

# Public Roads on Private Lands: Land Costs and Optimal Road Improvements in Urban Uganda

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## Abstract

High land acquisition costs often undermine the net returns and hinder the implementation of high-benefit infrastructure projects in developing cities. I examine this issue in Kampala, Uganda, using new survey data on real-estate brokers and landowners to evaluate the costs and benefits of 140 km of road improvements implemented since 2017. Leveraging the timing of improvements, I estimate their local benefits, and I develop a quantitative spatial model to account for the general equilibrium gains, including spillovers and the fiscal cost of land acquisition. Weak property rights enable the government to acquire land at reduced cost, acting as a subsidy to the government. Combined with a high cost of raising domestic funds, these weak property rights support more extensive road improvements with city-wide benefits. In turn, the realized road improvements yield net welfare gains equivalent to 99 USD per Kampala resident, compared to 40 USD if land had been acquired at market value, as legally mandated under eminent domain. Finally, I solve for the optimal road improvements under different institutional settings and funding restrictions, offering insights for policy design to enhance allocative efficiency and welfare outcomes.

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# 1 Introduction

In African cities, only one-third of roads are paved (Kumar and Barrett 2008). Given the link between urban mobility and economic development (Akbar et al. 2023), road improvements may yield significant welfare benefits. Consequently, transportation infrastructure upgrades have become a priority for both domestic governments and international agencies. However, for African governments attempting to improve these roads, such projects also entail substantial costs. For example, while international agencies often fund construction expenditures, domestic governments struggle to raise funds to acquire the underlying land on which the roads are built from private landowners.

Despite the importance of these costs, they have often been overlooked by the literature and by policy makers designing these projects. This oversight has had major consequences. Governments' inability to raise the funds needed to acquire this underlying land has led to costly delays, project revisions, and diminished net returns (World-Bank 1996). Better understanding the drivers of land acquisition costs in African cities is crucial to explaining the lack or suboptimal allocation of high-benefit infrastructure projects.

In this paper, I address this gap by examining a series of road improvement projects in Kampala, Uganda. I analyze how local government land acquisition practices affect the net returns to these improvements. With less than 45% of its major roads paved, Kampala resembles many other African cities. To enhance the quality of the road network, 140 km of road upgrades have been rolled out since 2017. Construction has been primarily funded by the World Bank and the African Development Bank, while the domestic government has been responsible for acquiring the underlying land. Consistent with a high wedge on tax revenues (Regan and Manwaring 2024), the government cited concerns that compensating landowners at market value, as required by the eminent domain legal framework, would jeopardize the project's full implementation (World-Bank 2023), and asked owners to voluntarily give up small portions of their land without compensation. Under this "*voluntary land take approach*", owners are only compensated if they incur a fixed cost to negotiate with the government. However, the success of this approach varied across the city, with owners in areas with stronger property rights less likely to participate and more inclined to seek the financial compensation to which they are legally entitled.

To study the impacts of the road improvement projects and the associated land acquisition challenges in this data-scarce environment, I begin by integrating several data sources. I collect new data on the property-level benefits and land costs of road improvements throughout the city, including two novel surveys that I collected with real estate brokers and landowners. I then use data on commuting patterns from a local ride sharing company and traffic data from Google Maps to estimate the reduced-form benefits of road improvements initiated in 2017, exploiting variation in the timing of the policy. I leverage the coexistence of three property right regimes in the city (Bird and Venables 2020) to characterize how land costs are influenced by land institutions. Next, I build and estimate a quantitative spatial model that accounts for changes in traffic flows and adjustments in the locations where residents live and work to measure the city-level returns of road improvements. Finally, I solve the model for the optimal road improvements under different institutional frameworks and in the absence of restrictions on the use of external funds for land acquisition, to evaluate how these institutional differences affect the location of

road improvements.

The main result of my paper is to show how weak property rights enable governments to provide more transportation infrastructures with large positive externalities, increasing net welfare gains, a result consistent with theoretical work on the conflict between private property rights and public good provision (Acemoglu and Robinson 2012, Posner and Weyl 2017). I first show that the road improvement projects in Kampala yielded large benefits in terms of decreases in commuting time and increases in property values. These high benefits translated into positive net returns only because the government was able to acquire most of the underlying land without compensating landowners, instead relying on weak property rights to facilitate cheaper and easier land acquisition. Because any funds the government aims to raise incur losses, weak property rights act as a subsidy for the government, enabling it to construct more roads that generate large positive externalities on city-level welfare. Moreover, in Kampala, the areas hosting road improvements with the largest benefits tend to correspond with those that have the strongest property rights regimes. Consequently, the local structure of property rights impacts both the level and the distribution of land costs within the city, changing not only the level but also the location of optimal investments.

To show how the net returns to road improvements are larger in the presence of weak property rights, I start by collecting novel datasets. I first conduct a survey with 377 real estate brokers, which includes a retrospective panel of transactions. I show that if the government was to pay the market value of residential properties under an eminent domain regime, these costs would constitute 80% of overall project costs. Next, I interview 548 landowners whose properties border the upgraded roads in Kampala. I find that under the existing *voluntary land take approach*, 80% of landowners consented to forfeit their land without compensation, averaging 73 square meters of land per property. These interviews also show that the decision of landowners to forfeit their land is partly motivated by the high cost of negotiating with the government and by the difficulties of obtaining official copies of ownership documents. These challenges are heterogeneous along the lines of Kampala's three main property right regimes: leasehold, freehold and mailo. *Leasehold* land features limited term ownership with a strong record of property titles. *Freehold* land features perpetual ownership but land titles are poorly tracked. *Mailo* land is characterized by double legal rights - landowner and legal occupant - over a single plot of land (Bird and Venables 2020). I find that leasehold landowners, who report the lowest cost of getting a copy of their ownership documents, are 55% more likely to negotiate than mailo owners, and 75% more likely to negotiate than freehold owners. Consequently, unlike under eminent domain, this approach ties land costs, and therefore the likelihood of road improvements, to the different property right regimes in the city.

I then estimate the local benefits of the road improvement projects in Kampala. I start by calculating the impact of road improvements on traffic speed by leveraging information from the Google Maps queries. I use variation in the timing of the policy and compare traffic speeds on the roads upgraded at the start and end of the policy to isolate the effects of the improvements. I estimate that the intervention increased uncongested local traffic speeds by 4.1 km/h (16% faster than baseline speed). Consistent with this change in local speed, I also show that trips across pairs of neighborhoods became more likely to take upgraded roads over time.

Road improvements also increased local property values. Using the appraisal of a standardized hypothetical property from the broker survey, I find that the assessed sales price of a residential property increases by 25% if the road in its immediate vicinity is improved. Extrapolating this estimated increase in sale prices to all properties that border upgraded roads implies a total increase in local property values of 76 million USD, less than the 80 million USD of road construction costs, implying that the project would not have net positive returns if benefits were only local. I also estimate this effect using a retrospective panel of transactions from the same survey and I find that improving roads in a parish (neighborhood) leads to a 19% increase in the sale price of local properties.

While informative, these reduced-form estimates do not capture the full net returns of the implemented policy. Indeed, road improvements also have benefits in distant locations through the rerouting of traffic patterns, and the location choices of residents and firms in equilibrium. In addition, using residential land for roads has an opportunity cost, as it reduces the land available for housing, which strains the stock of valuable residential land and increases prices. To capture these effects, I build a static quantitative spatial model of a closed city. This model includes standard elements as workers freely choose residential and workplace locations trading off between high commuting costs and high rents (Redding and Rossi-Hansberg 2017, Allen and Arkolakis 2022). In addition, I explicitly model the competition for land between residential and road uses and assume that residential land is owned by immobile private landowners.

