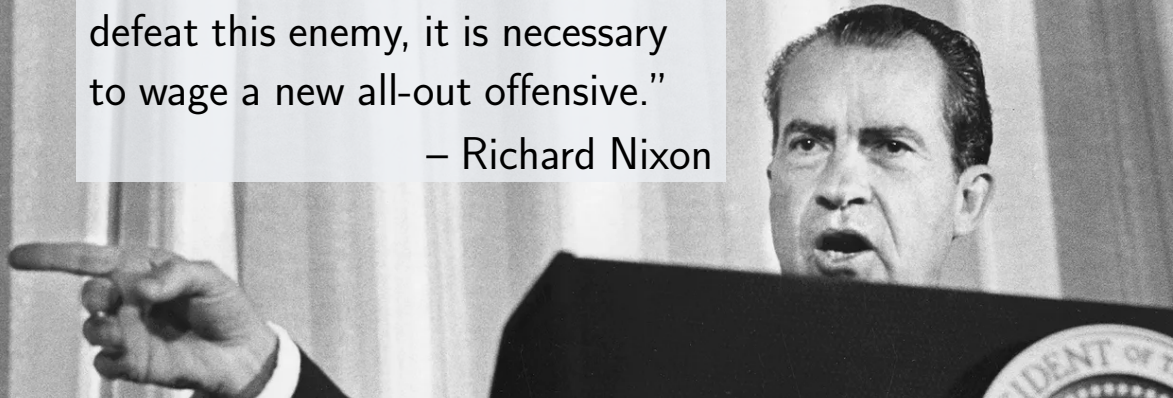
Argument: agency – “able and *sometimes* willing”

## Political incentives of elected leaders shape enforcement

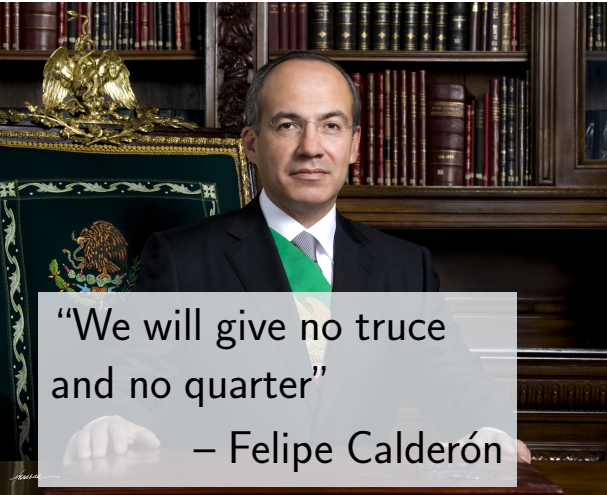
Building on: Bobo and Thompson (2006); Durán-Martínez (2018); Fagan et al. (2010); Felbab-Brown (2009); Freedman, Owens and Christopher (2022); Holland (2017); Jones and Olken (2005); Linnemann and Kurtz (2014); Lynch et al. (2013); Mitchell (2005); Mitchell and Caudy (2015); Musto (1999); Provine (2008); Trejo and Ley (2020)

# Elected Leaders Shape Drug Enforcement

“Public enemy number one [...] is drug abuse. In order to fight and defeat this enemy, it is necessary to wage a new all-out offensive.”  
– Richard Nixon



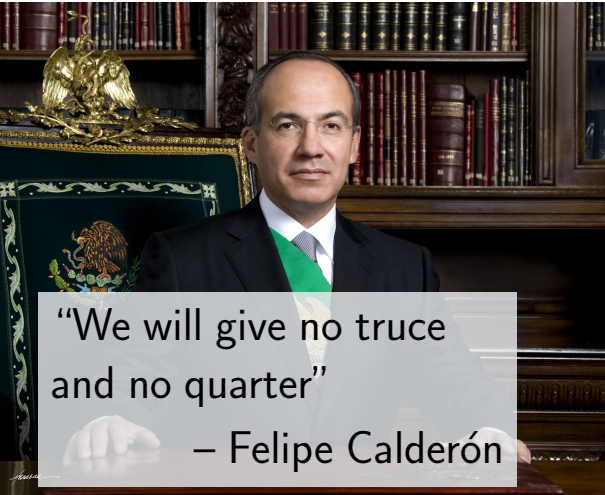
# Elected Leaders Shape Drug Enforcement



“We will give no truce  
and no quarter”  
– Felipe Calderón

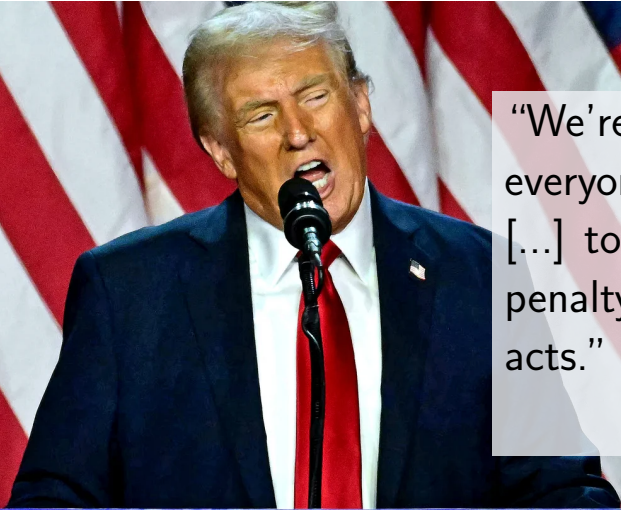
(Fondevila and Quintana-Navarrete, 2015; Lessing, 2017)

# Elected Leaders Shape Drug Enforcement



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# Elected Leaders Shape Drug Enforcement



“We’re going to be asking everyone who sells drugs [...] to receive the death penalty for their heinous acts.”

– Donald Trump

★★★★★  
**TRUMP WILL FIX IT**

# Elected Leaders Shape Drug Enforcement

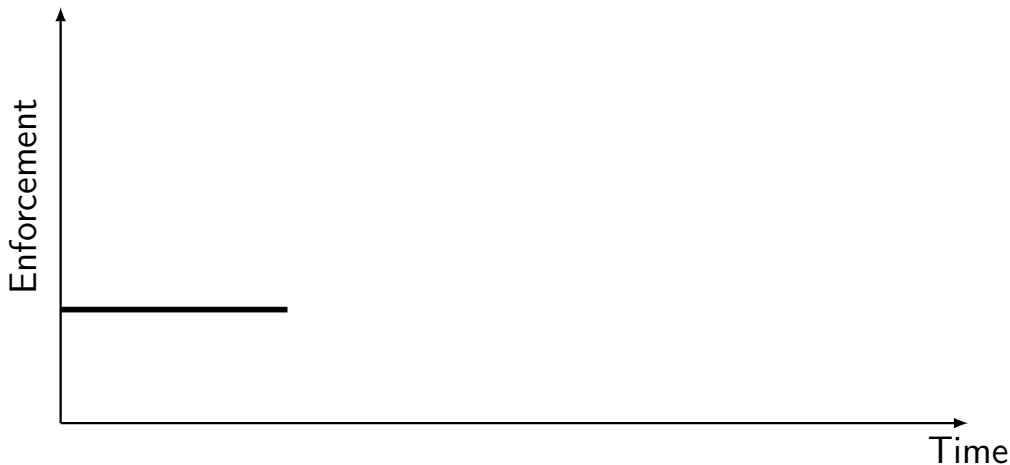


Alvaro Uribe,  
President of  
Colombia:  
2002-2010



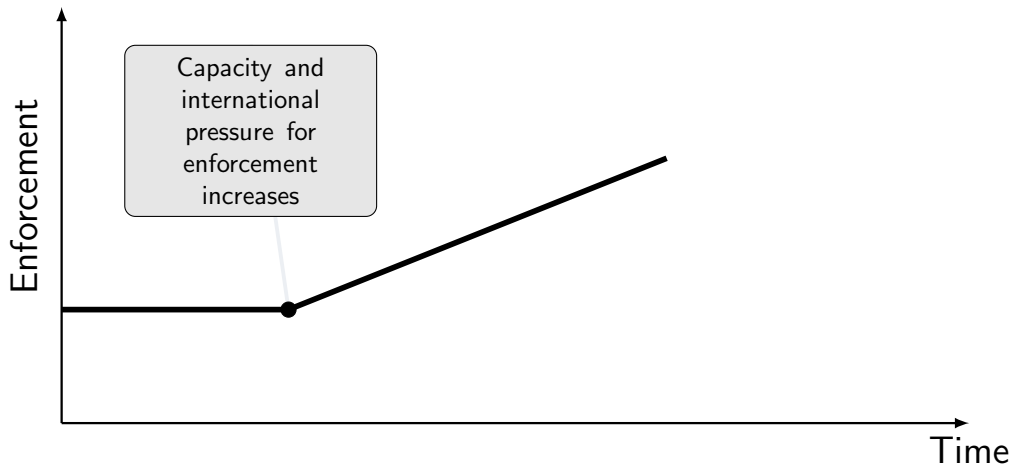
# Selective Enforcement

Influence of criminal groups creates differential incentives for crackdowns



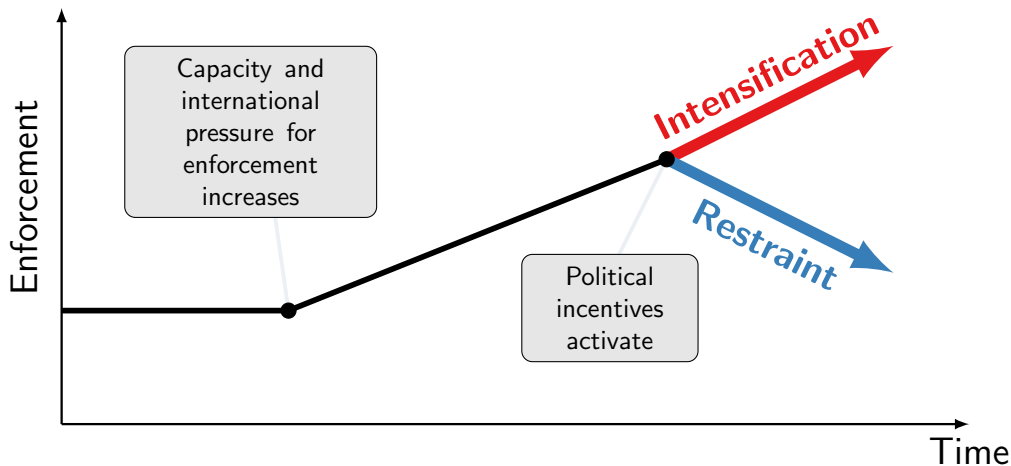
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# Road Map

- 1 Introduction
- 2 Theory: The Politics of Counternarcotics
- 3 Context, Research Design, and Data
- 4 Results: Political Incentives and Selective Enforcement
- 5 Mechanisms: Paramilitary Demobilization
- 6 Conclusion

A photograph of Bill Clinton and Vicente Fox. Bill Clinton is on the right, wearing a dark suit and a patterned tie, smiling and looking down at a book. Vicente Fox is on the left, wearing a dark suit and a blue tie, also smiling and looking at the book. An American flag is visible in the background on the left, and a chandelier is on the wall behind them.

