

Connivance and coercion *

Melissa Pavlik[†]

July 4, 2025

Abstract

Why do states often fail to enforce their own policies, especially those governing the poor and vulnerable? How do state policies and their enforcement shape coercion by extralegal groups? This paper investigates these questions in the context of modern-day Nigeria, a decades-old democracy which nevertheless features high levels of violence and exploitation. I provide a conceptual framework which shows how enabling such exploitation is functional for democratic states insofar as it empowers non-state allies who benefit from the vulnerability of regulated populations. Diverging from existing approaches, which characterize uneven enforcement against the poor as benevolent, I detail how states ‘produce precarity’ in vulnerable citizens through uneven enforcement of their own policies. I focus on the role of transport unions in Lagos, Nigeria—extractive actors who exploit drivers, and work as purveyors of electoral violence for the ruling party—in determining state enforcement of a ban on okada motorcycle taxis in the state. First, using evidence from several months fieldwork in Lagos; as well as an original networked dataset of the Lagos transport union, I show how the Lagos State Government’s selective enforcement of its ban on motorcycle taxis was preceded by driver threats of secession amidst union extortion. Second, I use original geocoded data on enforcement locations, union territory, and traffic patterns along the Lagos road network to show how the political geography of the ban’s enforcement displaces riders into areas controlled by the union. Third and finally, I exploit the timing of a surprising election result to show how a shock to state reliance on the union affects enforcement patterns. This paper not only explores how states can exploit an understudied ‘enforcement lever’ to usurp democratic institutions and redistribute to their allies, but how powerful—but not necessarily criminal—groups can trade extralegal violence for such redistribution.

***Draft; do not circulate.** This project was supported by Yale’s MacMillian Center as well as APSA’s Doctoral Dissertation Research Improvement Grant. I would like to thank my committee Kate Baldwin, Gerard Padro i Miquel Elisabeth Wood, and Sarah Bush; as well as the many colleagues who offered thoughts on early drafts, especially David Cerero Guerra, Jessica Hickle, and Ryan Pike. Thank you also to audiences at and organizers of Yale’s Leitner Program on Political Economy, APSA 2024, APSA Comparative Labor Politics 2025, NEWEPS 2025, and BWGAPE 2025. I am very grateful to all those who worked beside me in Lagos, or who offered words of wisdom on navigating that incredible city. All errors are my own.

[†]Yale University, Political Science: melissa.pavlik@yale.edu; www.melissapavlik.com

1 Why do states often fail to completely enforce their own policies, especially those governing
2 the poor and vulnerable? All over the world, state actors propose, debate, and pass laws targeting
3 informal economies and populations: examples include bans on hawking, squatting, driving an
4 unregistered taxi, or crossing borders without documentation. However, enforcement of these
5 policies often varies significantly over time and space—even from one week to the next, even
6 within a single city. Most scholarship chalks up gaps in policy enforcement to gaps in state
7 capacity; more recent scholarship has advanced alternative explanations for *forbearance*, defined
8 by Holland as “intentional and revocable government leniency towards violations of the law”
9 (Holland, 2016, 233). These explanations show how governments can refuse to enforce policies
10 targeting the poor as a form of redistribution.

11 However, incomplete enforcement cannot always be attributed either to state failure or to state
12 sympathy with its most vulnerable populations. In many cases, infringing populations are not key
13 to the selectorate: undocumented migrants for example, who cannot vote either for or against the
14 incumbent and have few allies among registered voters. And while forbearance is often described
15 in benevolent terms, incomplete enforcement does not always benefit offenders. When a politician
16 blackmails Romani populations about tax nonpayment in direct exchange for votes—as described
17 by Mares and Young (2019)—the state, alongside the apparent concession, is extending not a favor
18 but a threat. When U.S. police across the Jim Crow South declined to enforce racist laws, only for
19 lawbreakers to be attacked by extrajudicial militias, the state was not *conceding* anything to Black
20 populations.¹

21 In this paper, I argue that gaps in enforcement can be strategically used to ‘produce precarity’ in
22 offenders and enhance opportunities for their exploitation, especially by third parties. I argue that
23 allowing—and even enabling—exploitation in this way is functional for democratic states insofar

¹One such example occurred during nonviolent sit-ins by Black college students at segregated lunch counters in Nashville, Tennessee in 1960. Several days into the demonstration, on February 27 of that year, the police began to engage—not just with arrests, but by standing aside as violent proxies attacked protesters. One demonstrator Bernard Layfayette noted that on that day “we were told in advance... that they are gonna allow the hoodlums to beat us up, and *then* the police were gonna arrest us.” Once the sit-in began, “as predicted, the police held back for the first fifteen minutes,” resulting in severe violence against the Black students. For more, see (York, 1999, minutes 0:35:22–0:36:22).

¹ as it empowers non-state allies who benefit from the vulnerability of those being regulated. I refer
² to this strategy as *connivance*: a coercive, rather than concessional, form of forbearance. I add to
³ existing conceptualizations of forbearance by integrating the concept with theories of repression,
⁴ and describe how selective enforcement can act as a strategy of political control (Hassan et al.,
⁵ 2021) and consolidate relationships between state agents and extortionate, exploitative, and often
⁶ violent third-party actors.

⁷ I elucidate the concept of connivance against the backdrop of the informal transport industry
⁸ in Nigeria. Specifically, I analyze the 2022 selective enforcement of a ban on ‘okada’ motorcycle
⁹ taxis in Lagos, Africa’s largest metropolitan area. Okada riders² are usually economic migrants
¹⁰ and internally displaced persons from the North of Nigeria, often fleeing climate-induced land
¹¹ scarcity and political violence (Grasse and Pavlik, 2025). While the Lagos okada ban applied
¹² to all major roads in the state, it was enforced only across select bridges, intersections, streets,
¹³ and neighborhoods. Capacity-based explanations do not convincingly explain the ban’s uneven
¹⁴ enforcement; Lagos streets are heavily policed by at least a half a dozen branches of law enforcement.
¹⁵ The ban’s passage and partial enforcement also drew outrage from okada riders, calling into
¹⁶ question any benevolent motivation by the recalcitrant state.

¹⁷ I use original micro-level geospatial data on enforcement and extortion points on the Lagos
¹⁸ roadways, traffic patterns, and informal transport density, as well as qualitative data both from
¹⁹ fieldwork and archival news reports, to show that selective enforcement of the okada ban is a
²⁰ strategy of connivance by the Lagos State Government. The ban’s incomplete enforcement was
²¹ not for the purpose of redistribution to vulnerable riders, but to ease their extortion by extralegal
²² violence entrepreneurs—in this case, the Lagos branch of the National Union of Road and Transport
²³ Workers (NURTW; heretofore referred to as ‘the Union’): a mafia-like organization that behaves
²⁴ as a key source of thugs-for-hire for Lagos’s leading political party, the All Progressives Congress
²⁵ (APC). Drawing from participant observation on Lagos roadways and amongst Union members

²Here in keeping with local parlance, I refer to the drivers of the motorcycles as ‘riders,’ and the consumers of the service as ‘passengers.’

1 and okada riders, over 75 stakeholder interviews, and a near-comprehensive collection of local
2 newspaper reports spanning decades of Union activity, I describe how enforcement timing corresponded
3 with okada rider resistance to Union extortion, prompting state intervention. Using a ‘randomization
4 inference’ simulation-based technique on the Lagos road network, I demonstrate that the political
5 geography of the ban’s enforcement served to displace riders into areas of the city controlled
6 by the Union, increasing its ability to exploit riders during this tumultuous period. I show that
7 spatial patterns of ban enforcement does not cleanly vary with other potential explanations, such
8 as geographic variation in state capacity or traffic congestion. Finally, I exploit a surprising election
9 result from the year following the ban, which served as a shock to the state government’s perception
10 of the reliability of the Union. Using fine-grained satellite-based measurements of commercial
11 transport presence, I find that when state reliance on the Union is shaken, enforcement diminishes.

12 This paper contributes to several substantive agendas in the study of politics. I contribute to an
13 increasingly common focus on spatial economics in urban centers of low-income countries (Bryan,
14 Frye and Morten, 2025). More specifically, I add to the growing wealth of literature on the informal
15 transport sector and traffic politics (Agbiboa, 2022; Goodfellow, 2015; Fourchard, 2023) as well
16 as the road networks and their disruptions (Nathan, 2023; Schouten, 2022; Dell, 2015; Sánchez
17 De La Sierra et al., 2024) which characterize so many major cities in the Global South broadly,
18 and on the continent of Africa in particular. While there exist a number of studies examining
19 roads and their construction, including in studies of colonial expansion (Cowen, 2020), conflict
20 dynamics (Zhukov, 2012), and large-scale infrastructure corruption (Williams, 2017), modern laws
21 governing the roads get less attention.³ This is despite the fact that in urban areas worldwide,
22 parking tickets, traffic fines, and checkpoints are among the most frequent ways in which citizens
23 directly interact with the state. In what follows, I show how states can use instruments as seemingly
24 banal as traffic policy enforcement to advance the interests of exploitative third parties and alter a
25 city’s political geography. In doing so, I build on work in urban economics and in political science

³With notable exceptions: See Su and Buerger (2024) and Ben-Menachem and Morris (2023) for two examples in the U.S. context; and Sánchez De La Sierra et al. (2024) for a detailed dive into the Congo’s traffic agency.

¹ by Xu (2023) and others on how spatial externalities foster inequality in such contexts. Moreover,
² I add to existing work on the role of non-state actors in weakly-institutionalized democracies;
³ bridging the gap between work on civil society groups like unions (Dean, 2022; Hassan, Mattingly
⁴ and Nugent, 2022) and violence entrepreneurs including gangs, mafias, and political militias (Carey,
⁵ Mitchell and Scharpf, 2022; Tapscott, 2021; Acemoglu, Robinson and Santos, 2009). Indeed, in
⁶ this context, they are one and the same.

⁷ Most fundamentally, I expand our understanding of the concept of forbearance by exploring
⁸ how non-enforcement against precarious populations can counterintuitively be a tool of political
⁹ control, rather than a form of benevolent redistribution. In doing so, I expand both on studies
¹⁰ of forbearance and of repression in the context of informality. Most theories of repression and
¹¹ coercion focus on the state's ability to inflict physical violence (Thachil, 2020), maintain high
¹² levels of censorship and control over speech (Sullivan and Davenport, 2018), or even actively
¹³ engage in distribution for coercive purposes (Albertus, Fenner and Slater, 2018; Pan, 2020). Although
¹⁴ the study of how states inflict violence has recently expanded to include subtler modes of political
¹⁵ control (Hassan, Mattingly and Nugent, 2022), repression is still largely assumed to be a function
¹⁶ of a state's capacity to enforce its own repressive policies. As such, studies of repression have
¹⁷ incompletely grappled with the coercion produced where the state selectively does *not* engage in
¹⁸ enforcement. I show how in some contexts, uneven governance and policing patterns should be
¹⁹ interpreted not as lapses in capacity, but as purposeful tactics used to empower extra-state violence
²⁰ entrepreneurs and enable exploitation of the vulnerable. These findings not only elaborate on
²¹ existing scholarly accounts on policing (Eck, Conrad and Crabtree, 2021) and strategic non-enforcement
²² (Holland, 2017; Cunningham and Owens, 2020; Su and Buerger, 2024), but have significant implications
²³ for our understanding of a state's willingness to concede its monopoly on enforcement and violence
²⁴ (Acemoglu, Robinson and Santos, 2009; Carey, Mitchell and Scharpf, 2022). This paper answers a
²⁵ call by Hassan, Mattingly and Nugent (2022) to expand considerations of state repression "beyond
²⁶ capacity to also consider intent" (Hassan, Mattingly and Nugent, 2022, 157).

¹ 1 Connivance

² When forbearance is coercive

³ Why do agents of the state enforce certain policies and not others, and what explains vast variation
⁴ in enforcement across time and space? Most work tends to blame enforcement gaps on a lack of
⁵ state capacity, a concept at the center of explaining both how states form and how they operate—especially
⁶ how they utilize and build their coercive power (Tilly et al., 1992). ⁴ However, the concept of state
⁷ capacity often bundles distinct qualities of bureaucratic performance and policy implementation,
⁸ leading to over-generalized theories which do not account for idiosyncratic constraints and incentives
⁹ (Williams, 2021). In response, scholars have increasingly explored cases where governments
¹⁰ are able, but unwilling, to comprehensively enforce a policy—*refusing to use* the capacity they
¹¹ possess. These efforts have resulted in alternative ‘agent-based’ approaches, which theorize that
¹² enforcement will vary not just with state capacity, but with interests. For instance, governments
¹³ may neglect to enforce violations where non-enforcement is popularly preferred.⁵

¹⁴ These agent-based explanations largely attribute non-enforcement to state incentives to appease
¹⁵ the population breaking the law. In other words, when offenders are key allies or members of
¹⁶ the electorate, state agents may refrain from enforcement in order to retain this population’s
¹⁷ support: a *concessional* logic of non-enforcement. The literature has considered two varieties of
¹⁸ this phenomenon. Contingent non-enforcement occurs in exchange for favors from the infringing
¹⁹ population, as detailed in the extensive literatures on corrupt and clientelistic exchanges. However

⁴In fact, the concept of state capacity is often defined explicitly in terms of the state’s ability to enforce its laws; including by monopolizing coercive power (Weber, 1978) and expending resources to regulate relationships (Migdal, 1988). In a recent review of the concept, (Hanson and Sigman, 2021) define a state’s capacity as a function of its power, or in terms of Dahl’s pluralist ‘first face,’ the ability of the state to get its citizens to do things they would not otherwise do—the power to enforce its preferences (Hanson and Sigman, 2021; Dahl, 2007). See Cingolani (2013) for a detailed review of the intellectual history of the concept of state capacity. For a recent example of a study linking capacity with enforcement, see Cook and Fortunato (2023) on state legislative capacity and the enforcement of data provision laws.

⁵One example of this motivation finds that enforcement of traffic violations falls dramatically in the lead-up to sheriff’s races in the U.S.(Su and Buerger, 2024).

¹ in her seminal book Holland (2017) shows how politicians decline to *broadly* enforce certain
² policies targeting the poor in order to purposefully engage in “welfarist” forbearance (Holland,
³ 2016). Specifically in the context of Latin America, Holland (2017) shows how forbearance
⁴ towards squatting and street-vending violations depends on the importance of poor voters to the
⁵ ruling party. However, not all forbearance is so benevolent. In a more sobering example, Wilkinson
⁶ (2006) describes how Hindu politicians in India decline to call in law enforcement to suppress
⁷ militia violence against Muslim protesters during elections.⁶ Forbearance can also be conceptualized
⁸ in terms of regulatory capture⁷ on the *enforcement* lever, rather than the *policymaking* lever.

⁹ Building on these existing explanations for selective and incomplete state enforcement, I use
¹⁰ the term *connivance* to describe the strategic, selective enforcement of law-breaking through which
¹¹ the state ‘produces precarity’⁸ in the infringing population and renders it vulnerable to exploitation.
¹² Connivance—a coercive form of forbearance—occurs when the state is within its legal right (and
¹³ obligation) to enforce a law, but instead alters enforcement patterns *for the purpose of* expanding
¹⁴ opportunities for offenders’ exploitation. With connivance, I offer a framework for integrating
¹⁵ the concept of forbearance with the vulnerability imposed on offenders through the selective
¹⁶ enforcement of *de jure* policies. A core characteristic of forbearance is that it is revocable; that is,
¹⁷ a policy forbids the offending behavior, the state has the right to enforce it, and offenders believe
¹⁸ that sanctions are possible (Holland, 2016, 234). In concessional forbearance, revocability solves
¹⁹ the credibility problem of non-contingent clientelistic exchanges (Holland, 2016, 236). But this
²⁰ characteristic not only “cements political dependency” to the forbearing politician (Holland, 2016,
²¹ 236) but exacerbates precarity in the population violating the policy. This precarity is distinctly
²² *not* revocable. And while one of its advantages as a distributive tool is that it is used outside the

⁶In these accounts, the incentives to engage in forbearance are electoral. But concessional forbearance is not exclusively tied to electoral considerations. In a very different context—highly industrialized countries of Western Europe—Dewey and Di Carlo (2022) shows how states engage in ‘regulatory’ forbearance as a form of industrial policy; declining to enforce firm tax violations as a way to shape markets and favor specific producer groups (rather than groups of voters).

⁷Dal Bó (2006) defines regulatory capture broadly as “the process through which special interests affect state intervention in any of its forms” (Dal Bó, 2006)[203]

⁸Precarity here refers to “a situation lacking in predictability, security or material and social welfare” (A. et al., 2022)

¹ formal lawmaking system (Holland, 2016; Cunningham and Owens, 2020, 236), forbearance in
² fact requires a policy pass through this system before it is viable strategy. Revoking the precarity
³ produced by non-enforcement therefore requires a shift in policy, not enforcement—meaning this
⁴ vulnerability is intrinsic to forbearance as a strategy.

⁵ Connivance, then, rests on the precarity manufactured by states through the gap between law
⁶ passage and enforcement. These paired conditions of an existing law and the choice to not enforce
⁷ it create a dynamic closely captured by Lovett's conceptualization of domination, combining close
⁸ dependency and the arbitrary wielding of power (Lovett, 2001). It invokes a very specific type
⁹ of state power beyond its 'first face' (Lukes, 2021) and invites parallels with the concept of
¹⁰ coercive control in literature on domestic violence.⁹ Mann (1984) distinguishes between the state's
¹¹ power to impose mandates (despotic power) and to penetrate territory and implement policies
¹² (infrastructural power) (Mann, 1984, 190). Many studies of repression, even those encompassing
¹³ 'softer' and subtler forms, reference only the former; indeed Hassan, Mattingly and Nugent (2022)
¹⁴ explicitly define political control in terms of this despotic power of the state. The concept of
¹⁵ connivance instead focuses on how power is exerted through selective and instrumental use of the
¹⁶ state's infrastructural power in the presence of despotic mandates.

¹⁷ Crucially, while forbearance can be coercive in many ways, the concept of connivance requires
¹⁸ an intention to produce precarity in the infringing population such that it can be exploited by
¹⁹ the state or, in most cases, by its third-party state allies. An extensive literature shows how
²⁰ selective non-enforcement can empower criminal groups (Sobering and Auyero, 2019; Dewey,
²¹ 2012; Wilkinson, 2006), enable mass inefficiencies (Mahadevan, 2024), and produce negative
²² externalities such as environmental degradation (Dipoppa and Gulzar, 2023; Harding et al., 2024)—or
²³ as in the case of state failure to enforce the mafia's dumping of toxic waste across Southern Italy, all

⁹The concept of *coercive control* represents attempt to broaden understandings of domestic abuse beyond those of physical violence, and has been defined by Stark (2007, 228) as combined coercion—"the use of force or threats to compel or dispel a particular response" and control—"structural forms of deprivation, exploitation, and command that compel obedience indirectly" (Stark, 2007, 229) resulting in "a condition of unfreedom" (Stark, 2007, 205) experienced as entrapment.

¹ three (Walters, 2013; D’Alisa et al., 2010). Distinct from these cases, connivance does not simply
² imply non-enforcement against ‘bad’ actors or with ‘bad’ consequences. Instead, it requires that
³ the state purposefully look away, not from its criminal co-conspirators or key constituencies, but
⁴ from populations in whose vulnerability the state or its allies have a vested interest.

⁵ **Cui Bono? Third parties as profiteers of precarity**

⁶ If connivance is used to produce precarity and facilitate predation, understanding its use as a
⁷ strategy requires answering the question: Predation by whom? Who benefits¹⁰ from this coercive
⁸ forbearance? The majority of studies of forbearance focus on the relationship between the state and
⁹ the population being enforced (or not). And while connivance may be used to facilitate exploitation
¹⁰ by the state itself,¹¹ in most cases it requires evaluating the incentives and distributional consequences
¹¹ for more than just the infringing population and the state. I focus on the role of ‘third party’ actors,
¹² other than the state or the population directly affected, who are nevertheless interested in a law’s
¹³ enforcement. One straightforward example is those who profit off the illicit labor of lawbreakers,
¹⁴ such as mafias, militias, or exploitative employers.

¹⁵ How can ruling parties effectively foster and maintain relationships with such third parties?
¹⁶ Table 1 presents two dimensions along which state support to non-state allies can vary. Rows
¹⁷ represent whether the inducement is a positive expenditure (e.g., providing cash or contracts) or a
¹⁸ negative withdrawal of resources (e.g., revoking overburdensome regulations). Columns represent
¹⁹ the target beneficiary—whether it is an explicit allocation towards the ally in question, or is targeted
²⁰ towards other, related actors such as the ally’s competitors, labor pool, or consumers.

²¹ A great deal of work in the social sciences focuses on the first column, in particular direct
²² support from the state in the form of bribery or other incentives. However recent work from

¹⁰‘Cui bono’

¹¹Such is the case of ‘blackmail’ as described by Mares and Young (2019), in which state actors fail to enforce tax nonpayment by traveller populations in Hungary and Romania. While on its face a concession, these same state actors make a point to remind workers of their nonpayment in the lead-up to elections. In this case then, forbearance is not concessional but coercive. Its purpose is not conciliatory, but to foster vulnerability that can be exploited to gain votes.

	<i>Direct beneficiary</i>	<i>Indirect beneficiary</i>
<i>Expenditure</i>	Support (i.e., clientelism/corruption)	Burdening (i.e., regulation of competitors)
<i>Withdrawal</i>	Leniency (i.e., concessional forbearance)	Connivance (i.e., coercive forbearance)

Table 1: *Types of state assistance for a third party ally.* Columns are beneficiary type, and clarify whether the allocation is targeted towards the ally in question (making them a direct beneficiary), or towards other actors such as the ally’s competitors (making them an indirect beneficiary). Rows are inducement type, and represent whether the strategy involves the ‘positive’ extension of resources (expenditure) or the ‘negative’ removal of resources (withdrawal).

¹ Holland (2017); Cunningham and Owens (2020); Harding et al. (2024); Wilkinson (2006) and
² others focus on strategic *leniency*, where state actors support non-state allies by declining to
³ enforce their myriad offenses; respectively poor voters (engaged in squatting), specific sectors and
⁴ firms (engaged in tax evasion), construction contractors (engaged in deforestation), and communal
⁵ militias (engaged in electoral violence). With some groups however, states have an incentive to
⁶ avoid explicit or direct favoritism, especially when these groups are engaged in violence (Carey,
⁷ Mitchell and Scharpf, 2022). In such instances, the state may opt for indirect forms of support,
⁸ and avoid traceable resource transfers. A positive but indirect strategy of ‘burdening,’ may involve
⁹ targeting of rival groups with excessive regulation or repression. Finally, connivance refers to state
¹⁰ assistance characterized by a negative expenditure targeted towards an indirect beneficiary. It is
¹¹ a coercive form of forbearance, where the state facilitates exploitation—through inaction—at the
¹² behest of its non-state allies.

¹³ By considering the role of these exploitative third parties in enforcement decisions, I expand
¹⁴ the population of cases in which forbearance is plausibly purposeful and redistributive; but far
¹⁵ from concessional. I contend that connivance becomes an especially viable strategy when there is
¹⁶ significant state reliance on a extortionate third party who benefits from the enhanced vulnerability
¹⁷ of affected lawbreakers. Four stylized—and necessarily brief—potential examples of connivance
¹⁸ are laid out in Table 2.

Table 2: Empirical examples of connivance

<i>The Example</i>	<i>The Policy</i>	<i>The Actors</i>	<i>Connivance</i>
In the midst of an ongoing crackdown on undocumented immigrants, the U.S. federal government directed enforcement agencies to selectively pause raids and arrests; specifically targeting the agricultural industry.	A January 20, 2025 Executive Order by President Donald Trump “Protecting the American People Against Invasion” (The White House, 2025); Immigration & Nationality Act (INA).	Enforcer: The U.S. federal government; U.S. Immigration and Customs Enforcement (ICE); local enforcement partners Infringing population: Undocumented and unregistered immigrants in the U.S. Third party: U.S. agricultural employers	The abrupt shift of the U.S. administration’s ongoing mass deportation campaign towards <i>selective non-enforcement</i> of undocumented immigrants working on farms was aimed at avoiding “hurting industries...[President Trump] does not want to lose” (Aleaziz and Kanno-Youngs, 2025). President Trump himself acknowledged this in a social media post: “Our great Farmers...have been stating that our very aggressive policy on immigration is taking...workers away from them, with those jobs being almost impossible to replace” (Aleaziz and Kanno-Youngs, 2025). In lobbying for this position, U.S. Secretary of Agriculture Brooke Rollins reportedly explained that “farmers rely on immigrants to work long hours”(Pager et al., 2025), because they “cannot find Americans willing to do the physically onerous work” (Pager et al., 2025). By exercising forbearance only selectively, against immigrant farmworkers, the Trump administration has engaged in connivance—consolidating the vulnerability of undocumented workers in the interest of agricultural employers who rely on the “cheap and disposable labor” (Ngai, 2014, 3) made possible by these workers’ precarious legal position. While enforcement writ large continues, selective non-enforcement was exercised against the populations on whose labor Trump’s allies exploit.
Despite laws against settling on public land, officials in Nairobi, Kenya have selectively declined to enforce laws against the habitation of slums and sub-standard shacks and tenement buildings which proliferate in the city’s periphery.	Kenya’s Land Act of 2017 requires that authorities “shall issue a notice to the unlawful occupiers of public land to vacate the land” and evict tenants (Government of Kenya, 2017, Part VIII Section 64.(1), 2612)	Enforcer: Politicians, chiefs, and officials responsible for enforcing land grabs in Nairobi, Kenya Infringing population: Informal settlers on Nairobi’s public land Third party: Well-connected landlords overseeing construction and habitation in these informal slums	Urban land markets in Nairobi are “heavily influenced by what the local government does (or fails to do)...and the extent to which these rules are applied and for whom they are enforced” (Earle and Grant, 2019, 2). Landlords in Nairobi profit enormously from illicit settlements erected on public land: as noted by (Mwau and Sverdlik, 2020, 487), “hazardous informal shelter in Nairobi can be highly lucrative,” with up to four times higher returns than middle- and high-income housing. Kenyan authorities selectively engage in a “a lack of approval or enforcement...a “silent” policy of enabling tenement production,” (Mwau and Sverdlik, 2020, 495) which enables landlords and structure owners to extract from the city’s most vulnerable communities in the form of from informal rent, illicitly collected. Political appointees, including chiefs, have private interests which “can mean some criminal activity is tolerated due to the income it generates” (Price et al., 2016, 15). In addition to corrupt kickbacks, tenants often maintain coercive patronage relationships with structure-owners and landlords in which tenants are expected to support “the political interests of the latter” (Rigon, 2015, 2768). A strategy of connivance by local officials towards residents of informal areas run by their allies benefits these specific landlords and structure-owners, who both pass on profits and votes to (non-)enforcers.