The model described above has a number of parameters that need to be estimated in order to compute the benefits and land costs of road improvements. To estimate the elasticity of commuting flows on commuting times, I partnered with a local prominent ride hailing company with more than 160,000 monthly users. Using the universe of rides (flows) by users of the app for a random sample of weeks from 2019 to 2024, I estimate that the number of commuters between two locations decreases by 0.33 percent for every one percent increase in commuting times. This elasticity is in the bottom half of existing estimates in middle and high income country settings and consistent with the few existing estimates in Low Income Country (LIC) cities (Balboni et al. 2020, Kreindler and Miyauchi 2023). I use my reduced-form estimates to recover the elasticity of road speed on road infrastructure and owners' negotiation costs.<sup>1</sup> I calibrate the rest of the parameters using public data for Kampala or from the literature.

Equipped with this model and the parameters described above, I compute the benefits of road improvements implemented in Kampala since 2017. I solve for workers' new equilibrium residential and workplace locations to estimate that workers' average commuting time decreased by 7%, and that total property values in the city increased by 1.48% as compared to the period before road improvements. Using the model, I solve for the compensating differential transfer that would need to be made to each resident to reach the same level of welfare gains realized by the road upgrades. Abstracting from land payments, I estimate that this one time transfer is equal to 120 USD, or 164% of the median monthly wage of a worker in Kampala. Summing over all residents, this implies 108 million USD net welfare gains in the absence of land payments (but accounting for construction costs). Comparing this number to the 66 million USD increase in local property values from the brokers' appraisal exercise highlights the importance of account-

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<sup>1</sup>I estimate that increasing road width by 10% decreases average trip time by 0.39%. This estimated elasticity of travel time on road infrastructure is on the upper end of existing estimates, all of them in high or middle income countries (Couture et al. 2018, Fajgelbaum and Schaal 2020, Bordeu 2024).

ing for city-level benefits and general equilibrium effects when assessing the impacts of road improvements on property values.

Payments for land may decrease the net returns to road improvements in the presence of large costs of levying taxes to raise domestic funds. I solve for the wedge on tax revenues rationalizing the government’s claim that acquiring land at market value would threaten the viability of the project (World-Bank 2023). I find that acquiring all land at market value would lead to negative net welfare gains of the realized road upgrades if the tax revenue wedge was above 0.7. This number is consistent with Regan and Manwaring (2024), who find that for every 1 dollar due in property taxes, the Kampala capital city government recovers only 39 cents, or a wedge of 0.61. Under this high tax revenue wedge, the net welfare gains of the policy under the existing *voluntary land take* approach are equivalent to a one time 99 USD lump-sum transfer to all Kampala residents. Importantly, I find a positive correlation between the realized upgrades and the corresponding net welfare gains from specific link-level improvements predicted by my model, consistent with the realized improvements being relatively well allocated given the existing costs and benefits. I also find a negative correlation between local land costs and road upgrades, consistent with the model capturing property right regimes as a relevant feature of the city government’s decision.

I then use the model to study how land institutions impact the allocation and welfare gains of road improvements by conducting counterfactual exercises with alternative land cost structures. I fix the maximum total kilometers of roads improved to match the policy and I solve for the welfare-maximizing (optimal) road improvements. Under compensation at market value, I estimate net welfare gains equivalent to a one-time transfer to each resident of 114 USD, or 156% of the median monthly wage. Under the voluntary land take approach, net welfare gains of the optimal policy would instead be equivalent to a 397 USD transfer by resident. Because of the large tax revenue wedge, weak property rights act as a subsidy, enabling the government to improve more high benefit roads and reach higher welfare gains than under market value compensation. Yet, this voluntary land take approach also ties land payments to spatially heterogeneous property rights, affecting the spatial distribution of optimal investments. If these property rights were homogeneous (fixing the fiscal cost as under the voluntary land take approach), net welfare gains would be 419 USD per resident as improvements would be allocated towards the highest benefit locations, leading to a larger decrease in commuting times and larger increase in property values.

Finally, given the key role played by the tax revenue wedge in driving these results, I investigate the impact of a policy removing existing restrictions on the use of external funds. Currently, the city government can only use funds from the World Bank and the African Development Bank for road construction, not for land acquisition. I solve for the optimal improvements removing this restriction. I find that under the voluntary land take approach in place, the net welfare gains would be 6% larger in the absence of fund use restrictions than under the existing restrictions. Under compensation at market value, the net welfare gains would be 125% larger in the absence of fund use restrictions than under the existing ones. Both results hold despite fewer roads being improved, as some external funds are instead being used for land acquisition. The existing restrictions are likely driven by corruption concerns, which are outside the scope of this paper. Yet, given the World Bank’s stated objective to ensure that owners are “*provided prompt and effective compensation [...] for losses of assets attributable directly to the project.*” (World Bank OP 4.12), these numbers should be used to benchmark whether the anticorruption benefits outweigh the potential

benefits in terms of better compensating affected owners, and allocating improvements towards higher benefit locations.

## Related Literature

My work contributes to a growing literature at the intersection of development and urban economics.<sup>2</sup>

First, my paper is the first measure the net returns to road improvements in a LIC city, accounting for both benefits and land costs. I find that while road improvements had large impacts on local speed and local property values, as well as on city-level average commuting time and property values through general equilibrium effects, they also had high land costs. In turn, the structure of these costs affect both the net returns of existing improvements, and the optimal location of road improvements. On the benefits side, existing evidence on the impact of road quality on speed is either in the United-States (US) (Duranton and Turner 2011, Currier et al. 2023) or across cities worldwide (Akbar et al. 2023). One of the few within-city studies on the benefits of road paving in low or middle income country cities comes from Gonzalez-Navarro and Quintana-Domeque (2016), who randomize the rollout of road paving in Acayucan, Mexico and study the impact on local property values, but do not account for the city-level impacts. Instead, I build on a literature that has developed quantitative spatial models (Ahlfeldt et al. 2015, Allen and Li 2015, Redding and Rossi-Hansberg 2017, Monte et al. 2018, Heblich et al. 2020, Severen 2023, Almagro et al. 2024) to study the benefits of improved transportation infrastructure, and add competition between private and public land use while papers usually model competition between residential and business land uses (Lucas and Rossi-Hansberg 2002). Within-city studies using quantitative spatial models in low and middle income countries have focused on the impact of Bus Rapid Transit (BRTs) (Majid et al. 2018, Balboni et al. 2020, Tsivanidis 2023, Zarate 2024, Kreindler et al. 2023), and very few estimates of the models' key elasticities exist in LIC cities (Kreindler and Miyauchi 2023).<sup>3</sup> In Kampala, as in the majority of LIC cities, road is the only transportation mode, so that the improvements will be experienced by all commuters.<sup>4</sup> On the costs, high land costs have been shown to have increased the costs of transportation infrastructure since 1960s in the US (Brooks and Liscow 2023), and a few policy reports have discussed their role in LICs (World-Bank (1996)). Collier and Venables (2016) discuss the cost of land use for LIC infrastructures but I am the first to formalize and quantify this mechanism through my quantitative spatial model of optimal road improvements, building on the recent methodologies developed by Allen and Arkolakis (2022), Fajgelbaum and Schaal (2020) and Bordeu (2024). I micro-found land cost heterogeneity in Kampala's historical property right regimes under the voluntary land take approach, and show how they affect the optimal location of road investments by shifting the distribution of relative costs.

Second, I show how, in the presence of a high tax revenue wedge, weak property rights can act as a

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<sup>2</sup>Collier and Venables (2016) and Glaeser and Henderson (2017) highlight that this intersection has historically been overlooked by both development and urban economists, partly to the absence of detailed data needed to estimate quantitative spatial models.

<sup>3</sup>Works on between-city and rural impact of road paving in LIC countries include Storeygard (2016), Aggarwal (2018) Gertler et al. (2024a), Graff (2024).