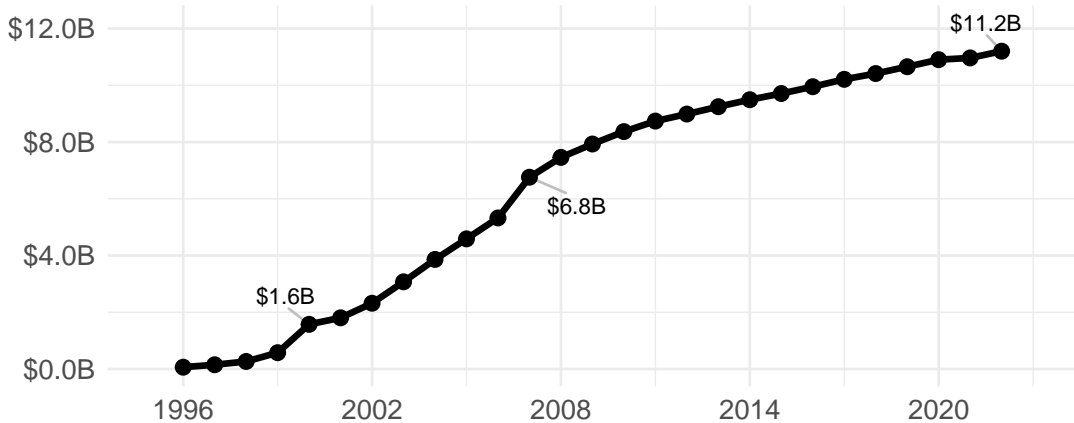
# The Politics of Counternarcotics

# Leaders Control, Benefit from Enforcement

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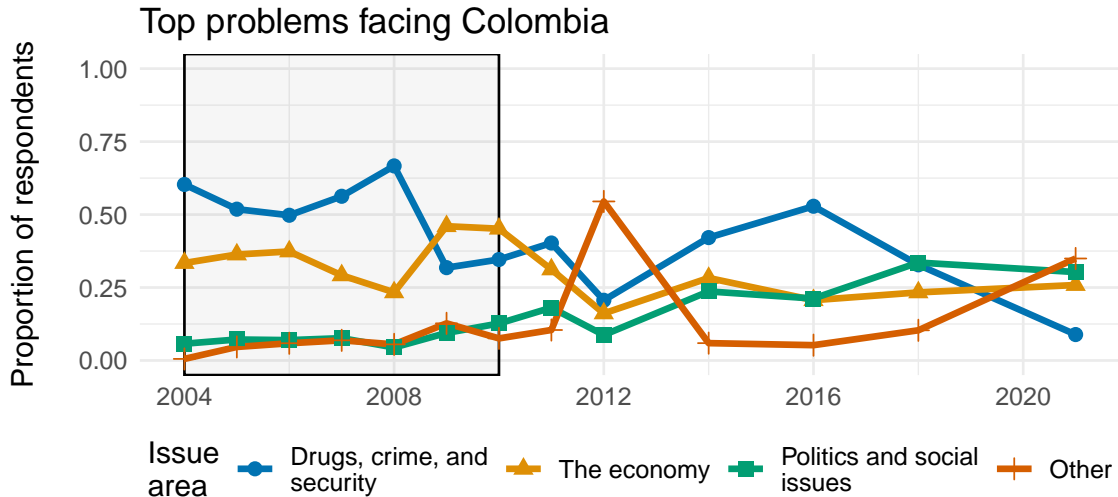
International politics, U.S. bilateral aid

Cumulative U.S. security aid to Colombia (1996–2022)



# Leaders Control, Benefit from Enforcement

Salient domestic issue in elections





# The Burden of Enforcement is Local

International considerations less relevant at the micro level



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Affected voters (*campesinos*) are politically marginalized

Like many Latin American countries, Colombia is very urban (80%)

$\frac{2}{3}$  turnout in largest cities,  $\frac{1}{3}$  in rural municipalities



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$\frac{2}{3}$  turnout in largest cities,  $\frac{1}{3}$  in rural municipalities

Consequences of counternarcotics strategies highly local

(Abadie et al., 2014; Camacho and Mejía, 2017; Calderón et al., 2015; Dell, 2015; Lessing, 2017; Phillips, 2015;

Ramírez, 2011; Rincón-Ruiz et al., 2016; Rozo, 2014)



# Leaders Respond to Political Incentives

International and domestic pressure for enforcement

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Criminal groups involved in illicit markets and exert political influence

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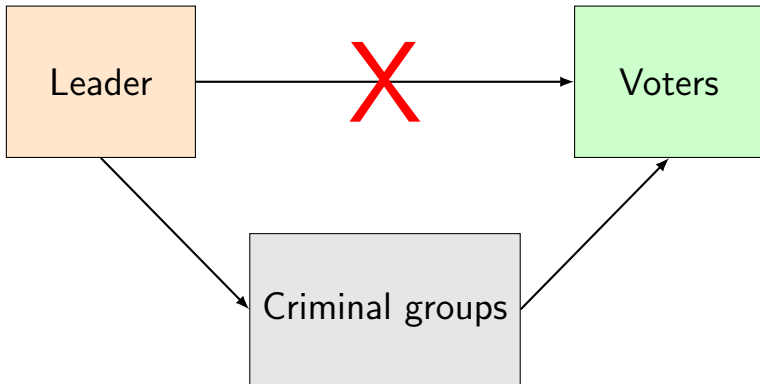
(Acemoglu, Robinson and Santos, 2013; Ch et al., 2018; Barnes, 2017; Dipoppa, 2021)

Selective enforcement depending on **criminal group** influence



# Electoral Oversight with Criminal Groups

**Criminal groups** mediate the distribution of enforcement

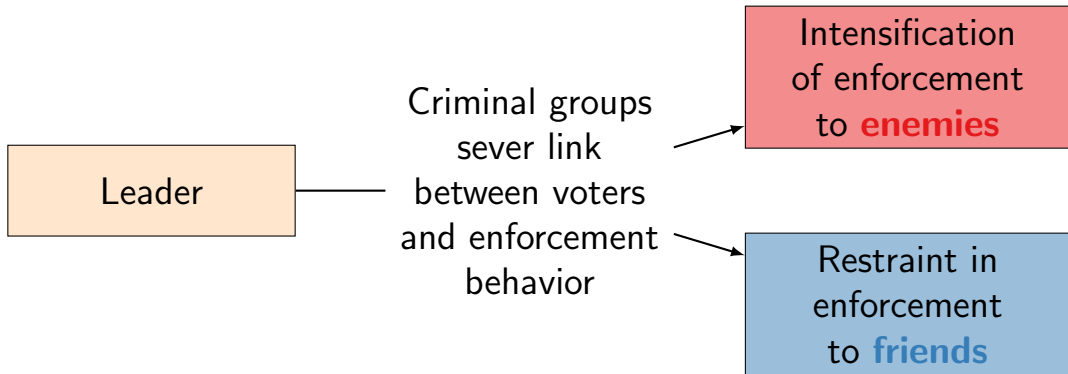


Leader acts on behalf of **criminal group** preferences for enforcement



# Leaders Target Foes, Leave Friends Alone

**Criminal group influence** creates differential incentives for crackdowns

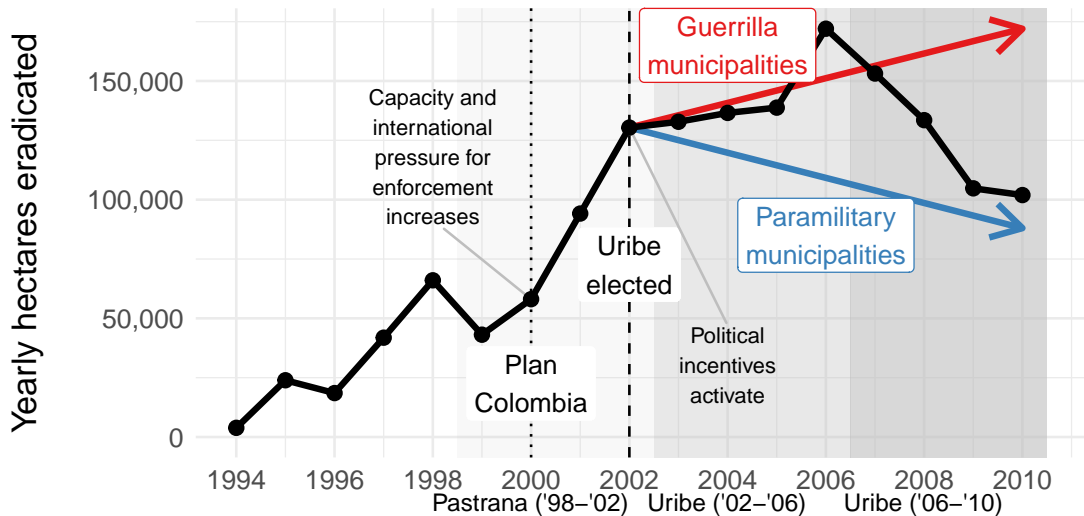




**PRESIDENTE: vamos por la  
tercera , en esa los acabamos.**

## Context, Research Design, and Data

# Aerial Crop Eradication in Colombia



# Criminal Groups and Illicit Markets

Characteristics of cocaine create opportunities for organized crime:

- Illegality

- High lootability

- Low obstructability



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Main group types:

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# Criminal Groups and Illicit Markets

Characteristics of cocaine create opportunities for organized crime:

- Illegality

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Main group types:

- guerrilla groups and

- paramilitary groups

Leverage coercive power for influence

(Ch et al., 2018; Fergusson et al., 2021; Hirschel-Burns, 2021)



# Criminal Groups and Politics

**Guerrilla groups** (e.g., FARC, ELN) opposed the state, Uribe  
1960s: Foundation of FARC, ELN as left-wing insurgent groups





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2000s-2016: Local dominance, criminal governance

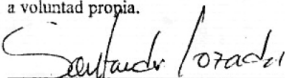


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**Paramilitary groups** (e.g., AUC) favored Uribe

2001: Secret pact between AUC and politicians, “refounding”

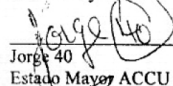
En este documento queda constancia de los asistentes a esta reunión, firman a voluntad propia.



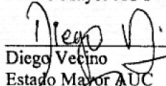
Santander Losada  
Estado Mayor AUC



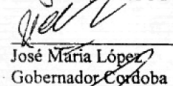
Adolfo Paz  
Estado Mayor AUC



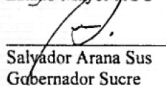
Jorge 40  
Estado Mayor ACCU



Diego Vecino  
Estado Mayor AUC



José María López  
Gobernador Córdoba



Salvador Arana Sus  
Gobernador Sucre

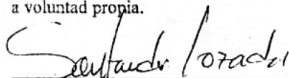
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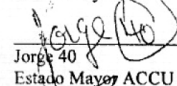
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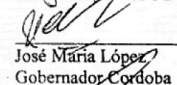
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Santander Losada  
Estado Mayor AUC



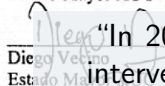
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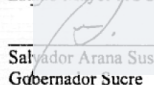
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“In 2002, the AUC directly intervened in the elections.”

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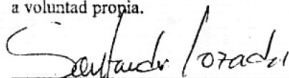
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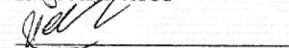
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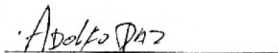
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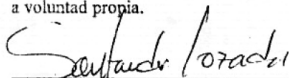
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2006-present: *Parapolítica* scandal, dozens investigated, convicted

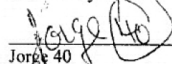
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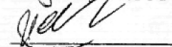
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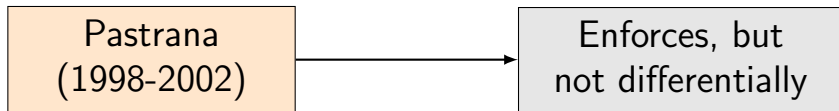
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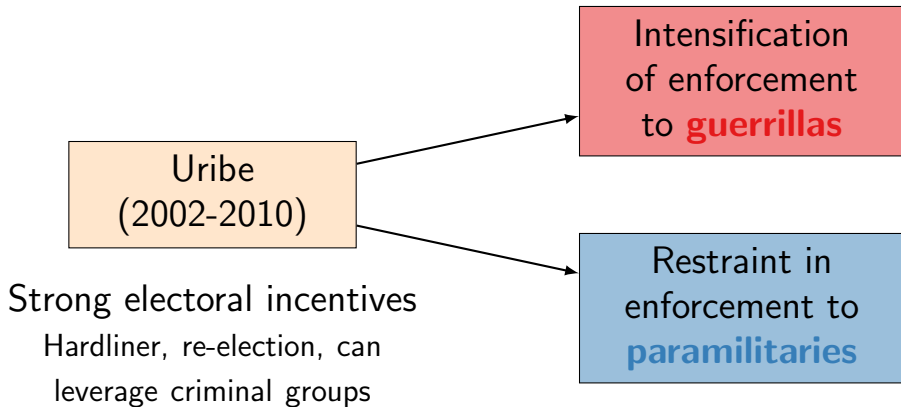


# Differential Incentives for Eradication

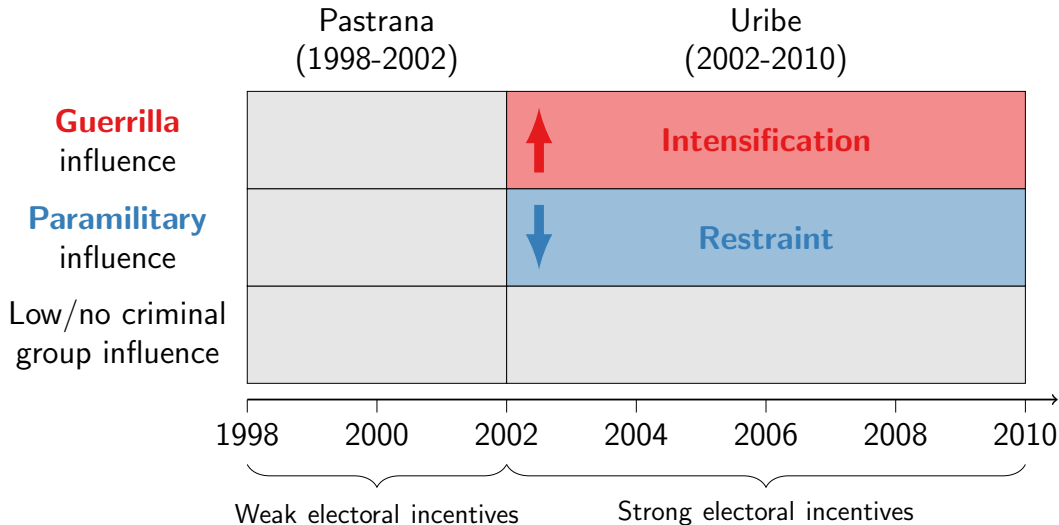


Weak electoral incentives  
Softer, no re-election incentives,  
weaker ties with criminal groups