<i>The Example</i>	<i>The Policy</i>	<i>The Actors</i>	<i>Connivance</i>
The Colombian government has failed to consistently enforce artisanal and small-scale mining (ASM) of gold along the pacific coast, especially in territory populated by indigenous and Afro-Colombian populations.	Decree 2235 in 2012 instructed police to destroy untitled mining operations, effectively criminalizing long-informal ASM operations (Mindefensa, 2012).	Enforcer: Colombian law enforcement agencies Infringing population: Informal, small-scale subsistence gold miners (<i>barequeros</i>) Third party: Armed actors including Ejército de Liberación Nacional (ELN)	Between 80% and 90% of gold in Colombia comes from seldom-enforced ASM (Massé and Munevar', 2017). This “uneven and insufficient” enforcement (Martínez-Fernández, 2019, 7) is not due to an inability to locate mines, which often operate in full view of authorities (Massé and Munevar', 2017; ABColombia, 2012). Instead, it is a function of the selective complicity of local authorities towards armed groups who extort illicit mining operations for profit (Massé and Munevar', 2017). A common complaint is that “all the world sees [them] coming and going...except public officials” (Translated by author; Massé and Camargo', 2012, 41), and security forces have been accused of withholding protection of miners in order to facilitate their extortion (Massé and Munevar', 2017, 17). Gold has surpassed cocaine as the main source of armed group revenues in Colombia (Berg, Ziemer and Kohan, 2021), and the ELN in particular relies on its highly profitable extortion—not operation—of illegal mines (Massé and Munevar', 2017; Martínez-Fernández, 2019). These Colombian authorities are engaged in connivance: perpetuating, through selective non-enforcement, the vulnerability of local miners in order to enable their exploitation by armed groups in return for kickbacks.
City officials in Madrid, Spain failed to enforce extensive illicit electricity hookups by residents of a specific area of the city: Cañada Real, often called “Europe’s largest shantytown” (Gil, 2022), a 15km strip of informal housing on public land to the city’s Southeast.	Spain’s Criminal Code Chapter VI Subchapter 3 Article 255 prohibits illegal connections to the electrical grid, and provides guidelines for penalties and enforcement of the offense (Ministerio de Justicia, 2016)	Enforcer: Law enforcement officials in the Autonomous Community of Madrid, Spain Infringing population: Residents of Cañada Real informal settlements with illicit connections to the electrical grid (the majority) Third party: Commercial real estate developers in Madrid	In 2020, electricity supplier Naturgy cut off electricity supply to Cañada Real, forcing most residents—largely migrants, most of whom are children—to abandon their powerless homes (Jones, 2021). At the time of the shutdown, only 4 out of 1,800 total connections to the electrical grid in the settlement were legal (Duran, 2021). The state’s restraint cannot be attributed to lack of capacity (EFE, 2021), popular will or ignorance of the situation (EFE, 2020), ambiguous jurisdiction (Sánchez, 2022) and certainly not a desire to appease the population of the neighborhood (Braddock, 2021; Quesada, 2023). A representative for Cañada Real argued in a complaint that the state’s restraint was a form of “direct coercion for the most vulnerable to abandon their homes ... [with the purpose of benefiting] the urban developments of southeast” Madrid (Marín, 2022). The beneficiaries of this state restraint certainly included urban real estate developers, whose sprawling developments in eyesight of the settlement have been continually stalled due to complaints from residents (Montaner, 2021). Benefiting too is Madrid Community President Isabela Díaz Ayuso’s administration, whose reelection campaign proposed ambitious urban development projects and expansions to Madrid’s Metro System southeast, in the area surrounding the settlement (Comunidad de Madrid, 2022). In declining to enforce its own policies, her administration engaged in connivance, ultimately passively displacing the population in order to benefit its own political agenda.

1 It is beyond the scope of this article to lay each case in Table 2 out in its full complexity
2 and detail, but examining key differences and similarities is instructive. First, there is important
3 variation between these cases. These four examples take place across multiple regions of the world,
4 from generally high-capacity and income contexts such as the U.S. and Spain to the middle- and
5 lower-income states of Colombia and Kenya. The third-party actors vary as well; from industrial
6 farms and urban developers in the West, to armed actors in Latin America, and landlords in Africa.
7 The ignored infraction in each case varies widely as well, encompassing illicit electricity hookups,
8 border crossings, informal settlements, and artisanal mining.

9 What remains consistent however is the nature of the relationships between the three primary
10 actors featured in each case, as summarized in Figure 1. Connivance requires that the enforcing
11 actor actively restrains from enforcement, either in part or in total; this feature it shares with more
12 traditional ideas of forbearance as described by Holland (2016) and others. The key distinguishing
13 feature of connivance is the introduction of a third-party interest, with its own relationships to
14 the enforcer and the infringing population. In a framework of connivance, the third party has
15 some form of alliance with relevant state actors, be it through the provision of votes, bribes, or
16 other forms of support. Moreover, the third party also has exploitative power over the infringing
17 population, and stands to benefit from their precarity. It is this exploitative relationship which
18 makes connivance a potent tool for state actors seeking to strengthen or maintain their alliance
19 with its third-party allies.

20 There are other commonalities between the cases which help elucidate the advantages of
21 connivance either as a substitute or complement to other strategies of state support (see Table
22 1). Just as concessional forbearance is cheaper than welfare through official state programs such as
23 social security (Holland, 2017), connivance is cheaper than bribes, physical repression, crackdowns,
24 or evictions. It is also notable that each of these stylized cases take place in more or less democratic
25 contexts. This is not necessarily a strict scope condition, and it is beyond the scope of this article to
26 definitively state democracy's role in enabling connivance as a strategy. However, democracy may

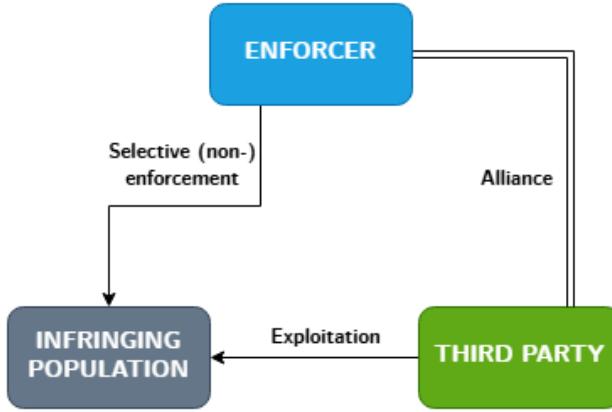


Figure 1: Relationships between key actors in connivance

¹ foster the relationship between the state and third parties which is itself a necessary pre-requisite
² for connivance. When the state is not all-encompassing or autocratic, parties and politicians may
³ increase their reliance on non-state allies in their quest to gain and maintain power. Moreover,
⁴ democracy may foster a situation in which state actors are unwilling to be seen as actively favoring
⁵ specific third parties over others, making connivance—an indirect and ‘negative’ strategy, as described
⁶ in Table 1—attractive. This is especially true in contexts where the third party is itself violent or
⁷ illicit (Carey, Mitchell and Scharpf, 2022).

⁸ Another similarity between the cases is the scope, if not the nature, of the infringement for
⁹ which the state exercises forbearance. All of these are vast informal enterprises which are likely
¹⁰ difficult to enforce entirely in the first place. The state’s ability to plausibly disguise enforcement
¹¹ variation as a state capacity issue may make it easier for the state to portray these patterns as
¹² benign. Just as perceptions of capacity matter for deterrence in a positive sense, perceptions of low
¹³ state capacity may help states ‘get away with’ connivance. [?] show that when local governments
¹⁴ suspected of Mafia infiltration in Italy are dismissed, the resulting fall in petty crimes is due
¹⁵ less to high capacity for enforcement, but to perceptions that the state is more powerful than
¹⁶ it is. This mechanism can work in the other direction as well. Finally, when the populations
¹⁷ responsible for law violations are an outgroup, or already vulnerable—as is true in every case

¹ described above—connivance can, in a phrase, ‘kill two birds with one stone,’ and foster increased
² control over suspect populations, while simultaneously depressing their ability to advocate publicly
³ for themselves. Not only are state actors able to provide assistance to valuable allies, but such
⁴ vulnerable populations are already collectively weak and often socially ostracized, reducing the
⁵ likelihood of societal backlash.

⁶ The remainder of this article explores a fifth example: the use of connivance by the state
⁷ government of Lagos, Nigeria through the selective non-enforcement of the state’s 2022 ban on
⁸ okada motorcycle taxis.

⁹ **2 Connivance in the Lagos transport sector**

¹⁰ In what follows, I argue that the state government of Lagos, Nigeria engaged in a strategy of
¹¹ *connivance* in its selective enforcement of a ban on ‘okada’ motorcycle taxis which began in
¹² mid-2022, as illustrated in Figure 2. The state (enforcer) chose to enforce the ban at such a
¹³ time—and in such a way—so as to increase the vulnerability of okada operators (infringing population)
¹⁴ and ease their exploitation by a key ally of the state: The Lagos transport Union (third party). This
¹⁵ Union operates a complex system of coerced taxation through controlling motorparks along Lagos
¹⁶ roadways, where they extort passing informal commercial transit operators. Around elections, the
¹⁷ Union also mobilizes voters and commits violence in support of Lagos’s ruling party. With an
¹⁸ election six months away, and okada riders in the midst of mobilizing against Union extortion,
¹⁹ the state acted. The strategic timing and geography of the ban’s enforcement not only dulled the
²⁰ existential threat okada riders posed to the Union, but forced riders into motorpark-laden areas and
²¹ therefore increased the Union’s opportunities to exploit them.

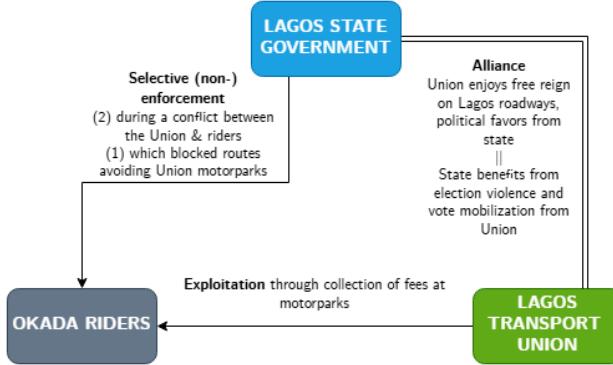


Figure 2: Relationships between key actors in connivance: The case of the Lagos okada ban

¹ ‘Eko traffic’ and the okada ban

² Lagos is Nigeria’s (and Africa’s) largest city, and the center of power in the southwest of the
³ country—almost the furthest southwest you can get without crossing the border into Benin, or
⁴ wading into the Gulf of Guinea (see Figure 3a and 3b). Lagos houses 10% of the country’s massive
⁵ and rapidly growing population, crammed into only 0.5% of its physical landmass. And though it
⁶ is no longer the federal capital¹² it remains a crucial economic and social powerhouse, and its state
⁷ politics consistently reverberate to the federal level.

⁸ Like the country, there are strong and evident political geographies in Lagos. And like the
⁹ country, Lagos wealth and (formal) employment is densely concentrated in the south, especially
¹⁰ on small islands near the port. Meanwhile the vast majority of residents live on the mainland. The
¹¹ islands have the marina, the beaches, the city’s main tourist attractions, the fanciest hotels, and the
¹² multinational corporate headquarters. But as in many major cities, the residents of Lagos generally
¹³ live and work in geographically separate areas.

¹⁴ Lagos’s economic geography means that Lagos traffic is infamous; a trip of ten kilometers
¹⁵ can easily take two or three hours, and sometimes up to eight.¹³ Public transportation options

¹²Negotiations between the country’s highly disparate North (the country’s most populous and poorest region, largely Muslim and Fulani or Hausa) and South (largely Christian and Igbo in the Southeast, and Yoruba in the Southwest) in the early 1990s moved the capital to the more centrally located and neutral planned city of Abuja

¹³‘Eko Traffic’ (Lagos Traffic) is the name of a cocktail I enjoyed at three different bars across three different neighborhoods of Lagos in the summer of 2023, it is so ubiquitous of a joke. While the precise recipe differed from



Figure 4: Photo: Okada riders, Lagos. Image sourced from <https://punchng.com/lagos-renews-campaign-against-okada/>

1 However on May 18, 2022 Lagos State Governor Babajide Sanwo-Olu held a press conference
2 in which he announced that starting on June 1, his administration would begin enforcing a ban
3 on okada motorcycle taxis originally passed in 2018 (Adelagun, 2022; Osazuwa, 2017). The
4 announcement of the ban's enforcement provoked high anxiety across Lagos, as individuals feared
5 its effect on their already intolerably long and expensive commutes (Enimola and Joseph, 2022;
6 Ige, 2022; Mosadioluwa, 2022; Enimola, 2022). However, the state government stood firm; according
7 to Governor Sanwo-Olu and various spokespersons, the ban was necessary for crime prevention
8 (Adelagun, 2022; Reporters, 2022), traffic control (Adekoya, 2022), and for reducing environmental
9 pollution and degradation across the state (Badmus, 2022). But there was one curious aspect of
10 the 2022 Lagos okada ban's enforcement: namely, its selectivity. While the ban being enforced
11 technically applied to all major roads across Lagos State (Lagos State of Nigeria, 2018), the
12 government in 2022 detailed a specific list of areas in which it intended to enforce the ban (Reuters,
13 2022).¹⁴ This left many parts of Lagos under the letter of the law, but outside the scope of state
14 enforcement.

¹⁴See SI §section B.4 for more details on the technicalities of this ban, as well as on the history of okada bans within and beyond Lagos.

1 The okada ban is a prime example of a traffic restriction which maps neatly onto pre-existing
2 socio-political cleavages. Many okada riders are from Muslim and Hausa or Fulani-majority
3 northern states who have traveled to Lagos in search of work, often as a result of conflict and
4 climate change in the country's North. In particular, there exist differences in ethnic group, region,
5 language, and religion between native Lagosians and many okada riders, which adds a communal
6 dimension to the ban's enactment. News reports on the ban tend to be tinged with inter-ethnic
7 suspicion, often conflating okada riders with perpetrators of jihadist violence in the North such
8 as Boko Haram (Odunsi, 2022), as well as crime writ large (Mosadioluwa, 2022; Tribune Online,
9 2022). Riding an okada is an extremely vulnerable occupation and is viewed with hostility and
10 suspicion by many Nigerians, even as they make daily use of the services these riders offer.

11 So why did the Lagos government selectively enforce a controversial and restrictive policy
12 such as the okada ban? I argue that understanding selective enforcement of the okada ban requires
13 understanding the interests and incentives of not only the Lagos State Government and okada riders
14 themselves, but additional interested third parties. In particular, I argue that the state selectively
15 enforced the okada ban in order prop up its preferred faction of Lagos's transport unions amidst an
16 internal power struggle, through indirectly increasing the Union's ability to engage in extraction.

17 **The third party: Touts and transit policy in Lagos**

18 In the late 1970s, as the population grew and Lagos traffic steadily became more congested, the
19 burgeoning informal transport industry began to organize. Given the rising transportation needs of
20 the city, the sector became a primary growth industry—but also one where extractive actors stood
21 to make enormous profit and gain enormous power.

22 Today, transport unions are commonplace: extractors from commercial drivers, rather than
23 representative unions *of* drivers. Transport unions set up shop along the Lagos road network, in
24 areas called 'motorparks,' where they charge every commercial driver that passes a ticket for usage
25 of the road. The money unions collect from drivers does not go towards the improvement of public

1 infrastructure, or to the state at all, and Lagos drivers and riders see no assistance or benefits from
2 ‘their’ union. The unions monitor, manipulate, tax, and coerce commercial drivers of all types
3 of vehicles on a daily basis; in this way, it is perhaps as accurate to describe Lagos transport
4 *mafias*, rather than transport unions. It is estimated the unions bring in many billions of naira
5 every year through their extraction from drivers,¹⁵ and over its roughly 50-year existence, these
6 mafias have become some of the most powerful players in Lagos politics. This is in part because
7 ground-level ‘bureaucrats’ of the unions—also referred to as ‘touts,’ ‘agberos,’ ‘area boys,’¹⁶
8 ‘thugs,’ or ‘hoodlums’—are crucial tools for politicians around election time for the purpose of
9 voter intimidation and coercion (Fourchard, 2023). Agbiboa (2022) provides a succinct summary
10 of the most powerful branch of the transport unions:

11 “Consider the NURTW branch in Lagos. Founded in 1978, the NURTW
12 constitutes the primary support base for the Lagos state governor during
13 election campaigns. The state is often unwilling or unable to rein in the
14 union’s predatory treatment of its workers. The Union routinely engages
15 in patronage politics and voter mobilization to support various parties and
16 candidates in return for permission to levy taxes on informal transport operators
17 in public spaces. ‘The NURTW is a law unto itself,’ said a danfo driver.”
18 (Agbiboa, 2022, 179)

19 During elections, Union touts—at the behest of their powerful, well-connected chairmen—block
20 voting places, beat voters of the opposition party, participate in rallies, and perform other services
21 for the dominant All Progressives Congress (APC) political party.¹⁷ The rest of the time, they set
22 up fiefdoms in motorparks and adjacent bus terminals, their primary points of extraction, where
23 they collect their daily wares from informal commercial transport operators (see Figure 5). Union

¹⁵See investigative reports from the International Centre for Investigative Reporting (ICIR) Nigeria, including Odinaka Anudu’s “Money for the boys: How ‘agberos’ pocket billions of Lagos transport revenue.” ICIR. <https://shorturl.at/pho6a>

¹⁶I use these terms interchangeably for the sake of simplicity; but there are some subtle differences in their local use. In particular, ‘agberos’ sometimes implies the area boys who have ‘made it.’

¹⁷See SI §B.2 for more on the APC and the political context of Lagos and Nigeria.

¹ touts are a source of constant strife for many commercial transport operators, who find themselves
² at the whim of these henchmen on a day-to-day basis.

³ Union-led motorparks are thriving centers of commerce and chaos. They vary in their formality;
⁴ some are large tracts of land along the side of a road, surrounded by a fence; others are less
⁵ officially delineated, and seem to consist largely of the people and vehicles stretching across the
⁶ roads and spilling over the side, as opposed to any physical ‘park.’ Often, the parks and their
⁷ personnel (touts, drivers) stretch across multiple lanes of a road, blocking or slowing passage
⁸ of all passing vehicles, and extracting ‘ticket fares’ from all commercial drivers who pass. The
⁹ conglomeration of passengers boarding and alighting, drivers passing through, and touts milling
¹⁰ about collecting their dues means that motorparks and their surrounding areas are packed with
¹¹ people. The masses in turn attract street vendors and hawkers, so many motorparks also feature
¹² makeshift markets of stalls selling grilled maize, apparel, or gin in small hot sauce-like packets.
¹³ Representatives of various branches of law enforcement are very often spotted ‘hanging out’ in the
¹⁴ park, chatting to touts, admonishing drivers, or simply surveying the scene.

¹⁵ ‘Omo Area’ (Area boy), or agbero¹⁸ culture is a distinct Southwestern Nigeria phenomenon,
¹⁶ sharing some similarities with gang culture. It connotes the thousands of (mostly young, mostly
¹⁷ male) individuals crowding bus stops, stations, and motorparks demanding money from, cajoling,
¹⁸ and harassing commercial transportation drivers, often violently—the foot-soldiers of transport
¹⁹ unions. Agberos are there to ensure that every commercial transport driver who comes through the
²⁰ park that day buys a ticket.

²¹ Lagos transport unions do not fit neatly into the typologies social scientists have offered categorizing
²² non-state, third-party actors in politics. While they are a union in both name and technically
²³ in function, using this label may fail to communicate the real behavior and intention of these
²⁴ groups. They are sector-constrained interest groups; they don uniforms, collect dues, and any
²⁵ litigation by or against them is handled by the industrial courts of Nigeria rather than other judicial

¹⁸Agberos comes from Yoruba and the literal translation refers to someone who beckons or herds (agbo) a group of passengers onto a bus (akero).



Figure 5: Example of Lagos motorpark (photo by author, July 10, 2023)

systems. However, they are largely unrepresentative of the workers they claim to represent; and
2 do not engage in collective bargaining with employers on behalf of transport operators. And while
3 violence is certainly not foreign to union activity around the world, their particular relationship
4 with violence and with the political machine of Lagos make them unique. For these reasons,
5 the transport unions also share some similarities with criminal organizations and gangs, most
6 commonly explored in the context of Latin America (Lessing, 2024; Arias, 2017; Blattman et al.,
7 2025; Feldmann and Luna, 2022) as well as the mafias of southern Europe (Dipoppa, 2025;
8 Calderoni, 2011). Namely, Lagos transport unions exert occasional fatal violence and frequent
9 extortive violence over civilian actors, and maintain control over particular industries as well as
10 pockets of urban territory. However, while the transport unions engage in a variety of activities
11 that are not by any means strictly legal; their main source of income is not from protection
12 rackets, drug smuggling, human trafficking, illicit natural resource or mining operations, or any
13 other inherently criminal enterprise. They are formally recognized interest groups whose income

¹ comes from the collection of worker fees and tickets to pass through motorparks—higher and more
² frequent fees than necessary perhaps, and with limited services offered in return, but nevertheless
³ a legal and legitimate source of income. They also share some commonalities with paramilitary
⁴ groups and pro-government militias (Carey, Mitchell and Scharpf, 2022; Acemoglu, Robinson and
⁵ Santos, 2009). Like these groups, Lagos transport unions can and do influence elections through
⁶ interference and violence, and there certainly exists a symbiotic relationship between politicians
⁷ and transport unions. However unlike paramilitaries, these groups' primary function is certainly
⁸ not broad or targeted political violence, and they are not, in fact, 'armed' groups in any sense of
⁹ the word. It is highly atypical for members to carry firearms, for example.

¹⁰ **Argument and hypotheses**

¹¹ My contention is that the selective enforcement of the Lagos okada ban—over both time and
¹² space—is an example of *connivance*, and its patterns are motivated by the interests not of the
¹³ lawbreakers themselves (okada riders) but of exploitative third parties on whom the state relies
¹⁴ (Lagos transport unions). The okada ban's incomplete enforcement served to 'produce precarity'
¹⁵ in riders whose livelihoods were made illicit by the ban, making it more difficult for the riders
¹⁶ to organize in opposition to their exploiters.¹⁹ I contend that *variation* in enforcement of the ban
¹⁷ across areas in Lagos served to displace riders to Union-led motorparks, where they do most of
¹⁸ their extraction. The state therefore selectively enforced policies aimed at okada riders in order to
¹⁹ ease their exploitation by the Union, the state's third-party ally. My principal assertion, then, is
²⁰ that:

²¹ **Argument** The Lagos State Government varies enforcement of the okada ban in order to appease
²² the Lagos transport Union, riders' primary extractors.

²³ With the okada ban, as with many laws, enforcement is not all or nothing: there was neither

¹⁹SI §E introduces a simple sequential model formalizing this logic. See SI §§ for a more detailed qualitative analysis of okada rider organization in the lead-up and aftermath of the ban.

¹ a completely absent nor a completely comprehensive enforcement regime. In fact, I contend that
² there is strategic variation in enforcement which neatly aligns with Union interests. Specifically,
³ my theory predicts that enforcement patterns will vary *spatially* according to where Union extraction
⁴ takes place, and *temporally* according to the Lagos State Government's reliance on the Union to
⁵ maintain power. This leads to two more specific hypotheses regarding variation in enforcement:

⁶ **H1.** Enforcement of the okada ban varies *spatially* so as to displace okada into Union-controlled
⁷ motorparks.

⁸ **H2.** Enforcement of the okada ban varies *temporally* according to state perception of Union
⁹ reliability.

¹⁰ In what follows, I describe how the state government selectively enforced the okada ban in
¹¹ order to assist the Union's ability to extract from okada drivers, and to address the looming threat
¹² these riders posed.

¹³ 3 Empirical Strategy

¹⁴ I probe these hypotheses across the three subsequent sections. In Section 4, I trace the law's initial
¹⁵ enactment and enforcement with data obtained through participant observation on Lagos
¹⁶ motorways and by interviewing Union members, okada riders, and political and law enforcement
¹⁷ officials; as well as with original data compiled by consulting archival newspaper reports. I show
¹⁸ that okada riders in the leadup to the ban were creating issues for Union leadership, culminating

¹⁹ In Section 5 I address hypothesis 1, or the *spatial* variation of the state's selective enforcement.

²⁰ I use original data collected on ban enforcement locations and motorpark locations, as well as
²¹ Very High Resolution (VHR) remote sensed imagery to count okada motorbikes on individual
²² road segments across multiple dates both before and after the ban's proposed enforcement. I show
²³ that not only did the number of okada (as expected) decrease slightly in enforced areas after June

1 2022, but increased substantially on road segments featuring union motorparks. I then employ a
2 detailed quantitative analysis of the spatial distribution of Union territory and state enforcement
3 areas to show that okada riders' displacement to motorparks was *uniquely* driven by the particular
4 selective enforcement areas chosen by the state, as compared to other possible enforcement areas.
5 Specifically, I transform maps of the Lagos road system into a connected spatial network. I then
6 simulate 10,000 routes between random points on the network, and show that blocking segments
7 enforced by the state disproportionately drive riders into Union territory. I then collect data
8 on other potential areas where enforcement might have been prioritized, based on Lagos state
9 government statements motivating the ban—including traffic congestion, pollution, income, police
10 presence, and crime—to construct potential *counterfactual* enforcement patterns. Running these
11 same simulated routes under these counterfactual enforcement patterns, I find that none of them
12 have the same effect of pushing okada riders into Union motorparks.

13 Finally, in Section 6, I address hypothesis 2, or the *temporal* variation in state enforcement. I
14 utilize an unexpected and close election result in the Lagos presidential elections of 2023, in which
15 former Lagos State Governor—and primary Union patron—Bola Tinubu of the APC lost in his
16 home state to upstart challenger Peter Obi of the Labour Party. I argue this close election reduced
17 confidence of the APC and the Lagos State Government in the Union's ability to sway elections on
18 their behalf. While the Union turned this around with rejuvenated efforts during the Gubernatorial
19 election about a month later (which was marked by heavily increased electoral interference, and
20 ultimately an APC victory for the incumbent Lagos governor), I show that this inter-election period
21 was marked by a significant decrease in enforcement by the state, in particular affecting the number
22 of okada present in motorparks according to VHR estimates. I treat the Presidential election as a
23 shock to the state's interest in fostering riders' precarity on behalf of the Unions, and argue that
24 resulting changes in enforcement during the inter-election period were due to perceived Union
25 defection during the Presidential election.

26 Before turning to this evidence, I briefly detail the original data collection supporting these

¹ efforts.

² Geolocated data along the Lagos road network

³ Locations of ban enforcement

⁴ I consult original policy documents to code the areas in which the okada ban was selectively
⁵ enforced in the period following June 2022. The government announced ahead of time the areas
⁶ of the state and city that it would actively enforce.

⁷ Examples of enforced areas announced by Lagos State Government 2022

⁸ Agege Motor Road/Oshodi Loop, Oshodi, Ikeja/Mushin Local Government
Mushin/Isolo Link Bridge, Mushin Oshodi Local Government
Dorman Long Bridge Surulere/Lagos Mainland Local Government
Ojuelegba Bridge, Surulere/Lagos Mainland Local Government
National Stadium Flyover, Surulere Local Government
Iganmu/Funsho Williams Bridge Surulere Local Government

¹⁰ The ban's (selective) enforcement went into effect on 1 June 2022. The state listed roughly 60

¹¹ stretches of road, highway, or specific bridges or flyovers; as well as over a dozen local government
¹² areas and specific councils and neighborhoods, in which the government would enforce the ban.

¹³ I hand-code these individually and merge this dataset with geospatial road data in Lagos²⁰ to get
¹⁴ a full universe of areas in which the state could choose to enforce, but with an indicator of where
¹⁵ they announced their intention to. I refer to these areas as 'intention to enforce' or ITE areas. This
¹⁶ is the primary dataset I use to map cross-sectional state enforcement (see Figure6 for a map of ITE
¹⁷ areas, marked with ×).