<sup>4</sup>Most existing studies on road improvements in LIC countries focus on rural or cross-city infrastructure (Baum-Snow et al. 2017, Asher and Novosad 2020, Alder et al. 2022, Balboni 2023, Gertler et al. 2024b, Morten and Oliveira 2024, Herzog et al. 2024). There are also several studies on the impact of railroad development on across-city transportation costs and migration, including Gollin and Rogerson (2010), Faber (2014), Ghani et al. (2016), Jedwab and Moradi (2016), Donaldson (2018) and Bryan and Morten (2019).

subsidy for investments in public infrastructure, increasing city-level welfare and the net returns from road improvements. On the one hand, I estimate that any tax revenue wedge above 0.7 would lead to negative net welfare gains from road improvements if land was acquired at market value, in line with estimates for Kampala by [Regan and Manwaring \(2024\)](#) and consistent with a dense literature on the widespread challenges faced by LICs to levy domestic and property taxes ([Traxler 2010](#), [Besley and Persson 2014](#), [Knebelmann 2019](#), [Bergeron et al. 2023](#), [Brockmeyer et al. 2023](#)). On the other hand, I provide new evidence on the relationship between property rights structure and efficiency of investments, through the channel of public good provision. Unclear property rights yield higher risks of expropriation, associated with lower economic development ([North 1990](#), [Acemoglu et al. 2001](#), [Besley and Ghatak 2010](#)), including in Kampala ([Bird and Venables 2020](#)). However, in the context of public goods provision, weak property rights can act as a subsidy, as strong private property rights conflict with public interests ([Acemoglu and Robinson 2012](#), [Posner and Weyl 2017](#)), which in turn, justifies eminent domain ([Munch 1976](#), [Shavell 2010](#), [Jeong et al. 2016](#)) for public purposes. I further show how, in LICs, eminent domain may lead to suboptimal investments, given the high costs of raising public funds. As in [Holland \(2023\)](#), who studies the role of strong property rights in shaping opportunistic behavior by private owners in Colombia, I find that payments are increasing with the strength of owners' property rights. My contribution is to show how this relationship can affect the optimal amount of high benefit infrastructure improvements and, in the presence of spatially heterogeneous property rights, shift their location away from the highest benefit areas.

The rest of the paper is organized as follows. Section 2 describes the context and the data I collected. Section 3 details how I estimate the reduced-form benefits and land costs of road improvements. In Section 4, I build a quantitative spatial model to study the city-level impacts of these improvements and study the welfare effects of spatially heterogeneous land costs. In Section 5, I estimate the model on Kampala, characterize the net returns of the road improvements. In Section 6, I solve for the optimal road improvements to quantify the welfare consequences of the existing property right regime, land acquisition rule and fund use restrictions. Section 7 concludes.

## 2 Context and Data

### 2.1 Context

#### 2.1.1 Road Improvements in Kampala

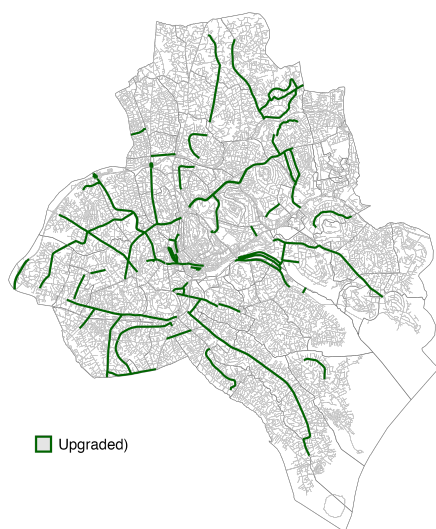
Kampala, Uganda's capital, is a fast growing city, with approximately 1.9 million inhabitants in 2024. While Kampala hosts an increasing share of Uganda's population and GDP, its road infrastructure is of low quality: out of 4,622 km of roads and paths recorded by Open Street Map (OSM), only 742 km (16%) are major roads, and less than 43% of these major roads are paved (7% of all roads).<sup>5</sup>

In the past decade, relatively large amounts have been invested to improve the quality of the existing road network. In addition to some domestic investments, most funds come from the World Bank (WB) and the African Development Bank (AfDB), under the umbrella of two projects, the Second Kampala Institutional

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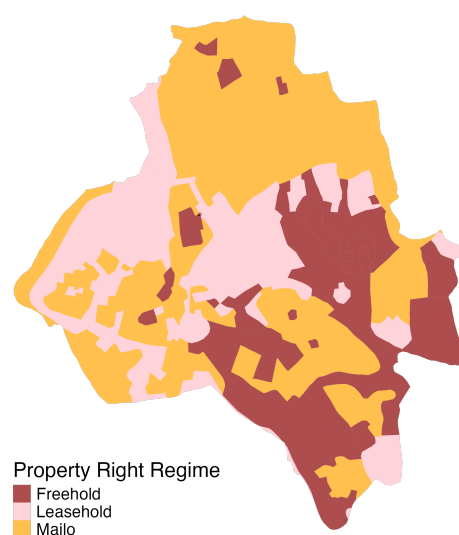
<sup>5</sup>[Akbar et al. \(2023\)](#) define major roads as roads falling under the motorway, primary, secondary or tertiary road classes on OSM. For reference the mean lane km of major urban road in US MSAs in 2008 was 14,000 km ([Couture et al. 2018](#)).

Figure 1: Road Improvements in Kampala



Notes: Road improvements in Kampala since 2017, including domestically-funded improvements, as well as improvements parts of the projects funded by the WB and the AfDB. Sources: KCCA, WB, AfDB.

Figure 2: Property Right Regimes in Kampala



Notes: This map displays the three dominant land tenure systems in Kampala, together with the parish administrative boundaries. Source: GIS Unit at KCCA.

and Infrastructure Development Project and the Kampala City Road Rehabilitation Project. Estimated construction costs, excluding land costs, sum to 80 million USD (572 thousand USD per kilometer on average as per the AfDB) and are part of larger investments worth almost 500 million USD. First planned in 2013, road improvements under these two projects started in 2017 and the last roads are expected to be completed by the end of 2026. The few upgrades funded by the Government of Uganda during that period are also included in the analysis. Not all roads mentioned in the initial plan got upgraded. In total, major road improvements targeted about 140 km of roads since 2017, including the upgrades to be completed by the end of 2026, equivalent to almost 20% of the major roads in the city.<sup>6</sup> Figure 1 displays a map of these upgrades and a timeline of the upgrades can be found on Appendix Figure A3.

## 2.1.2 Land Acquisition for Infrastructure Projects in Kampala

### Widespread Issues to Acquire the Underlying Land

Upgrading road networks commonly requires additional land. Widespread challenges to acquire that land have been documented. In Kampala, while the studied road network upgrades have mostly involved improving the existing roads, additional land is still needed as these roads must be widened to either add traffic lanes or build sidewalks and drainage channels. The latter are necessary to absorb the frequent and important volumes of water during the rainy seasons, so as not to damage road pavement. Analogous land acquisition has been challenging worldwide, including for the U.S. Interstate highway system (Brooks and Liscow 2023). In WB-funded projects, these challenges are further fueled by domestic governments' inability to use these international funds for land acquisition. A 1996 report estimates that while the WB's legal framework requires compensation, less than 15% of the projects reviewed included funds for

<sup>6</sup>This number is in addition to regular maintenance, which is out of the scope of this paper.



resettlement activities, and more than 30% of the projects had been delayed by an average of two years as a result. In Kampala, the completion of the WB project was delayed by 3.5 years due in part “to delays in acquisition of rights of way” (World-Bank (2023))

## Legal and Practical Land Acquisition Frameworks

The legal framework for land acquisition in Uganda - governed by article 26 of the Constitution of the Republic of Uganda (1995), section 42 of the Land Act (1998), as well as the WB’s OP 4.12 - is eminent domain with compensation at market value. Yet, in practice, shortly after the start of the WB-funded project, the city government argued that the project’s land costs were too high under eminent domain and threatened the project, since most owners were only minimally affected as only one or two meters of land were taken. Thus, it adopted a “*voluntary land take approach*”, where owners were asked to forfeit a small piece of land for the road to be upgraded, without receiving compensation.