# Differential Incentives for Eradication

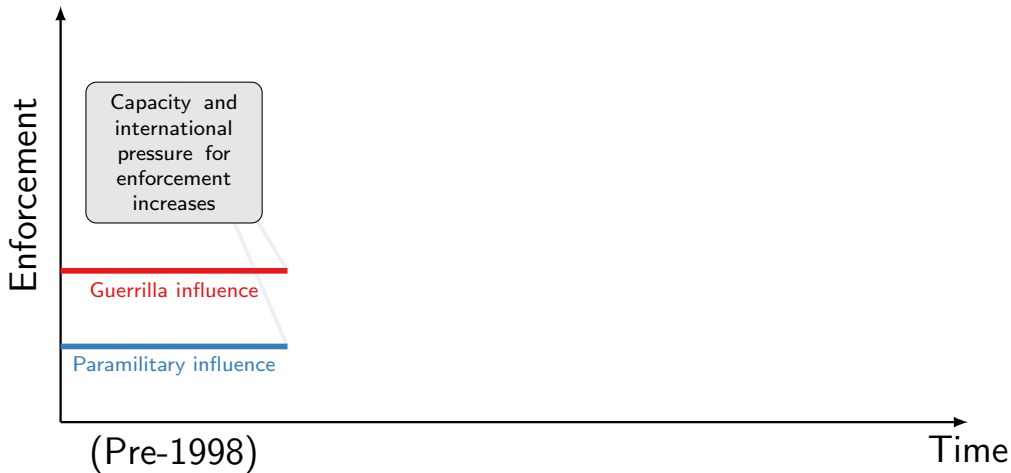


# Difference-in-Differences Design

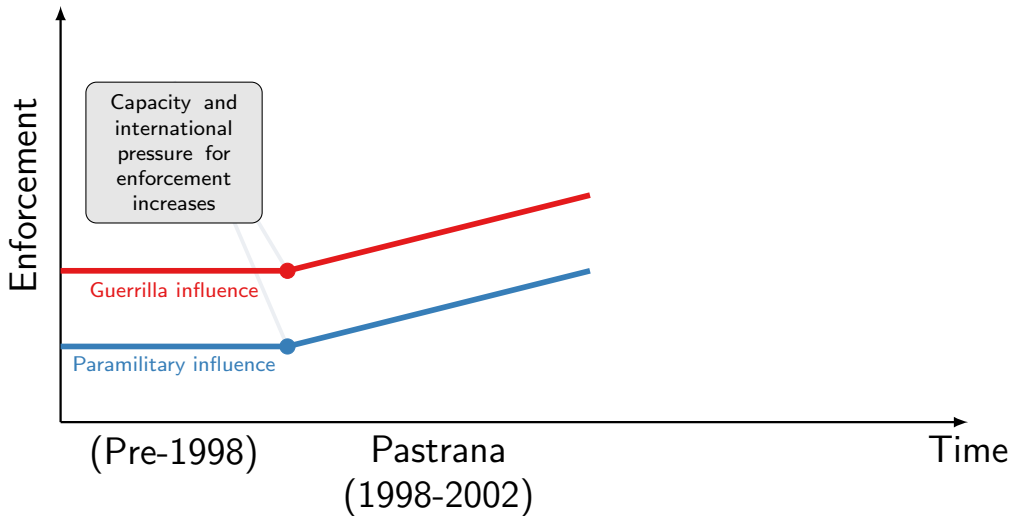




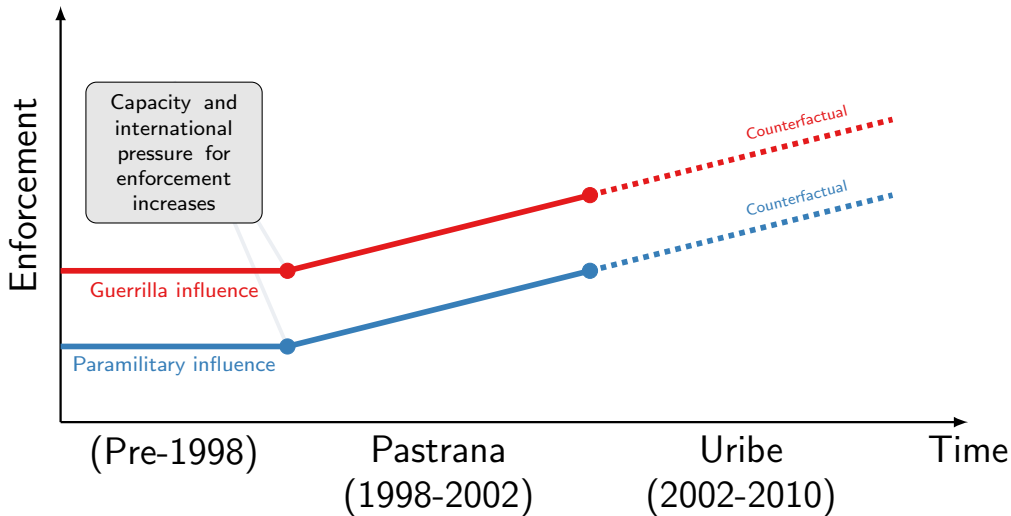
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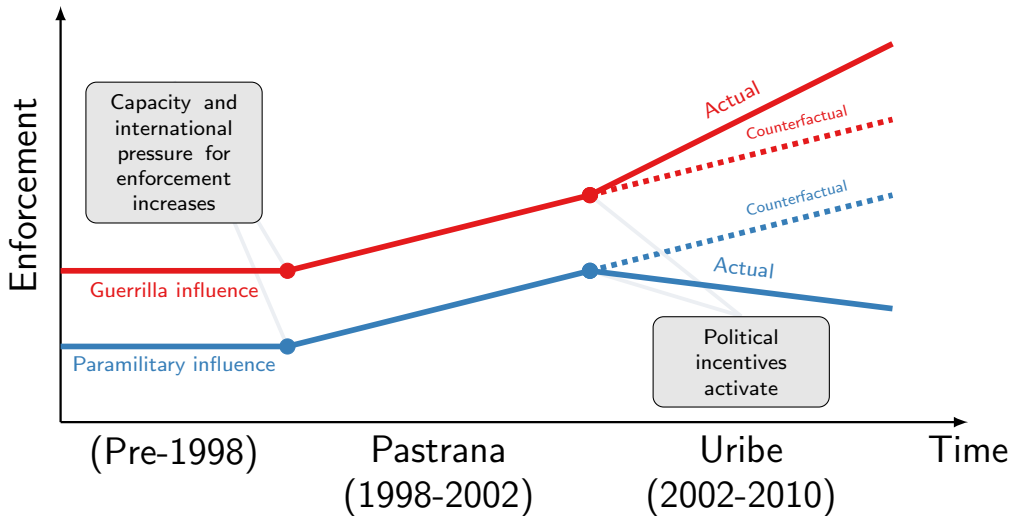
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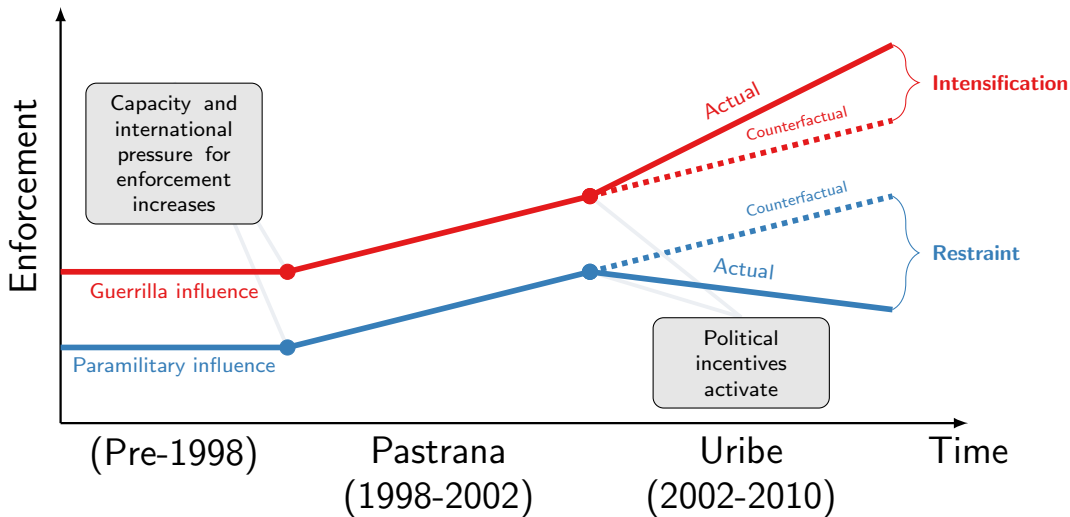
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# Data: Monthly Municipal Panel (1998-2010)

Crop eradication data sourced via information request

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Criminal influence proxied by historical violence Violence Map

(Acemoglu, Robinson and Santos, 2013; Aponte González, Hirschel-Burns and Uribe, 2024; Ch et al., 2018)

Data from CINEP (Colombian NGO)

Time-invariant: creates geographic variation

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Data from CINEP (Colombian NGO)

Time-invariant: creates geographic variation

Estimation sample: 318 of 1,122 municipalities [Sample Map](#)



# Testing Differential Enforcement

$$\underline{\textit{Eradication}}_{i,t} = \beta_1 G_i \times \mathbb{1}[2002-2010] + \beta_2 P_i \times \mathbb{1}[2002-2010] + \mathbf{X}_{i,t}^\top \gamma + \delta_i + \zeta_t + \epsilon_{i,t}$$

*Eradication*<sub>*i,t*</sub>: eradication in municipality *i* in year-month *t*

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$\mathbb{1}[2002-2010]$ : indicator for Uribe's presidential terms

# Testing Differential Enforcement

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$\mathbb{1}[2002-2010]$ : indicator for Uribe's presidential terms

$\mathbf{X}_{i,t}^\top$ : controls;  $\delta_i$  and  $\zeta_t$ : municipality and year-month fixed effects

# Testing Differential Enforcement

$$\text{Eradication}_{i,t} = \beta_1 G_i \times \mathbb{1}[2002-2010] + \beta_2 P_i \times \mathbb{1}[2002-2010] + \mathbf{X}_{i,t}^\top \gamma + \delta_i + \zeta_t + \epsilon_{i,t}$$

$\text{Eradication}_{i,t}$ : eradication in municipality  $i$  in year-month  $t$

$G_i$ : historical **guerrilla** attacks in municipality  $i$

$P_i$ : historical **paramilitary** attacks in municipality  $i$

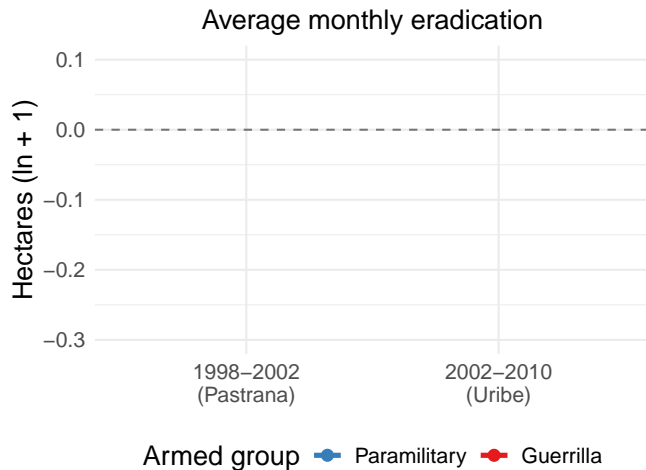
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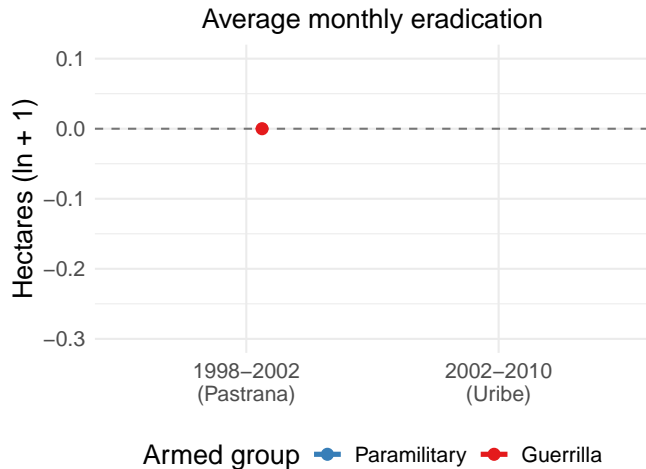
# Political Incentives and Selective Enforcement