²⁰Data available at https://data.humdata.org/dataset/hotosm_ng_a_roads?

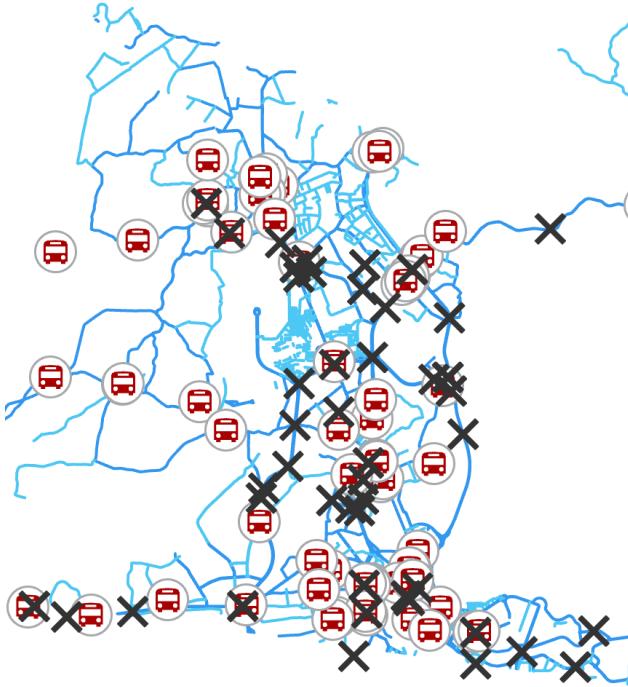


Figure 6: Motorpark and okada ban ITE locations in Lagos, Nigeria (mapped by author). Motorparks are represented by red bus icons; ITE areas are represented by ×.

¹ Motorpark locations in Lagos

² Measuring the geographic presence of motorpark operations is slightly more complicated; transport
³ unions in Lagos border between state-sanctioned and illicit, and their activities are largely informal.
⁴ Nevertheless, motorparks are palpably visible (see Figure 5 for an example of a motorpark). They
⁵ are also mostly permanent; while new parks sometimes do crop up, change hands, or split, the
⁶ general locations of major parks have mostly stayed the same over the decades.

⁷ I utilize a list in an Appendix of a transport statistics report produced by former Lagos Governor
⁸ Ambode's administration in 2019²¹. This strategy is useful in that it avoids any post-treatment bias,
⁹ as it was put together in the years before the okada ban's enforcement. I am left with a list of over

²¹See <https://mepb.lagosstate.gov.ng/wp-content/uploads/sites/29/2022/02/Transport-Statistics-2019.pdf>

¹ 100 motorparks, which represent the primary locations of Union extortion of okada riders in Lagos
² state. Figure 6 shows maps these park locations (represented by red bus icons).

³ **Satellite and Streetview-based informal transport density estimates**

⁴ Where are informal commercial transport operators—particularly okada riders—located? My
⁵ primary measure of this is constructed through analysis of very high resolution (VHR) remote-sensing
⁶ imagery. I combine measurements from publicly available Google Earth Historical Imagery and
⁷ supplementary, commercially available imagery from Airbus and Maxar VHR satellite images,
⁸ which have a spatial resolution of 30-50cm—meaning each pixel in the resulting geo-located
⁹ image measures less than half a meter ²². At this resolution, it is possible to distinguish individual
¹⁰ commercial vehicles, including okada motorcycles, and therefore to create measures of okada
¹¹ presence over space and time on the Lagos roadway. I utilize both manual coding and computer
¹² vision models to count of the number of motorcycles present per segment of road over time on
¹³ the Lagos road network. I supplement this with (temporally sparser but more precise) Google
¹⁴ Streetview data to establish baseline counts, as well as lower-resolution but more readily available
¹⁵ satellite imagery, for which I utilize methods of Bayesian hierarchical spatial disaggregation to
¹⁶ infer okada prevalence in images where the resolution is not high enough to visually distinguish
¹⁷ them manually, as described in SI §D.1. My VHR data consist of roughly 10 observations between
¹⁸ 2021 and 2023, covering a continuous 75km² stretch of Lagos; including roughly a quarter of
¹⁹ banned areas and motorparks in the state.

²²See §D.1 for more details

1 Interviews, participant observation, newspaper archives

2 Fieldwork and interviews

3 I spent about 6 months observing the aftermath of the ban while living in Lagos. My fieldwork
4 began in the summer of 2023 and continued in the spring and summer of 2024 and 2025. My
5 fieldwork²³ encompassed participant observation including in motorparks, at stops and intersections,
6 and in transit; as well as semi-structured and unstructured interviews with over 75 stakeholders,
7 ranging from the highest-ranking officials in the Lagos Ministry of Transportation and law enforcement
8 agencies, to well connected Union chairmen and logistics operators, to okada riders themselves.

9 Participant observation focused on observing the inner workings of the Lagos transportation
10 system as a curious passenger. I conservatively estimate that I took 400 separate trips through
11 Lagos traffic, to all corners of the city, including via taxis and Uber (as well as competitor Bolt),
12 informal transit such as keke, danfo, and okada, and public transit options such as the BRT.
13 Many of my most meaningful conversations were with drivers as we were in transit, whether
14 to the local supermarket or to a far-flung motorpark. I also spent time sitting in or walking
15 through busy bus stops, markets, and motorparks; observing the behavior of informal transport
16 operators, Union touts, and law enforcement officials. My sampling strategy was partly based on
17 convenience—where I believed I could safely go, or had a connection—but purposefully spanned
18 the entire geography of Lagos, from the northern border with Ogun state, to the marina on the Gulf
19 of Guinea, and from Ikorodu in the far northeast of the city to Badagry in the far southwest.

20 Beyond my participant observation, I also conducted 75 semi-structured interviews with multiple
21 types of relevant stakeholders. Interviews largely took place in person, at locations ranging from
22 government offices, police stations, in abandoned corners of motorparks or unregistered watering
23 holes down the road from them, and in the lobbies of immaculate hotels. Roughly half of these
24 interviews were with okada riders themselves. The other half were divided between Union officials

²³Approved by Yale Institutional Review Board (IRB) protocol #2000035418.

Type	Number
State and law enforcement officials	10
NURTW Union officials (all factions)	17
Okada riders	38
Other transport operators	10
<i>Total</i>	75

Table 3: List of interviews. I conducted interviews 75 with stakeholders between June 1, 2023 and the present while based in Lagos, Nigeria. The table denotes the number of interviews conducted across five broad categories of interviewees.

¹ across the organization’s complex hierarchy (about half), and the remainder consisted of a roughly
² even split between high-ranking policy and law enforcement officials, and other informal transit
³ drivers, private operators or logistics business owners and entrepreneurs (see Table 3).

⁴ **Newspaper and social media archives**

⁵ Utilizing archives of mostly English-language Nigerian daily newspapers²⁴ over the past three
⁶ decades, as well as social media content, official press releases and documents, and other archival
⁷ and primary material, I create a localized event and network dataset for Lagos, which combines
⁸ and links information on hundreds of relevant actors, places, and events, covering both recent
⁹ internal political dynamics of the Union and events such as passage and enforcement of the okada
¹⁰ ban, protests by vehicle operators, battles between rival Union factions and political factions, and
¹¹ interpersonal rivalries and relationships. These data allow me to analyze trends in the political
¹² dynamics between the Union, the Lagos State Government, and informal commercial transport
¹³ operators qualitatively.²⁵

²⁴See SI §?? for a list

²⁵In the text, I refer to these data collectively as Transport Operator & Union Tracking (TOUT) data for short. See SI §?? for more on the specific record-gathering process and other examples.

4 Evidence: The role of the Union in the ban's enactment

2 My argument requires analyzing the interactions between three actors: the Lagos State Government,
3 the Union, and informal transport operators in Lagos, particularly okada riders. I draw on newspaper
4 and social media ('TOUT') data as well as first-hand participant observation and interviews to
5 investigate whether and how:

6 **Argument** The Lagos State Government varies enforcement of the okada ban in response to the
7 interests of the Lagos transport Union, riders' primary extractors.

8 Politicking with gangsters: Motorparks and the power of a union

9 Because the Lagos informal transport sector is extremely lucrative, transport unions have over
10 the years developed rigorous hierarchies and distinct organizational cultures (Fourchard, 2023;
11 Agbiboa, 2022). The national level NURTW presides over state-level chapters, themselves affiliated
12 with sub-unions which represent particular types of transport workers. For example, affiliated
13 with NURTW are the Tricycle Owners and Operators Association of Nigeria (TOOAN) which
14 manages tricycle ('keke') riders, and the Motorcycle Operators Association of Lagos (MOALS)
15 which manages okada riders. The precise nature of the relationship between TOOAN, MOALS,
16 and NURTW chapters at the state level has been the source of considerable strife, as I detail in the
17 following section.

18 The foot soldiers of the Union are the touts working in the motorparks. Touts are there to
19 bring some level of organization to the chaos, mainly via collecting ticket fees from commercial
20 drivers who pass through— fees which have steadily increased in price over the last couple of
21 years.²⁶ At the time of the ban's enforcement, it seemed to range from around NGN 200 per ticket
22 to nearly NGN 700.²⁷ Area boys weave in and out of lines of commercial vehicles, often with

²⁶Interviews with author, including July 22, 2023

²⁷See SI §?? for an example of a motorpark ticket one okada rider showed me.

1 hands firmly on doors shouting back and forth with the drivers. Touts operate with remarkable
2 impunity, and often with the visible and tacit acknowledgment of state officials. At one motorpark
3 on the mainland, I was struck when I noticed a small group of Lagos State Traffic Management
4 Authority (LASTMA) agency officials—obvious in their yellow, red-trimmed uniforms—laughing
5 and looking on as a tout, yelling over the chaos of the park, chased after a slow-moving danfo
6 mini-bus, whipping the already-cracked windows with a long wooden rod, a wad of apparently
7 not-enough cash clutched in his other fist. Touts' jobs are violent and often dangerous; though
8 they are also important community members—more than once, an agbero kindly helped direct me
9 when I was lost.²⁸

10 Touts are presided over by ‘unit’ chairmen who rule over a particular junction, intersection, or
11 park. They in turn are directed by ‘area’ chairmen, who are directed by zonal chairmen; meanwhile
12 the chairman of the state-level chapter of the union exerts authority over the whole operation. The
13 work of chairmen at all levels is very political, dangerous, and delicate. When I interviewed one
14 unit chairman representing tricycle riders on the mainland, he spent 15 minutes (in Yoruba) at
15 the beginning of the interview interrogating my colleague, a Nigerian, about his background and
16 purpose for being there, in order to make sure that he was not a spy for another faction or for rivals
17 who sought the chairman’s coveted position. “Chairman work is very dangerous,” he explained to
18 me apologetically afterwards, “we have many enemies”.²⁹ There are often clashes, as rival factions
19 fight over control of particularly hotly contested motorparks.³⁰

20 However, the dangerous political maneuverings of Union leadership only escalate as they
21 ascend the ladder, and as the stakes get correspondingly higher. The chairman of the NURTW
22 Lagos branch (as of 2019) MC Oluomo is a longstanding ally of President (as of 2023) Bola
23 Tinubu, the first governor of Lagos after the fall of military rule. Tinubu has long exercised control

²⁸See SI §§? for social media posts on this particular issue.

²⁹Author interview, July 14, 2024

³⁰Indeed once, after one long afternoon of chatting to reticent touts in the back of a beat-up minibus parked in a motorpark in the North, a battle for supremacy over the park broke out between loyalists of the park’s general chairman and rival foot soldiers in a nearby neighborhood. About twenty minutes after I had vacated the park, the entire stretch of road leading to the roundabout was deserted, as touts battled with guns, cutlasses, bottles and stones.

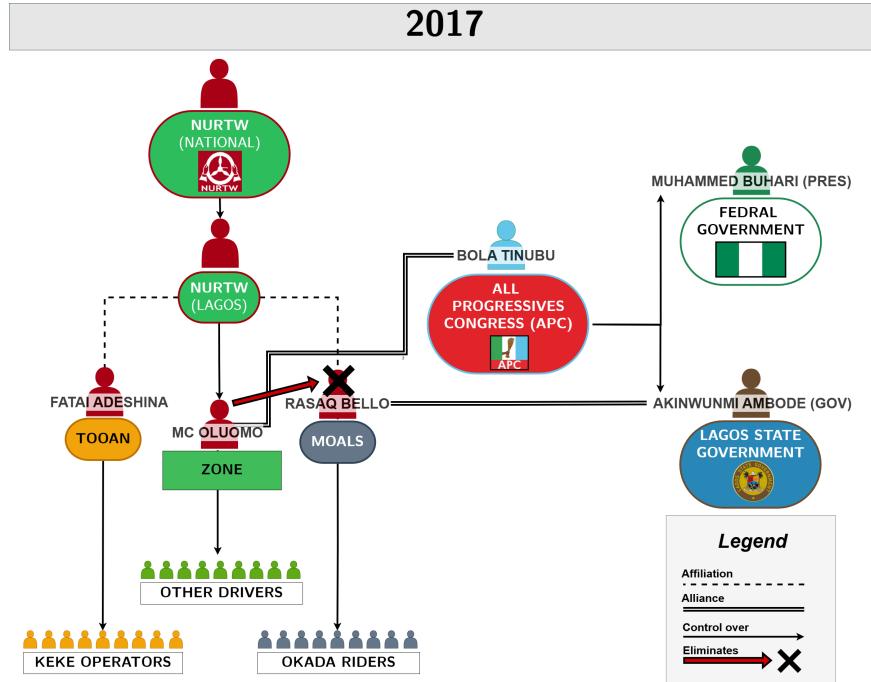
1 over the state politics of Lagos as a sort of “godfather.” Tinubu’s unique hold on formal institutions
2 in Lagos has its roots in his earlier life and in his Lagos governorship, in which he cultivated and
3 strengthened ties with informal networks of power such as the Union (Whiteman, 2013). Most
4 relevant for our purposes, Tinubu and MC Oluomo were both reportedly area boys in their youth
5 (Momoh, 2000, 188). During his governorship of Lagos, Tinubu was frequently referred to as the
6 “Area Boy Governor”, and is quoted by prominent journalist Kaye Whiteman as saying of the area
7 boys “These are my boys; I care for them.” (Whiteman, 2013, 217)

8 MC Oluomo’s rise to power in the Union was riddled with violence and with interference
9 from Tinubu, though this is not to underestimate his vast and loyal following. In particular, he has
10 historically been in conflict with drivers of keke and okada, and publicly rivals with the leaders of
11 two- and three-wheeler operations. When he took over as Lagos NURTW Chairman in 2019, then
12 leader of TOOAN (tricycle union) Fatai Adeshina³¹ was also vying for the top spot in the Lagos
13 NURTW chapter. Reports suggest Bola Tinubu personally called him and demanded he cede the
14 nomination to MC Oluomo (see 7).

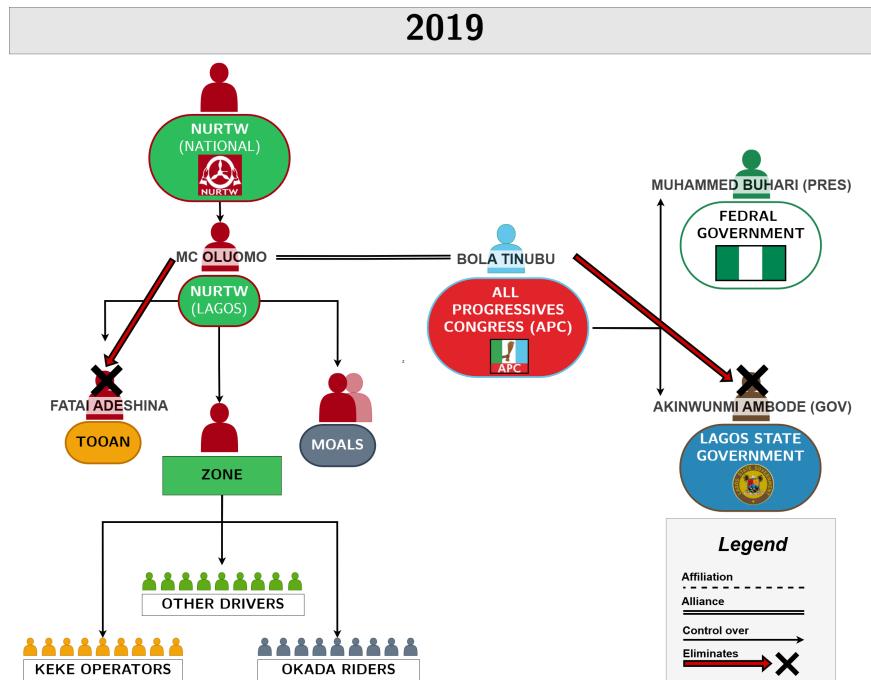
15 He fared better than his MOALS (okada union) colleague, however, who crossed paths with
16 MC Oluomo two years earlier. Rasaq Bello, who was rumored to support a different APC faction
17 than MC—and who operated in the same neighborhood, Oshodi—was shot to death in 2017,
18 almost certainly by MC’s men (Akinkuotu, 2019a; Akinsanmi, 2017; Odesola, 2022). The assassination
19 caused then Lagos State Governor Ambode to temporarily suspend NURTW from operating in
20 Lagos (Bankole, 2017; Akinsanmi, 2017), leading to an ambitious transport sector reform law
21 which attempted to regain state control over the industry (Infrastructure News, 2017; Olawoyin,
22 2017). Shortly thereafter, Tinubu, then leader of the APC, refused to allow the party to nominate
23 Governor Ambode for a second term (Akinsanmi, 2018; Abiodun, 2019), earning public support
24 from MC Oluomo on social media. Ambode’s reputation amongst people I spoke with remains
25 one of a man who got too close to disrupting big corruption, and whose second term candidacy

³¹Interviewed by author, July 10, 2023

Figure 7: Recent Union politics



(a) NURTW Lagos politics: 2017



(b) NURTW Lagos politics: 2019

¹ Tinubu halted as a result (Abiodun, 2019). MC Oluomo's role in this rearrangement of Lagos politics—albeit in a way that suited him quite well—is less clear. What is clear is that at least some Ambode supporters connected the dots. During an APC rally for Sanwo-Olu, the *new* nominee for governor to replace Ambode (and Lagos's current governor, reelected in 2023), MC Oluomo was stabbed by an assailant who reportedly blamed him for Ambode's ruined political fortunes (Egbas, 2019). MC's political allies allegedly ensured he got medical treatment abroad (Elezuo, 2019).

⁷ When MC recovered, Tinubu ushered him into the Union's top spot, over the incumbent chairman, an Ambode loyalist (Akinkuotu, 2019b; Inyang, 2019; Oyero, 2022b; Society Now, 2019).

⁹ As is evidenced throughout the decades, the relationship between Tinubu's APC and Oluomo's Union has been integral to the success of both. NURTW Lagos is not just thugs on demand for the APC, and the APC is not just the political pawn of the Union. Theirs is a complex and dynamic relationship; the internal politics of the APC are intricately related to the internal politics of the Union, and the political success of particular factions. They have separate but also overlapping goals, and their centers of power depend on each other—and, in the Union's case, on commercial drivers whom they extort for billions.

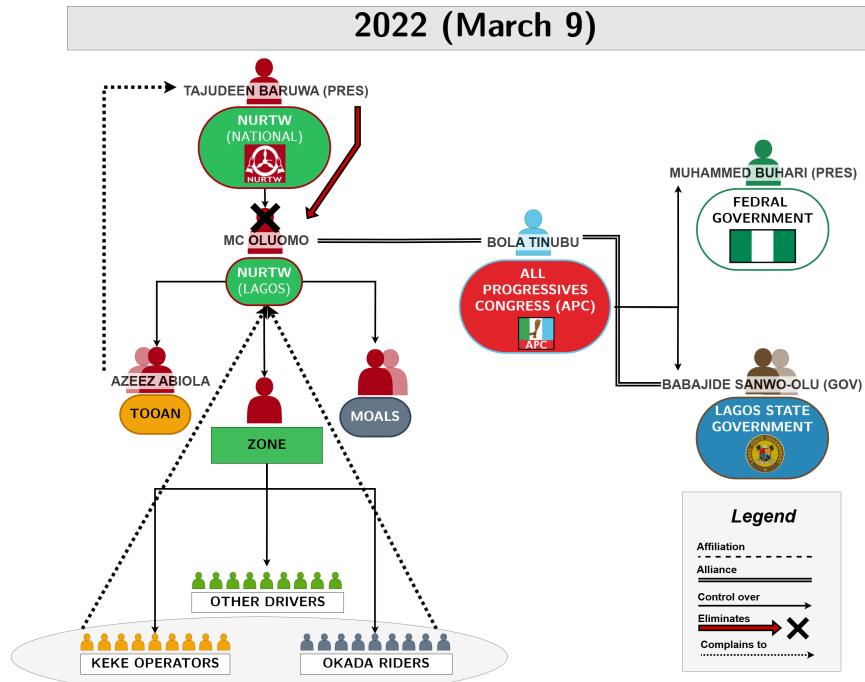
¹⁶ **The lead-up to the ban's enforcement**

¹⁷ Shortly after MC Oluomo overtook leadership of the Union, he made moves to integrate okada and keke operators directly under his control, usurping their existing unions—which had previously operated independently, if under the same umbrella (Oboagwina, 2022). As a result, okada and keke riders began to publicly complain about the rise in extortion at Lagos motorparks (Daily Trust, 2022b; Edema, 2022). In late 2021, this culminated in the then leader of TOOAN³² announcing that excessive extortion from MC Oluomo's NURTW meant that he was looking to form an alternative union of these operators, outside the purview of NURTW (Premium Times, 2022).

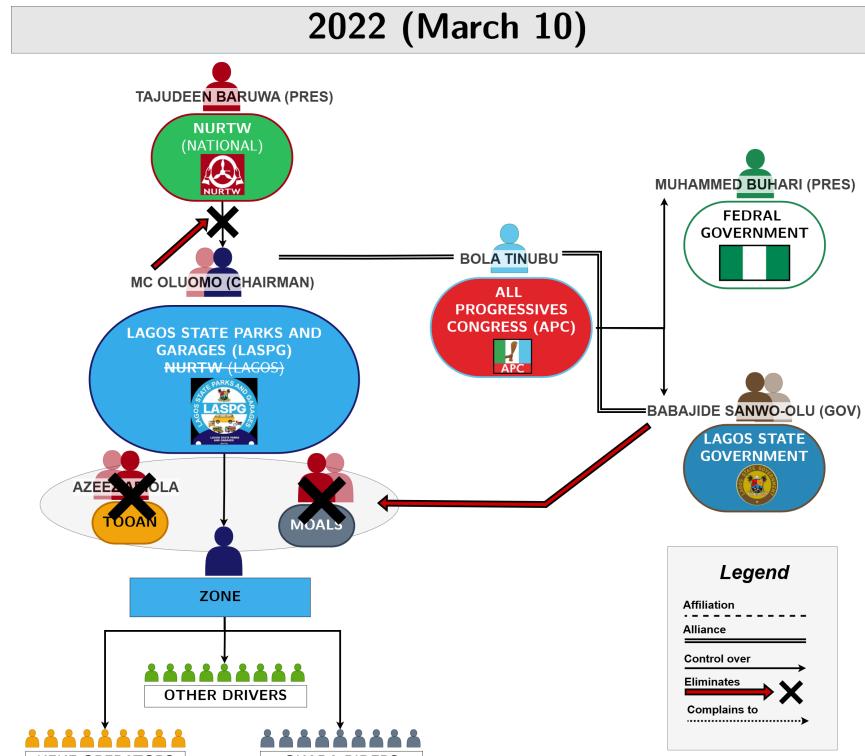
²⁴ He then went over MC Oluomo's head, directly to the national NURTW president, to complain of

³²Interview with author August 28, 2023

Figure 8: Internal union developments in the lead-up to the okada ban.



(a) NURTW political developments: 9 March 2022



(b) NURTW political developments: 10 March 2022

¹ the extortion from MC’s regime (Olaoluwa, 2022). The National President of NURTW demanded
² MC Oluomo work to prevent extortion of these riders (Olanrewaju, 2022). MC Oluomo refused
³ (Faith, 2022; Balogun, 2022). On March 9, 2022, the National NURTW umbrella suspended MC
⁴ Oluomo as chairman of the Lagos State NURTW Branch (Olaoluwa, 2022), and appointed in his
⁵ stead Fatai Adeshina, former TOOAN leader and MC Oluomo rival (?).

⁶ A rapid series of developments occurred in quick succession—in fact, all on the same day. MC
⁷ Olumo withdrew his membership from NURTW, and announced that not only was he still head
⁸ of Lagos State NURTW chapter, but that the chapter was breaking away from the national union
⁹ entirely (Akoni, 2022). That evening, Lagos Governor Sanwo-Olu announced that all parks and
¹⁰ garage management would now be under the purview of MC Oluomo’s Union—now to be dubbed
¹¹ Lagos State Parks and Garage Management (LASPG) (Alilyu, 2022; Society Now, 2022; Oyero,
¹² 2022b). They also announced that it was suspending the activity of the national union, NURTW
¹³ from the state entirely (Oyero, 2022a); leaving MC Oluomo’s faction as the only one operating
¹⁴ with the state’s blessing.

¹⁵ These developments did not unfold without protest. NGOs and rights organizations publicly
¹⁶ called out the role of the LSG in manipulating the law over the course of the leadership tussle;
¹⁷ one organization claimed that over 1,000 of these operators’ vehicles were vandalized by Union
¹⁸ members over the course of one week in April (Sahara Reporters, 2022). Multiple motorparks
¹⁹ were forcibly overtaken (Oyero, 2022c; Daily Trust, 2022a), extortion by touts continued (Agha
²⁰ and Aliyu, 2022), and violence erupted between area boys and okada riders that resulted in multiple
²¹ fatalities (Oyeleke, 2022; Lambo, 2022). And on May 8, a Justice of the Federal High Court in
²² Lagos ruled that MC Oluomo and his Union stop levying fees against riders (Titlola, 2022).

²³ After having promised MC’s men that he would “act on [their] grievances” (The Nation, 2022),
²⁴ Governor Sanwo-Olu announced a week later that his government would pursue imminent—and
²⁵ selective—enforcement of an okada ban, originally passed in 2018 (?). Okada riders and their
²⁶ tricycle-riding counterparts had long been a thorn in MC’s side, and they bore the brunt of the

¹ extortion from a union that they did not feel they were members of. They had begun the protesting
² excessive extortion in motorparks across the city, and in fact were at the root of MC Oluomo's—one
³ of the most powerful and connected men in Lagos—ongoing beleaguerment.

⁴ 5 Evidence: The political geography of selective enforcement

⁶ My theory requires that selective enforcement of the okada ban will vary in ways that benefit
⁷ the Union; allowing them to more effectively exert control over (and profit from) okada riders.
⁸ Therefore we should expect *spatial* variation in the ban's enforcement according to Hypothesis 1:

⁹ **H1** Enforcement of the okada ban varies *spatially* so as to displace okada into Union-controlled
¹⁰ motorparks.

¹¹ Hypothesis 1 implies two component hypotheses. First, it implies that enforcement along the
¹² ban's ITE routes *causes* more okada to be present at motorparks after enforcement commences:

¹³ **H1a** After the ban, the number of okada in motorparks *increases* compared to before the ban.