A typical land acquisition timeline goes as follows: First, the government decides which roads will be upgraded based on a collection of factors, including potentially overall land availability. Once the road has been chosen, an engineering study is conducted to assess the optimal road design and affected land, under the supervision of the chief government valuer. The government then reaches out to local leaders and affected owners to inform them about the road project, and to obtain their consent. These owners decide whether to accept to give the required amount of land, or to negotiate over the compensation. In this context, negotiations can take the form of discussions with city council members and local leaders, involvement of lawyers, grievances, and court proceedings. The eminent domain legal framework de facto imposes an upper bound on the compensation that can be claimed by affected owners.<sup>7</sup> Finally, once all land parcels have been secured, the road construction can start.

Thus, the final price of land depends on both the market value of this land, to be paid under eminent domain, and the share of owners who get compensated. While the voluntary land take approach is a more cost effective strategy from the perspective of the government, concerns have been raised that owners may not be informed and able to negotiate effectively (World-Bank 2023), potentially because of unclear property rights as land disputes and issues of multiple land claimants per plot would have to be sorted out prior to compensation.

### 2.1.3 Coexistence of Multiple Property Right Regimes in Kampala

The strength of property rights is heterogeneous within Kampala because land falls under three main types of property right regimes, inherited from the 1900 Buganda Agreement between the British Protectorate and the Buganda Kingdom. Bird and Venables (2020) showed that this distribution of land ownership systems in the city, depicted in Figure 2 and which has not changed much in the past 50 years, has influenced patterns of land use.

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<sup>7</sup>Eminent domain is a well studied solution to holdout problems, where private sellers may strategically wait and exert their monopoly power (increasing as they become the last holdout) to seek higher compensations, leading to project delays and inflated costs (Posner and Weyl 2017). In this context, owners may still holdout until they get the eminent domain payment, but they would not be able to extract more than the ex-ante market value of their property. I abstract from the time dimension in this analysis.

Leasehold is dominant at the center and west part of the city and it is the clearest property right regime, as the registry of active leases is maintained by local land boards. Property right clarity is weaker under freehold and mailo regimes. Freehold land is mostly concentrated on the South-East part of the city, less central and closer to Lake Victoria. Freehold land titles are the least restrictive (single ownership and unlimited term), but the accountability of land titles is poor.<sup>8</sup> On mailo land, predominant on the North and South-West parts of the city, issues of multiple land claimants per plots are common, as both landowners and occupants have historically received rights and protections. The history of these property right regimes is described in more details in Appendix A.

## 2.2 Data Sources

To study the benefits and costs of road improvements in Kampala, and later estimate the main model parameters, I gather detailed information on traffic speed and flows, property values, and the land acquisition process via google maps queries, a partnership with a local ride-hailing company, a survey with real estate brokers, and a survey with landowners.

### 2.2.1 Google Maps API Data

To study changes in traffic speed following road improvements in the absence of good traffic-related administrative data, I query Google Maps Direction API almost every month between March 2023 and September 2024 to constitute a panel of about 180,000 Google Maps trips. More details on these queries can be found in Appendix E. Each query includes information on the trip’s time, distance, straight line distance, traffic at the time of query, and average duration. In addition, Google Maps Direction API includes a list of coordinates approximately mapping the path between origin and destination. I sample trips with three distinct goals

1. Impact of road upgrades on local speed (road level): 248 roads ( $248 \times 2 = 596$  unique trips, both ways) were selected to compare car speed on upgraded and to be upgraded roads. These roads include 46 roads targeted by the policy and used in the main analysis, as well as as a subset of roads mentioned in earlier road improvement plans but not ultimately upgraded.<sup>9</sup>
2. Average speed by road category across the city: trips between neighboring  $1000 \times 1000$  meter grid cells in Kampala were mapped to the underlying OSM road network to recover the average speed by road category (results presented in Table A2). The short distance of these trips between neighboring locations limits optimization through route choice and ensures that my trip sample covers all road types in Kampala.
3. Change in shortest path in response to road upgrades: I mapped trips between all pairs of Kampala parish centroids (9,216 unique trips) to the underlying OSM road network to provide evidence that a road is more likely to be used for a given trip after being upgraded. In Appendix D.2, I show how the share of a trip’s length happening on roads in the later wave of the policy, which start to get

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<sup>8</sup>The Uganda government started to digitize the national land registry to decrease the number of land fraud cases in 2013, but, as of 2023, this process had not been finalized yet.

<sup>9</sup>These roads were selected based on information available as of March 2023, and thus may not include roads whose upgrade was decided later.

upgraded over the course of the Google Maps sample, increases over time, while no such relationship exists for roads upgraded before the start of the sample.

I recover that the average speed in Kampala across all pairs of parishes (unweighted) is 24.4 km/h (26.8 km/h at non-rush hour and 22.0 km/h at rush hour). This corresponds to cities at the 25<sup>th</sup> percentile of the world speed city distribution (Akbar et al. 2023). Improving road speed is thus a first order concern in Kampala.

### 2.2.2 Ride-Sharing Data

To study how workers will adjust their location choice in response to a change in commuting time, I partner with a local ride hailing company running a popular app with more than 30,000 rides per day.<sup>10</sup> I have access to the universe of motorcycle trips for a random sample of weeks between 2019 and 2024. As described in Table 1. My sample covers 95 days (5% of the period) and includes 2.3 million trips and more than 330 thousand unique users, or about 38% of the adult population in Kampala’s 2014 Population Census. These users take an average of 6.9 trips on the app, but the standard deviation is very large (104), indicative of a subset of riders using the app daily. 39% of users take at least 5 trips on the app. The average trip costs 1.47 USD, lasts 21 minutes for a straight line distance of 5 kilometers. In addition, as displayed on Appendix Figure A6, most trips at morning commuting hours (6 am - 10 am) originate from the outskirts of the city to end at the city center. This commuting pattern towards the city center is consistent with the location of jobs from the Uganda Census of Business of Establishments displayed on Figure 6 Panel (b), but residents originating the northern part of the city are overrepresented among morning commuters, compared to the universe of Kampala residents (Figure 6, Panel a).

### 2.2.3 New Broker Survey

Data on property values is not usually readily available in developing country cities like Kampala, so I collect my own data by interviewing 377 real estate brokers in 89 of Kampala’s 96 parishes in March-April 2024, as described in Appendix Figure A9. My respondents are brokers with an average of 14 years of experience, and 76% of them operate in multiple parishes (neighborhoods). The average broker rented 11.3 properties in the 3 months preceding the interview, and sold 1.9 properties over the same timeframe. Descriptive statistics are displayed in Table 2. The survey is designed with two goals in mind.

First, to characterize property values over space and track their changes over time, I ask brokers about retrospective transactions (sales and rentals). The final database includes about 3,250 properties, transacted between 2018 and 2024 all over Kampala. For each property, available information includes the transaction date (month, year), location (parish, village, closest road, distance to closest road) and a comprehensive set of property characteristics. The average bare land property in my sample is sold for 59,262 USD, which corresponds to an average price/m<sup>2</sup> of 142 USD, or about 1/4<sup>th</sup> of the rate per m<sup>2</sup> for land in Delhi (Statista). There is a lot of variation in the sales prices and monthly rents of properties across parishes. To build confidence in my data, I also compare these past transactions to scraped online data of high-end

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<sup>10</sup>Safeboda has been operating in Kampala since 2019, and dominates the app-based ride hailing market for motorcycle rides.

Table 1: Ride-Hailing Descriptive Statistics

	Total		Per Month	
	Count / Mean	sd	Count / Mean	sd
<i>Sample</i>				
Number of Days Covered	95			
Share of Days Covered	0.05			
<i>Trips</i>				
Number of Trips	2,284,986		163,213	57,747
Length of Trip (min)	20.66	12.6	20.79	0.72
Straight Line distance (km)	5.01	2.89	5.15	0.47
Price (USD)	1.47	1.63	1.59	0.37
<i>Users</i>				
Number of Users	330,426		163,213	57,747
Avg Nb Trips per User	6.9	104.2	2.7	31.3
Share of users w/ more than 5 trips	0.39		0.15	

Notes: Descriptive statistics from the ride hailing data. The sample covers a random sample of weeks between 2019 and 2024. Motorcycle rides only are included in the sample.

properties transacted online (Appendix E). I show that my novel survey data is positively correlated with data from the online platform, but has a better spatial coverage and is more representative.