# More for Guerrillas, Less for Paramilitaries

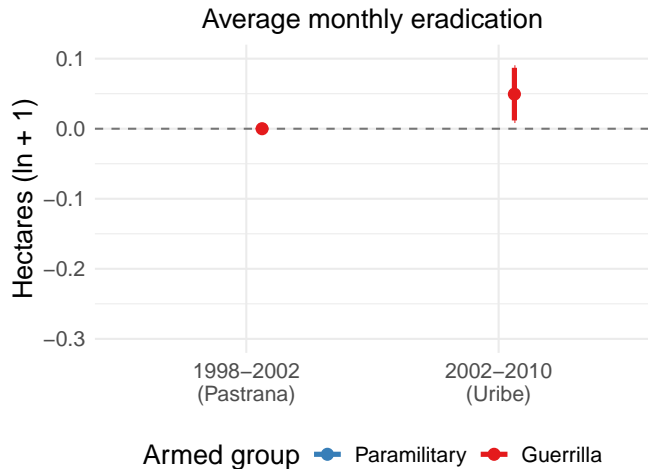




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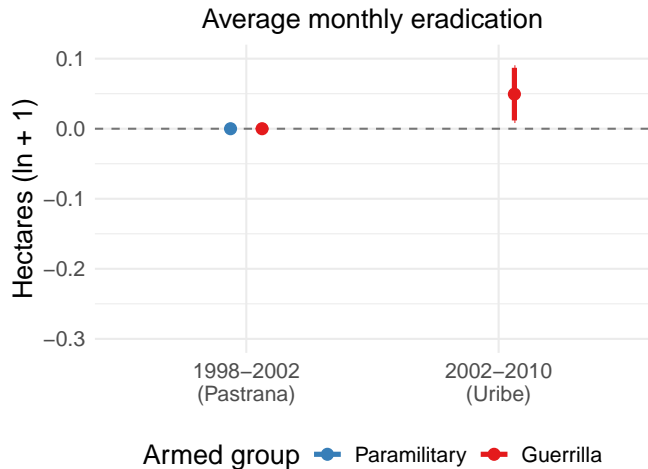


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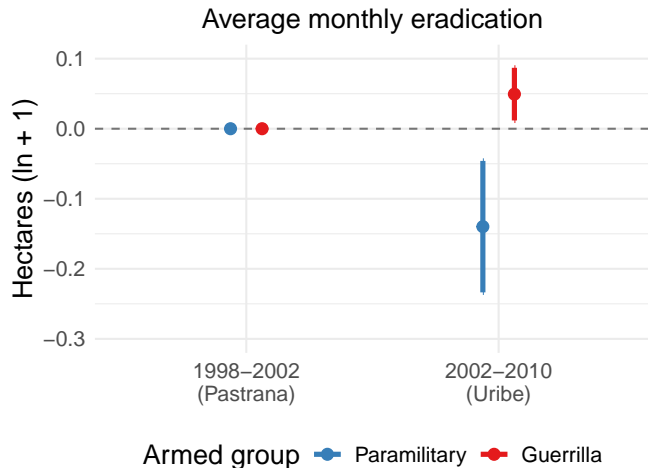
Guerrilla areas experienced an **increase** in fumigation (400 hectares over 4 years)

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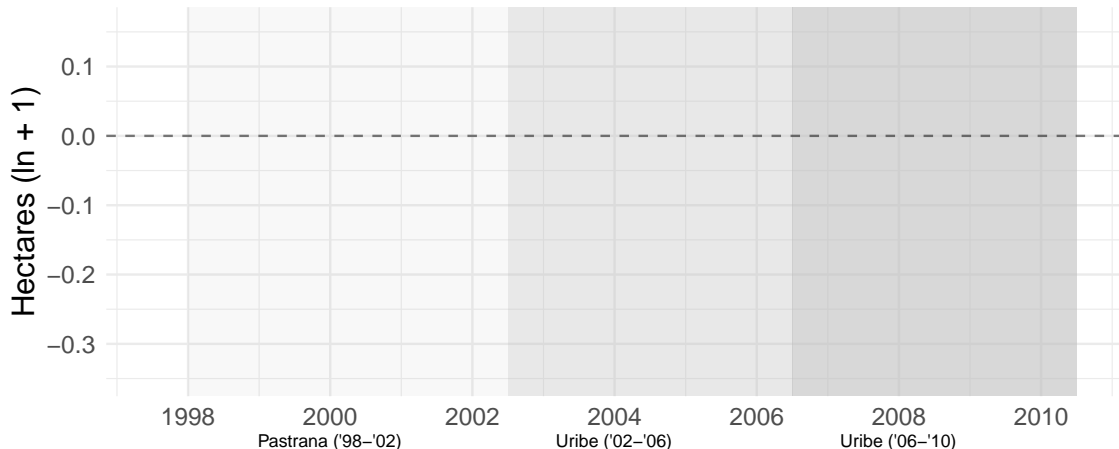
Guerrilla areas experienced an **increase** in fumigation  
(400 hectares over 4 years)

Paramilitary areas experienced a **decrease** in fumigation  
(500 hectares over 4 years)