¹⁴ I test this assertion using a panel of satellite-based measures along road segments before and
¹⁵ after the ban. I show that okada density shifted in the aftermath of the ban; and that it shifted
¹⁶ differentially for (i) ITE areas vs. non-ITE areas; and (ii) for motorparks vs. non-motorparks.

¹⁷ Second, it implies that this choice of ITE areas is driven, at least in part, *in order to bring about*
¹⁸ this result:

¹⁹ **H1b** Enforcement areas (ITE) will increase okada density in motorparks *more* compared to other
²⁰ possible enforcement areas.

1 I use randomization inference and simulations on the Lagos road network to not only show that
2 enforcement along ITE areas increase okada presence in motorparks, but that it does so significantly
3 more than other plausible counterfactual enforcement patterns.

4 These propositions were validated by some news reports (Olasupo, 2022), as well as by many
5 of the conversations I had with okada riders throughout Lagos. When I asked riders specifically
6 whether the Union, allegedly their representatives, were troubled by the ban, I got more than one
7 laugh of derision. One rider in the southwest said:

8 “The ban … did not disturb [the Union] much. Why? Because if you cannot
9 go [on enforced routes], you go to the streets. That is where they collect
10 their dues … so while you struggle, you still give them their dues.” (Okada
11 rider, Badagry, 20 July 2023)

12 Every rider I spoke to opined about the Union’s increased collection, even though what they were
13 collecting on was technically an illegal activity. They often cited the Union’s collusion with the
14 state. In Ikorodu, northeast Lagos, one rider stated “… all money they are collecting is illegal …
15 they know that Lagos state has stopped bike. They stop okada. And they are *still* selling ticket, and
16 Lagos State did not stop them … they are selling it, and we are buying it.”³³

17 Other riders confirmed the increased fees and interaction with the Union faced after the ban:

18 “The problem is the ticket[s you must buy to pass through the motorparks].
19 You buy one, 200 naira. You buy another, 200… [Now] you have to
20 buy four, five, six tickets. You cannot now ride bike [without passing a
21 park]…So you have to go more, more, more—but then you pass more and
22 pay more.” (Okada rider, Mile 2, 23 June 2023)

23 Some riders mentioned to me how the okada ban’s enforcement shifted how things were done in
24 terms of ticket pricing. Some noted that their Union membership ID card no longer exempted them
25 from paying ticket fees.³⁴ Moreover, many riders mentioned how many new okada riders had been

³³Interview with author on June 12, 2023

³⁴Interview with author July 27, 2024

¹ crowded into pockets where okada riding was relatively safe from state enforcement, resulting in
² increased competition and rate depression.³⁵

³ Okada presence before and after the ban

⁴ **H1(a)** After the ban, the number of okada in motorparks *increases* compared to before the ban.

⁵ I begin by establishing that the ban was enforced as intended; namely by estimating the ban's
⁶ differential effect across enforced (ITE) areas, and non-enforced ITE areas. Using the satellite-based
⁷ measurement strategy described above, I run a standard two-way fixed effects model with two-periods,
⁸ which estimates the *differential* effect of the ban between ITE and non-ITE areas:

$$(1) \quad \text{Okada}_{it} = \gamma_i + \lambda_t + \tau(\text{ITE}_i \times \text{Post}_t) + \varepsilon_{it}$$

⁹ Where Okada_{it} is the number of okada counted in satellite imagery from date t on road segment
¹⁰ i . γ_i and λ_t account (respectively) for time-invariant, road segment (Edge ID) specific fixed effects
¹¹ and space-invariant, date-specific fixed effects. Given that each date represents a different satellite
¹² image, λ_t also act as controls for image-specific variation such as percentage cloud cover. ITE_i
¹³ represents whether i road segment is expected to be enforced (1) or not (0), and Post_t whether t is
¹⁴ before (0) or after (1) the ban's announced enforcement date. τ then represents the *differential*
¹⁵ effect of the ban's enforcement on ITE areas compared to non-ITE areas. Errors are clustered by
¹⁶ road segment.

¹⁷ Next, I test H1(a) directly by investigating the ban's differential effect, if any, on motorparks,
¹⁸ using the same satellite-based measurement strategy. I investigate the I run the same standard
¹⁹ two-way fixed effects model with two-periods as in Equation 1, substituting Motorpark_i for ITE_i .

³⁵ Interview with author on August 15, 2024

	Num. okada	
Model:	(1)	(2)
Post × ITE	-4.560*** (1.677)	
Post × Motorpark		2.991** (1.200)
<i>Fixed-effects</i>		
Edge ID	Yes	Yes
Time period	Yes	Yes

Clustered (Edge ID) standard-errors in parentheses

*Signif. Codes: ***: 0.01, **: 0.05, *: 0.1*

2 Unsurprisingly, and as shown in Model (1) of Table 5, the presence of okada significantly
 3 decreases in the aftermath of the okada ban in areas where there is stated enforcement, by about
 4 4 okada per road segment. Meanwhile, and in line with H1(a), the number of okada significantly
 5 increase after the ban on road segments which feature motorparks, compared to before (by about 3
 6 okada per road segment). This is consistent with the theory that okada are displaced after the ban's
 7 enforcement to areas controlled by the union via motorparks, enabling their increased extortion by
 8 the Union.

9 **Randomization on a road network**

10 **H1b** Enforcement areas (ITE) will increase okada density in motorparks *more* compared to other
 11 possible enforcement areas.

12 Above, I show that the number of okada increased around motorparks after the ban's enforcement.

1 Next, I contend that this was the result of a purposeful strategy of displacement by the state (H1b).
2 Figure 9 lays out a simplified example of the process by which this may occur.³⁶ The top panel (a)
3 shows an abstracted road map of a particular neighborhood of Lagos, with two points marked A
4 and B. Union-run motorparks are marked with red buses. Before the ban, if one attempted to travel
5 via okada from point A to point B along the road network, the quickest route goes straight north
6 and east (marked in green). Notice on this route, no motorparks are passed.

7 In the bottom panel (b), areas enforced during the okada ban—bridges, streets, and intersections—are
8 denoted with red lines. These represent areas which, after the ban, are untraversable by okada
9 riders. Adapting to avoid the enforced areas, the green line in this bottom panel shows the new
10 shortest route from A to B riding an okada. Not only is the route longer and more circuitous, but
11 the okada is now forced to pass two motorparks.

12 This figure represents the type of displacement I suggest occurred on a systematic scale after
13 the okada ban. It demonstrates that while okada can still traverse through many areas, even in a
14 neighborhoods directly affected by the ban's enforcement, the risk of arrest and bike confiscation
15 at key points on the road network forces riders to alter their routes, landing them in the hands of
16 Union touts who can exploit their precarity to extort them.

17 To investigate whether the chosen enforcement areas have this systematic effect, I built an
18 original *R* package³⁷ which uses Open Street Maps (OSM) geospatial road data to construct a
19 networked dataset, where roads ('edges') are connected by intersections ('nodes'). This allows me
20 to flexibly simulate travel through the Lagos road network under a variety of circumstances—including
21 before and after the ban took place.³⁸ This process transformed all Lagos streets and roads—approximately
22 18,000 km in total—into 304,207 edges separated by 320,460 nodes.

23 I complete the setup of the road network by merging it with the above-described motorpark
24 locations data, as well as enforcement locations named in the 2022 okada ban. I use this networked

³⁶Based, incidentally, on the route I took from my residence (A) to the supermarket (B) during fieldwork.

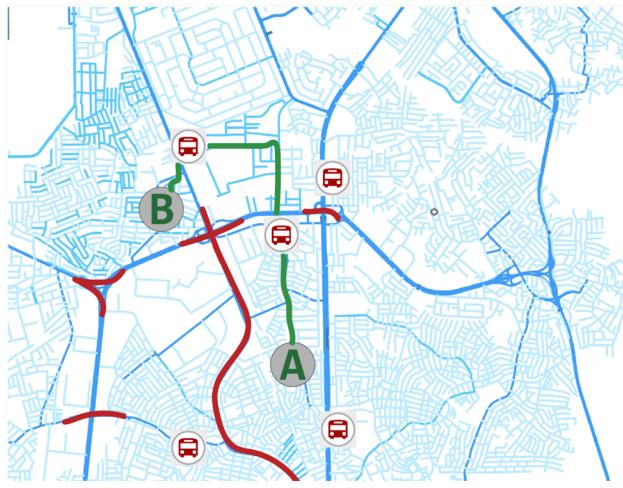
³⁷tRaffic, forthcoming on GitHub.

³⁸For more details on the technical approach of this package and how it is used in this approach, see SI §D.4.

Figure 9: Example of affected route. Illustration of how a okada rider's route may be affected by the okada ban.



(a) Example okada route (in green) from point A to point B, before okada ban enforcement. Union-led motorparks represented by red bus symbols. The route outlined passes no motorparks.



(b) Example okada route (in green) from point A to point B, after okada ban enforcement. Union-led motorparks represented by red bus symbols. The route outlined passes two motorparks.

¹ dataset to calculate features of the paths traversed by riders between multiple points. SI §D.4
² contains extensive details of this analysis process.

³ Results from 10,000 simulations provide suggestive evidence that the specific geography of the
⁴ okada ban enforcement served the Union’s interests by displacing okada routes into areas in which
⁵ the Union runs motorparks.

⁶ To begin with, analysis shows that okada riders face significant curtailing of freedom of movement
⁷ as a result of ban enforcement in ITE areas. Just over $\frac{1}{3}$ of randomly selected routes become
⁸ intraversable after the okada ban; that is, the okada ban prohibits passage from point A to point
⁹ B at all, by any route. Moreover, enforcement in these areas results in a significant increase in
¹⁰ cost. If we approximate that 1km of traversing the road costs NGN 100,³⁹ and each motorpark
¹¹ adds an additional NGN 500,⁴⁰ the average route’s cost after the ban’s enforcement is at least
¹² NGN 1,192.48 more than before the ban, either because of added mileage due to circumventing
¹³ ban enforcement, or because of passing extra motorparks. This is equivalent to nearly 12km extra
¹⁴ miles per route, or passing at least two additional motorparks on a given journey. While it is
¹⁵ unsurprising that no routes will see a *decrease* in cost,⁴¹ the scale of the impact is significant. Over
¹⁶ 75% of routes saw a more than NGN 500 cost increase, and the route with the maximum added
¹⁷ cost totals more than NGN 5,500, more than 50% higher than before the ban was enacted.

¹⁸ While it is evident that the okada ban’s enforcement resulted in significant costs to okada riders,
¹⁹ my primary analysis concerns the degree to which the Union benefited from the displacement of
²⁰ okada riders prompted by this selective enforcement. Indeed, simulations suggest that the Union
²¹ benefited enormously. The majority of okada routes pass at least one additional motorpark after
²² the okada ban prior to before, with an average of approximately 1 additional motorpark per route.
²³ Less than 2% of routes pass fewer motorparks after the ban than before (see Figure ??).

²⁴ Importantly, the increase in ‘motorparks passed’ holds *even* when extending the sample to

³⁹Such as for fuel and ‘wear and tear’ motorbike expenses.

⁴⁰A standard price of a motorpark ticket around the time of the ban, evidenced via fieldwork by author

⁴¹Mechanically, the shortest route between points A and B after enacting barriers in the form of ban enforcement will either remain unaffected or get longer.

Joint Distribution of Route Cost and Motorparks Passed Enforcement Based on Okada Ban ITE Areas

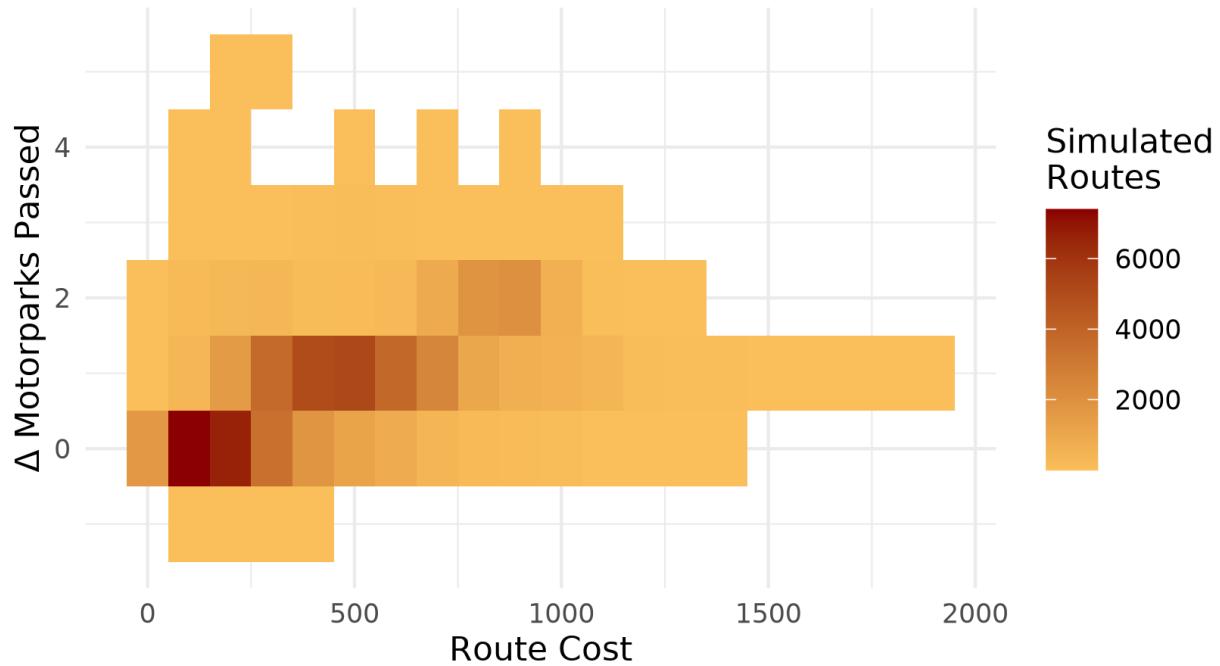


Figure 10: Density plot of 100,000 simulated routes given ITE enforcement areas. The y-axis is the change in the number of motorparks passed on a given route, compared to no ban enforcement/roadway impediments. The x-axis the approximate ‘cost’ of the route in NGN, determined as described above. The darkness of the shaded area indicates frequency of cost/motorpark combinations. Inspiration for this plot comes from ?.

¹ all routes traversable in the pre-period. That is to say, even across $\frac{1}{3}$ *fewer* rides, the number of
² motorparks passed increases nearly 5.5 times after the enforcement of the okada ban on ITE areas.
³ Translating into profit, this means that the Union's profit over 10,000 routes increases over 430%.

⁴ The total portion of routes which pass a motorpark before the ban's enforcement is also significantly
⁵ less. Before okada ban enforcement on ITE areas, only 3% of simulated routes pass motorparks.
⁶ In contrast, after okada ban enforcement, a majority of the traversable routes, nearly 61%, pass
⁷ motorparks. This large difference attests to the significant displacement of okada riders into areas
⁸ where the Union can take advantage of the ban's enforcement to extort from them. This trend holds
⁹ *even* considering the larger sample of all routes, including those intraversable after the ban; over
¹⁰ all simulated routes, less than 10% pass motorparks before the ban is enforced, compared to 36%
¹¹ afterwards.

¹² **Are these effects unique?**

¹³ **H1b** Enforcement areas (ITE) will increase okada density in motorparks *more* compared to other
¹⁴ possible enforcement areas.

¹⁵ The previous analysis makes clear that the ban's enforcement along stated ITE routes serves to
¹⁶ displace okada riders into motorpark-laden areas, increasing the power wielded by the Union over
¹⁷ these riders. However, this alone does not imply that these ITE routes were chosen (at least in part)
¹⁸ in order to produce this effect.

¹⁹ Therefore, I now turn to additional analyses which make use of *counterfactual* ITE areas
²⁰ which could have—given the purported goals of the ban—reasonably been chosen as enforcement
²¹ locations instead. There are a variety of potential logics which *could* have guided the government's
²² choice of enforcement priorities. These include logics drawn directly from the LSG's stated
²³ reasons for the ban in first place; such as easing traffic congestion, preventing road accidents,
²⁴ or reducing pollution. We can also imagine alternative logics less likely to be stated in so many

¹ words by state agents, such as enforcing the ‘eyesore’ of okada riders in the richest areas of the
² city, or prioritizing areas where the traffic agencies had the highest pre-existing capacity.

³ I propose a variety of potential logics which could have reasonably guided enforcement decisions,
⁴ summarized in Table 4. I then construct *counterfactual* ITE areas based on these logics by selecting
⁵ approximately 60 areas of the road network chosen from appropriately-weighted distributions. I
⁶ then re-run the simulations described above, using these hypothetical ITE configurations instead
⁷ of the actual ITE areas analyzed in the previous section. My goal with this exercise is to determine
⁸ whether these counterfactual ITE configurations have effects similar to the actual ITE areas chosen
⁹ by the state. My theory suggests that these alternative hypothetical enforcement logics will have
¹⁰ substantively different results than those produced by enforcement along real ITE areas; and
¹¹ especially, that choosing enforcement areas in this way would have benefited the Union substantially
¹² *less* than the actual enforcement areas chosen by the state.

¹³ The results of this analysis indeed show that the effects of the okada ban’s enforcement along
¹⁴ ITE routes is substantively unique, especially in its effects on the number of motorparks passed
¹⁵ by riders. To take one example: one purported logic behind the okada ban’s enforcement was
¹⁶ an effort to reduce traffic congestion. One potential enforcement logic is, therefore, to prioritize
¹⁷ enforcement in areas with high levels of traffic. I therefore construct a counterfactual ITE area by
¹⁸ randomly selecting from a distribution of heavy-traffic areas, measured by an area’s mention in
¹⁹ pre-ban 96.1FM traffic radio announcements (See SI §D.4 for more details). I then run simulations
²⁰ along the same start- and end-points used in the previous section.

²¹ For starters, ‘traffic-driven’ hypothetical enforcement areas make more routes intraversable to
²² riders. While the ban’s *actually* enforced areas prevent $\frac{1}{3}$ of simulated routes from being completed
²³ at all, traffic-driven counterfactual enforcement areas on average prevent a much higher percentage
²⁴ of routes from being traversed. Using this hypothetical ITE configuration, okada riders are able to
²⁵ make only 65% of all simulated journeys, significantly and more efficiently curtailing okada use
²⁶ with identical capacity.

Joint Distribution of Route Cost and Motorparks Passed Counterfactual Enforcement Based on Traffic Congestion

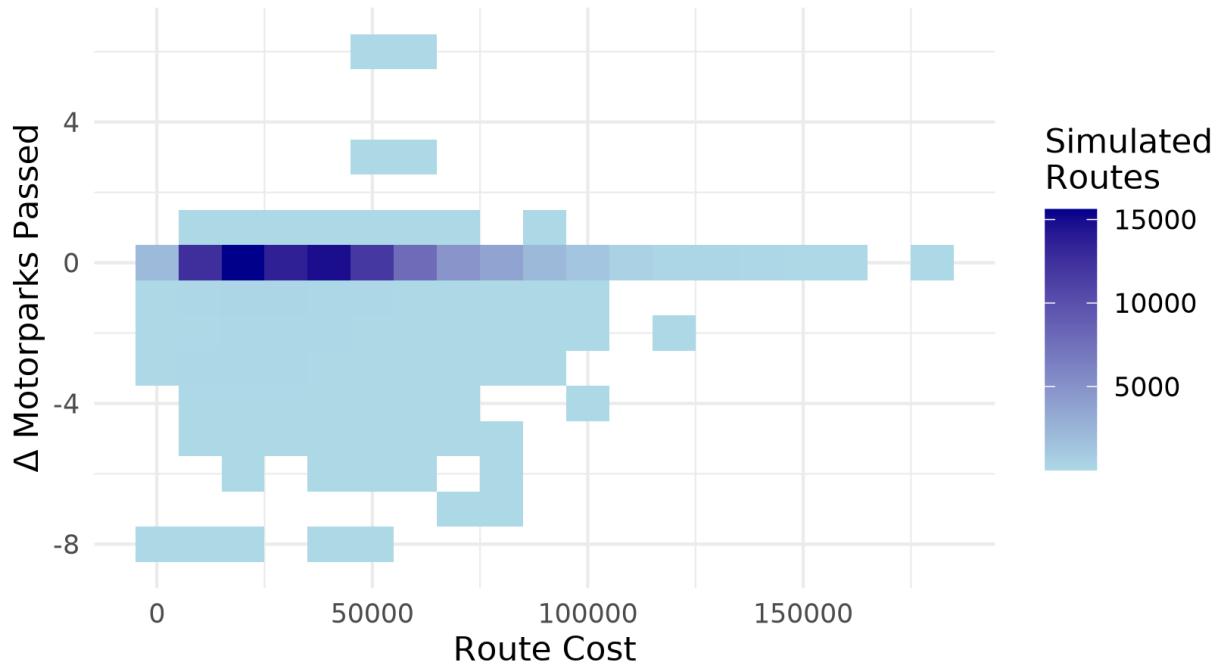


Figure 11: Density plot of 100,000 simulated routes given counterfactual enforcement based on traffic congestion. The y-axis is the change in the number of motorparks passed on a given route, compared to no ban enforcement/roadway impediments. The x-axis the approximate ‘cost’ of the route in NGN, determined as described above. The darkness of the shaded area indicates frequency of cost/motorpark combinations. Inspiration for this plot comes from ?.

1 Most crucially is the difference in the effect on motorparks passed. The ‘true’ ITE configuration
2 prompts riders to pass roughly one additional motorpark per journey on average. Meanwhile,
3 traffic-driven hypothetical bans have the opposite effect, and in fact *reduce* the number of motorparks
4 okada pass on their routes. Hypothetical bans on average result in a net and per-route loss in
5 the number of motorparks passed on an average route; and the majority of routes pass fewer
6 motorparks under this enforcement logic than without it (see Figure 11). SI §D.4 replicates this
7 analysis for other potential motivators of enforcement areas—such as locations of police stations;
8 locations most commonly associated with okada riding; and most reported accidents—with broadly
9 similar effects. These effects are summarized in Table 4.

10 This analysis confirms not only that most counterfactual enforcement patterns fail to benefit
11 the Union, but also that the actual enforcement patterns followed by the LSG *uniquely* benefit the
12 Union amidst a variety of plausible alternatives. This lends some credibility to my assertion that
13 the enforcement patterns of the ban were intended, at least in part, to prop up the Union’s ability
14 to exert coercive control over rebellious riders—even at the expense of more efficiently reducing
15 okada presence on the roadway.

16 **6 Evidence: Sudden disillusionment as a shock**

17 In previous sections, I described the peculiar political circumstances surrounding the conflict
18 between okada riders and the Lagos Union in the lead-up to an announcement of the ban’s enforcement
19 (Section 4), and demonstrate, consistent with H1, that spatial variation in the ban’s enforcement
20 served to displace riders into areas where the Union can extort them (Section ??). I show that this
21 pattern is not replicated when considering counterfactual enforcement areas based on traffic report
22 density, law enforcement presence, or other potential motivations for enforcement configurations
23 (Section 5. I now turn to an analysis of the *temporal* variation of okada ban enforcement, where I
24 predict that:

<i>Enforcement logic</i>	<i>Justification</i>	<i>Operationalization</i>	<i>Effects</i>
Actual ITE areas	Actual LSG enforced areas	Coded from policy announcement	67% routes traversable; 87 more motorparks
High traffic congestion	Ease traffic flow	Highest traffic incident density (via 96.1FM data)	35% routes traversable; 21 fewer motorparks
High state capacity	Utilize existing capacity	Closest to police stations (via OSM)	61% routes traversable; same number of motorparks
High offense areas	Target areas of high-okada density	Highest pre-ban okada level (via Streetview)	55% routes traversable; 4 more motorparks
High income areas	Appease wealthy constituents	Highest nightlight density (via EOG)	91% routes traversable; 8 fewer motorparks
High priority areas	Prioritize public safety	Closest to primary schools	84% routes traversable; 51 fewer motorparks
High crime areas	Target crime-dense areas	Highest level of crime (via LSG)	<i>Forthcoming</i>
High violence areas	Target hotbeds of political violence	Highest density of political violence (via ACLED)	81% routes traversable; 44 fewer motorparks
High pollution areas	Reduce pollution by okada	Highest particulate matter density (via NASA)	54% routes traversable; 23 fewer motorparks
Most efficient areas	Maximize enforcement efficiency	Areas with highest betweenness centrality on Lagos road network	27% routes traversable; 12 more motorparks
Concede to commuters	Minimize disruption to commuters	Areas with <i>lowest</i> pre-ban okada density (via Streetview)	<i>Forthcoming</i>
Ideal road conditions	Preserve okada availability where popular	Areas with most developed roads	<i>Forthcoming</i>

Table 4: Counterfactual enforcement patterns and their effects. Results are based on 10,000 simulated different start and end points randomly chosen from a uniform distribution across the Lagos road network. Number of motorparks passed is per 100 simulated routes, compared to pre-ban counts.

¹ **H2** Enforcement of the okada ban varies *temporally* according to state perception of Union
² reliability.

³ The evidence for this hypothesis comes from exploiting a shift in the state's reliance on the
⁴ Union, stemming from a surprising national election result at the state level. The Nigerian Presidential
⁵ and Gubernatorial elections of 2023, held about six months after enforcement of the ban began,
⁶ did not happen simultaneously. Federal elections—including for the President—were held on 25
⁷ February 2023. Meanwhile Lagos state elections—including for the governor, as well as for seats
⁸ in the state assembly—were held about a month later, on 18 March 2023.

⁹ The national election results were in one sense unsurprising; the incumbent APC's candidate
¹⁰ Bola Tinubu was elected against rivals Atiku Abubakar of the Peoples Democratic Party (PDP), and
¹¹ Peter Obi of the Labour Party (LP), the upstart populist candidate who captivated young Nigerian
¹² voters in particular. The APC's stronghold is in the Southwestern Nigeria, especially in Lagos
¹³ State, where the party has won every election it has ever stood.

¹⁴ Therefore, one unexpected result of the 2023 presidential election was that the people of
¹⁵ Lagos state voted en masse not for their former governor Bola Tinubu, but for the charismatic
¹⁶ and populist LP candidate Peter Obi. While the Lagos results did not ultimately sway the election
¹⁷ in Obi's favor, they nevertheless shocked the political elite of the state, who had been expecting
¹⁸ a comfortable APC victory, guaranteed (as usual) by the efforts of the Union. Multiple sources,
¹⁹ including a former Lagos commissioner of transport under the APC, told me the national election
²⁰ sent reverberations through the Lagos political system, asking me: "Did you see the election? the
²¹ Union didn't deliver. Their days are numbered."⁴²

²² I contend that the national election provided a shock to Lagos state leadership's perspective of
²³ the Union's reliability in guaranteeing an election victory for the incumbent APC, and therefore a
²⁴ shock to their willingness to cater to Union demands.