Second, I obtain additional reduced-form and descriptive evidence of the local impacts of road improvement and distance from roads on property values from an appraisal exercise. Brokers were presented with a standardized property (business or residential) and were asked to estimate its sales price varying one characteristic at the time, including whether the road was upgraded or not. In Appendix E, I further describe the sampling strategy and sample characteristics.

#### 2.2.4 New Owner Survey

The second component of land costs is the share of owners getting compensated. As I do not have access to the comprehensive land acquisition data from the Kampala Capital City Authority, I collected data on the land acquisition process by surveying land and property owners (hereafter owners) along segments of 87 road in Kampala. I interviewed a random sample of owners on upgraded roads and roads in the process of being upgraded.<sup>11</sup> More details on the selection of road segments and the sampling strategy are available in Appendix C. Descriptive statistics about the data are displayed in Table 3.

A total of 772 respondents were interviewed on roads targeted by upgrades since 2017.. Among them, 612 were on the sides of roads upgraded in the past and 160 were on the sides of roads to be upgraded in the near future, and owners with plots under each of the three property right regimes (leasehold, freehold, mailo) were interviewed. The average respondent had owned that property for 36 years, and 41% of them

<sup>11</sup>Because of the different land tenure systems in Kampala, not all house owners own the underlying land. Yet, under the legal regime, these owners must be compensated (or consent) if their property is affected by public work, and are thus valid respondents for our study.

Table 2: Broker Survey's Descriptive Statistics

	Mean	Sd	Median
Number of brokers	377		
Number of brokers per parish	3.92	1.79	4
<i>Panel A: Broker characteristics</i>			
Age	42	8.9	
Years operating as brokers	14	10.1	
Share of brokers operating in multiple parishes	0.76		
<i>Panel B: Activity</i>			
Number of properties rented in the past 3 months in Kampala	11.32	18.09	
Number of properties sold in the past 3 months in Kampala	1.89	2.68	
<i>Past Transactions (USD)</i>			
Price - rented residential	206	335	108
Price - rented business	218	337	
Price - sold residential	70472	60550	54054
Price - sold business	78019	72479	
Price - sold land	59262	55035	40541
Price / m2 - sold residential	203	384	131
Price / m2 - sold business	330	554	184
Price / m2 - sold land	142	206	97
<i>Appraisal Exercise - Standardized res property near road</i>			
Relative sales price post upgrade	1.318	0.58	1.25

Notes: The descriptive statistics in this table come from the real estate broker survey. Panels A and B describe the characteristics of the respondents. Panel C describes the properties transacted by the respondents and panel D includes information about the impact of road upgrade on the hypothetical price of the standardized property.

had purchased the property, as opposed to inherited it. The large share of inherited property, higher for freehold (0.64) and mailo (0.6) than for leasehold (0.5) is consistent with the very high reported cost of obtaining a copy of ownership documents, which is also higher for freehold and mailo than for leaseholds. This cost indeed often entails hiring a lawyer, going to court, obtaining certified copies, recovering old land titles that had never been accounted for etc.

The property of 548 of these owners had been or was being affected by the road upgrade, on average by 73  $m^2$ . 80% of respondents consented to give this land without getting compensated, and this probability is higher for freehold and mailo owners than for leasehold, who were more likely to negotiate to get compensated. The average owner (including both owners who consented, and those who did not) believed that they could get a higher compensation by not consenting for at least six months with probability 0.31 and this probability is the highest for leasehold owners.

In the survey, respondents were also asked about current and historical road quality, past and ongoing road updates, land acquisition, as well as property and neighborhood characteristics.

Table 3: Owner Survey’s Descriptive Statistics

	Overall	Leasehold	Freehold	Mailo
<i>Panel A: Sample Characteristics</i>				
Number of owners	548	127	112	293
Share on already upgraded roads	0.78	0.68	0.63	0.87
Car ownership (share)	0.4	0.25	0.48	0.42
<i>Panel B: Property Characteristics</i>				
Purchased property (share)	0.41	0.5	0.36	0.4
Tenure (years)	35.5	35.2	38.9	34
Cost Copy of Ownership Documents (USD)	1750	1111	2651	1653
<i>Panel C: Land Acquisition</i>				
Land Affected (m2)	72.7	50.33	79.68	80.62
Negotiated Over Initial Compensation Offer (share)	0.2	0.28	0.16	0.18
Perceived Probability to Increase Compensation if Negotiation	0.31	0.38	0.24	0.3

Notes: Descriptive statistics are from the owner survey, restricted to the 548 respondents (out of 772) whose property or land was affected by the road upgrade.

### 3 Empirical Evidence on the Benefits and Costs of Road Improvements

Equipped with this new data and using variation in the timing of road upgrades and heterogeneity in property right regimes in the city, I estimate the local benefits and land costs of road improvements in Kampala. I show that road upgrades increased local speed and local property values. The associated land costs were high, but heterogeneous across locations.

#### 3.1 Local Benefits of Road Improvements

##### 3.1.1 Impact of Road Improvements on Local Road Speed

The impact of road improvement on local speed is an important outcome of interest for two reasons. First, it is a reduced-form measure of the most direct impact of the intervention, which may be very different than in the U.S. given the prevalence of unpaved roads at baseline in African cities. Second, it is a key input into the model that I estimate in Section 5 to recover the full returns of the policy.

To recover the impact of road improvements on local speed, I compare speed on upgraded and non-upgraded roads. I restrict my sample to roads being upgraded as part of the policy, which includes roads upgraded before the start of my data in March 2023, roads upgraded between the start and end of my data in August 2024, and roads whose upgrade hasn’t started yet.<sup>12</sup> I run the following regression:

$$s_{h,d,r} = \alpha + \beta \times D_{d,r} + \beta \times 1_{h,d,r} + \mathbf{X}'_r \psi^x + \gamma_h + \gamma_d + \epsilon_{h,d,r} \quad (1)$$

<sup>12</sup>I do not have road-level trips for a representative sample of all roads in Kampala, and I cannot control well for road quality beyond its OSM category and a dummy for whether that road is paved. Thus, to isolate the impact of the policy on speed, rather than confounding the effect of selection into the policy, I only include roads part of the policy.

where the dependent variable  $s_{h,d,r}$  is the GoogleMap estimated traffic speed on road  $r$ , day  $d$  at hour  $h$  in km/h,  $D_{d,r}$  is a dummy variable equal to 1 if road  $r$  was upgraded by day  $d$ ,  $l_{h,d,r}$  is the length of the trip (in km),  $\mathbf{X}_r$  is a vector of road-level characteristics, including road origin and destination coordinates or road fixed effect. I also control for whether the road was under construction on day  $d$ . I include hour and day fixed effects to control for traffic patterns being highly variable over time.<sup>13</sup> Standard errors are clustered at the road and day of query levels to reflect sampling. To isolate the direct impact of the road improvement, I focus on trips queried outside of traffic rush hours, and I discuss the impact on speed for rush-hour trips in Appendix A5.

Because my Google Maps data covers the period between March 2023 and August 2024 and many roads were already upgraded by March 2023, most of the variation used to identify  $\beta$  is cross-sectional and comes from the comparison of roads that were upgraded early (before the start of the Google Maps sample), and roads upgraded late (during or after the Google Maps sample). Thus, in the main specification, to interpret  $\beta$  as the average treatment effect on the treated (ATT), I need to assume that roads were not selected into early or late upgrade based on the predicted increase in free flow speed, conditional on observable characteristics. Note that this assumption does not require that the selection of roads into early vs late wave is orthogonal to the predicted impact of the upgrade on economic activity or overall traffic flows, but instead that the timing was not based on the predicted impact on local speed in non-rush hour.<sup>14</sup> A subset of roads got upgraded between the start and end of my panel, allowing me to run a specification with road fixed effects. Standard errors increase because the identifying variation comes only from a subset of roads, but the ATT identification assumption is weaker than in the absence of road fixed effects and only relies on the exact timing of the specific road upgrade (e.g. segment) being conditionally random. To further interpret  $\beta$  as the Average Treatment Effect (ATE), I further need to assume that the above assumptions would hold for the average road in Kampala.