# Year-by-Year Results

Estimating Equation

Average monthly eradication

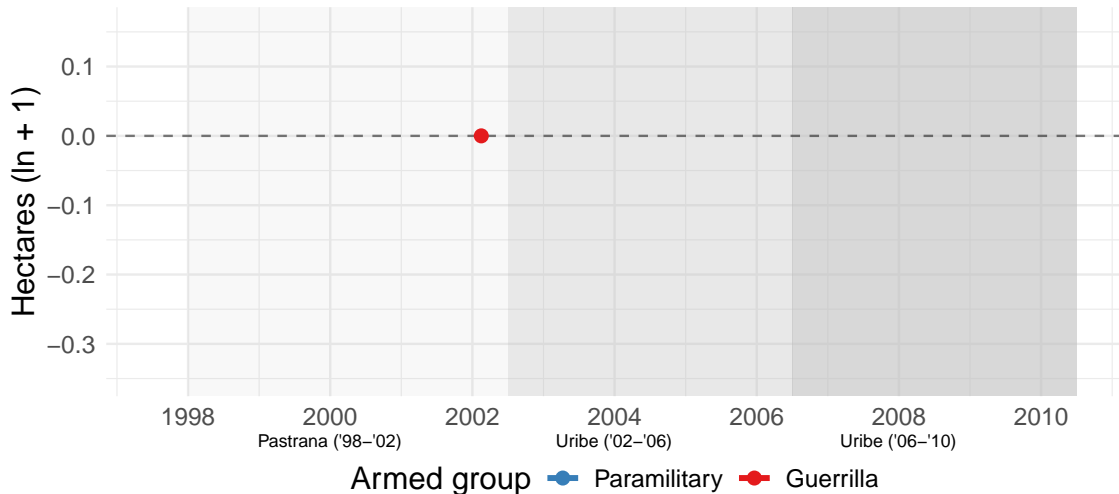


Armed group — Paramilitary — Guerrilla

# Year-by-Year Results

Estimating Equation

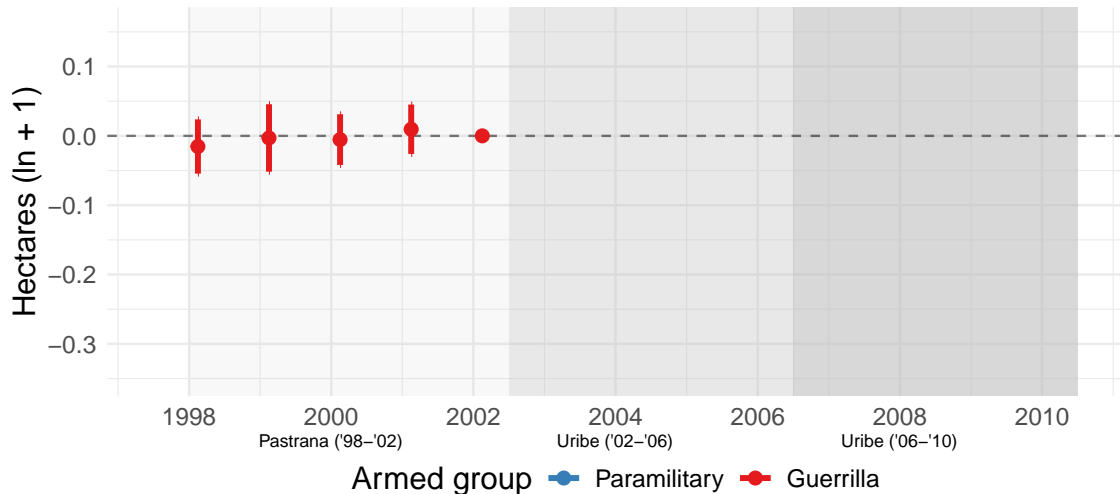
Average monthly eradication



# Year-by-Year Results

Estimating Equation

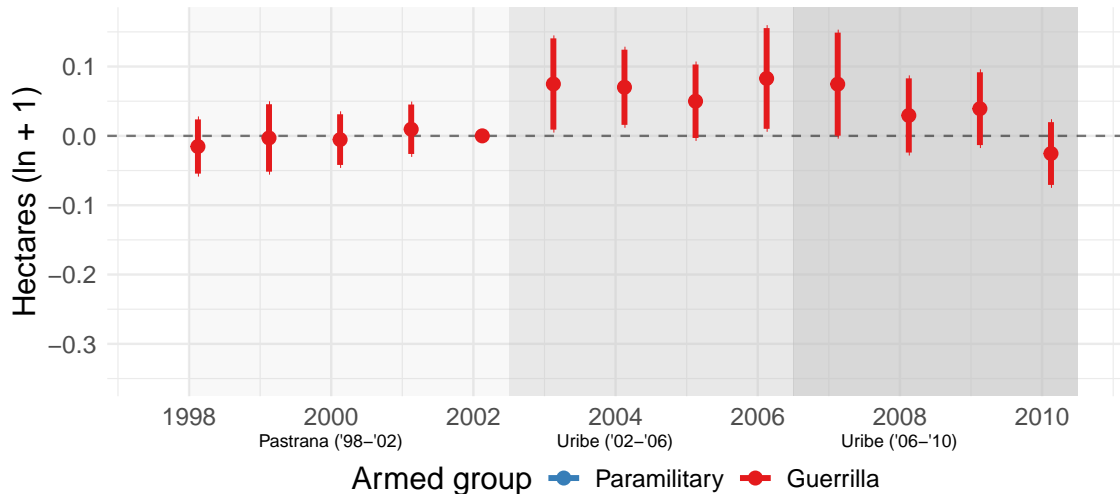
Average monthly eradication



# Year-by-Year Results

Estimating Equation

Average monthly eradication

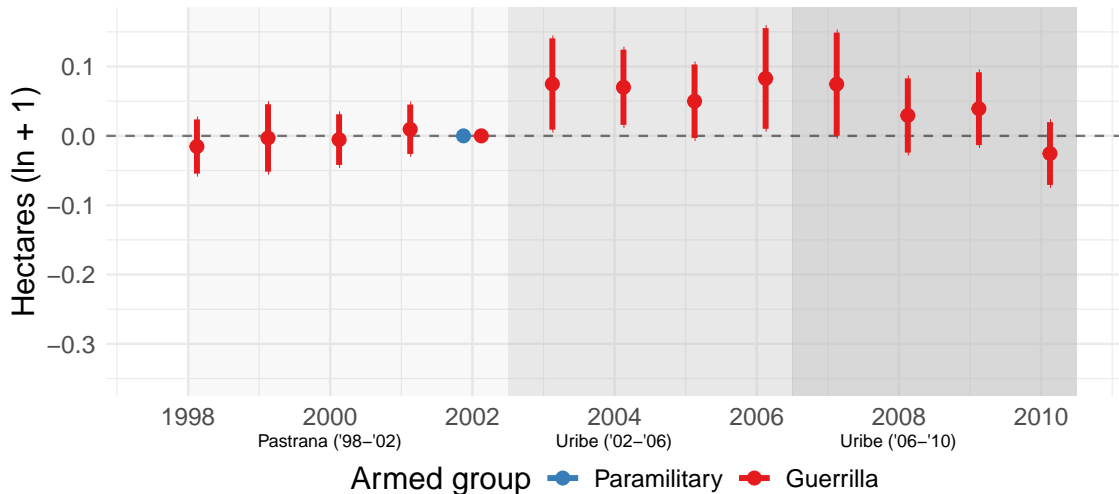




# Year-by-Year Results

Estimating Equation

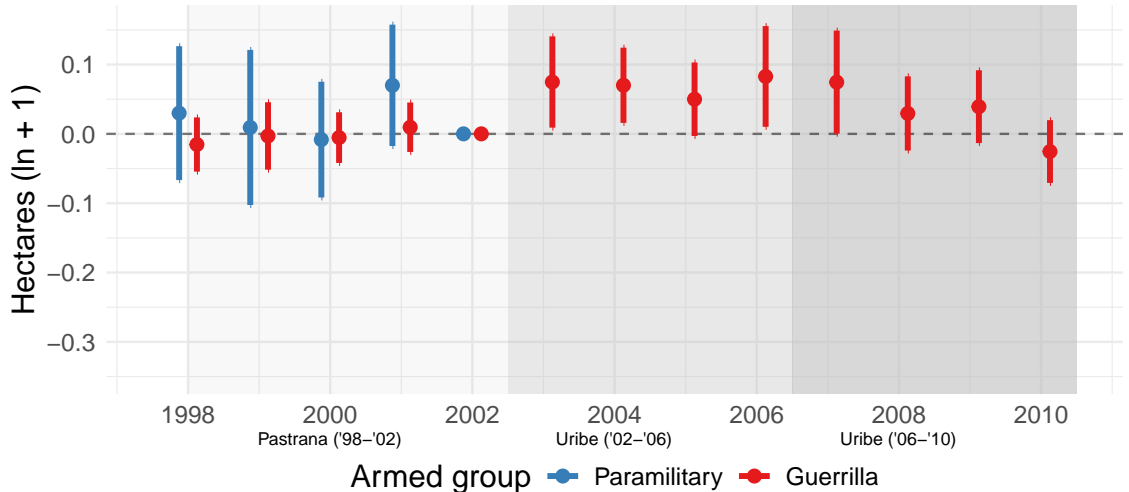
Average monthly eradication



# Year-by-Year Results

Estimating Equation

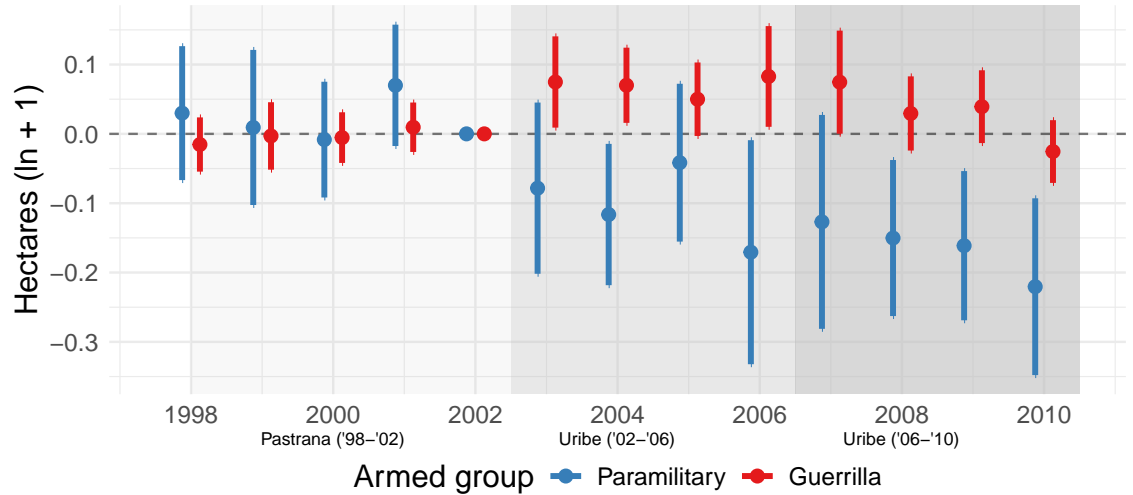
Average monthly eradication



# Year-by-Year Results

Estimating Equation

Average monthly eradication



# Selective Drug Enforcement

Eradication patterns under Uribe reflect differential enforcement

**Decreased** eradication where greater past **paramilitary** violence

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Results not determined by differential coca cultivation Baseline Lagged Proportion

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...or by local partisanship

Additional Covariates

Mayoral Regression Discontinuity Design

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Results robust to different measurement, functional forms

Appendix Contents



# Paramilitary Demobilization

# Paramilitaries Mostly Demobilized by 2005

Favorable conditions, despite accounting for  $\approx 9,000$  civilian deaths  
Eight year sentence max, sentences could be served on private farms,  
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Favorable conditions, despite accounting for  $\approx 9,000$  civilian deaths

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could keep profits from criminal activities

Groups have less direct electoral influence post-demobilization



# Electoral Overperformance

Leader holds back on enforcement to friendly criminal groups

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**Paramilitaries deliver votes → receive relief from enforcement**

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Examine Uribe's overperformance in paramilitary areas and eradication

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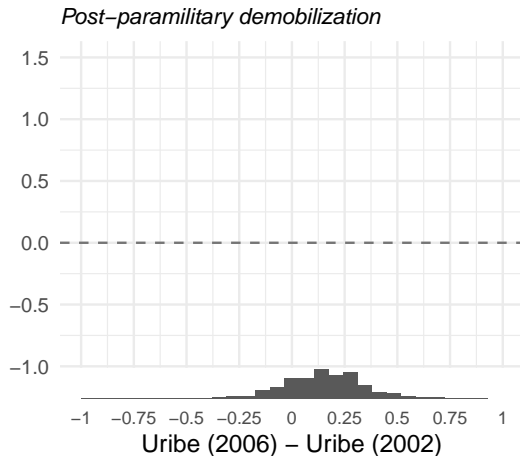
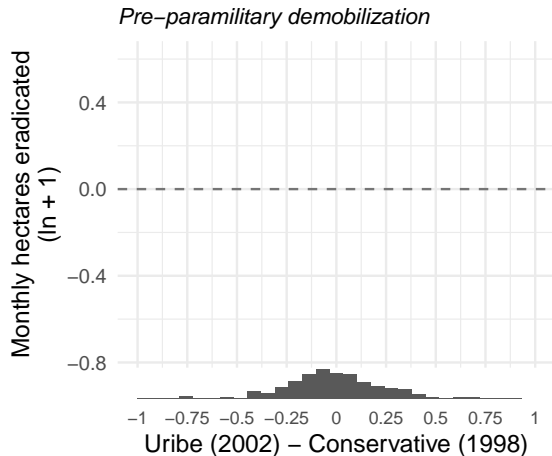
**Paramilitaries deliver votes → receive relief from enforcement**

Examine Uribe's overperformance in paramilitary areas and eradication

<b>Vote Difference</b>	<b>Used to Predict</b>	<b>Restraint</b>
2002 relative to 1998	Eradication 2002-2006	More likely
2006 relative to 2002	Eradication 2006-2010	Less likely

Expect **restraint** in **paramilitary** areas with **2002** overperformance

# Elections and Demobilization

[More Results](#)[Electoral Violence](#)

Paramilitary attacks



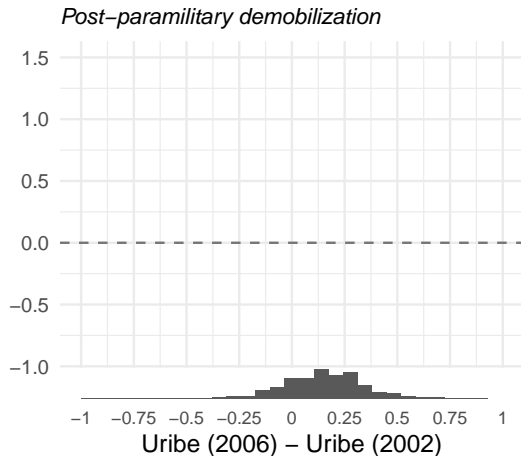
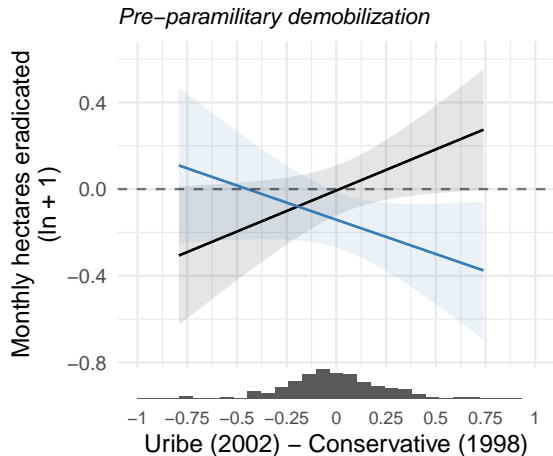
0



+2 SD



# Elections and Demobilization

[More Results](#)[Electoral Violence](#)

Paramilitary attacks

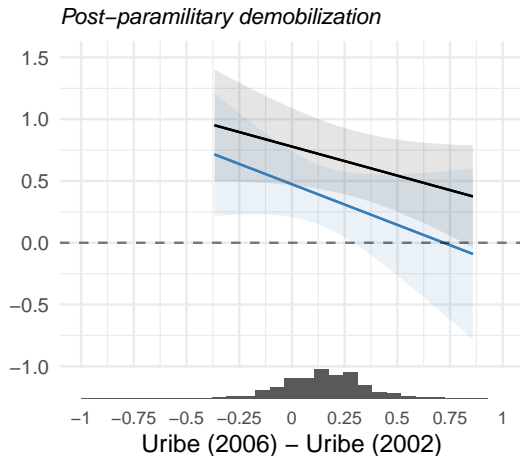
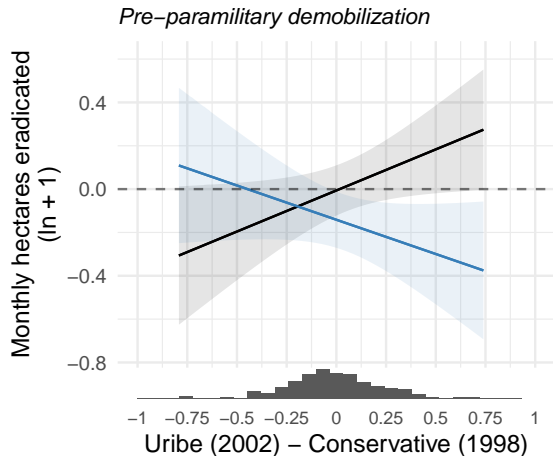


0



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# Elections and Demobilization

[More Results](#)[Electoral Violence](#)

Paramilitary attacks



0

+2 SD

# Conclusion

# Discussion: Theory

When does the state target competitors to its monopoly on violence?

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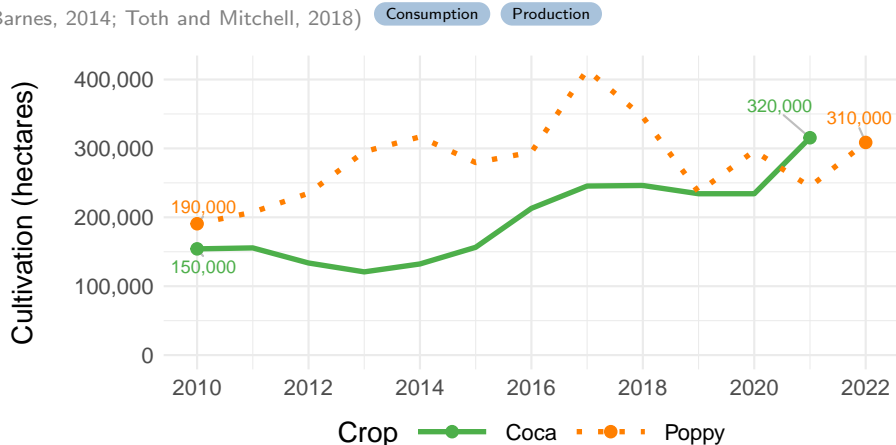
*“The purpose of this process is achieving national peace through the strengthening of democratic governance and the restoration of the monopoly on violence to the State.”*

— Santa Fe de Ralito agreement between Colombian national government and AUC

# Discussion: Policy

## Harshly punitive policy has **limited proven effectiveness**

(Baumgartner et al., 2021; Blair and Weintraub, 2023; Brinks, 2007; Geller et al., 2014; Flores-Macías and Zarkin, 2021; Kleck and Barnes, 2014; Toth and Mitchell, 2018)



# Discussion: Policy

Harsh punitive policy has **limited proven effectiveness**  
...but preventive policy demonstrates **increasing promise**

(Prieto-Curiel, Campedelli and Hope, 2023; Blattman, Jamison and Sheridan, 2017; García-Ponce et al., 2023; Heller et al., 2017; Mitchell, Wilson and MacKenzie, 2012)

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(Prieto-Curiel, Campedelli and Hope, 2023; Blattman, Jamison and Sheridan, 2017; García-Ponce et al., 2023; Heller et al., 2017; Mitchell, Wilson and MacKenzie, 2012)