²⁵ Thus, as with H1, H2 can be broken into two component observational implications. First:

⁴²Interview with author on June 18, 2023

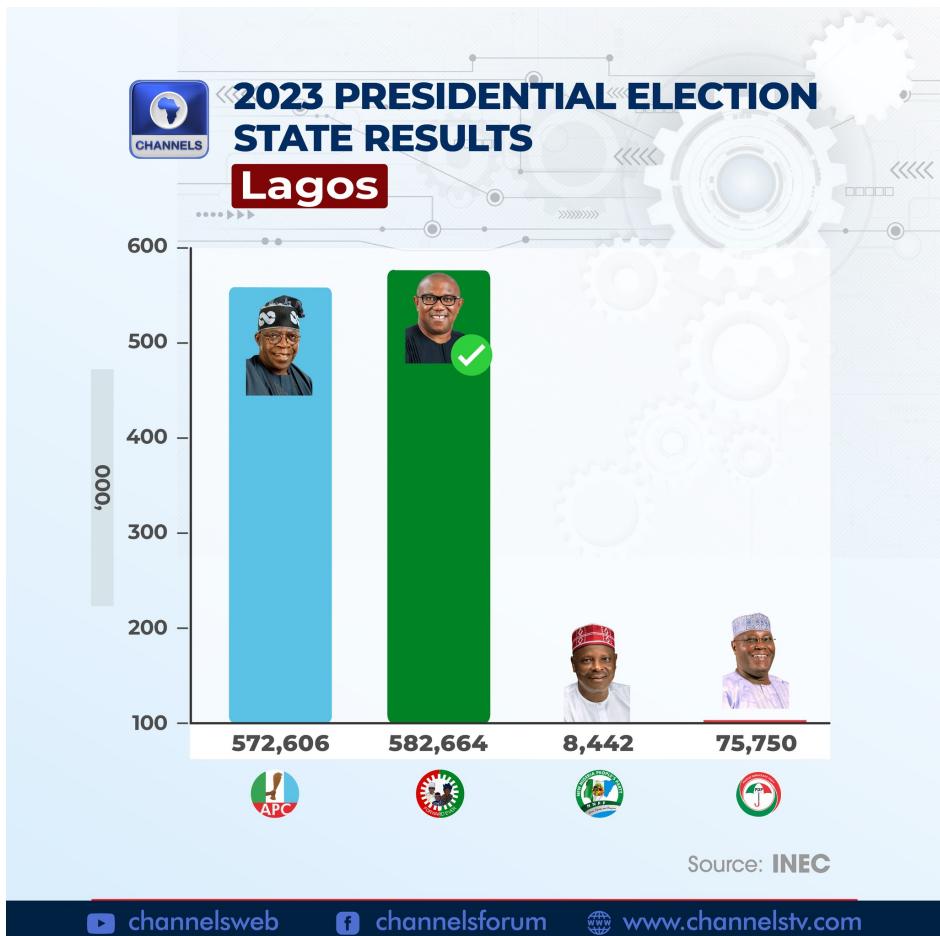


Figure 12: 2023 Nigerian Presidential Election Results for Lagos State. Image sourced from <https://tinyurl.com/ycympm3r>

¹ **H2a** Okada density *increases* after the Presidential election on previously enforced routes (ITE
² areas).

³ **H2a** Okada density *decreases* after the Presidential election around motorparks.

⁴ My strategy involves testing whether state enforcement of the ban changed after the national
⁵ election, in the lead-up to the Gubernatorial election a month later. If the LSG is selectively
⁶ enforcing the ban in order to produce precarity in okada riders for exploitation by the Union, we
⁷ should expect a shock to the state's trust in the Union to fundamentally alter this calculus, and
⁸ shift enforcement patterns accordingly. Interviews, including with okada riders, substantiate this
⁹ empirical strategy. Several (unprompted) brought up abrupt changes in enforcement during the
¹⁰ inter-election period.⁴³

¹¹ I therefore define:

$$(2) \quad \text{InterElec}_t = \text{PresElec}_t - \text{GovElec}_t$$

¹² Here as above, i denotes the road segment and t denotes the date of the measurement. The
¹³ variable PresElec_t takes the value of 1 when t is after the date of the Nigerian Presidential Election,
¹⁴ and GovElec_t when t is after the date of the Lagos Gubernatorial election. Therefore $\text{PresElec}_t -$
¹⁵ $\text{GovElec}_t = 1$ if t denotes a time period between the presidential and gubernatorial elections. I
¹⁶ measure the change during the inter-election period because, as I analyze further in SI §D.7), the
¹⁷ Union's behavior changed significantly during the gubernatorial election. After the ire from their
¹⁸ patron expressed after the Presidential election, the Union increased its election interference and
¹⁹ mobilization, meaning that I expect the enforcement dynamic to return to pre-Presidential election
²⁰ patterns after the Gubernatorial election is complete.

²¹ I use this variable to estimate two two-way fixed effects linear models:

⁴³Interviews with author, including 15 June, 2023

$$(3) \quad \text{Okada}_{it} = \gamma_i + \lambda_t + \tau(\text{ITE}_i \times \text{InterElec}_t) + \varepsilon_{it}$$

1

$$(4) \quad \text{Okada}_{it} = \gamma_i + \lambda_t + \tau(\text{Motorpark}_i \times \text{InterElec}_t) + \varepsilon_{it}$$

2 As above, Okada_{it} is the number of okada counted via satellite imagery for i road segment
 3 (edge) square at t date. Motorpark_i is a spatially-varying indicator of whether road segment i
 4 features a motorpark; while ITE_i is an indicator for whether road segment i is affected by okada
 5 ban enforcement. $(\text{ITE}_i \times \text{InterElec}_t)$ denotes whether a given i, t is an observation from an ITE
 6 area in the inter-election period.⁴⁴

7 Results are given in Table 5. Compared to the time before and after the inter-election period,
 8 ITE areas saw a dramatic increase in the number of okada per road segment, with nearly 20
 9 additional okada per quarter-kilometer of coded area—suggestive of a temporary drop in enforcement
 10 during the inter-election period and an increase in okada in previously-enforced areas. Meanwhile
 11 and relatedly, the number of okada in motorparks during the inter-election period decreased substantially.
 12 Okada presence varies much as we might expect: Namely, okada presence decreases during this
 13 inter-election period in motorparks, and increases across areas where okada presence is usually
 14 enforced. These trends lend credence to my theory that the state reacted to the Union’s failure
 15 to sway the election for the APC by reducing selective enforcement; itself an indication that
 16 enforcement patterns were driven by Union interests.

⁴⁴See SI §D.6 for additional specifications; as well as for assessment of assumptions necessary for causal interpretation of these results.

Table 5: Inter-election trends in okada density

	Number of Okada	
	(1)	(2)
<i>Variables</i>		
ITE × (InterElec)	19.18*** (6.624)	
Motorpark × (InterElec)		-29.39*** (10.31)
<i>Fixed-effects</i>		
Edge ID	Yes	Yes
Time period	Yes	Yes

Significance Codes: ***: 0.01, **: 0.05, *: 0.1

Clustered (Edge ID) standard-errors in parentheses

1 7 Conclusion

- 2 In this paper, I argue that states can selectively enforce policies aimed at vulnerable populations
 3 in order to ease their exploitation by important third-party groups. I examine the case of uneven
 4 enforcement of a ban on specific types of informal commercial transport in Africa’s largest city,
 5 and provide evidence that uneven enforcement of this ban is in the interest of appeasing powerful
 6 transport mafias, political allies of the regime who are relied upon around election time to provide
 7 political violence on demand in favor of the ruling party. I demonstrate first, that the period before
 8 enforcement was rift with intra-union politics, and specifically unrest from commercial drivers
 9 fighting back against their exploitation by the Union. Next, I demonstrate that variation in okada
 10 ban ITE areas *systematically* and *uniquely* alters okada rider routes towards Union-led motorparks,
 11 where they face increased exploitation and extortion. Finally, I use VHR remote-sensing and
 12 Streetview data to count the number of okada motorcycles present in quarter-kilometer squares
 13 over time, specifically in the lead-up to and in the immediate aftermath of a shocking Lagos-level
 14 result in the Presidential election. I find that okada presence *increases* in areas enforced by the

¹ ban (and decreases in areas with motorparks) in the period after the Presidential and before the
² Gubernatorial elections, a period of immense tension between the Lagos State Government and
³ the Lagos transport unions.

⁴ This project, while focusing on the micro-level dynamics of a single policy enacted in a
⁵ single city, has implications far beyond the streets of Lagos. My findings underline the necessity
⁶ of integrating third parties—beyond state agents and directly regulated populations—into our
⁷ understandings of why states pursue certain policies and their enforcement (Artabe et al., 2023;
⁸ Albertus, Fenner and Slater, 2018). I build upon prevailing understandings of repression and
⁹ forbearance by integrating them into a broader framework of how states assist their non-state
¹⁰ allies. My focus on *connivance* as a strategy emphasizes how governments may aid allies not
¹¹ only through direct concessions, but through manipulating enforcement to produce vulnerability
¹² in populations challenging these allies' power, or whose exploitation underwrites that power—or
¹³ both. Moreover, my focus on these third-party state allies shows the ‘flipside’ of a phenomenon
¹⁴ of great interest to the study of politics: election violence. I examine the supply-side dynamics
¹⁵ of election interference and violence by non-state groups, clarifying the independent motivations,
¹⁶ incentive structures, and interests of such groups. Moreover I provide a portrait of the ‘day jobs’
¹⁷ of election thugs, emphasizing how these actors exist within—and influence—political structures
¹⁸ even between elections. In doing so, I emphasize the need for research on the relationships between
¹⁹ governments and allied ‘violence entrepreneurs’ not only around election times, but in the years
²⁰ and months beforehand.

²¹ Finally, I contend that even nominally benign policies (and their level of enforcement) are
²² motivated by complex, and often corrupt, political arrangements. Like Holland (2016), I caution
²³ against interpreting a particular policy’s lack of enforcement as necessarily implying an *inability* to
²⁴ enforce. As I reinforce above, a government’s recalcitrance may instead be strategic; and—crucially—is
²⁵ often coercive. Future work should take these lessons to heart, especially in probing why governments
²⁶ tolerate illicit and informal industries; or engage only half-heartedly in enforcing violations such

¹ as irregular migration. As this paper demonstrates, such light-handedness is not always out of
² benevolence towards policy infringers.

³ Policy enforcement—like its passage—is highly political with distinct distributive consequences.
⁴ And indeed, enforcement is often the policy stage in which many citizens first convey their demands
⁵ and interact with the state, especially in the Global South (Scott, 1969, 1142). This article deepens
⁶ our understanding of how democracies might use this ‘enforcement lever’ to contribute to the
⁷ oppression of vulnerable populations, and to the empowerment of exploitative third parties. In
⁸ this way it reminds us that non-enforcement is not merely the absence of power, but a method of
⁹ wielding it; and that examining its patterns may reveal not the limits of the state, but the logic of
¹⁰ its alliances.

¹ References

- ² A., Schaap, Weeks K., Maiguascha B., Barvosa E. Bassel L. and Apostolidis P. T. 2022. “The politics of precarity.” *Contemporary Political Theory* 21(1):142–73.
- ⁴ ABColombia. 2012. “Giving It Away: The Consequences of an Unsustainable Mining Policy in Colombia.” *ABC* Colombia November. https://www.abcolombia.org.uk/wp-content/uploads/2017/06/Giving_it_Away_mining_report_ABColombia-2012-ENG.pdf.
- ⁷ Abiodun, Eromosele. 2019. “Of Touts, Danfo Drivers and Illegal Taxes.” *This Day* October 25. <https://www.thisdaylive.com/2019/10/25/of-touts-danfo-drivers-and-illegal-taxes/>.
- ¹⁰ Acemoglu, Daron, James A Robinson and Rafael Santos. 2009. “The monopoly of violence: evidence from Colombia.” *National Bureau of Economic Research* .
- ¹² Adekoya, Adekunle. 2022. “The June 1 Date For Okada In Lagos.” *Vanguard* May 27. <https://www.vanguardngr.com/2022/05/the-june-1-date-for-okada-in-lagos/>.
- ¹⁴ Adelagun, Oluwakemi. 2022. “Again, Lagos Bans Commercial Motorcycles.” *Premium Times* May 18. <https://www.premiumtimesng.com/news/top-news/530827-again-lagos-bans-commercial-motorcyclists.html?tztc=1>.
- ¹⁷ Agbiboa, Daniel E. 2020. “How informal transport systems drive African cities.” *Current History* 119(817):175–181.
- ¹⁹ Agbiboa, Daniel E. 2022. *They eat our sweat: Transport labor, corruption, and everyday survival in Urban Nigeria*. Oxford University Press.
- ²¹ Agha, Eugene and Abdullateef Aliyu. 2022. “Despite NURTW Suspension, ‘Extortion’ Of Drivers Persists In Lagos.” *Daily Trust* April 7.
- ²³ Akinkuotu, Eniola. 2019a. “MC Oluomo hails Tinubu, Sanwo-Olu over NURTW job.” *The Punch* . September 7. <https://punchng.com/mc-oluomo-hails-tinubu-sanwo-olu-over-nurtw-job/>.
- ²⁶ Akinkuotu, Eniola. 2019b. “Support Oluomo’s NURTW chairmanship bid, APC tells Tinubu.” *The Punch* September 1. <https://punchng.com/support-oluomos-nurtw-chairmanship-bid-apc-tells-tinubu/>.
- ²⁹ Akinsanmi, Gboyega. 2017. “Lagos Suspends NURTW over Murder of Leader of Commercial Motorcyclists.” *This Day* June 6. <https://www.thisdaylive.com/index.php/2017/06/06/lagos-suspends-nurtw-over-murder-of-leader-of-commercial-motorcyclists>.

- ¹ Akinsanmi, Gboyega. 2018. “Ambode Fights Back, Moves to Neutralise Mandate Group, Rallies Abuja for Support.” *This Day* September 12. <https://www.thisdaylive.com/2018/09/12/ambode-fights-back-moves-to-neutralise-mandate-group-rallies-abuja-for-support/>.
- ⁴ Akoni, Olasunkanmi. 2022. “NURTW crisis: MC Oluomo, Lagos council pull out of national body.” *Vanguard* March 10. https://www.vanguardngr.com/2022/03/nurtw-crisis-mc-oluomo-lagos-council-pulls-out-of-national-body/#google_vignette.
- ⁸ Albertus, Michael, Sofia Fenner and Dan Slater. 2018. *Coercive Distribution*. Cambridge University Press.
- ¹⁰ Aleaziz, Hamed and Zolan Kanno-Youngs. 2025. “Trump Shifts Deportation Focus, Pausing Most Raids on Farms, Hotels and Eateries.” *The New York Times* June 13. <https://www.nytimes.com/2025/06/13/us/politics/trump-ice-raids-farms-hotels.html?smid=nytcore-ios-share&referringSource=articleShare>.
- ¹⁴ Alilyu, Abdullateef. 2022. “NURTW: Facing Existential Crisis?” *Daily Trust* April 28. <https://dailytrust.com/nurtw-facing-existential-crisis/>.
- ¹⁶ Arias, Enrique Desmond. 2017. *Criminal enterprises and governance in Latin America and the Caribbean*. Cambridge University Press.
- ¹⁸ Artabe, Eugenia, Samantha Chapa, Leah Sparkman and Patrick E. Shea. 2023. “External Threats, Capacity, and Repression: How the Threat of War Affects Political Development and Physical Integrity Rights.” *British Journal of Political Science* 53(4):1311–1327.
- ²¹ Badmus, Bola. 2022. “Okada ban aligns with democratic wishes of Lagosians—CPPM.” *Nigerian Tribune* August 21. <https://tribuneonlineng.com/okada-ban-aligns-with-democratic-wishes-of-lagosians-%E2%80%95-cppm/>.
- ²⁴ Balogun, Ishola. 2022. “I’ll never bring NURTW to disrepute, MC Oluomo replies National body.” *Vanguard* . March 8. <https://www.vanguardngr.com/2022/03/i-ll-never-bring-nurtw-to-disrepute-mc-oluomo-replies-national-body/>.
- ²⁷ Bankole, Taofik. 2017. “NURTW Tension Ambode orders suspension of union following Hamburger murder.” *NEUSROOM* June 5. <https://neusroom.com/nurtw-tension-ambode-orders-suspension-union-following-hamburger-murder/>.
- ³⁰ Ben-Menachem, Jonathan and Kevin T Morris. 2023. “Ticketing and turnout: the participatory consequences of low-level police contact.” *American political science review* 117(3):822–834.
- ³² Berg, Ryan C., Henry Ziemer and Arianna Kohan. 2021. “A Closer Look at Colombia’s Illegal, Artisanal, and Small-Scale Mining.” *Center for Strategic & International Studies (CSIS)* December 20. <https://www.csis.org/analysis/closer-look-colombias-illegal-artisanal-and-small-scale-mining>.

- ¹ Blattman, Christopher, Gustavo Duncan, Benjamin Lessing and Santiago Tobón. 2025. “Gang rule: Understanding and countering criminal governance.” *Review of Economic Studies* 92(3):1497–1531.
- ⁴ Bradock, James. 2021. “Madrid offers emergency shelter to 4,000 slum dwellers without electricity in record cold snap.” *The Telegraph* January 11.<https://www.telegraph.co.uk/news/2021/01/11/madrid-offers-emergency-shelter-4000-slum-dwellers-without-electricity/>.
- ⁸ Bryan, Gharad T, Kyra Frye and Melanie Morten. 2025. “Spatial economics for low-and middle-income countries.” *National Bureau of Economic Research* .
- ¹⁰ Calderoni, Francesco. 2011. “Where is the mafia in Italy? Measuring the presence of the mafia across Italian provinces.” *Global Crime* 12(1):41–69.
- ¹² Carey, Sabine C., Neil J. Mitchell and Adam Scharpf. 2022. “Pro-Government Militias and Conflict.” *Oxford Research Encyclopedia* .
- ¹⁴ Cingolani, L. 2013. “The State of State Capacity : a review of concepts, evidence and measures.” *MERIT Working Papers* .
- ¹⁶ Comunidad de Madrid. 2022. Díaz Ayuso anuncia una nueva estación de Metro en superficie en la línea 9 para dar cobertura a los nuevos desarrollos de Los Ahijones y Los Berrocales Comunidad de Madrid. Technical report Comunidad de Madrid. February 12. <https://www.comunidad.madrid/noticias/2022/10/04/diaz-ayuso-anuncia-nueva-estacion-metro-superficie-linea-9-dar-cobertura-nuevos-desar>
- ²¹ Cook, Scott J. and David Fortunato. 2023. “The politics of police data: State legislative capacity and the transparency of state and substate agencies.” *American political science review* 117(1):280–295.
- ²⁴ Cowen, Deborah. 2020. “Following the infrastructures of empire: Notes on cities, settler colonialism, and method.” *Urban Geography* 41(4):469–486.
- ²⁶ Cunningham, David and Peter B. Owens. 2020. “Enforcement networks and racial contention in civil rights-era Mississippi.” *Ruling by other means* pp. 110–139.
- ²⁸ Dahl, Robert A. 2007. “The concept of power.” *Behavioral Science* 2(3):201–215.
- ²⁹ Daily Trust. 2022a. “Gunshots As Warring Transporters Clash In Lagos.” *Daily Trust* April 23. <https://dailytrust.com/gunshots-as-warring-transporters-clash-in-lagos/>.
- ³¹ Daily Trust. 2022b. “Why N800 Daily Levy On Transporters Is Generating Ripples In Lagos.” *Daily Trust* February 9. <https://dailytrust.com/why-n800-daily-levy-on-transporters-is-generating-ripples-in-lagos/>.

- ¹ Dal Bó, Ernesto. 2006. “Regulatory capture: A review.” *Oxford review of economic policy* 22(2):203–225.
- ³ D’Alisa, Giacomo, David Burgalassi, Hali Healy and Mariana Walter. 2010. “Conflict in Campania: Waste emergency or crisis of democracy.” *Ecological economics* 70(2):239–249.
- ⁵ Dean, Adam. 2022. *Opening up by cracking down: labor repression and trade liberalization in democratic developing countries*. Cambridge University Press.
- ⁷ Dell, Melissa. 2015. “Trafficking networks and the Mexican drug war.” *American Economic Review* 105(6):1738–1779.
- ⁹ Dewey, Matías. 2012. “Illegal police protection and the market for stolen vehicles in Buenos Aires.” *Journal of Latin American Studies* 44(4):679–702.
- ¹¹ Dewey, Matías and Donato Di Carlo. 2022. “Governing through non-enforcement: Regulatory forbearance as industrial policy in advanced economies.” *Regulation & Governance* 16(3):930–950.
- ¹⁴ Dipoppa, Gemma. 2025. “How criminal organizations expand to strong states: Local agreements and migrant exploitation in Northern Italy.” *The Journal of Politics* 87(2):556–571.
- ¹⁶ Dipoppa, Gemma and Saad Gulzar. 2023. “No Smoke Without a Fire: Bureaucratic Incentives, Crop burning, and Air Pollution in South Asia.” *Working Paper*.
- ¹⁸ Duran, Luis F. 2021. “Electrical raid in Cañada Real: Police dismantle illegal power hookups in 80 houses in Madrid.” *El Mundo* November 11. <https://www.elmundo.es/madrid/2021/11/10/618c38f4fddffa93f8b4581.html>.
- ²¹ Earle, Lucy and Emma Grant. 2019. “Access to shelter and services for low-income groups: lessons from Hawassa, Mogadishu and Nairobi on the politics of informal settlements and shelter access.” *IIED Synthesis Reports* October. https://assets.publishing.service.gov.uk/media/5dea4411ed915d09cc5d2474/Synthesis_report.pdf.
- ²⁵ Eck, K., C. R. Conrad and C. Crabtree. 2021. “Policing and Political Violence.” *Journal of Conflict Resolution* 65(10):1641–1656.
- ²⁷ Edema, Grace. 2022. “NURTW Oppressing Us Say Tricycle Operators.” *The Punch* February 17. <https://punchng.com/nurtw-oppressing-us-say-tricycle-operators/>.
- ²⁹ EFE. 2020. “Ayuso, sobre la Cañada Real: “Para tener Porsches ahí aparcados, bien; para pagar las facturas, no”.” *El Español* December 10. https://www.elspanol.com/madrid/comunidad/20201210/ayuso-canada-real-porsches-aparcados-facturas-no/542446285_0.html.
- ³³ EFE. 2021. “Once meses sin luz en la Cañada y pocas soluciones para las familias.” *La Vanguardia* December 9. <https://www.lavanguardia.com/local/madrid/20210912/7715300/once-meses-luz-canada-pocas-soluciones-familias.html>.

- ¹ Egbas, Jude. 2019. "This is the full story of how MC Oluomo was
² stabbed with a poisoned knife at Lagos APC campaign rally." *The
3 Pulse* January 9. <https://www.pulse.ng/articles/news/politics/mc-oluomo-how-notorious-nurtw-official-was-stabbed-with-poisoned-knife-20240809200239>
- ⁵ Elezuo, Eric. 2019. "Oluomo Discharged from Lagos Hospital, May Have Jetted Out to the
6 US." *The Boss Newspaper* January 16. <https://thebossnewspapers.com/2019/01/16/oluomo-discharged-from-lagos-hospital-may-have-jetted-out-to-the-us/>.
- ⁸ Enimola, Olorundare. 2022. "Lagosians reminiscent Days Of
⁹ Swift Okada Rides As Traffic Gridlocks Blight Gains Of Ban."
¹⁰ *Daily Independent* . September 5. <https://independent.ng/lagosians-reminiscent-days-of-swift-okada-rides-as-traffic-gridlocks-blight-gains-of->
- ¹² Enimola, Olorundare and Titilope Joseph. 2022. "Lagos Okada ban Commuters Recount
¹³ Experiences As Enforcement Begins." *Daily Independent* June 2. <https://independent.ng/lagos-okada-ban-commuters-recount-experiences-as-enforcement-begins/>.
- ¹⁵ Faith, Adeoye. 2022. "MC Oluomo replies NURTW national body's query." *Nigerian Tribune*
¹⁶ March 1. [news_faith2022](#).
- ¹⁷ Feldmann, Andreas E and Juan Pablo Luna. 2022. "Criminal governance and the crisis of
¹⁸ contemporary Latin American states." *Annual Review of Sociology* 48(1):441–461.
- ¹⁹ Fourchard, Laurent. 2023. "Expanding profit and power: The National Union of Road Transport
²⁰ Workers in Nigeria." *Canadian Journal of Development Studies/Revue canadienne d'études du
21 développement* 44(1):97–112.
- ²² Gil, Andres. 2022. "Neighbors of Cañada Real denounce discrimination in Europe after two
²³ years without electricity." *El Diario* November 29. https://www.eldiario.es/madrid/vecinas-canada-real-claman-bruselas-queremos-luz_1_9754442.html.
- ²⁵ Goodfellow, Tom. 2015. "Taming the "rogue" sector: Studying state effectiveness in Africa
²⁶ through informal transport politics." *Comparative Politics* 47(2):127–147.
- ²⁷ Government of Kenya. 2017. "Legal Notice No. 280: The Land Act (No. 6 of 2012)." *Special
issue: Kenya Gazette Supplement No. 179, Legislative Supplement No. 88* November 24. https://kenyalaw.org/k1/fileadmin/pdfdownloads/LegalNotices/2017/LN280_2017.pdf.
- ³⁰ Grasse, Donald and Melissa Pavlik. 2025. "Conservation, Climate Change, and Conflict: Nigeria's
³¹ farmer-herder violence." *Working paper* .
- ³² Hanson, Jonathan K. and Rachel Sigman. 2021. "Leviathan's Latent Dimensions: Measuring State
³³ Capacity for Comparative Political Research." *The Journal of Politics* 83(4):1495–1510.

- ¹ Harding, Robin, Mounu Prem, Nelson A. Ruiz and David L. Vargas. 2024. “Buying a blind eye: Campaign donations, regulatory enforcement, and deforestation.” *American Political Science Review* 118(2):635–653.
- ⁴ Hassan, Mai, Daniel Mattingly and Elizabeth R. Nugent. 2022. “Political Control.” *Annual Review of Political Science* 25(1):155–174.
- ⁶ Hassan, Mohammed Salah, Raja Noriza Raja Ariffin, Norma Mansor and Hussam Al Halbusi. 2021. “The Moderating Role of Willingness to Implement Policy on Street-level Bureaucrats’ Multidimensional Enforcement Style and Discretion.” *International Journal of Public Administration* pp. 1–15.
- ¹⁰ Holland, Alisha C. 2016. “Forbearance.” *American Political Science Review* 110(2):232–246.
- ¹¹ Holland, Alisha C. 2017. *Forbearance as Redistribution*. Cambridge University Press.
- ¹² Ige, Segun. 2022. “Okada Ban Means Different Things to Different Lagos Residents. Many Think It’s Unfair.” *Fiji* October 2. <https://fij.ng/article/okada-ban-means-different-things-to-different-lagos-residents-many-think-its-unfair/>.
- ¹⁵ Infrastructure News. 2017. “Ambode tells NURTW to get prepared for a reform in transportation sector.” *Infrastructure News Nigeria* December 21. <https://www.infrastructurenews.ng/ambode-tells-nurtw-to-get-prepared-for-reform-in-transportation-sector/>.
- ¹⁸ Inyang, Ifreke. 2019. “What MC Oluomo said about Tinubu after emerging new Lagos NURTW chairman.” *Daily Post* September 6. <https://dailypost.ng/2019/09/06/mc-oluomo-said-tinubu-emerging-new-lagos-nurtw-chairman/>.
- ²¹ Jones, Sam. 2021. “‘You kind of die’: life without power in the Cañada Real, Spain.” *The Guardian* October 27. <https://www.theguardian.com/world/2021/oct/27/you-kind-of-die-life-without-power-in-the-canada-real-spain>.
- ²⁴ Lagos State of Nigeria. 2018. “A LAW TO CONSOLIDATE ALL LAWS RELATING TO THE TRANSPORT SECTOR, TO PROVIDE FOR THE DEVELOPMENT AND MANAGEMENT OF A SUSTAINABLE TRANSPORT SYSTEM IN LAGOS STATE AND FOR CONNECTED PURPOSES.” *Lagos State of Nigeria Official Gazette Extraordinary* 51. March 23. <https://archive.gazettes.africa/archive/ng-la/2018/ng-la-official-gazette-supplement-dated-2018-03-23-no-12.pdf>.
- ³⁰ Lambo, Deji. 2022. “Lagos Motorcycle Riders Clash As Man Stabbed During Street Fight Dies.” *The Punch* April 5.
- ³² Lessing, Benjamin. 2024. “Explaining Persistent Duopolies of Violence: How the State Gets Drug Gangs to Govern for It.” Available at SSRN .
- ³⁴ Lovett, Francis N. 2001. “Domination: A preliminary analysis.” *The Monist* 84(1):98–112.