I present regression results in Table 4. In columns 1, 2 and 3, I estimate the reduced-form impact of any road improvement (dummy variable) on traffic speed. In column 1, I do not include any road-level control. In column 2, I add road-level controls and in column 3, I replace these controls by road-fixed effects. I estimate a robust impact of road upgrade on free-flow speed of 4.96\*\*\*, 4.45\*\*\* and 4.35\*\* km/h, respectively, or 18% of the average speed in Kampala. This effect is large: if average speed in Kampala increased by 4.4 km/h, this would take Kampala from the 25<sup>th</sup> percentile of the speed distribution among cities worldwide, to the 50<sup>th</sup> percentile (Akbar et al. 2023). In columns 4 and 5, I estimate the elasticity of speed on road width using analogous specifications as in columns 1 to 3. This elasticity is a key input in my structural model and I discuss it in more details in Section 5.1.1.

### 3.1.2 Impact of Road Improvements on Local Property Values

**Property Appraisal:** Real estate broker survey respondents were asked to assess the contemporary market value of a standardized property (residential or business) where the property’s characteristics were

<sup>13</sup>In addition, my panel is unbalanced at the hour-day level, as I could not query trips for all roads for each hour-day due to restrictions on the frequency of Google Maps API queries. Thus, including hour and day fixed effects controls for potential sample bias across hours and days.

<sup>14</sup>In Appendix Table A5, where I also discuss the impact on speed for rush-hour trips, the assumption is stronger, as it now accounts for existing and predicted traffic flows (which may affect rush hour speed). In that case, if timing was based on existing flows - with the busiest roads upgraded first - the bias may go both ways, depending on the congestion technology.

Table 4: Impact of Road Improvement on Traffic Speed

	<i>Dependent variable:</i>				
	Traffic Speed (km/h)			Log Traffic Speed (km/h)	
	(1)	(2)	(3)	(4)	(5)
Upgraded (d)	4.964*** (1.156)	4.449*** (0.859)	4.351** (1.598)		
In Progress (d)	1.290 (1.479)	-1.319 (1.264)	-1.112 (1.267)		
Log Road Infrastructure (width m)				0.390*** (0.101)	0.313*** (0.108)
Sample	Non-Rush	Non-Rush	Non-Rush	Non-Rush	Non-Rush
Road Controls		Y		Y	
Road FE			Y		Y
Day + Hour FE		Y	Y	Y	Y
SE Clustered	road+day	road+day	road+day	road+day	road+day
Observations	682	682	682	632	632
R <sup>2</sup>	0.165	0.473	0.641	0.280	0.662

Notes: Standard errors are displayed in parentheses and clustered at the road and day levels, with significance levels  $*p < 0.1$ ;  $**p < 0.05$ ;  $***p < 0.01$ . Observations are at the road level (trips covering a single road) from Google Maps API queried between 2023/03/17 and 2024/06/30. Columns 2 and 4 include flexible geographic controls: longitude, longitude<sup>2</sup>, latitude, latitude<sup>2</sup> of the trip's origin and trip length. Rush hour is defined as 6 to 9am and 4 to 7pm. Columns 2 to 5 include hour and day fixed effects.

varied one at the time (full description in Appendix E). I compare the assessed sales price of that property in the immediate vicinity of a road if that road is not upgraded (e.g. unpaved), and if that road gets upgraded.<sup>15</sup> The regression results are displayed in columns 1 and 2 of Table 5, and the distribution of answers is relegated to Appendix Figure A7. Brokers estimate that a road upgrade increases the price of this standardized property by 25% on average for the standardized residential property on the side of the road (column 1) and 27% (column 2).

Focusing on residential properties, evaluated at the median property value in the past transaction database (54 thousand USD), and assuming that each property spans an average of 50 meters along these roads, 140 km of road upgrades would have increased the total value of properties by the side of the road by about 76 million (broker appraisal), which is slightly below the corresponding construction costs of 80 million USD.<sup>16</sup>

Importantly, this increase in local property values captures two potential impacts of road improvements: the change in the property's accessibility and the change in local amenities. Indeed, interviewed owners on the side of upgraded roads confirm that these improvements had positive impacts on road quality, but also decreases flooding and dust (Appendix Figure A10), all of these amenities being local. Therefore, while this is a helpful sufficient statistic to measure the local benefits of road upgrades, fixing property characteristics, it does not capture the impact of the road upgrade on farther away properties through spatial spillovers and general equilibrium effects. I estimate this full effect in Section 5.2 leveraging a quantitative spatial model.

<sup>15</sup>Brokers are asked to think of that property in a typical location of their parish, so the road upgrade is hypothetical.

<sup>16</sup>The average width spanned by properties in my owner survey is 48m. 572 thousand USD per km  $\times$  140 km.



Table 5: Broker Appraisal: Impact of Road Improvement on Local Property Values

	<i>Dependent variable:</i>					
	Log Price					
	(1)	(2)	(3)	(4)	(5)	(6)
Post road upgrade	0.224*** (0.028)	0.224*** (0.032)	0.224*** (0.028)	0.237* (0.129)	0.237*** (0.042)	0.237*** (0.036)
Property Type	Res	Res	Res	Bus	Bus	Bus
FE		Parish	Respondent		Parish	Respondent
Observations	354	354	354	304	304	304
R <sup>2</sup>	0.008	0.693	0.978	0.011	0.770	0.962
Adjusted R <sup>2</sup>	0.005	0.606	0.955	0.008	0.693	0.923
Residual Std. Error	1.235 (df = 352)	0.778 (df = 275)	0.262 (df = 176)	1.125 (df = 302)	0.626 (df = 227)	0.313 (df = 151)

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Notes: Standard errors are displayed in parentheses, with significance levels \*p < 0.1; \*\*p < 0.05; \*\*\*p < 0.01. Columns 1 and 2 include broker fixed effects. Brokers were randomly asked about either the standard residential or the standard business property.

**Transacted Properties:** Next, I leverage the database of past transactions from the broker survey to shed additional light on the impact of road improvements on property values. More specifically, I run the following regression:

$$\log Q_{i,b,p,t} = \alpha + \beta \times D_{p,t} + \mathbf{X}_i' \cdot \eta^{\mathbf{X}} + \mathbf{Z}_p \cdot \eta^{\mathbf{Z}} + \gamma_t + e_{i,b,p,t} \quad (2)$$

where  $Q_{i,b,p,t}$  is the sales price of property  $i$ , transacted by broker  $b$  in parish  $p$  at time  $t$ ,  $X_i$  is a vector of property  $i$ 's characteristics and  $Z_p$  is a vector of location  $p$ 's characteristics, potentially including division fixed effects or parish-level controls. Parish-level controls include log 2010 population, share of residents with permanent flooring (2014 Census), and the share of land surface built (2015 Global Human Settlement Layer).  $\gamma_t$  is a vector of time (month and year) fixed effects. Property characteristics  $X_i$  include {floor type, number of bathrooms, distance to the closest main road, log(land dimensions), formal (dummy), building type, presence of buildings in addition to the main one (dummy)}, as well as missing dummies for these controls. Standard errors are clustered as in [Conley \(1999\)](#).<sup>17</sup>  $D_{p,t}$  is a dummy variable equal to 1 if parish  $p$  had received a road improvement by time  $t$ . As variation in  $D_{p,t}$  is both temporal and spatial,  $\beta$  captures both the effect of a road being upgraded in parish  $p$ , as well as some general equilibrium impact on control locations. These general equilibrium effects may bias estimates down if the neighboring locations benefit from the road upgrade, or bias estimates up if the neighboring locations become less attractive.