Understanding **how these policies are implemented** and  
under what circumstances is vital



# Discussion: Policy

**Agency** of elected leaders matters for drug enforcement

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Leaders can use **criminal groups** to their advantage

Creating undesirable normative outcomes (Daly, 2022; Nieto-Matiz, 2023)

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Creating undesirable normative outcomes (Daly, 2022; Nieto-Matiz, 2023)

Instead make policy goals **incentive compatible** for leaders

# Conclusion

Political incentives of elected leaders shape drug enforcement

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Tested variation in political incentives to enforce:

**Increased** eradication where greater past **guerrilla** violence

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variation in criminal justice

**This paper**, *SPPQ* paper,  
in-progress book project,  
origins of U.S. foreign aid

# Conclusion

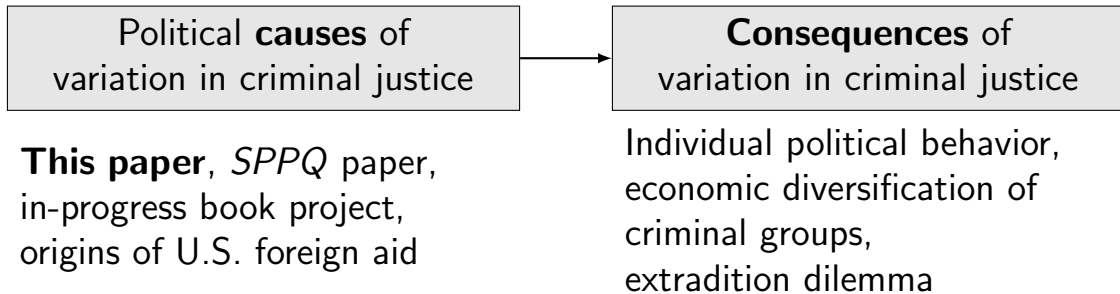
Political incentives of elected leaders shape drug enforcement

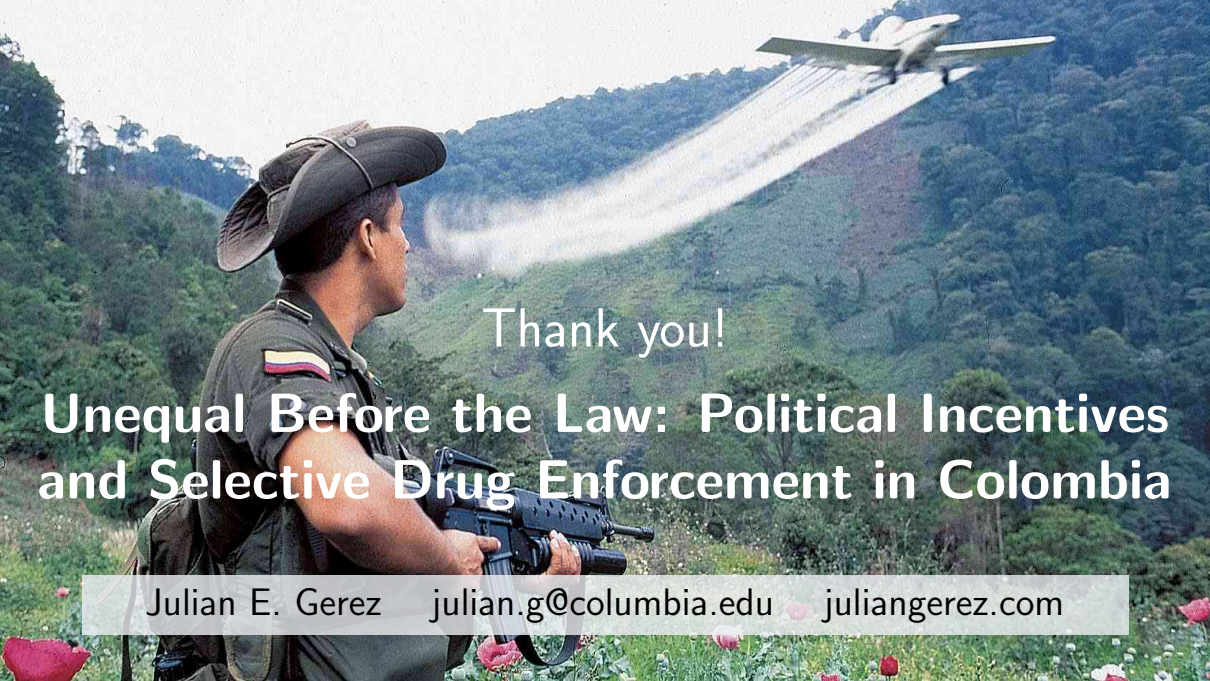
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Thank you!

# Unequal Before the Law: Political Incentives and Selective Drug Enforcement in Colombia

Julian E. Gerez   [julian.g@columbia.edu](mailto:julian.g@columbia.edu)   [juliangerez.com](http://juliangerez.com)

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- 3 Context, Research Design, and Data
- 4 Results: Political Incentives and Selective Enforcement
- 5 Mechanisms: Paramilitary Demobilization
- 6 Conclusion

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Event study estimating equation

Full set of results

Formal test of parallel trends

Cross-sectional results

Controlling for coca cultivation

Including time-varying covariates

Mayoral RDD

Pre-baseline violence data (1988-1997)

Binary measure of violence data

$\ln + 1$  transformed violence data

Squaring violence measures

Interacting violence measures

Electoral overperformance table

Electoral violence results

Global drug production and consumption

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[Pre-baseline violence data](#)

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[\$\ln + 1\$  treatment](#)

[Squaring violence measures](#)

[Interacting violence measures](#)

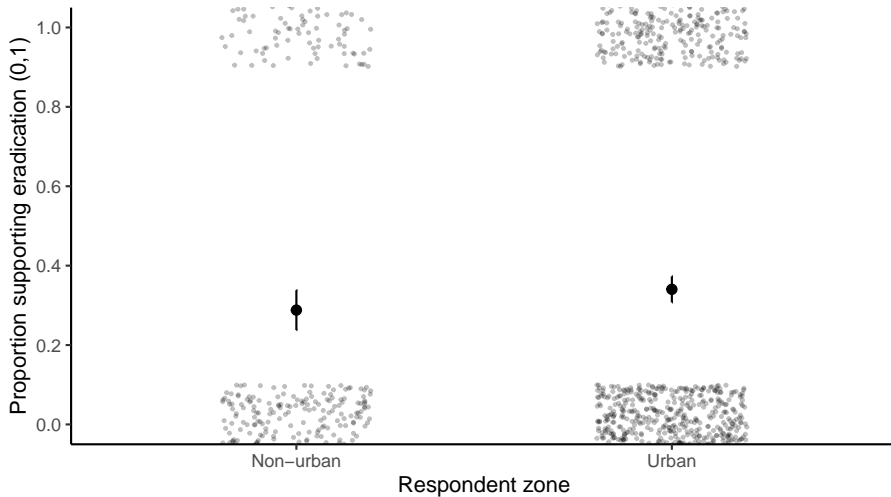
[Additional results overperformance](#)

[Additional results violence](#)

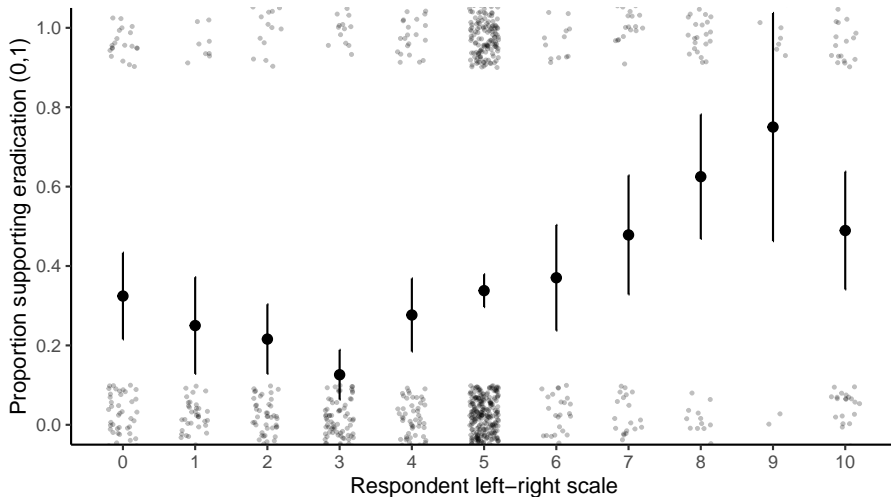
[Consumption](#)

[Production](#)

# Eradication and Area

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# Ideology and Candidates

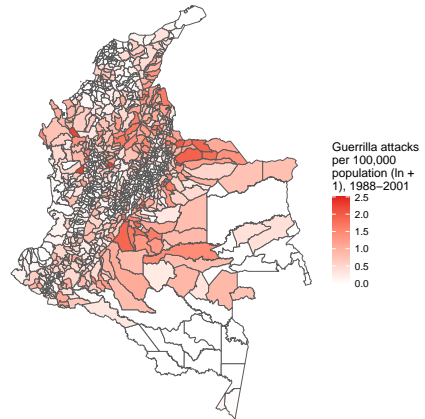
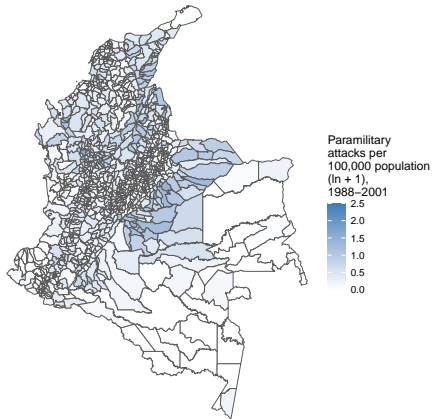
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**Table:** Relationship between ideology and political opinions.

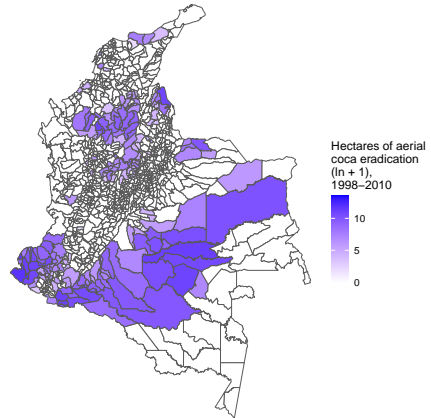
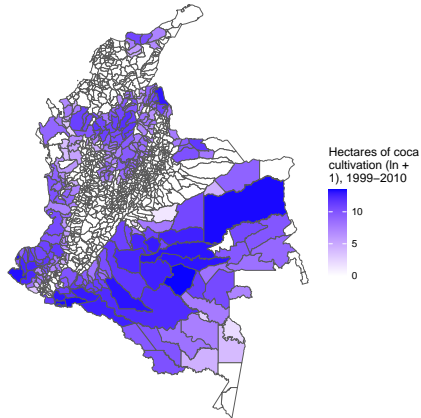
<i>Outcomes:</i>	Approval	Self-reported vote: Round 1				Vote intention: Round 2	
	Duque	Any	Petro	Fico	Hernández	Petro	Hernández
	<i>Right</i>		<i>Left</i>	<i>Right</i>	<i>Right</i>	<i>Left</i>	<i>Right</i>
	<i>Incumbent</i>						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
<b>Predictor is respondent self-reported left-right scale placement, range: {0,1,2,3,4,5,6,7,8,9,10}</b>							
Ideology	0.219*** (0.016)	-0.006 (0.004)	-0.104*** (0.006)	0.055*** (0.005)	0.030*** (0.005)	-0.106*** (0.005)	0.086*** (0.005)
R <sup>2</sup>	0.17	0.001	0.22	0.13	0.03	0.23	0.19
Observations	1,132	1,109	966	966	966	1,128	1,128

*Notes:* All specifications are estimated using OLS. Heteroskedasticity-robust standard errors in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Geographies of Violence

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# Geographies of Coca

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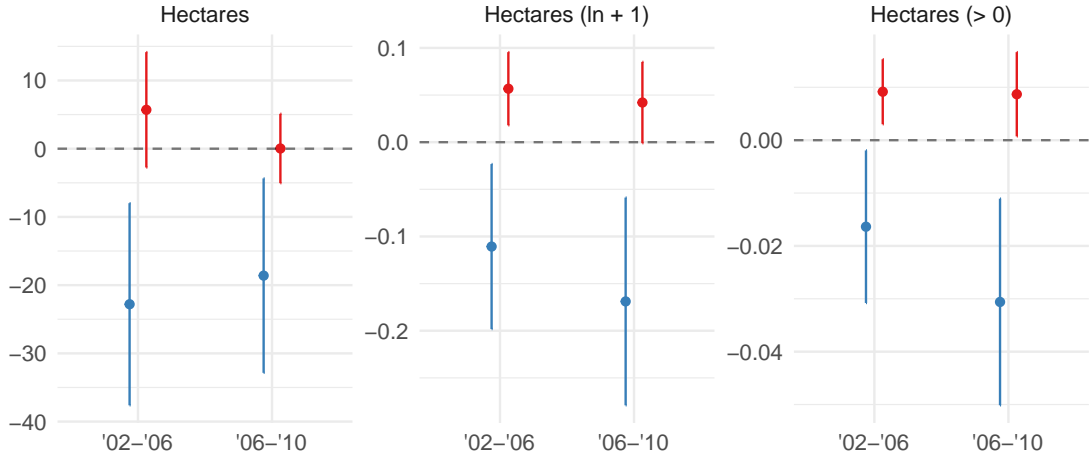
# Event Study Estimation