- ¹ Lukes, Steven. 2021. *Power: A radical view*. Bloomsbury Publishing.
- ² Mahadevan, Meera. 2024. “The Price of Power: Costs of Political Corruption in Indian Electricity.” *American Economic Review* 114:3314–3344.
- ³
- ⁴ Mann, Michael. 1984. “The autonomous power of the state: its origins, mechanisms and results.” *European Journal of Sociology/Archives européennes de sociologie* 25(2):185–213.
- ⁵
- ⁶ Mares, Isabela and Lauren E. Young. 2019. *Conditionality & Coercion: Electoral Clientelism in Eastern Europe*. Oxford University Press.
- ⁷
- ⁸ Marín, Elena. 2022. “Las administraciones se dan un plazo de ocho años para terminar con los realojos de la Cañada Real vert El Periódico de España.” *Epe* December 12. <https://www.epe.es/es/madrid/20221212/administraciones-plazo-realojo-vecinos-canada-real-madrid-79611832>.
- ⁹
- ¹⁰
- ¹¹
- ¹² Martínez-Fernández, Andrés. 2019. “The National Liberation Army in Colombia and Venezuela: Illicit Finance Challenges Stemming from Illegal Mining.” *American Enterprise Institute (AEI)* July. <https://www.aei.org/wp-content/uploads/2019/07/The-National-Liberation-Army-in-Colombia-and-Venezuela.pdf?x85095>.
- ¹³
- ¹⁴
- ¹⁵
- ¹⁶ Massé, Frédéric and Johanna Camargo’. 2012. “Actores Armados Ilegales y Sector Extractivo en Colombia.” *CITpax Colombia* https://www.academia.edu/9300978/Actores_armados_ilegales_y_sector_extractivo_en_Colombia.
- ¹⁷
- ¹⁸
- ¹⁹ Massé, Frédéric and Juan Munavar’. 2017. “Responsible Business Conduct: Due Diligence in Colombia’s Gold Supply Chain Overview.” *Organization for Economic Cooperation and Development (OECD) Due Diligence Guidance* <https://mneguidelines.oecd.org/Colombia-gold-supply-chain-overview.pdf>.
- ²⁰
- ²¹
- ²²
- ²³ Migdal, Joel S. 1988. *Strong societies and weak states: state-society relations and state capabilities in the Third World*. Princeton University Press.
- ²⁴
- ²⁵ Mindefensa. 2012. “Decree 2235 of 2012.” *Diario Oficial. AÑo CXLVII* 48599:15. October 29. <https://www.suin-juriscol.gov.co/viewDocument.asp?id=1808886>.
- ²⁶
- ²⁷ Ministerio de Justicia. 2016. “Código Penal/Criminal Code.” *Subdirección General de Documentación y Publicaciones* https://www.mjusticia.gob.es/es/AreaTematica/DocumentacionPublicaciones/Documents/Criminal_Code_2016.pdf.
- ²⁸
- ²⁹
- ³⁰ Momoh, Abubakar. 2000. Youth Culture and Area Boys in Lagos. In *Identity Transformation and Identity Politics under Structural Adjustment in Nigeria*, ed. Attahiru Jega. Nordiska Afrikainstitutet and Centre for Research and Documentation pp. 181–203.
- ³¹
- ³²
- ³³ Montaner, Núria Rius. 2021. “La Cañada Real, the largest irregular settlement in Europe that resists in the dark.” *ara* January 31. https://en.ara.cat/society/canada-real-the-largest-irregular-settlement-in-europe-that-resists-in-the-dark_130_3840774.html.
- ³⁴
- ³⁵
- ³⁶

- ¹ Mosadioluwa, Adam. 2022. “Lagosians Lament As Full Implementation Of Okada Ban Commences.” *Tribune Online* September 1. <https://tribuneonlineng.com/lagosians-lament-as-full-implementation-of-okada-ban-commences/>.
- ⁴ Mwau, Baraka and Alice Sverdlik. 2020. “High rises and low-quality shelter: rental housing dynamics in Mathare Valley, Nairobi.” *Environment & Urbanization* 32.
- ⁶ Nathan, Noah L. 2023. “Do grids demobilize? How street networks, social networks, and political networks intersect.” *American Journal of Political Science* .
- ⁸ Ngai, Mae M. 2014. *Impossible subjects: Illegal aliens and the making of modern America*. Princeton University Press.
- ¹⁰ Oboagwina, Felix. 2022. “Didn’t Tinubu Just Goof On MC Oluomo?” *This Day* April 15. <https://www.thisdaylive.com/2022/04/15/didnt-tinubu-just-goof-on-mc-oluomo/>.
- ¹² Odesola, Runde. 2022. “Obaship Ifa Rejects MC Oluomo.” *The Punch* Feburary 13. <https://punchng.com/obaship-if-a-rejects-mc-oluomo-1/>.
- ¹⁴ Odunsi, Wale. 2022. “Lagos get tougher on okada as terrorists target Nigeria’s biggest city.” *Daily Post* . August 20. <https://dailypost.ng/2022/08/20/lagos-get-tougher-on-okada-as-terrorists-target-nigerias-biggest-city/>.
- ¹⁷ Olanrewaju, Sulaimon. 2022. “NURTW queries MC Oluomo over insubordination, anti-union activities.” *Nigerian Tribune* February 28. <https://tribuneonlineng.com/nurtw-queries-mc-oluomo-over-insubordination-anti-union-activities/>.
- ²⁰ Olaoluwa, Joseph. 2022. “NURTW suspends MC Oluomo.” *International Centre for Investigative Reporting (ICIR)* March 10. <https://www.icirnigeria.org/nurtw-suspends-mc-oluomo/>.
- ²³ Olasupo, Ayoola. 2022. “Ban Forces Lagos Commercial Motorcyclists to Inner Routes, Ogun.” *The Punch* . September 10. <https://punchng.com/ban-forces-lagos-commercial-motorcyclists-to-inner-routes-ogun-communities/>.
- ²⁶ Olawoyin, Oladeinde. 2017. “Fear grips Lagos drivers, conductors as government plans to ban danfo buses.” *Premium Times* February 19. <https://www.premiumtimesng.com/news/headlines/223919-fear-grips-lagos-drivers-conductors-government-plans-ban-danfo-buses.html>.
- ³¹ Osazuwa, Job. 2017. “Commercial motorbike operators dare Ambode.” *The Sun* March 16. <https://thesun.ng/commercial-motorbike-operators-dare-ambode/>.
- ³³ Oyeleke, Sodiq. 2022. “Just In: One Killed As Okada Riders, Hoodlums Clash in Lagos.”” *The Punch* April 4.

- ¹ Oyero, Kayode. 2022a. “BREAKING: Crisis: Lagos backs MC Oluomo, suspends NURTW.” *The Punch* March 10.
- ³ Oyero, Kayode. 2022b. “MC Oluomo hijacked Lagos NURTW from me, snatched my second wife – Fatai Adeshina, Lagos caretaker chairman.” *The Punch* May 21. <https://punchng.com/mc-oluomo-hijacked-lagos-nurtw-from-me-snatched-my-second-wife-adeshina-lagos-caretaker-chairman/>
- ⁶ Oyero, Kayode. 2022c. “Park Takeover: Soldiers, policemen storm Agbado as MC Oluomo, Istijabah boys draw battle line.” *The Punch* April 18.
- ⁸ Pager, Tyler, Miriam Jordan, Hamed Aleaziz and Zolan Kanno-Youngs. 2025. “Inside Trump’s Extraordinary Turnaround on Immigration Raids.” *The New York Times* June 14. <https://www.nytimes.com/2025/06/14/us/politics/trump-immigration-raids-workers.html?smid=nytcore-ios-share&referringSource=articleShare>.
- ¹² Pan, Jennifer. 2020. *Welfare for autocrats: How social assistance in China cares for its rulers*. Oxford University Press, USA.
- ¹⁴ Premium Times. 2022. “Tricycle riders seek independence from NURTW.” *Premium Times* February 12. <https://www.premiumtimesng.com/news/more-news/511162-tricycle-riders-seek-independence-from-nurtw.html>.
- ¹⁷ Price, Megan, Peter Albrecht, Francesco Colona, Lisa Denney and Wangui Kimari. 2016. “Hustling for Security: Managing plural security in Nairobi’s poor urban settlements.” *Clingendael Conflict Research Unit Plural Security Insights* June. https://pure.diis.dk/ws/files/581493/2016_Nairobi_plural_security_report_.pdf.
- ²¹ Quesada, Juan Diego. 2023. ““No queremos un refugio, queremos luz”.” *El país*
²² URL: January 9.
- ²³ Reporters, Sahara. 2022. “Most Commercial Motorcyclists In Lagos Are Criminals, Many Caught With Pistols, Police Commissioner Backs ‘Okada’ Ban.” *Sahara Reporters* . May 19. <https://saharareporters.com/2022/05/19/most-commercial-motorcyclists-lagos-are-criminals-many-caught-pistols-police-commissioner-backs-okada-ban/>
- ²⁷ Reuters. 2022. “Nigeria’s Lagos bans ‘nuisance’ motorbike taxis from most roads.” *Reuters* . May 19. [https://www.reuters.com/world/africa/nigerias-lagos-bans-nuisance-motorbike-taxis-most-roads-2022-05-18/#:~:text=LAGOS%20May%2018%20\(Reuters\),mode%20of%20transport%20for%20residents.](https://www.reuters.com/world/africa/nigerias-lagos-bans-nuisance-motorbike-taxis-most-roads-2022-05-18/#:~:text=LAGOS%20May%2018%20(Reuters),mode%20of%20transport%20for%20residents.)
- ³² Rigon, Andrea. 2015. “Collective or individual titles? Conflict over tenure regularisation in a Kenyan informal settlement.” *Urban Studies* 53.
- ³⁴ Sahara Reporters. 2022. “NURTW Crisis: Lagos Government Sank Low By Doing Transporter, MC Oluomo’s Bidding, Over 1000 Vehicles Vandalised In Four Days—Group.”

- 1 *Sahara Reporters* April 29. <https://saharareporters.com/2022/04/29/nurtw-crisis-lagos-government-sank-low-doing-transporter-mc-oluomo%E2%80%99s-bidding-over-1000>.
- 4 Sánchez De La Sierra, Raúl, Kristof Titeca, Haoyang Xie, Aimable Amani Lameke and Albert Malukisa Nkuku. 2024. “The real state: Inside the congo’s traffic police agency.” *American Economic Review* 114(12):3976–4014.
- 7 Sánchez, Fernando. 2022. “Almeida fija el ”camino más corto” para devolver luz a La Cañada en “acabar con plantaciones de marihuana”.” *Ultima Hora* November 23. <https://www.ultimahora.es/noticias/comunidades/2022/11/23/1833965/almeida-fija-camino-mas-corto-para-devolver-luz-canada-acabar-plantaciones-marihuana.html>.
- 12 Schouten, Peer. 2022. *Roadblock politics: the origins of violence in Central Africa*. Cambridge University Press.
- 14 Scott, James C. 1969. “Corruption, machine politics, and political change.” *American political science review* 63(4):1142–1158.
- 16 Sobering, Katherine and Javier Auyero. 2019. “Collusion and cynicism at the urban margins.” *Latin American Research Review* 54(1):222–236.
- 18 Society Now. 2019. “Ambode’s Loyalist Agbede Surrenders, Raises The Hand Of MC Oluomo—Tinubu’s Man—As New Transport Union—NURTW—Chairman In Lagos State.” *Society Now* . September 9. <https://societynow.ng/detailsambodes-loyalist-agbede-surrenders-raises-tinubus-man-mc-oluomos-hand-as-new-t>
- 22 Society Now. 2022. “The Statement MC Oluomo Issued That Got Him Suspended.” *Society Now* March 10. <https://societynow.ng/the-statement-mc-oluomo-issued-that-got-him-suspended/>.
- 25 Stark, Evan. 2007. *Coercive control: How men entrap women in personal life*. Oxford University Press.
- 27 Su, Min and Christian Buerger. 2024. “Playing politics with traffic fines: Sheriff elections and political cycles in traffic fines revenue.” *American Journal of Political Science* .
- 29 Sullivan, Christopher M. and Christian Davenport. 2018. “Resistance is mobile.” *Journal of Peace Research* 55(2):175–189.
- 31 Tapscott, Rebecca. 2021. *Arbitrary States*. Oxford University.
- 32 Thachil, Tariq. 2020. “Does Police Repression Spur Everyday Cooperation? Evidence from Urban India.” *The Journal of Politics* 82(4):1474–1489.

- ¹ The Nation. 2022. “Sanwo-Olu to protesting NURTW: I’ll act on grievances.” *The Nation* March 6. <https://thenationonlineng.net/sanwo-olu-to-protesting-nurtw-ill-act-on-grievances/>.
- ⁴ The White House. 2025. “Protecting the American People Against Invasion.” *Executive Order (EO) 14159*. January 20. <https://www.whitehouse.gov/presidential-actions/2025/01/protecting-the-american-people-against-invasion/>.
- ⁷ Tilly, Charles et al. 1992. *Coercion, capital, and European states, AD 990-1992*. Blackwell Oxford.
- ⁹ Titlola, Babatunde. 2022. “Court orders MC Oluomo, Lagos, NURTW to stop levying drivers.” *The Punch* May 25. <https://punchng.com/court-orders-mc-oluomo-lagos-nurtw-to-stop-levying-drivers/>.
- ¹² Tribune Online. 2022. “Okada ban: Fear of attacks by foreigners spreads in lagos.” *Nigerian Tribune* . May 21. <https://tribuneonlineng.com/okada-ban-fear-of-attacks-by-foreigners-spreads-in-lagos/>.
- ¹⁵ Walters, Reece. 2013. Eco mafia and environmental crime. In *Crime, justice and social democracy: International perspectives*. Springer pp. 281–294.
- ¹⁷ Weber, Max. 1978. *Economy and society: An outline of interpretive sociology*. Vol. 2 University of California press.
- ¹⁹ Whiteman, Kaye. 2013. *Lagos: a cultural and historical imagination*. Exeter, UK.
- ²⁰ Wilkinson, Steven. 2006. *Votes and violence: Electoral competition and ethnic riots in India*. Cambridge University Press.
- ²² Williams, Martin J. 2017. “The political economy of unfinished development projects: Corruption, clientelism, or collective choice?” *American Political Science Review* 111(4):705–723.
- ²⁴ Williams, Martin J. 2021. “Beyond state capacity: bureaucratic performance, policy implementation and reform.” *Journal of Institutional Economics* 17(2):339–357.
- ²⁶ Xu, Alice Z. 2023. “Segregation and the spatial externalities of inequality: A theory of interdependence and public goods in cities.” *American Political Science Review* pp. 1–18.
- ²⁸ York, Steve. 1999. *A Force More Powerful*. International Center on Nonviolent Conflict. Episode 1 available at <https://youtu.be/04dDVeAU3u4>.
- ³⁰ Zhukov, Yuri M. 2012. “Roads and the diffusion of insurgent violence: The logistics of conflict in Russia’s North Caucasus.” *Political Geography* 31(3):144–156.

1

Supplemental Information

2 Table of Contents

3	A ETHICAL CONSIDERATIONS	2
4	A.1 Positionality	2
5	A.2 Qualitative data collection: Semi-structured interviews	2
6	A.3 Qualitative data collection: Participant observation	2
7	A.4 Qualitative data collection: TOUT data and examining social networks	2
8	A.5 Quantitative data: High-resolution	2
9	B BACKGROUND AND CASE APPENDIX	2
10	B.1 List of acronyms	2
11	B.2 Primer on Nigerian and Lagos politics	2
12	B.3 Primer on (Transport) Union Politics in Nigeria	5
13	B.4 Okada bans within and beyond Lagos	6
14	C QUALITATIVE DATA APPENDIX	7
15	C.1 Transport Operators Union Tracking (TOUT) data	7
16	C.2 Interviews and selection strategy	11
17	C.3 Participant observation; fieldwork sites and details	11
18	D QUANTITATIVE DATA APPENDIX	12
19	D.1 Satellite imagery analysis	12
20	D.2 Streetview imagery analysis	18
21	D.3 Alternative explanations and specifications for satellite imagery analysis	19
22	D.4 Road network analysis	20

1	D.5 Additional Twitter analyses and descriptive statistics	30
2	D.6 Alternative measurement strategies and additional analyses	30
3	D.7 Election analysis	31
4	E MODEL APPENDIX	32
5	E.1 Actors and preferences	32
6	E.2 Model setup	33
7	E.3 Actor incentive constraints and equilibrium profiles	35
8	E.4 With Grim-Trigger Punishment	38

A ETHICAL CONSIDERATIONS

¹ Considering the potential ethical ramifications of this work has been at the forefront throughout
³ the course of this project. There are five primary areas which require additional discussion.

A.1 Positionality

⁵ This project entailed going into spaces where I was not only an outsider, but quite an obvious
⁶ outsider.

A.2 Qualitative data collection: Semi-structured interviews

A.3 Qualitative data collection: Participant observation

A.4 Qualitative data collection: TOUT data and examining social networks

A.5 Quantitative data: High-resolution

B BACKGROUND AND CASE APPENDIX

B.1 List of acronyms

B.2 Primer on Nigerian and Lagos politics

¹⁵ In contrast to other West African countries Guinea, Burkina Faso, Cote d'Ivoire, Gambia, Senegal,
¹⁶ and Togo, no Nigerian chief executive has yet prompted a crisis by attempting to extend their
¹⁷ term past constitutionally-protected limits. In contrast to Burkina Faso, Mali, Niger, Chad, and
¹⁸ Guinea, Nigeria's democracy has not fallen to a coup in the past three years. In contrast to every

Table B.1: List of acronyms

APC	All Progressives' Congress
LASPG	Lagos State Parks and Garages
LSG	Lagos State Government
MOALS	Motorcycle Owners and Operators Association of Lagos State
NGN	Nigerian Naira
NURTW	National Union of Road and Transport Workers
PDP	People's Democratic Party
RTEAN	Road Transport Employers' Association of Nigeria
TOOAN	Tricycle Owners and Operators Association of Nigeria

¹ other country in the region, Nigeria's system is deeply federalist, in which state governments and
² autonomous communities flourish as powerful centers of government. I illustrate this logic in
³ the empirical context of Nigeria, a federalist, weakly-institutionalized democracy with a long (if
⁴ frequently interrupted) history of elections.

⁵ Compared to many of its neighbors in West Africa and the Sahel, Nigeria's current institutions—formed
⁶ with the advent of the Fourth Republic in 1999—have withstood considerable pressure. Yet more
⁷ than quarter-century after its adoption, Nigeria's democracy remains exceptionally violent and
⁸ corrupt. Beyond the high-profile episodes of violence perpetrated by armed groups, Nigerians also
⁹ face daily exploitation, coercion, and repression at the hands of non-state actors who use force to
¹⁰ exert power and extract profit, permeating everyday life with extortion and violence. In Nigeria,
¹¹ the advent of democratic institutions has not effectively protected the country's most vulnerable
¹² from victimization by such groups.—one that saw over 3,000 civilians killed in 2024 by myriad
¹³ rebel groups, communal and political militias, and the state.⁴⁵

¹⁴ Nigeria is the 6th most populous country on earth, with more than 230 million people who
¹⁵ speak over 500 different languages. It is a weakly institutionalized democracy with high inequality
¹⁶ and extensive corruption across all levels of government. It features a federal system in which
¹⁷ states hold immense power, and in which local-level politics often ‘trickle upward,’ making salient
¹⁸ overlapping regional, ethnic, religious, and linguistic divides. This is especially so between the

⁴⁵ Analysis conducted by author using ACLED data, see ?.

¹ rural, poorer, more populous North—largely Muslim, and Hausa or Fulani—and the richer South,
² primarily Igbo in the Southeast and Yoruba to the Southwest.⁴⁶

³ Nigeria has seen long and varied periods of military rule, oscillating with short-lived democracies
⁴ and brutal civil wars. The current iteration, the Fourth Nigerian Republic, began in 1999. Nevertheless,
⁵ Nigeria's first President to be voted out of office only stepped down in 2015, when the All Progressives
⁶ Congress (APC) unseated the Peoples Democratic Party (PDP) in Nigeria's first – and so far only
⁷ – peaceful transfer of power to an opposition party. The APC, at the time a newly formed coalition
⁸ merging several opposition parties and PDP defectors, remains in power to this day– led former
⁹ governor of Lagos Bola Tinubu who was elected president in the most recent 2023 general election.
¹⁰ President Tinubu was inaugurated two days before my arrival in Lagos for fieldwork; the 2023
¹¹ election was the first in the country in which one of the major candidates was *not* a former military
¹² ruler.

¹³ President Tinubu's reign makes especially evident the key role of Lagos politics in the national
¹⁴ political scene. President Bola Tinubu is a native Lagosian, and indeed was born to the family
¹⁵ of a traditional leader (chief) of the Yoruba people in Lagos, the *iyálòrún* (a title his daughter
¹⁶ now controversially occupies). Tinubu was governor of Lagos for two terms in the immediate
¹⁷ aftermath of the transition to the fourth republic, and it is thought that he amassed most of his
¹⁸ significant wealth through his near-total control of the state during this time. He has since played
¹⁹ the role of 'kingmaker' or 'godfather' of Lagos politics. Lagos has been firmly in the APC's hands
²⁰ since its founding (in which Tinubu played a part) and Tinubu has hand-selected every governor
²¹ of the state since his own rule ended. Tinubu's influence on Lagos politics is palatable; Indeed,
²² while in Lagos I lived in an apartment on "Tinubu Rd" on the mainland – one of at least 6 streets I
²³ encountered named after the former governor, and current President.

⁴⁶Lagos is both a state and a city; Lagos *state* is broken up into 20 local government areas (LGAs) and 37 local council development areas (LCDAs), while the Lagos metropolitan area encompasses 13 of these LGAs grouped in the center, generally considered the heart of the *city* of Lagos. In keeping with general colloquialisms, and in the interest of parsimony, my references to Lagos are to the state writ large.

1 B.3 Primer on (Transport) Union Politics in Nigeria

2 Interest groups—especially organized labor and sociocultural organizations—began to emerge as
3 formidable political players in Nigeria during industrialization in the first half of the 20th century.
4 Unions in particular played a large role in Nigeria’s successful bid for independence from the
5 British in 1960 (??). In the period between decolonization and the adoption of its current system
6 in 1999, Nigeria oscillated between military rule and short-lived democratic projects. Because of
7 a concerted effort by both the British and by post-independence regimes to repress, reorganize,
8 and split unions among ethnic lines—and because regional divisions overlap significantly with
9 occupational patterns—the landscape of labor and sociocultural organizations in Nigeria is dense
10 and highly localized (??). Especially after the advent of presidential federalism in 1978, individual
11 chapters, informal community-based advocacy networks, and breakaway factions established varying
12 levels of presence and power across Nigeria’s highly disparate regions. And a perhaps surprising
13 number of these organizations act not only as advocates for their members’ interests in bargains
14 with employers and with the state, but also as what we might call ‘violence entrepreneurs;’ non-state
15 brokers who use and sell violence, extortion, or exploitation to achieve political and economic
16 ends.⁴⁷

17 The right to a union is one that is protected directly by the Nigerian constitution (TK). Literature
18 on trade union activism in developing democracies largely focuses on the combative and repressive
19 relationship between the government and trade unions (Dean, 2022); or on state infiltration into
20 unions as a form of political control (?). While umbrella unions in Nigeria often call general
21 strikes or publicly protest against government policies at the national level, this paper illuminates
22 how localized factions of labor unions can also act as strategic partners to local parties, often in
23 open defiance to parent labor organizations. Given these unions’ role as violence entrepreneurs of

⁴⁷For example, the infamous Bakassi boys of Southeast Nigeria, eventually a militia group wanted for mass extrajudicial killings, was originally organized by the Shoemakers Industrial Union as a means of protecting Ariaria Market in Aba from criminal elements (?).

¹ the state, I add to a substantial literature detailing how political militias (Carey, Mitchell and
² Scharpf, 2022; ?; Acemoglu, Robinson and Santos, 2009) and criminal groups interact with the
³ state.

⁴ The transport unions of Lagos have been powerful political players for half a century, organizations
⁵ older than Nigeria's current republic. Transportation in Lagos for many years was a public enterprise
⁶ run by the Lagos State Transport Corporation (LSTC), begun in 1958. In particular, LSTC provided
⁷ bus coverage across Lagos road networks at the end of the country's First Republic,

⁸ There are two broad 'umbrella' unions in the country: the National Labour Congress (NLC)
⁹ which generally represents *employees* across industries, and the Trade Union Congress (TUC)
¹⁰ which generally represents *employers* across industries. As such there are two major transport
¹¹ unions, each affiliated with one of these umbrellas. In general, unions in Nigeria are complex
¹² hierarchical structures. The first is the National Union of Transport and Road Workers (NURTW)
¹³ associated with the NLC, and the second is the Road Transport Employers' Association of Nigeria
¹⁴ (RTEAN) associated with the TUC. Both are national unions with state chapters across the country,
¹⁵ especially in the southwest, and *especially* in Lagos. NURTW⁴⁸ and RTEAN Lagos state chapters
¹⁶ coexist, but their operations are not divvied up as might be suggested by their umbrella organization.
¹⁷ Their membership pools are rather idiosyncratically assigned; sometimes based on this division,
¹⁸ sometimes based on location.

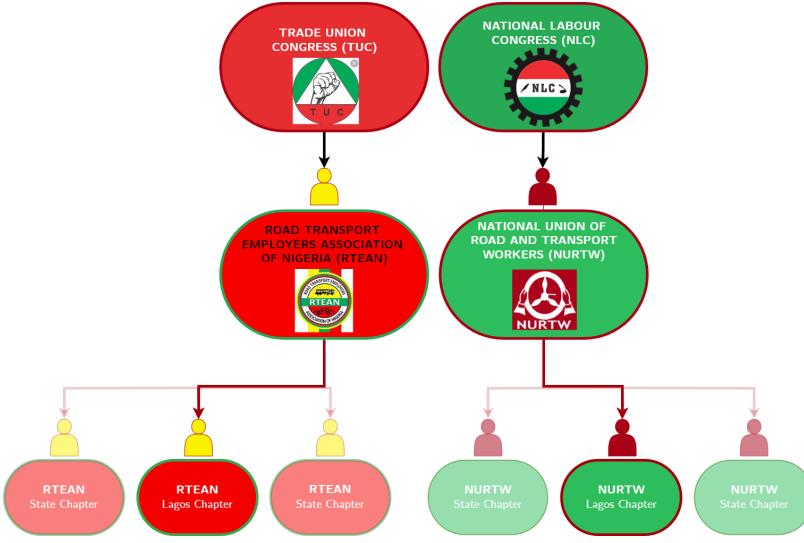
¹⁹ **B.4 Okada bans within and beyond Lagos**

²⁰ In the recent 2023 election, it was big news when one newly elected member of the parliament
²¹ admitted during his campaign that he had at one point been 'an okada man.' Many Lagosians I
²² spoke to claimed that the okada riders are "not even Nigerians," but foreigners from Niger and the
²³ Sahel who fled past Nigeria's porous borders to make their way to Lagos.⁴⁹ It is all but impossible

⁴⁸The more powerful of these unions tends to be NURTW, and when I say 'the Union,' I refer to the Lagos branch of NURTW.