To interpret  $\beta$  as the causal effect of a road improvement in parish  $p$  on property values in that parish, I need to assume that road improvements did not target parishes with differential trends in property values (no selection) and did not come with other place-based policies. This condition would be violated if the specific road improvements targeted specific locations based on returns, but, reassuringly, policy-documents emphasize the goal of improving traffic flows and connectedness in the city, rather than road upgrades being used as place-based policy. Still, I include parish socio-economic characteristics from the 2014 Census to control for selection based on observable characteristics. Further, in column 4, I restrict my sample to parishes that received a road improvement at some point (ever-treated), so as to leverage the exact timing of these road improvements rather than their location and address concerns that parishes

<sup>17</sup>Each observation's location is defined at the centroid of its parish. I define a 3000 meter threshold, so that on average, one parish has 18% of the other parishes within that cutoff

receiving road upgrades would be different along unobservable characteristics. Ideally, I would want to run an event study at the parish level, including parish fixed effects. However, I am not powered to do so, as I have too few properties transacted by parish in my database.

I present the results in Table 6. In column 1, I do not include any parish or location control and I estimate that properties sold after a road upgrade in the parish (dummy) commanded a 17% premium (0.157\*\*\*). This estimate is relatively robust to adding socio-economic controls at the parish level (column 2) and division fixed effects (column 3). To address the concern that locations targeted by road upgrades were fundamentally different than locations that did not receive road upgrades, in column 4 I restrict my sample to parishes that received a road upgrade since 2017 (*ever-treated*), de facto comparing properties sold in parishes with early vs late upgrades. The identification assumption in column 4 is thus that there is no selection on the timing of road upgrades. This parish-level estimate is about 60% of the estimate from the broker appraisal exercise on properties on the side of upgraded, which is consistent with some of these local benefits being local and only affecting properties in the immediate vicinity of the road.

Table 6: Past Transactions: Impact of Road Improvements on Property Values

	<i>Dependent variable:</i>			
	Log Sales Price (USD)			
	(1)	(2)	(3)	(4)
Post Road Upgrade (dummy)	0.157*** (0.006)	0.142*** (0.007)	0.156*** (0.003)	0.188*** (0.019)
Division FE			Y	Y
Parish controls		Y	Y	Y
Sample	All	All	All	Ever-treated
Mean Sales Price (USD)	77,500	77,500	77,500	75,884
Observations	1184	988	952	234
R2	0.43	0.44	0.49	0.58

Notes: Conley SHAC Standard errors are displayed in parentheses, with significance levels  $*p < 0.1$ ;  $**p < 0.05$ ;  $***p < 0.01$ . I use 3000 meters as the cutoff distance for spatial correlation so that on average, one parish has 18% of the other parishes within that cutoff. All specifications include controls for property characteristics {floor type, number of bathrooms, distance to the closest main road, log(land dimensions), formal (dummy), building type, presence of buildings in addition to the main one (dummy)}, as well as missing dummies for these controls. Parish-level controls in columns 2 to 4 include flexible geographical controls {longitude, latitude, longitude<sup>2</sup>, latitude<sup>2</sup>} and pre-policy socio-economic controls {log 2010 population, share of residents with permanent flooring (2014 Census), share of land surface built (2015 Global Human Settlement Layer)}. There are 5 divisions in Kampala and 96 parishes. The sample of transacted properties span 86 parishes.

## 3.2 High and Heterogeneous Land Costs

While road improvements had benefits, they also had high and spatially heterogeneous land acquisition costs, driven by property values and by the probability that owners get paid. The latter depends on both the amount of land affected and owners' property right regimes. I characterize these two components using data from both the broker and owner surveys.

### 3.2.1 High and Heterogeneous Market Value of Affected Land

Under eminent domain, owners would be compensated at market value. In my survey, road improvements claimed a portion of the land or property of 548 owners. These owners report an average of 73m<sup>2</sup> of their

land being taken up by the road upgrade (Table 3). Given that the average property in the survey spans 48 meters along the road, this is equivalent to about 1.5 meter of land being taken on each side of the road, about one additional lane of average width 3.2 meters (KCCA), and a total of approximately 5,600 owners affected in total.

The market value of that land is substantial. From the broker survey, I estimate that the average sales price per m<sup>2</sup> for residential land properties is 171.7 USD (median 152.3 USD). To recover the total market value of affected land, I take the average land market rate for each parish and I assume that 3 meters of land were affected on average per owner. The corresponding parish-level predicted affected land market value per owner is displayed on Appendix Figure A14. Summing over all upgrades, this corresponds to 210 thousand m<sup>2</sup> of land, 65 million USD], or about 2/3 of the construction costs and 9,600 USD for the median owner.

### 3.2.2 Heterogeneous Share of Owners Giving Their Land Without Compensation

Under the *voluntary land take* approach, owners do not get paid if they consent to donate their piece of land without compensation. Among the 548 affected owners in my survey, 77% reported not being initially offered a financial compensation, and yet, 79% of them consented to forfeit their land within less than 6 months (hereafter “*accepted*” as opposed to “*negotiated*”).<sup>18</sup> There are two types of reasons given by respondents for accepting to forfeit a piece of their land (on average 73m<sup>2</sup>). First, are genuine benefits from the road improvements, either for the community (69%) or in terms of improved local amenities (44%). Second, however, 1/3<sup>rd</sup> of owners report not negotiating because they were powerless, which is consistent with only 31% of owners believing that they would be able to get a compensation by negotiating (Table 3). Respondents’ emphasis on their low ability to get compensated is consistent with property rights being weak on average, as highlighted in multiple policy reports. On the other hand, the top three reasons given by owners negotiating are: being entitled by law (55%), wanting to get a “piece of the pie” (34%), and being asked to give too much land without compensation (27%). The full distributions of reasons given by owners to negotiate or consent are displayed in Appendix Figures A11 and A12.

There is a significant heterogeneity in where these negotiations are happening, and one dimension of heterogeneity is the plot’s property right regime. Indeed, to get compensated, owners must prove their ownership rights, but doing so is costly and this cost differs across the different property right regimes in Kampala. As summarized in Table 3, owners on leasehold land report a cost 33% lower than mailo and 58% lower than freehold owners. In turn, leasehold owners are 55% and 75% more likely to negotiate than mailo and freehold owners, respectively.

To isolate the impact of each property right from other factors like the amount of land being affected, I model the probability that owner  $o$  in location  $i$  negotiates in order to get compensated  $\gamma_{oi}^N$ , as opposed to consenting to donate their piece of land without compensation. I model this decision as the probability that the potential gains from negotiating - the market rate  $q_i$  times the amount of land being taken

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<sup>18</sup>These numbers are consistent with administrative data I have access to on a subset of roads. Matching the universe of affected owners to dated consents, I find that 22% (33%) of affected owners had consented within 30 (90) days (from the day the first owner on that road consented). This number is likely biased downwards as not all consents in the database could be matched to the universe of affected owners.

$\Delta H_{oi}$ , as per the eminent domain legal framework - is greater than the costs of negotiating. The average residential property rate per m<sup>2</sup>,  $q_i$ , is recovered from the broker survey. This cost depends on owners' property right regime  $Z_{oi}$ , and some idiosyncratic cost  $\epsilon_{oi}$ , such that

$$\gamma_{oi}^N = \mathbf{P} \left( \alpha_1 \times q_i \times \Delta H_{oi} - \alpha_0 T_{oi}^0 \geq Z'_{oi} \mu_Z + \epsilon_{oi} \right), \quad (3)$$

where I control for whether owners were initially offered some financial compensation  $T_{oi}^0$ . I assume that  $\epsilon_{ok}$  is logistic distributed (0,1) and I estimate the parameters of the model  $\{\alpha_1, \alpha_0, \mu_Z\}$  through maximum likelihood. The eminent domain imposes an upper bound on compensation, ruling out owner holdout. This is consistent with evidence from the survey where owners were asked how likely they believed the road would be upgraded on a scale of 0 to 5. 96% and 89% of owners stated that the road had more than 80% (4 out of 5) changes to be upgraded if they accepted the government's offer right away, or negotiated, respectively. The full distribution of answers is displayed in Appendix Figure A13.