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$$Eradication_{i,t} = \sum_{j \neq 2002} \beta_j P_i \times \mathbb{1}[y = j] + \sum_{j \neq 2002} \zeta_j G_i \times \mathbb{1}[y = j] + \gamma_i + \delta_t + \epsilon_{i,t}$$

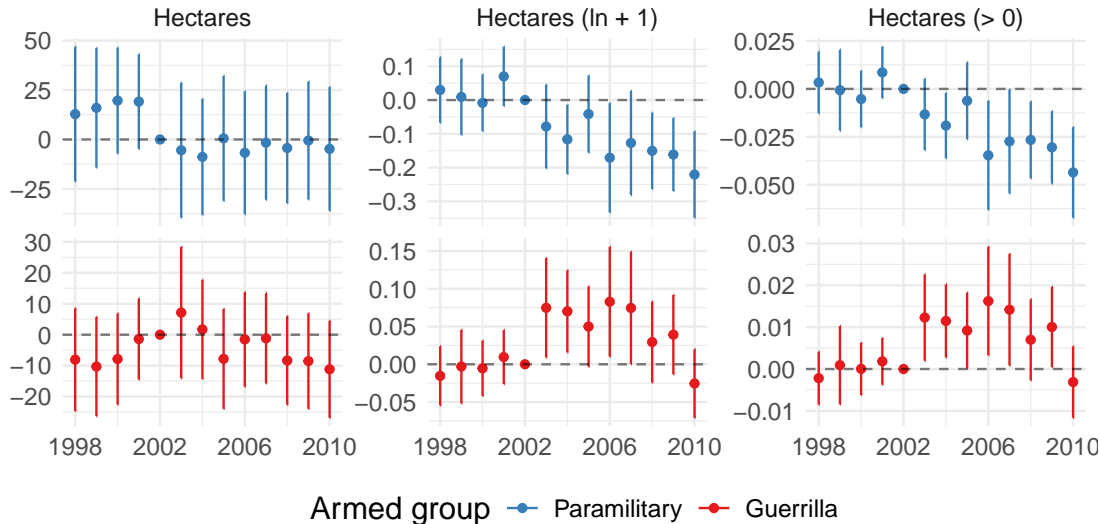
- $Eradication_{i,t}$ : measure of eradication in municipality  $i$  in year-month  $t$
- $P_i$ : time-invariant measure of paramilitary attacks
- $G_i$ : time-invariant measure of guerrilla attacks
- $\mathbb{1}[y = j]$ : year indicators
- $\gamma_i$ : municipality fixed effects
- $\delta_t$ : year  $\times$  month fixed effects

# Full Main Results

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Armed group — Paramilitary — Guerrilla

# Full Event Study Results

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**Table:** Formal test for parallel trends (Plan Colombia placebo), continuous treatment.

	Hectares (1)	Hectares (ln +1) (2)	Hectares (> 0) (3)
Paramilitary attacks $\times$ 2001-2002	-7.338 (6.095)	-0.009 (0.036)	0.001 (0.007)
Guerrilla attacks $\times$ 2001-2002	3.021 (2.572)	0.007 (0.014)	0.000 (0.003)
R <sup>2</sup>	0.10	0.28	0.29
Observations	15,264	15,264	15,264
Municipalities	318	318	318

*Notes:* All specifications are estimated using OLS and include municipality and year  $\times$  month fixed effects. Robust standard errors clustered by municipality are in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

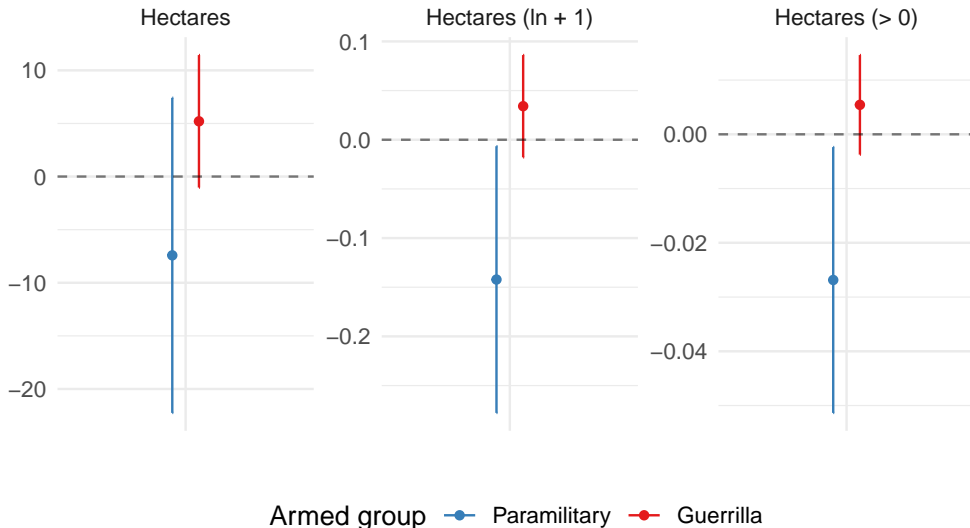


**Table:** Formal test for parallel trends (July 2000 placebo), continuous treatment.

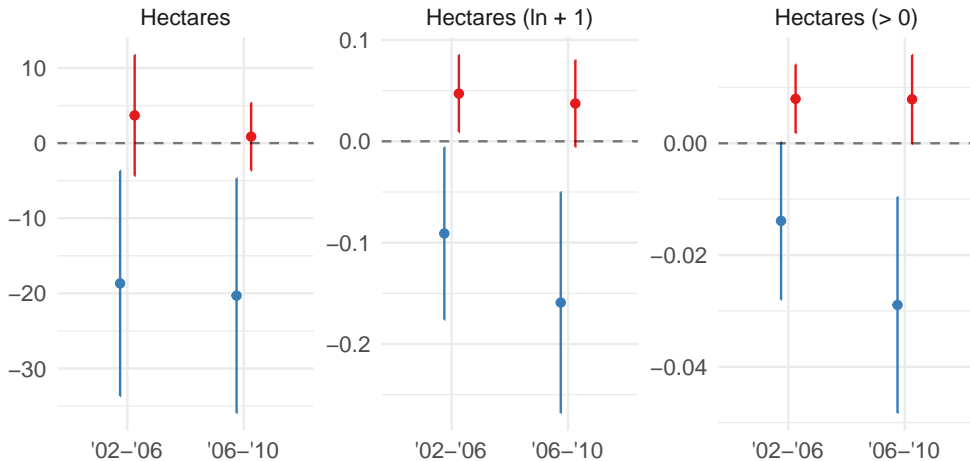
	Hectares (1)	Hectares (ln +1) (2)	Hectares (> 0) (3)
Paramilitary attacks $\times$ 2000-2002	-4.924 (7.335)	0.017 (0.039)	0.004 (0.007)
Guerrilla attacks $\times$ 2000-2002	6.046* (3.602)	0.013 (0.017)	0.001 (0.003)
R <sup>2</sup>	0.10	0.28	0.29
Observations	15,264	15,264	15,264
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# Cross-sectional Results

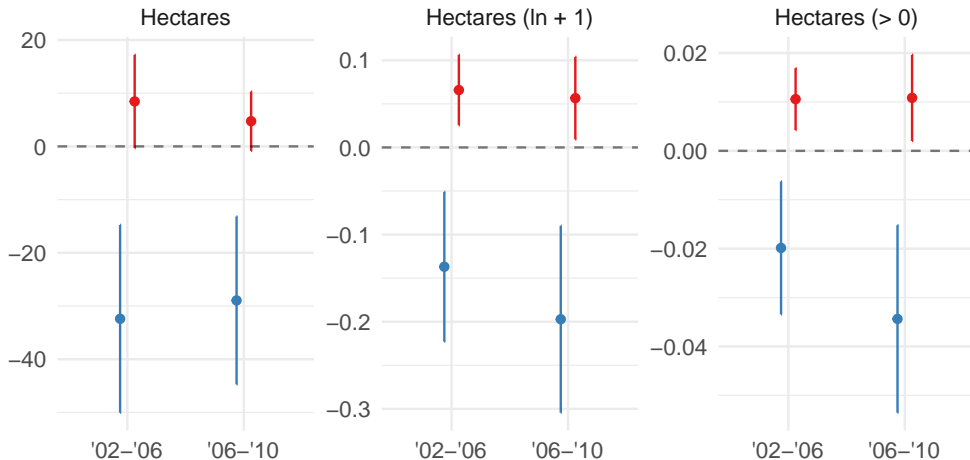
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# Baseline Cultivation

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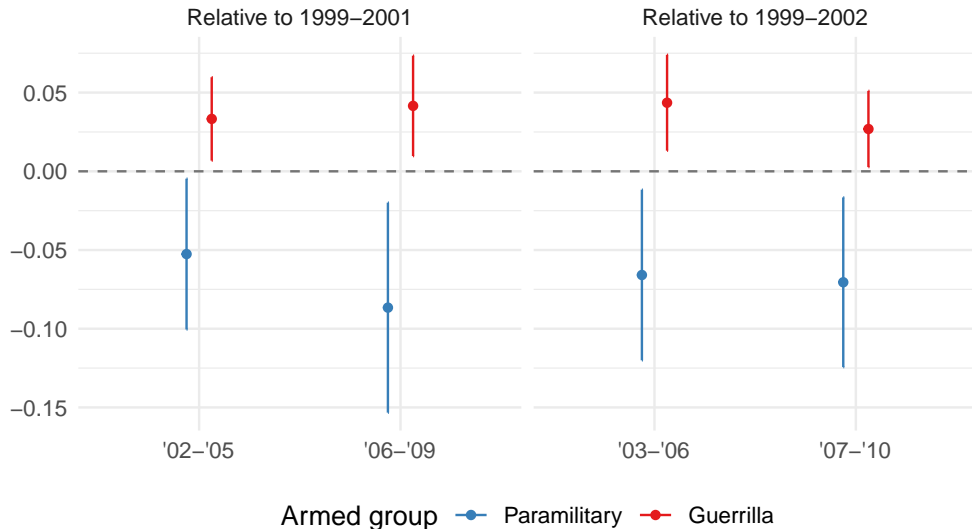
Armed group — Paramilitary — Guerrilla

# Lagged Cultivation

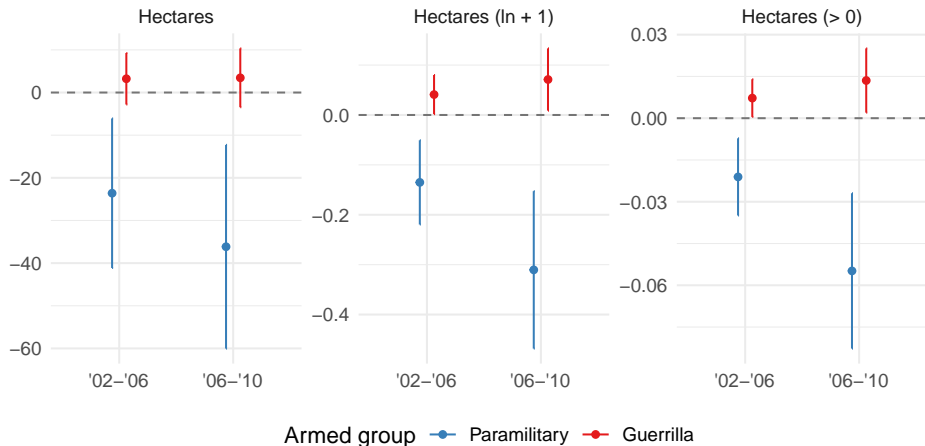
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Armed group — Paramilitary — Guerrilla