⁴⁹Interviews with author; including on June 14, 2023

Figure B.1: Transport union organizational chart



- ¹ to verify this claim. Nevertheless, its omnipresence speaks to the general disdain the riders face
- ² from the population they serve. Typically, okada operators do not own the bikes they ride, but
- ³ instead borrow in exchange for a portion of their daily earnings.
- ⁴ The various fees associated with moving throughout the city mean that okada riders make a
- ⁵ meager living, and are offered no social protection or security. In general, their wages by the end
- ⁶ of the day will last them only that day.⁵⁰

⁷ C QUALITATIVE DATA APPENDIX

⁸ C.1 Transport Operators Union Tracking (TOUT) data

- ⁹ Searching online archives of mostly English-language Nigerian daily newspapers (see Table ??
- ¹⁰ over the past decade, as well as social media content, official press releases and documents, and
- ¹¹ other archival and primary material; I collect over 2,500 articles and reports regarding the politics
- ¹² of the Union and its allies over the past decade, and hundreds more pieces of primary material

⁵⁰Interviews with author, including July 7, 2023

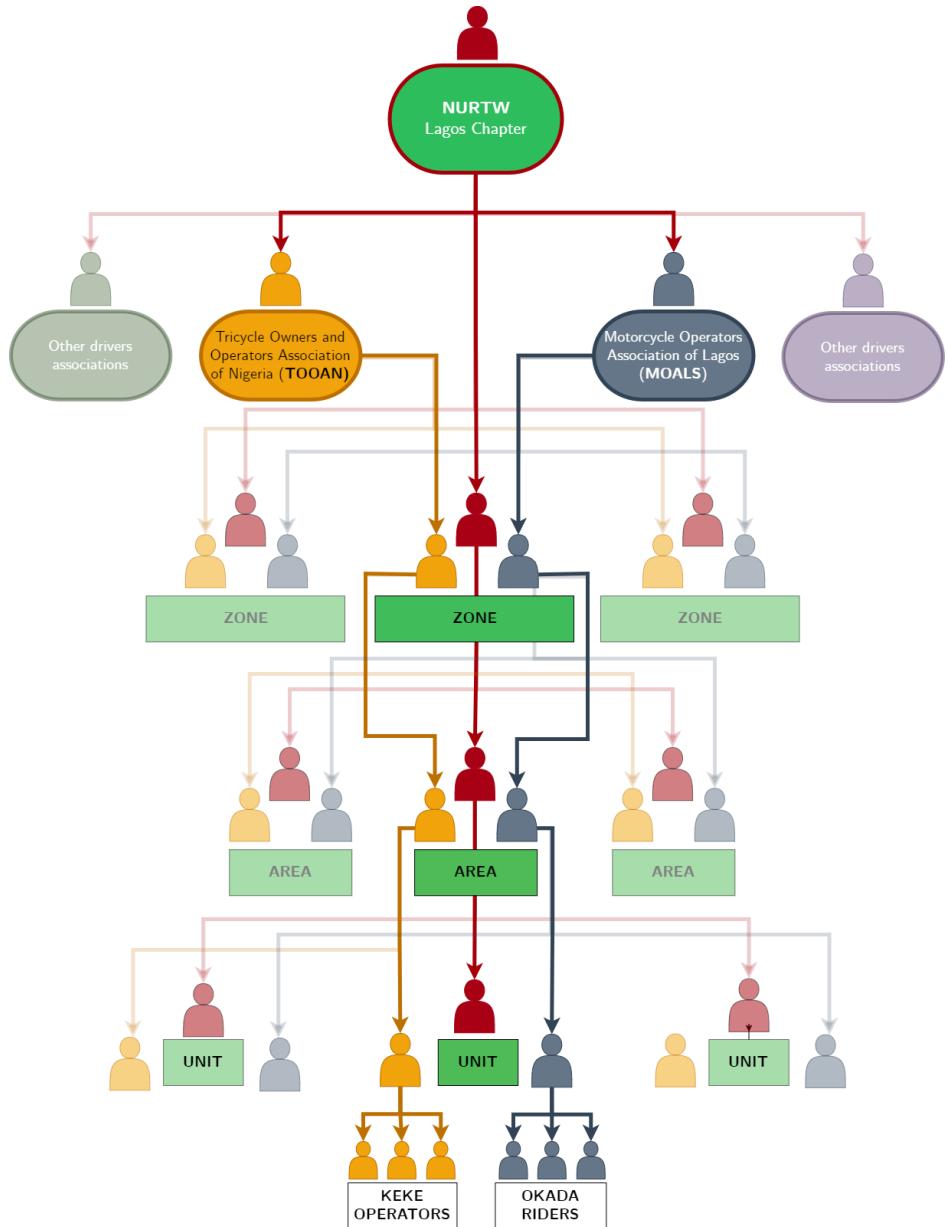


Figure B.2: Union chapter organizational chart. Each Union chapter is led by a chairman, who resides over zonal chairmen of both the parent organization (NURTW) and sub-unions (TOOAN, MOALS). Zonal chairmen exert authority over ‘area’ chairmen, and area chairmen over ‘unit’ chairmen, who rule a particular junction, intersection, or park.

2000	Tinubu comes to power, allegedly with the help of electoral interference by the Union.
2013	MC Oluomo first mentioned in Nigerian dailies as rising Union leader.
2017	MOALS (okada union) leader and Ambode supporter Rasaq Bello killed by MC's men.

Figure B.3: Timeline of (selected) pertinent events in Union history

¹ and documentary evidence. Using these sources, I build a localized event and network dataset
² for Lagos, which combines and links information on hundreds of relevant actors, places, and
³ events, covering both recent internal political dynamics of the union and events such as passage and
⁴ enforcement of the okada ban, protests by vehicle operators, battles between rival union factions,
⁵ and interpersonal rivalries and relationships. These data allow me to analyze trends in the political
⁶ dynamics between the union, the Lagos State Government, and informal commercial transport
⁷ operators qualitatively. I also utilize I refer to this database as the Transport Operator & Union
⁸ Tracking (TOUT) dataset.

⁹ A number of major newspapers, especially Nigerian dailies, were used in the construction of
¹⁰ this dataset. These include (but are not limited to): *PM News*, *This Day*, *Daily Independent*, *The*
¹¹ *Nation*, *The Will*, *The Punch*, *The Nation*, *Nigerian Tribune*, *Weekly Trust*, *International Centre*
¹² *for Investigative Reporting (ICIR)*, *Naija News*, *Daily Post*, *Times Watch*, *The Guardian*, *Sahara*
¹³ *Reporters*, *TBI Africa*, *The Independent*, *The Premium Times*, *Peoples Gazette*, *Foundation for*
¹⁴ *Investigative Journalism*, *Vanguard*, *The Daily Trust*, *News Agency of Nigeria*, *The Guild*, *British*
¹⁵ *Broadcasting Corporation (BBC)*, and *Echo News*.

¹⁶ Many sources were obtained through targeted boolean searches on LexisNexis from 2022

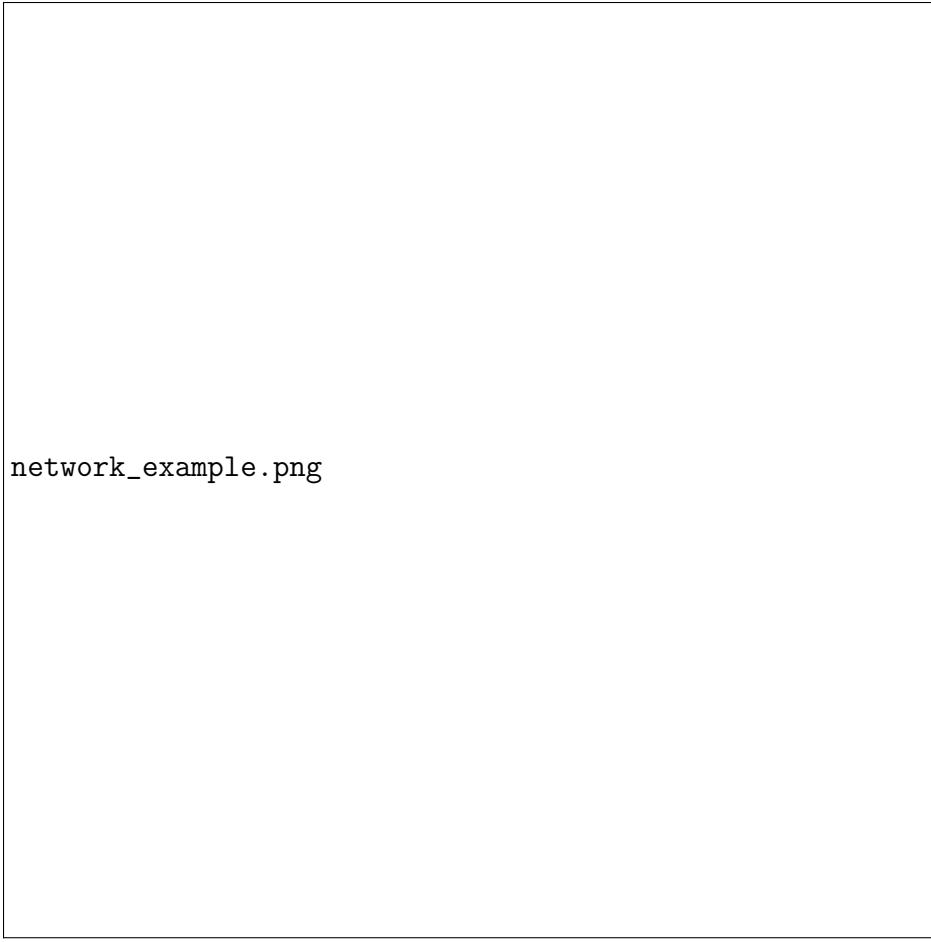


Figure C.1: A section of the networked TOUT dataset. Yellow dots represent events; blue places; red people; and white, groups.

1 through the present⁵¹. Others were provided by interview subjects or obtained during fieldwork,
2 such as videos or photographs of Union attacks on transport operators, or tickets provided to
3 operators from the Union. I also obtained some data from social media posts by both the Union's
4 official social media channels and from Union leadership; including through occasional live-streamed
5 meetings and ceremonies for which I was in (virtual) attendance.

6 The dataset includes (as of **February 2025**) roughly 200 people, 100 events and 50 organizations,
7 agencies, or groups.

⁵¹E.g., “lagos” AND “okada” OR ”motorbike“ OR “motorcycle”

¹ **Selected primary source material**

² Some multimedia sources contributed to the construction of the dataset, including posts from social
³ media and photographs taken during fieldwork. Below, I include a selection of those referenced in
⁴ the main text.

⁵ **FORTHCOMING:** Appendix In Progress.

⁶ **C.2 Interviews and selection strategy**

⁷ **FORTHCOMING:** Appendix In Progress.

⁸ Qualitative interviews about the political geography of Lagos lend sharp focus to the data
⁹ described above. Beyond that, interviews allow me to ask affected actors directly about their
¹⁰ preferences, concerns, and perceptions surrounding the ban and traffic politics in Lagos in general.
¹¹ Interviews come from directly contacting relevant individuals via email, WhatsApp, and other
¹² means; as well as from interacting with individuals over the course of their day-to-day life.⁵²

¹³ My time in Lagos yielded hundreds of anecdotes and observations, as well as informal conversations
¹⁴ with scores of individuals affected by the okada ban and transport issues more broadly in Lagos,
¹⁵ from commuters at bus stops to traffic enforcement officials at intersections.

¹⁶ **C.3 Participant observation; fieldwork sites and details**

¹⁷ **FORTHCOMING:** Appendix In Progress.

¹⁸ **Driver and rider collective action and organization**

¹⁹ **FORTHCOMING:** Appendix In Progress.

⁵²See SI §?? for more information on the qualitative data procedure I pursued, its ethical considerations, and its limitations.

¹ D QUANTITATIVE DATA APPENDIX

² D.1 Satellite imagery analysis

³ ONGOING: Appendix In Progress.

⁴ Imagery details

⁵ I purchased VHR satellite images as listed in Table D.1. Initial analysis areas are mapped in Figure
⁶ D.1.

Image	Area	Satellite	Date
1	Area 1	WV3	2022-06-13
2	Area 1	WV3	2023-03-24
3	Area 1	PNEO	2023-12-18
4	Area 1	WV3	2022-04-30
5	Area 1	WV3	2021-05-23
6	Area 2	WV2 HD	2023-02-2
7	Area 2	WV3	2023-03-24
8	Area 2	WV3	2023-03-17
9	Area 2	PNEO	2022-06-02
10	Area 2	WV3	2021-05-23
11	Area 2	PNEO	2023-12-13
12	Area 2	PNEO	2023-06-25
13	Area 2	PNEO	2023-04-20

Table D.1: Satellite images obtained (as of 4 September 2024)

⁷ Manual coding procedure

⁸ Counting okada – as discussed in the body of the paper – is difficult. Figure D.2 shows an example
⁹ of what okada look like on satellite images, with likely okada circled in green. There are several
¹⁰ things we can do to make sure that, in initial manual coding, we limit both ‘Type I’ and ‘Type II’
¹¹ errors in identification.

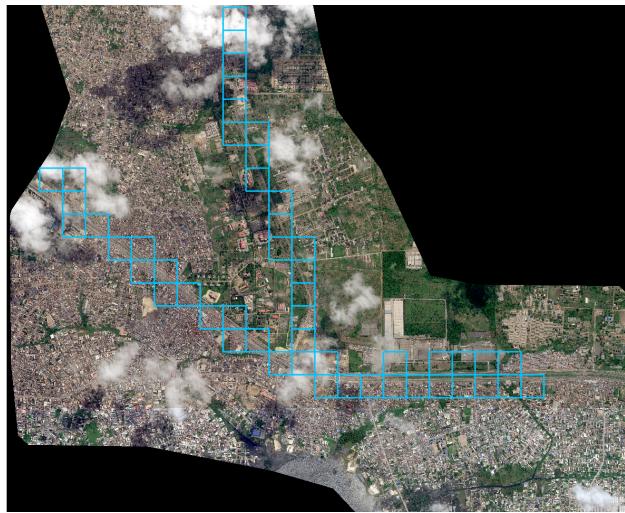


Figure D.1: Map of imagery area

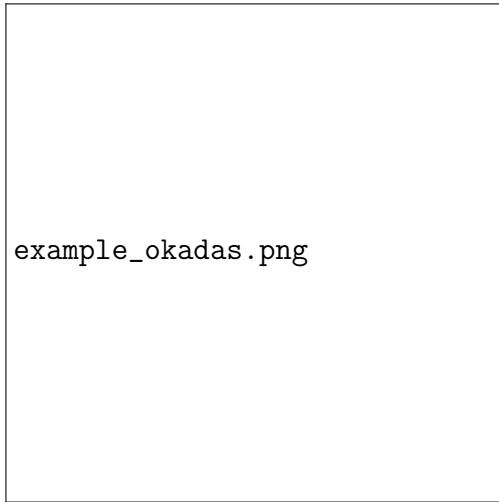


Figure D.2: Example of coding okada (over time) manually from satellite imagery.

1 First, okada are on the roads and just off them, limiting the coding scope to only the selected
2 grid squares (outlined in blue). Moreover, the size of the circled features is key. The typical
3 motorcycle in Nigeria measures 2-3m long. Given that the images pictured are 30cm per
4 pixel, the okada objects are roughly 6 pixels long. This distinguishes okada from other ‘dots on the
5 road.’ Additionally, other vehicles in the motorpark are all larger and more easily identified.⁵³ The
6 comparison helps clarify which features are indeed okada as opposed to other modes of informal
7 transit potentially located in the same area.

8 Additionally, fieldwork participant observation provided me with on-the-ground information
9 about how okada riders position themselves around the road network, and where in specific areas
10 the ‘okada zone’ is. Figure D.3 shows an image taken during fieldwork of okada riders along a road
11 in Lagos. As you can see, they form distinctive ‘clumps’ in particular areas of the road, including
12 along highway dividers and at intersections. The green-circled features take a similar formation.
13 These qualities help distinguish features that are likely okada during manual coding, from other
14 features likely not okada.

15 Relatedly, information obtained from Google Streetview data help validate this measurement
16 strategy (see Figure D.4) In particular, combinations between likely okada from VHR satellite
17 imagery are compared to similarly-dated streetview images of the same area. If, for example, a
18 streetview image shows a cluster of okada riders in a specific area, and this matches roughly the
19 cluster shown from above, it is helpful validation. Street-view imagery can also show roughly
20 where okada riders tend to be located in particular areas, and thus help adjudicate difficult cases.

21 Using these heuristics, I manually code 250 randomly selected⁵⁴ $0.5 \times 0.5\text{km}$ grid squares.
22 Selection is via random selection of grid square id numbers which contain some stretch of road or
23 off-road structure. Coding is done three separate times by me; without consulting previous totals
24 or with any knowledge (until the validation stage) of where precisely the grid-square is located or

⁵³FORTHCOMING: Particular dimensions and examples of coding these.

⁵⁴with the exception of the first 50 squares, chosen purposefully for variation in the variables of interest and for proof of concept



Figure D.3: (Blurry) photo of okada riders clustered around an intersection



Figure D.4: Example of Google Streetview image showing okada riders.

¹ the value of any relevant independent variables.⁵⁵

² In analyses, I drop grid square-dates where there is significant cloud cover. I also perform
³ additional specifications which include additional measures from Google Maps VHR satellite
⁴ imagery data.⁵⁶

⁵ Computer vision

⁶ FORTHCOMING: Appendix In Progress.

⁷ I adapt an existing algorithm FORTHCOMING to recognize features as okada automatically
⁸ based on the training set manually coded above.

⁹ Because of the rather complex and context-dependent nature of the coding of okada, there are
¹⁰ several restrictions on my ability to utilize computer vision models to code these data. For one,
¹¹ Google Earth Historical imagery is viewable, but not downloadable so all data labelled "historical
¹² imagery" was hand-coded by me.

¹³ I take my road network dataset, which as described in Section ?? consists of Open Street map
¹⁴ data separated into road segments according to junctions in the road. I 'buffer' this line data into
¹⁵ polygons, meaning that each road segment 'edge' becomes a polygon which goes five meters out
¹⁶ each side. This represents a pre-processing segmentation step, to avoid difficulties that can arise
¹⁷ from models incorrectly placing okada on areas that are not roads. Because okada are so small and
¹⁸ coding them is such a context-dependent procedure – they look at first glance like ants on the road
¹⁹ – this step is necessary for ensuring that other similar shapes in the satellite imagery are not falsely
²⁰ coded as okada despite not being on road segments.

²¹ I then overlay these polygons on satellite imagery. Purchased satellite imagery are in 'raster'
²² format, meaning they are detailed geo-coded mosaic images with several 'bands' merged together
²³ to create an image. I take these images and split them into smaller individual images, each
²⁴ associated with a specific road segment. These steps are all done in ArcGIS with ArcPy python

⁵⁵Namely; whether the grid square contains a motorpark or is an ITE-area of the okada ban.

⁵⁶FORTHCOMING: Results will be in Tables ?? and ??

¹ code as well as built-in Esri functions.

² Each of these individual images represent then a (literal) snapshot of a particular road segment
³ on the date the satellite image was taken. Using Label Studio, an open-source image annotation
⁴ software⁵⁷, I manually circle okada on a sample of the road segments collected. Image ?? shows
⁵ an example of this process, which is described further above.

⁶ I use this manually coded sample to train a detectron2⁵⁸ model to code the rest of the road
⁷ segment images for that date.

⁸ Because each satellite image has different qualities as described in Table ??, including different
⁹ times, reflectance, and resolution, images have slightly different qualities which may affect the
¹⁰ model's ability to interpret the image. Therefore, I repeat this process for every purchased satellite
¹¹ image, rather than combining them into one fully-trained mode. This means that each individual
¹² satellite image listed in table ?? has its own trained model. All satellite analysis includes image/date
¹³ fixed effects to account for any differences that arise in this regard.

¹⁴ Because of the complex technical nature of this exercise, as well as the 'black-box' processes
¹⁵ inherent in machine learning algorithms designed for image recognition, I also code 10km2 by
¹⁶ hand, and re-run analyses on this small subsample of images.

¹⁷ D.2 Streetview imagery analysis

¹⁸ I utilize Google Streetview historical and recent imagery to get extremely fine-grained geospatial
¹⁹ estimates of a few variables, including commercial transit;

²⁰ There are advantages and disadvantages to Streetview data as a source. For one, Streetview
²¹ requires Google staff to drive a car with a camera down every road, recording panorama images as
²² it goes. This requires that Google's vehicle can make it down the road; and that conditions are such
²³ that this is reasonably possible. This eliminates some roads nonrandomly from the sample; namely

⁵⁷

⁵⁸

¹ those which are smaller, in worse repair, or peripheral. Moreover, in part due to these restrictive
² scope conditions, Google streetview routes are not recorded often. Most locations in Lagos have
³ TK images associated from them, as much as ten years apart. While the data is rich in terms of
⁴ spatial granularity; it is sparse temporally.

⁵ What Streetview fails to provide in terms of comprehensiveness, it makes up for in the richness
⁶ of data it can provide at a hyper-granular spatial unit of analysis. Streetview panoramas. Previous
⁷ research in Political Science has used Streetview data to capture the locations of surveillance
⁸ cameras in China over time; other disciplines have used it to study TK, TK, TK.

⁹ **Imagery details**

¹⁰ I scrape Google Streetview imagery using Google's own API and streetview.

¹¹ **Manual coding procedure**

¹² I begin with manual coding of various particular characteristics

¹³ TK Example image one

¹⁴ TK Example image two

¹⁵ **OBIA algorithm**

¹⁶ **D.3 Alternative explanations and specifications for satellite imagery analysis**

¹⁸ **FORTHCOMING:** Appendix In Progress.

D.4 Road network analysis

Constructing the road network

I build and utilize an original R package (`road-networks`) in order to create a road network dataset based on Lagos road data.⁵⁹ The package takes data from Open Street Maps (OSM) open source geospatial road data, and transforms these data into a road network. OSM data provide information on the geography of roads of all types – from major highways to tertiary trails and footpaths – in the form of spatial files of polyline data, one line (row) for each road. OSM also provides information about each road, including classifications according to road type (primary, footway, path, motorway, etc.), road name, and other characteristics, including whether the road is a tunnel or a bridge, and whether traffic is bidirectional or one-way.

I utilize these latter three characteristics in particular to construct a spatial road network.⁶⁰ Namely, I abstract the map of roads in Lagos into a network of edges, representing the roads themselves, which are connected by nodes – in this case, placeholders for intersections and the ends and beginnings of particular streets. The network is determined entirely by geospatial characteristics: that is, the length of a road ‘edge’ between two intersection ‘nodes’ is simply the length of the road. Figure D.5 illustrates this process of segmenting the Lagos road network into edges and nodes.

However, the process of transforming OSM data into a network is not straightforward, and requires several discrete steps. Simply taking OSM data as they are and segmenting each road, at every place where they intersect or overlap, is insufficient. To see why, Figure D.6 shows this same 10km² OSM data displayed in Figure D.5, overlaid on a graphical representation of this road

⁵⁹This package is adapted from ?’s protocol in ArcGIS which creates an `arc.py` network dataset based on OSM street data.

⁶⁰By ‘road network,’ I mean an abstraction of the road polylines into ‘edges,’ connected to each other through ‘nodes.’ In a traditional network (or ‘graph’) set-up, nodes are objects of interest – for example, members of a community – and edges are connections between nodes – for example, relationships between community members. Both nodes and edges have characteristics which can be analyzed. For example, the edge between the nodes representing a parent *A* and child *B* might be short, indicating a very close relationship; while the edge between *A* and the local banker *C* might be longer, indicating a more distant connection. Nodes can also be connected to each other through other nodes; that is, the relationship between *A* and a second cousin *D* passes through another node: a shared great grandparent. Of course, networks are not always between people, and in what follows, the network I construct and examine is a spatial one.



(a) Road data, features abstracted, containing a multitude of separate roads connected through intersections.

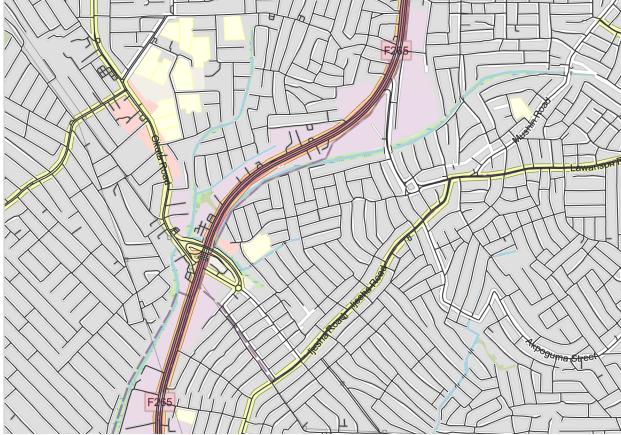


(b) This same road data, where separate roads are differently colored. Each separate road segment represents an ‘edge.’



(c) This same road data where ‘nodes’ are added for each location where edges start, end, or intersect with each other.

Figure D.5: Random selection of OSM road data in Lagos, Nigeria (roughly 10km^2), through the process of being constructed into a road network.



(a) Road data (black lines) over OSM basemap.



Figure D.6: Lagos roads with OSM basemap, zoomed in to one intersection. Each colored line (red, orange, etc.) represents a separate road in OSM data.

¹ network. For the sake of clarity, the second image zooms in further, to one particular intersection
² in mainland Lagos.

³ A naive division of the road data into edges and nodes would split roads at every location they
⁴ intersect with other roads. Figure D.7 shows the results of such a naive division, with nodes drawn
⁵ in blue. Such a division presents problems for analysis (?).

⁶ For example, take the node drawn furthest to top the top of the figure, drawn at the intersection
⁷ between the yellow road and the red road. If abstracted into a network, the creation of this node
⁸ would artificially create a connection between the red and yellow roads at this point, allowing travel

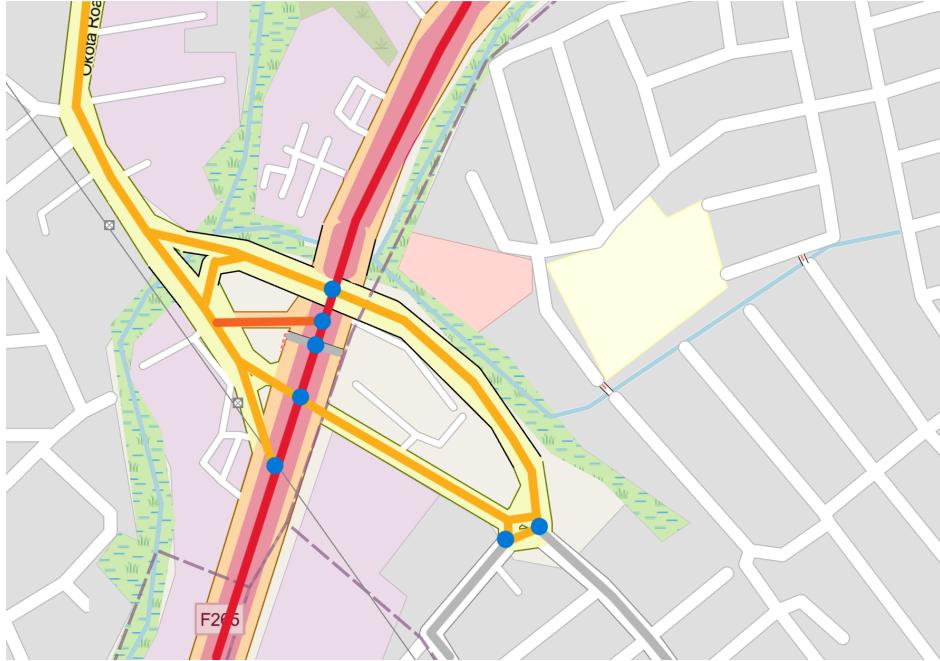


Figure D.7: Roads marked with where they intersect other roads (marked with blue dots)

1 on the network directly from the yellow road to the red road at this point; that is, travelling from
 2 west to east along the yellow road, this node would allow a traveller along the network to perform
 3 a sharp turn to the left to continue travel on the red road from this point. Figure D.8 shows this
 4 example, where the black line with arrows demonstrates how a traveller along the network could
 5 move between the roads at this node.

6 The underlying basemap illustrates why this is an issue: namely, despite the fact that the red
 7 and yellow roads technically intersect in a 2D space, in reality the yellow road represents a bridge
 8 *over* the red road. In reality, a traveller along the network could not pursue the route represented in
 9 D.8 without jumping off the yellow bridge, and onto the red road below. This node therefore falsely
 10 creates a direct path between these roads when in fact, none exist. In reality, a traveller would have
 11 to take a much more roundabout route in order to move from this node on the yellow bridge, to the
 12 red road below. Similar mistakes are made at other supposed intersections represented by the other
 13 blue nodes. D.9 Xes out such nodes, which are inappropriately created at several points in which
 14 a 2D representation of roads results in the creation of intersections which do not actually represent



Figure D.8: Example route if considering all intersections (blue dots) as ‘nodes’ in the road network

¹ connections between roads.

² As such, transforming OSM data into a road network requires differentiating between genuine
³ intersections in which roads are connected to each other, and ‘false’ intersections, where seemingly
⁴ overlapping roads in a 2D space are in fact not connected in a way that allows traversing the road
⁵ network through such an intersection.

⁶ The R package I have created⁶¹ addresses this issue. Specifically, the package separates OSM
⁷ data according to road classifications as bridges and tunnels. It then transforms each class of road
⁸ into nodes and edges separately, before combining (based on OSM locally-constructed ‘layer’)
⁹ data to ensure only genuine intersections) into one interconnected network dataset, resulting in a
¹⁰ network which does not include such ‘false intersections.’

¹¹ The final issue I resolve with this package is the taking into account one-way streets. Networks
¹² can be either ‘directed’ or ‘undirected’: ‘Undirected’ networks imply that edges between nodes

⁶¹ Adapted from the logic proposed by ? in their ArcGIS package

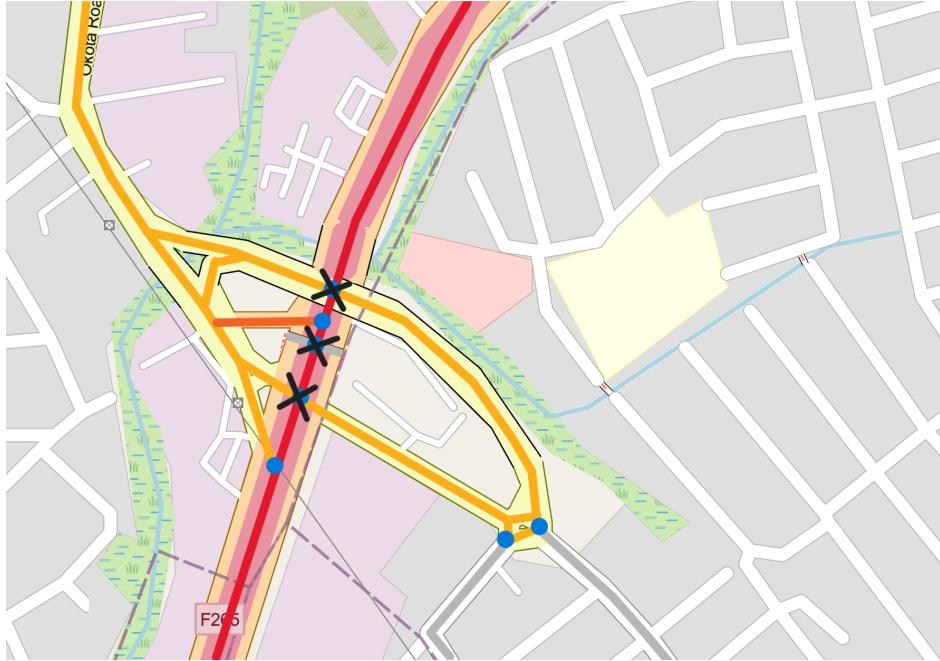


Figure D.9: Intersections (blue dots) xed out if they should not be considered ‘nodes’ in the road network.

¹ operate in both directions, and that a traveller along the network can move from point A to point B
² via the same route as from point B to point A, while ‘directed’ networks imply that each edge has
³ a start node and an end node, and cannot be traversed in the opposite direction. Road networks are
⁴ by their nature, complex, containing both directed edges (namely, one-way streets) and undirected
⁵ (two-way). I take this complexity into account by duplicating all edges that represent two-way
⁶ streets, reversing start and end nodes, allowing for the creation of a ‘directed’ graph in which
⁷ bidirectional travel is allowed for two-way roads through the creation of pairs of edges.

⁸ Road network randomization

⁹ My measurement strategy calculates the change in routes along the road network between pairs
¹⁰ of points before and after the okada ban, if simulating shortest distances between those points. I
¹¹ propose that the ban will a) displace riders from non-motorpark areas to motorparks; b) curtail
¹² freedom of movement generally; and c) increase the cost of the average ride. I pursue this strategy

¹ in several steps.

² The first step is to establish baseline routes undertaken by okada riders before the ban. To this
³ end, I select a random pair of discrete points from a uniform distribution along the road network,
⁴ simply meaning that every point on the road network has equal chance of being selected.⁶² These
⁵ points represent a potential ‘start point’ (i) and ‘end point’ (k) of an okada’s journey, between
⁶ which okada riders may be hired to take passengers. I utilize network analysis tools to evaluate
⁷ the shortest path⁶³ that can be taken between each of these start and end points for an okada rider
⁸ along the road network. I calculate the *cost* of each route, a function of both the length in meters
⁹ of the route, and the number of motorparks passed, as demonstrated below in Equation (5):

$$(5) \quad \text{Cost}_{i,k}^{pre} = \min_{P \in \mathcal{P}_{i,k}} \left(\sum_{e \in P} \ell(e) \times r + \sum_{e \in P \cap m} f \right)$$

¹⁰ Here, $\text{Cost}_{i,k}$ is the cost of traversing from point i to point k and $\mathcal{P}_{i,k}$ represents the set of all
¹¹ possible paths between point i and point k , where each path \mathcal{P} is made up of multiple edges e .
¹² $\ell(e)$ is the length, in meters, of a given edge. Simply, the formula clarifies that the *cost* of a journey
¹³ from point i to point k on the road network is the path which minimizes the length of the edges
¹⁴ traversed. An assumption here is that the cost to an okada rider of a particular route determined
¹⁵ by how far they have to travel along the road network. This cost-per-meter is represented by
¹⁶ ‘rate’ variable r . If $e \in m$, this simply means that that edge houses a motorpark. For every edge
¹⁷ which passes a motorpark in a given path P , an extra cost is incurred represented by ‘fee’ variable
¹⁸ f . Based on extensive interviews with okada riders, as well as rides on okada, I estimate the
¹⁹ approximate cost-per-kilometer travelled r at NGN 100 (₦; as of 23 April 2023 was approximately
²⁰ NGN 1,282.00 to USD 1.00). While the price of tickets f also vary over time, interviews suggest

⁶²**FORTHCOMING:** Additional sets of simulations outlined in section D.4 will relax this uniformity assumption in order to more closely mimic actual traffic and travel throughout Lagos. However, this initial uniform distribution assumption is not necessarily problematic; indeed, one of the points of okada is that they were commonly available all over the city to go all sorts of distances.

⁶³Specifically, I use `shortest_paths` and `distances` functions from R package `igraph`.

1 they seem to hover around NGN 500 per motorpark. This allows me to take into account the
 2 incentive okada riders have to avoid passage through motorparks, even if they lie on the shortest
 3 possible route between points.

4 The second step of my analysis takes explicit account of the okada ban. I rerun shortest distance
 5 calculations between these same random points i and k ; however this time I account for areas
 6 which are enforced during the course of the ban. Equation (??) simply states that in calculating the
 7 shortest possible path between points i, k after the ban, I discard paths in which any edges e are in
 8 the set b , where b is enforced areas ($e \notin b$).

$$(6) \quad \text{Cost}_{i,k}^{post} = \min_{P \in \mathcal{P}_{i,k}} \left\{ \left(\sum_{e \in P} \ell(e) \times r + \sum_{e \in P \cap m} f \right) : e \notin b \forall e \in P \right\}$$

9 Finally, I count the number of motorparks passed for each route i, k as demonstrated in (7).

$$(7) \quad \text{Count}_{i,k}^{pre(post)}(m) = \min_{P \in \mathcal{P}_{i,k}} \left(\sum_{e \in P \cap m} 1 \right)$$

10 I conduct the above analysis across $10,000^{64}$ simulated routes between different start and
 11 end points on the road network of Lagos, which allows me to compare several telling route
 12 characteristics between the period prior to and after the okada ban enforcement. Most simply, I can
 13 calculate the cost of the okada ban for riders, represented by (8), by simply taking the difference
 14 between the costs after the ban and the costs before the ban. This takes into account both the
 15 extra cost of additional motorpark tickets incurred by riders, and the extra mileage required to
 16 circumvent enforced areas.

⁶⁴**FORTHCOMING:** Analysis is ongoing: the simulation count currently is only 1,000 because of the computational intensity required.

$$(8) \quad \text{Cost of Okada Ban} = \sum_{i,k} \left(\text{Cost}_{i,k}^{post} - \text{Cost}_{i,k}^{pre} \right)$$

¹ More precisely related to my hypothesis, I calculate the difference in the number of motorparks
² passed post-ban from pre-ban (Equation (9)), as well as the share of all shortest routes between all
³ points i, k that pass at least one motorpark before and after the ban enforcement (Equation (10)).

$$(9) \quad \text{Difference in motorparks passed after ban} m = \sum_{i,k} \left(\text{Count}_{i,k}^{post}(m) - \text{Count}_{i,k}^{pre}(m) \right)$$

$$(10) \quad \text{Prop}^{pre(post)} = \frac{\sum_{i,k} \min_{P \in \mathcal{P}_{i,k}} (\mathbf{1}_{\{|P \cap m| > 0\}})}{\sum_{i,k} 1}$$

⁴ I also provide additional descriptive statistics, including the number of routes displaced entirely
⁵ by ban enforcement (i.e., a start and end point which one cannot travel between without passing an
⁶ enforced area), and geographic displacement of riders.

⁷ I complete these simulations using the actual locations of the okada ban, allowing me to
⁸ compute the ultimate effects of selective enforcement on okada rider welfare and union profit.
⁹ However, these results do not necessarily imply that the areas of okada ban enforcement were
¹⁰ selected *in order to* benefit the union; it is possible that any positive results for the union are an
¹¹ incidental consequence of enforcement locations that were chosen for other reasons. Along with
¹² other tests for possible alternative explanations, I guard against this possibility by re-running this
¹³ road network analysis, substituting for actual enforcement areas *hypothetical* possible locations of
¹⁴ okada ban enforcement.

¹⁵ More precisely, I select (for each iteration) roughly 50 intersections, streets, bridges, and areas

1 on the road network in which the okada ban could have been – but was not – enforced. I then
2 re-run the above analysis *as if* these areas were those enforced, in order to see whether other
3 areas of enforcement result in the same effects on okada riders. This allows me to calculate
4 whether the areas in which the ban was *actually* enforced *uniquely* benefit the union, compared
5 to other hypothetical enforcement areas. This approach is analogous to one of ‘randomization
6 inference,’ in which hypothetical treatment assignments are systematically shuffled among all
7 units in a study to robustly assess the causal impact of the actual treatment. This technique
8 helps in isolating the unique contribution of the treatment by comparing outcomes under real and
9 randomized enforcement scenarios, thereby enhancing the validity of the inferences drawn from
10 the analysis.

11 I compare the effects on shortest-route distances generated by the actual okada ban to the effects
12 of hypothetical bans on 1,000 sets of roads, intersections, or other potentially enforceable areas. I
13 choose the selection of roads not randomly, but based on 96.1 FM Twitter traffic data. I find the
14 sets of roads in the highest 5% of mentions of features, coded with reference to OSM data, roughly
15 385 features, all of which are mentioned at least 50 times in the universe of tweets. I then select 50
16 features (1,000 times) and run the same calculations as above.

17 **Alternative explanations and specifications for road network analysis**

18 **FORTHCOMING:** Appendix In Progress.

19 I intend to replicate road network simulations where instead of start-points and end points being
20 generated randomly according to a uniform distribution on the road network, we generate start and
21 end points randomly from a distribution of traffic-dense areas according to Lagos traffic twitter
22 data⁶⁵

23 I also select from different distributions of road areas for ‘hypothetical’ enforcement areas
24 which align with other potential reasons to vary enforcement geographically. This analysis is

⁶⁵See Appendix D.4 for more on this process.

1 aimed at addressing potential alternative explanations, including state capacity (or enforcement
2 power). I then compare hypothetical enforcement areas to actual enforcement areas both on their
3 own and in the context of results from the road network simulations. Table ?? will have these
4 results. The first column *capacity* randomly draws hypothetical ITE areas from roads surrounding
5 police stations⁶⁶ If the explanation for okada enforcement is one of limited capacity, it is plausible
6 that enforcement may be selective, and enforcement stronger around centers of law enforcement
7 power. Another capacity-based variation in ITE areas might be enforcement focused on areas
8 where okada are most common. The second column randomly draws hypothetical ITE areas from
9 locations in which, prior to the ban, were associated with a traffic incident involving okada riders⁶⁷.

10 **D.5 Additional Twitter analyses and descriptive statistics**

11 **FORTHCOMING:** Appendix In Progress.

12 **D.6 Alternative measurement strategies and additional analyses**

13 **Alternative measures of traffic**

14 **FORTHCOMING:** Appendix In Progress.

15 **Alternative measure of motorpark locations**

16 **ONGOING:** Appendix In Progress.

17 It is possible that official coverage is systematically biased against locations where motorparks
18 operate unofficially. In an attempt to overcome this issue, I utilize lists of motorparks and garages
19 as recorded by the Lagos Waste Management Agency (LAWMA) as part of organizing daily waste
20 collection routes. LAWMA is tasked with the waste and trash removal throughout all of Lagos, and
21 is organized into six divisions (East I/II, West I/II, Central I/II) with mutually exclusive operating

⁶⁶See www.humdata.org for data on locations of police stations in Lagos.

⁶⁷See Appendix D.5

1 routes. Part of the purview of this organization is to collect and remove waste from markets and
2 motorparks – both formal and informal. As such, LAWMA lists – unpublished and internal – do
3 not distinguish between informal and formal motorparks in the state. Their job is to collect waste
4 from both. I have individually obtained lists from each of these six divisions of LAWMA. Tables
5 ?? and ?? will show alternative analyses of political and temporal variation using this revised
6 measurement strategy.

7 **Measures of union power**

8 **FORTHCOMING:** Appendix In Progress.

9 My theory essentially requires that selective enforcement patterns are responsive to union
10 power. This is the basis behind the difference-in-differences strategy explored in section ???. Here,
11 I devise another way to measure union power using engagement with the social media posts of
12 the NURTW Lagos branch leader MC Oluomo. As described in Appendix ??, social media and
13 related channels are one way in which union officials engage with their members and the world;
14 and MC Oluomo has a particularly active instagram page. Sometimes he posts official statements;
15 more often they are happy birthday wishes to family members, friends, and colleagues. I roughly
16 proxy union ‘influence’ with the number of likes, shares, or views over time. Figure ?? will show
17 how engagement with this individual’s posts wax and wane over time.

18 **D.7 Election analysis**

19 Election results in Nigeria are collated by the Independent National Electoral Commission (INEC).
20 The 2023 elections introduced new technology to that data collation process. In particular, INEC
21 introduced a system by which individual polling stations (of which there are only 13,000 in Lagos
22 alone) could upload signed PDFs of the final vote tallies for that polling station, which were read
23 in using OCR and collated automatically.

24 However, polling unit-level data on Lagos governorship election, which took place on March

1 18 2023, are more difficult to access comprehensively. While some images appear to have been
2 captured by web crawlers from original uploads, no systematic collection seems to exist. I therefore
3 resort to idiosyncratic collection of governorship election results for those polling units I can find
4 online. This includes collection through internet archives, cached pages, old news reports, and
5 social media for any images portraying polling-unit level results. I find polling unit level results
6 for roughly 150 polling units , across every local government area in Lagos and almost 70 separate
7 voting wards. I compare these to the polling-unit level results from the Presidential race a month
8 earlier. I add polling station information (and results) to the Lagos road network dataset described
9 above.

10 **E MODEL APPENDIX**

11 My theory relies on three actors: the APC which controls the Lagos state government (S), the
12 Union (U), and okada riders (R).

13 The APC's main goal is retaining control of Lagos state. Their retention comes from two
14 channels: first, support and/or electoral interference from the Union; as well as some popularity
15 penalty from disenfranchised riders or consumers of okada rides.

16 I formalize this logic in the form of a simple model, where the players are the State (S), the
17 Union (U) and okada Riders (R). In keeping with the context, riders—being primarily internally
18 displaced persons without access to voting rights in Lagos—are not strategic actors. Their choice
19 is a deterministic response to enforcement.

20 **E.1 Actors and preferences**

21 The game is played as described in the following section, but in short encompasses two stages. In
22 the first ($t = 0$), the State decides whether or not to pass the ban. If they choose not to, the Union
23 proceeds to decide its level of election violence and the game ends. If they choose instead to pass

₁ the ban, a second stage begins at $t = 1$ in which the State chooses a level of enforcement of the
₂ ban, followed by Union decisions about the level of electoral violence to perpetrate. This second
₃ stage then infinitely repeats with a common discount rate. Each actor's payoff functions in every
₄ time period in the second stage, for all $t \geq 1$ is as follows. For the state:

$$\pi_S^t = V_0 + \alpha v_t - C(e_t)$$

₅ or the total number of votes they get at baseline (V_0), plus the total gain from the election violence
₆ perpetrated by the Union in this time period (αv_t), minus whatever they spend on enforcement in
₇ this period $C(e_t)$. The Union meanwhile has payoff function:

$$\pi_U^t = x(b, e_t) r^{in}(b, e_t) - K_{v_t}$$

₈ Which is given by the tax rate they charge per ride within their territory, based on ban passage and
₉ enforcement in that time period; times the number of rides within their territory taken by Riders in
₁₀ this time period, minus the total amount they spend on election violence in this time period.

₁₁ Riders, though they are not strategic actors in the game, have the following payoff function:

$$\pi_R^t = p[r_{in}(b, e_t) + r_{out}(b, e_t)] - x(b, e_t) r_{in}(b, e_t)$$

₁₂ Or the total amount they make per rides within and outside Union territory, minus the tax they pay
₁₃ per ride within Union territory.

₁₄ **E.2 Model setup**

₁₅ Consider a game that proceeds as follows:

1. The state S makes the first move, where it chooses whether or not to pass a law banning

okada riding across Lagos:

$$b \in \{0, 1\}$$

where b is the choice of whether or not to ban riders; $b = 0$ means the State does not pass the ban; and $b = 1$ means that it does.

If the state chooses $b = 0$, the game proceeds to the Union's decision on whether or not to engage in election violence $e \in [0, 1]$, described in Step 3 below. Otherwise, after the first one-time commitment stage (the State (S) choosing whether or not to implement a ban) the remaining stages of the game are infinitely repeated:

2. If $b = 1$ and the ban is passed, the State then chooses the intensity of its enforcement of the ban:

$$e \in [0, 1]$$

where $e = 0$ means the ban is not at all enforced; and $e = 1$ means the ban is fully enforced everywhere (with the full available capacity of the State). If $e = 1$, the number of rides taken by Riders is driven to 0. Where $0 < e < 1$, the State engages in *selective* enforcement of the ban.

3. Once the state chooses either not to pass the ban ($b = 0$) or chooses its level of enforcement once the ban is passed ($b = 1$ and $e \in [0, 1]$), the Union chooses its electoral violence effort on behalf of the State:

$$v \in \{0, 1\}$$

These stages 2,3 are infinitely repeated with a common discount factor of $\delta \in (0, 1)$. The payoffs for the first period are as follows. For the State (S):

$$\pi_S = V_0 + \alpha v - C(e)$$

- ¹ where V_0 is the state's baseline votes (assumed for simplicity here to be exogenously determined);
² $\alpha > 0$ is the marginal vote-gain from the Union (U)'s use of electoral violence $v = 1$, and $C(e)$ is
³ the cost to the State (S) of enforcing the okada ban.

For the Union (U):

$$\pi_U = x(b)r(e) - K_v$$

- ⁴ where, as stated above, $x(b)$ is the per-ride tax on every ride routed through Union (U) territory
⁵ and $r(e)$ is the number of rides riders take per period given the level of state enforcement e . Here
⁶ K_v represents the political or economic cost of perpetrating election violence on behalf of the state
⁷ ($v = 1$).

For Riders (R):

$$\pi_R = pr(e) - x(b)r(e)$$

- ⁸ where p is the per-ride profit times the number of rides $r(e)$, and $x(b)r(e)$ is the amount of taxes
⁹ paid by Riders to the Union (U).

¹⁰ E.3 Actor incentive constraints and equilibrium profiles

- ¹¹ At $t = 0$, the state chooses whether ($b = 1$) or not ($b = 0$) to ban okada by comparing the no-ban
¹² path and the ban path.

The Union (U) has an incentive to play $v = 1$ on-path. With cooperation (playing $v = 1$ in every period), the stream of extortion products will be equal to

$$\frac{Xr(e*)}{1 - \delta}$$

But if the Union (U) deviates by playing $v = 0$ once, they will gain $Xr(e*)$ in that period, but trigger punishment which will bring their future extortion down to 0. Their incentive constraint

Term	Type	Meaning, Value
$b \in \{0, 1\}$	Decision	Ban passage by State (S) in period 0, where $b = 0$ means no ban
$e_t \in [0, e*, 1]$	Decision	Level of enforcement chosen by State (S) in period t
$v_t \in \{0, 1\}$	Decision	Level of electoral violence perpetrated by Union (U) on behalf of State (S) in period t
$r(b, e)$	Decision (constrained)	Total number of rides taken by riders given ban and enforcement level (b, e)
$***R_{max}$	***	Maximum number of rides which can be taken by Riders (R) in a given period
$r^{in}(b, e)$	Decision (constrained)	Number of rides taken by Riders (R) <i>within</i> Union territory, given ban and enforcement level (b, e)
$r^{out}(b, e)$	Decision (constrained)	Number of rides taken by Riders (R) <i>outside</i> Union territory, given ban and enforcement level (b, e)
$***r(b, e)$	***	Number of rides taken by riders in a given period given ban and enforcement level (b, e)
r_0		Baseline total number of rides per rider before any ban or enforcement
$x(b, e)$		Per-ride tax levied by the Union (U) on rides routed through Union territory given ban and enforcement level (b, e)
x_0		Baseline per-ride tax levied by the Union before any ban or enforcement
V_0		The baseline level of votes for the State (S), before any ban or enforcement
α		The marginal vote gain for the State (S) from election violence per unit increase in enforcement level
$C(e)$		The cost to the State (S) of enforcement of election violence at level e
K_v		The cost to the Union (U) of exacting election violence on the State (S)
p		The per-ride profit to okada Riders (R) on each ride
δ		The common discount factor per-period

Table E.1: Caption

Case	Number of rides (r)	Per-ride tax by Union (x)
No ban ($b = 0$)	$r(0, 0) = r_0$	$x(0, 0) = x_0$
Ban, no enforcement ($b = 1, e = 0$)	$r(1, 0) = r_0$	$x(1, 0) > x_0$
Ban, selective enforcement ($b = 1, e = e^*$)	$r(1, e^*) > r_0$	$x(1, e^*) > x_0$
Ban, full enforcement ($b = 1, e = 1$)	$r(1, 1) = 0$	No rides; no tax

Table E.2: Caption

can therefore be formulated as:

$$IC_U : \frac{Xr(e*)}{1 - \delta} \geq Xr(e*) + \delta * 0 \Leftrightarrow \delta \geq 0$$

- ¹ This incentive constraint always holds as long as $\delta > 0$.

The State (S) has an incentive to play $b = 1, e = e^*$ on-path. When they cooperate (by passing the ban and engaging in selective enforcement), the state receives:

$$\frac{V_0 + \alpha - C(e^*)}{1 - \delta}$$

If the State deviates to no ban $b = 0$ or to some level of enforcement $e \neq e^*$ once, it triggers immediate per-period payoff changes where $v = 0$ in every period thereafter, costing α in every single future period. So the precise incentive constraint can be formulated as:

$$\frac{V_0 + \alpha - C(e^*)}{1 - \delta} \geq [V_0 - C(\hat{e})] + \frac{\delta(V_0 - C(1))}{1 - \delta}$$

- ² where \hat{e} is the deviating enforcement and $C(1)$ is the full-enforcement cost in the punishment phase.

Given this set-up, a selective-enforcement equilibrium is sustained by a grim-trigger strategy where in every period:

$$b = 1, e = e^* \in (0, 1)$$

where e^* is chosen to maximize the State's per-period payoff, subject to keeping Riders active so that $r(e^*) > 0$ and therefore the Union can still extort

$$e^* = \underset{e \in [0,1)}{\operatorname{argmax}} \{V_0 + \alpha - C(e)\}$$

- ⁴ subject to $r(e) > 0$.

¹ If the Union (U) ever plays $v = 0$, the state will revert to full enforcement $e = 1$ forever,
² meaning the Union's extortion revenue will be driven to 0. If the State (S) ever plays $b = 0$ or
³ picks $e \neq e^*$, the Union (U) will refuse to provide election violence ($v = 0$) forever, and the State
⁴ will lose α (the gains from this electoral violence) at every future period. Therefore:

$$SPNE : (b = 1, e = e^*, v = 1) \text{ on path (with grim-trigger punishments off-path)}$$

⁵ E.4 With Grim-Trigger Punishment

⁶ As long as $\alpha > C(e^*)$ and $K > 0$, Our grim-trigger punishment setup requires that after a deviation
⁷ by either actor, the other reverts to non-cooperation. For the Union, this means not participating in
⁸ election violence. For the State, this means not enforcing the ban; the cheapest option