# Proportion Eradicated

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# Additional Covariates

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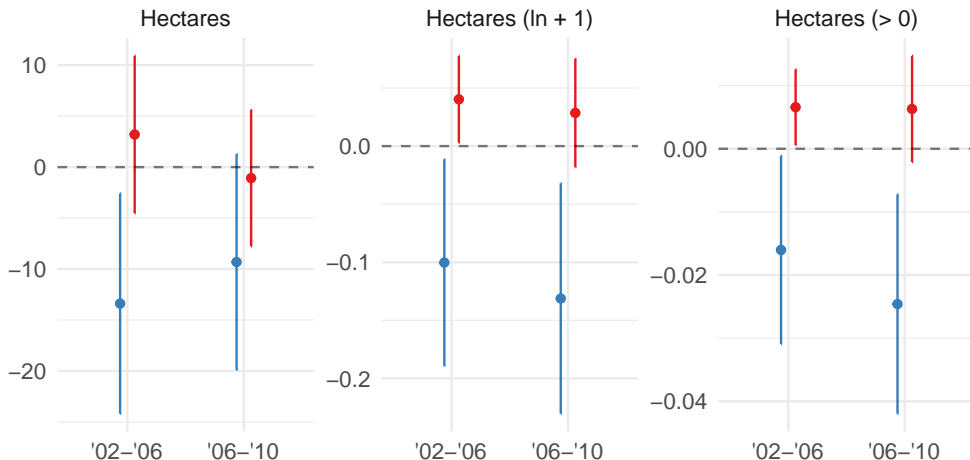
Year-month fixed effects interacted with department fixed effects, municipality area, coca suitability, altitude, and distance to Bogotá, as well as measures of the pre-violence right/left lean of each municipality

**Table:** Regression discontinuity: election of partisan mayors on eradication.

	Avg. yearly aerial eradication hectares (1)	Avg. yearly aerial eradication hectares (ln +1) (2)	Avg. yearly aerial eradication hectares (> 0) (3)
<b>Panel A: Right-wing mayor</b>			
Mayor elected	2.952 (18.430)	0.188 (0.406)	0.049 (0.087)
Observations	187, 172	187, 172	187, 172
Effective obs.	93, 86	102, 93	95, 88
Bandwidth	0.067, 0.067	0.076, 0.076	0.07, 0.07
<b>Panel B: Left-wing mayor</b>			
Mayor elected	414.558 (577.374)	-0.532 (1.062)	-0.175† (0.141)
Observations	41, 44	41, 44	41, 44
Effective obs.	13, 21	12, 20	10, 20
Bandwidth	0.06, 0.06	0.055, 0.055	0.049, 0.049

*Notes:* Standard errors are in parentheses. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

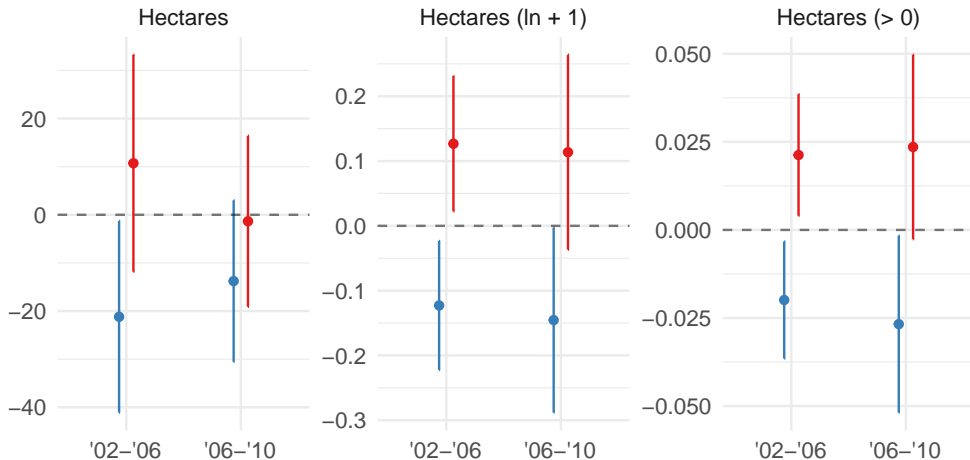
# Violence (1988-1997)

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Armed group — Paramilitary — Guerrilla

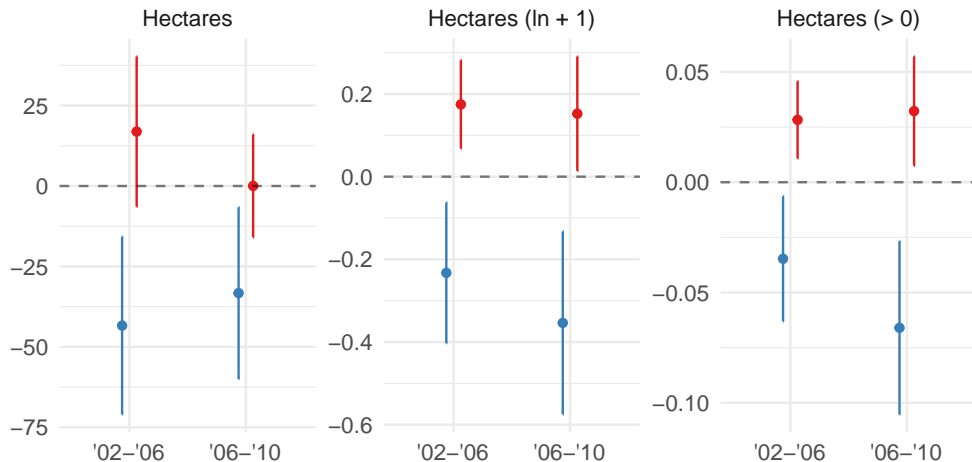


# Binary Violence

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Armed group — Paramilitary — Guerrilla

# ln + 1 Violence

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Armed group — Paramilitary — Guerrilla

# Squaring Violence

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	Hectares (1)	Hectares (ln +1) (2)	Hectares (> 0) (3)
Paramilitary attacks $\times$ 2002-2006	-39.628** (17.936)	-0.227* (0.124)	-0.033 (0.021)
Paramilitary attacks <sup>2</sup> $\times$ 2002-2006	7.634 (6.849)	0.049 (0.046)	0.007 (0.008)
Paramilitary attacks $\times$ 2006-2010	-30.790* (17.483)	-0.351** (0.170)	-0.065** (0.030)
Paramilitary attacks <sup>2</sup> $\times$ 2006-2010	5.924 (6.861)	0.076 (0.063)	0.014 (0.011)
Guerrilla attacks $\times$ 2002-2006	10.238 (8.640)	0.115*** (0.041)	0.019*** (0.007)
Guerrilla attacks <sup>2</sup> $\times$ 2002-2006	-0.697 (1.173)	-0.010* (0.006)	-0.002 (0.001)
Guerrilla attacks $\times$ 2006-2010	0.759 (6.632)	0.131** (0.058)	0.028*** (0.010)
Guerrilla attacks <sup>2</sup> $\times$ 2006-2010	-0.045 (0.819)	-0.015* (0.008)	-0.003** (0.001)
R <sup>2</sup>	0.12	0.22	0.21
Observations	45,792	45,792	45,792
Municipalities	318	318	318

*Notes:* All specifications are estimated using OLS and include municipality and year  $\times$  month fixed effects. Baseline category is Pastrana's term from 1998-2002. Robust standard errors clustered by municipality are in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Interacting Violence

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	Hectares (1)	Hectares (ln +1) (2)	Hectares (> 0) (3)
Paramilitary attacks × 2002-2006	-23.064* (10.492)	-0.115† (0.063)	-0.016 (0.010)
Paramilitary attacks × 2006-2010	-19.332* (9.436)	-0.191* (0.084)	-0.034* (0.014)
Guerrilla attacks × 2002-2006	5.557 (7.330)	0.054 (0.034)	0.009† (0.005)
Guerrilla attacks × 2006-2010	-0.368 (5.086)	0.030 (0.045)	0.007 (0.008)
Paramilitary attacks × guerrilla attacks × 2002-2006	0.165 (4.814)	0.003 (0.025)	0.000 (0.004)
Paramilitary attacks × guerrilla attacks × 2006-2010	0.454 (5.349)	0.013 (0.035)	0.002 (0.006)
R <sup>2</sup>	0.12	0.22	0.21
Observations	45,792	45,792	45,792
Municipalities	318	318	318

*Notes:* All specifications are estimated using OLS and include municipality and year × month fixed effects. Baseline category is Pastrana's term from 1998-2002. Robust standard errors clustered by municipality are in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Electoral Overperformance

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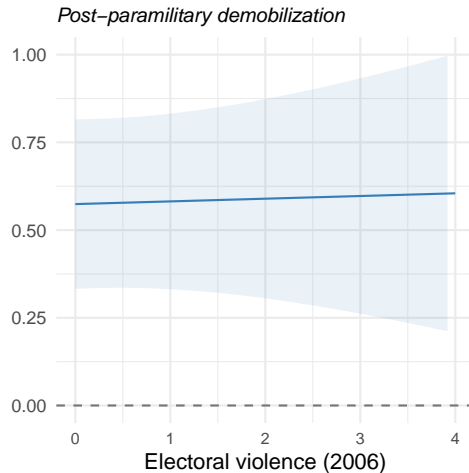
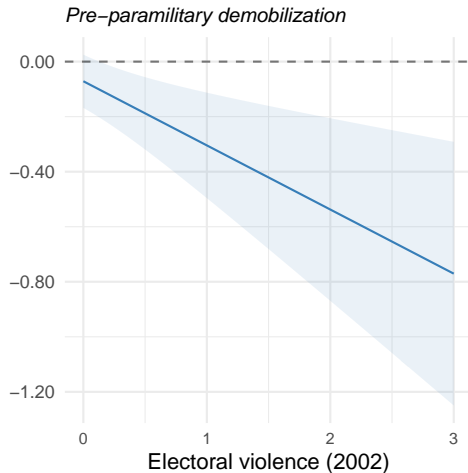
	Hectares (1)	Hectares (ln +1) (2)	Hectares (> 0) (3)
<b>Panel A: Aerial eradication (2002-2006)</b>			
$\Delta^{2002} \times$ Paramilitary attacks	-100.954* (58.624)	-0.419* (0.233)	-0.066* (0.036)
R <sup>2</sup>	0.04	0.10	0.09
<b>Panel B: Aerial eradication (2006-2010)</b>			
$\Delta^{2006} \times$ Paramilitary attacks	-1.33 (37.945)	-0.113 (0.427)	-0.018 (0.079)
R <sup>2</sup>	0.05	0.10	0.10
Observations	13,680	13,680	13,680
Municipalities	285	285	285

*Notes:* All specifications are estimated using OLS and include department and year  $\times$  month fixed effects. Robust standard errors clustered by municipality are in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Electoral Violence

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Outcome: Hectares (ln + 1)



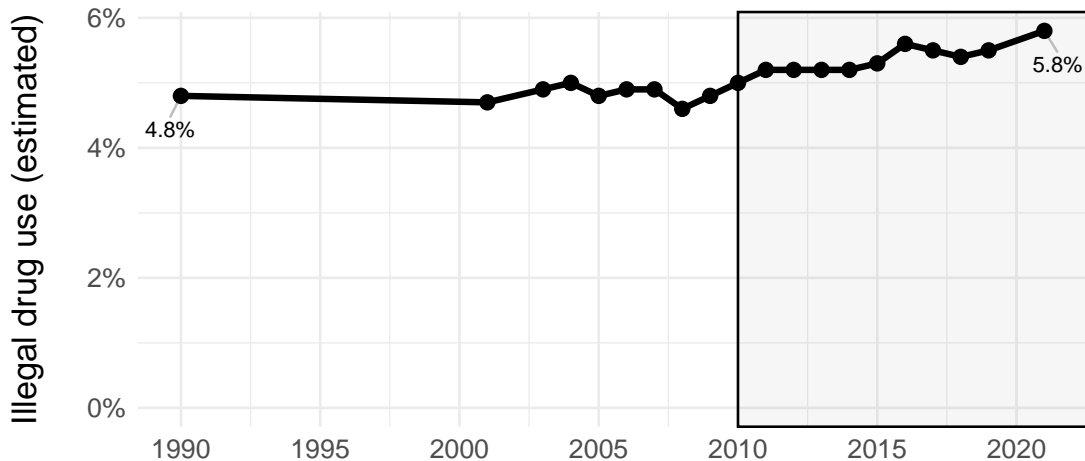
# Electoral Violence

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	Hectares (1)	Hectares (ln +1) (2)	Hectares (> 0) (3)
<b>Panel A: Aerial eradication (2002-2006)</b>			
Electoral violence (2002)	-42.064 (30.472)	-0.235** (0.092)	-0.037*** (0.014)
R <sup>2</sup>	0.04	0.11	0.10
<b>Panel B: Aerial eradication (2006-2010)</b>			
Electoral violence (2006)	-5.210 (4.591)	0.008 (0.050)	0.007 (0.010)
R <sup>2</sup>	0.05	0.10	0.10
Observations	14,208	14,208	14,208
Municipalities	296	296	296

*Notes:* All specifications are estimated using OLS and include department and year  $\times$  month fixed effects. Robust standard errors clustered by municipality are in parentheses. \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Global Drug Consumption

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# Global Drug Production

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