There are two key parameters to estimate. First,  $\alpha_1$  governs the elasticity of owners' negotiation probability  $\gamma_{oi}^N$  on the market value of the affected land, or the extent to which owners are going to respond to different values and amounts of their land being encroached upon. To causally estimate this parameter, I need to assume that there is no selection of  $\Delta H_{oi}$  or  $q_i$  conditional on  $Z_{oi}, T_{oi}$ . The amount of affected land  $\Delta H_{oi}$  is determined by engineering studies before the start of the road construction and to the extent that it is reported with measurement error by survey respondents, I assume that this measurement error is classical. In other words, accepting respondents do not systematically over or under estimate  $\Delta H_{oi}$  compared to negotiating respondents. In addition, I need to assume that there is no omitted variable bias at the parish level, so  $q_i$  is orthogonal to  $\epsilon_{oi}$ . Importantly, to identify  $\alpha_1$ , I use both across-road and within-road variation. In the former, both  $\Delta H_{oi}$  and  $q_i$  vary, while in the latter, all the variation comes from  $\Delta H_{oi}$  (e.g. differences in the amount of land affected across two neighboring owners on the same road) as  $q_i$  is defined at the parish level. Differences in the amount of land affected across two neighboring owners on the same road may come from uneven initial or final road width or from some plots encroaching on the existing road, which is common due to the unplanned nature of historical urban development in Kampala. If low property value  $q_i$  parishes were also home of owners less likely to negotiate, conditional on owner and parish characteristics including property right regimes, then my estimate of  $\alpha_1$  would be biased upward.

The second parameter of interest,  $\mu_z$ , governs the relationship between the different property right regimes and owners' negotiation. As the different regimes are pre-determined to the policy, the estimated coefficient captures their overall effect on owners' decision. As those regimes may be correlated with other observables likely to affect owners' negotiation, including owners' wealth and their social integration in the community, to isolate the impact of property right regimes on negotiation probability, I control for owners' wealth and social integration in some specifications. Overall, to interpret this coefficient as the causal effect of different levels of property right clarity, I need to assume that my controls - which include owner-level and parish-level socio-economic characteristics and geography - are capturing potential owner sorting, as well as all other aspects of the different regimes, including heterogeneous historical economic development (Bird and Venables (2020)).

The results are presented in Table 7, with non-dummy regressors standardized for interpretation. In column 1, without any parish level control, I estimate  $\hat{\alpha}_1 = 0.316^{***}$ , which corresponds to a 37% increase in the odds of negotiating for the average owner in response to a one standard deviation in the market value of owners' affected land. This number is robust to including parish-level geographical (latitude, longitude, latitude<sup>2</sup>, longitude<sup>2</sup>) and socio-economic (wealth index, 2014 population and the share of owner occupied properties) in column 2 and division fixed effects in columns 3 and 4. In column 5, when including parish fixed effect, all variation comes from within-parish differences in the amount of affected land (as  $q_i$  is fixed within parish) and, while I am less powered statistically, the estimate does not appear to be statistically different.

Moving on to the role of property right regimes, I find that compared to owners on leasehold land - the strongest property rights and reference category -, owners on Mailo and Freehold land had lower odds of negotiating, with estimated coefficients  $-0.673^{***}$  and  $-0.667^{**}$  in the absence of parish-level controls in column 1, respectively. The coefficient on mailo is robust to including parish-level controls (column 2), division fixed effects (column 3) and owner-level controls (column 4), and the coefficient on freehold jumps to  $-1.253^{***}$  (column 2) and up to  $-1.442^{**}$  (column 4), consistent with freehold neighborhoods having different socio-economic characteristics. Given the limited within-parish variation in property right regimes, I am not powered to estimate these coefficients precisely with parish fixed effects (column 4). Noticeably, the estimates on mailo and freehold are about 40% and 30% lower than the coefficients in columns 1-3. However, the identifying variation comes only 42% of the respondents in parishes with at least 2 distinct property rights regimes (29% of the parishes), and the estimated coefficients are not statistically different than when variation from the whole sample is used for identification.<sup>19</sup>

With these estimates, I predict the share of negotiating owners in each parish and the corresponding payment for owners. The distribution of owner-level payment plotted on Figure 3 highlights the shift in land acquisition costs under *voluntary land take* to the left (orange), compared to payment at market value (green). I estimate that under *voluntary land take*, total land costs are equal to 18 million USD.

Taking stock, I estimated that the benefits of road improvements in terms of property values by the side of improved roads was 67 million USD (broker appraisal). The market value of affected land acquired from private owners is around 65 million USD. Even if land was acquired under voluntary land take for 18 million USD, the construction costs of 80 million USD would imply negative net returns for these investments if the benefits were only local. However, road improvements have benefits throughout the whole city, as roads are used by workers to commute between their residence and workplace. Hence increasing local road infrastructure benefits workers all over the city by decreasing commuting costs. To properly account for spatial spillovers, and study how the costs and benefits of road improvements interact, in Section 4 I build a quantitative spatial model of optimal road improvement with land acquisition.

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<sup>19</sup>These findings are consistent with estimates from administrative data I have access to for a subset of roads part of the last wave of road upgrading (KCRRP). I find that the more land affected, the less likely owners are to have consented within 90 days (from the day the first owner on that road consented): going from the first to the last quintile in terms of amount of affected land decreases the probability to have consented within 90 days by 6 percentage points. I also find that owners under the "customary" (no land title) property right regimes are 13-19 percentage point more likely to have consented within 90 days than titled owners. Unfortunately, this administrative data does not include the exact location of plots, so I cannot match it to my data on property right regimes.

Table 7: Drivers of Negotiation and Delays - Owner Survey

	Propensity to Negotiate				
	(1)	(2)	(3)	(4)	(5)
Market Value Affected Land	0.316*** (0.106)	0.324*** (0.107)	0.326*** (0.108)	0.367*** (0.121)	0.251* (0.134)
Tenure Mailo	-0.673*** (0.249)	-0.689*** (0.255)	-0.684*** (0.257)	-0.681* (0.365)	-0.395 (0.516)
Tenure Freehold	-0.667** (0.327)	-1.253*** (0.4)	-1.285** (0.516)	-1.442** (0.62)	-0.874 (1.266)
Any Compensation Initially Offered	-0.541* (0.308)	-0.757** (0.319)	-0.813** (0.327)	-0.816** (0.362)	-0.456 (0.423)
Observations	544	544	544	540	443
Ref Tenure	Leasehold	Leasehold	Leasehold	Leasehold	Leasehold
Geo FE			Division	Division	Parish
Parish Geo Controls		Y	Y	Y	
Parish SocioEcon Controls		Y	Y	Y	
Owner SocioEcon Controls				Y	Y

Notes: The coefficients of equation (3), estimated through maximum likelihood are presented in this table. In column 1, I do not include any other control than the ones presented in the table. In columns 2,3 and 4, I include parish-level geographical (latitude, longitude, latitude<sup>2</sup>, longitude<sup>2</sup>) and socio-economic (wealth index, 2014 population and the share of owner occupied properties) controls. The parish-level wealth index is a standardized index made of the share of residents (above 18 y.o.) who are literate, the share of housing units with a permanent floor, the share of households owning a computer, the share of households owning a bicycle, the inverse of the share of households within distance of a secondary public school, the share of households who have had a bank account in the past 12 months and the share of households who reporting using any mosquito net. In column 4, I include an owner-level wealth index and an owner-level social-integration index to control for owners' covariates that may be correlated with property right regimes and influencing owners' decision. In column 5, the sample is restricted to respondents with non-missing parishes and in parishes with at least one observation. The market value of affected land is standardized so as to yield mean 0 and standard deviation 1. Source: own survey.

## 4 Model

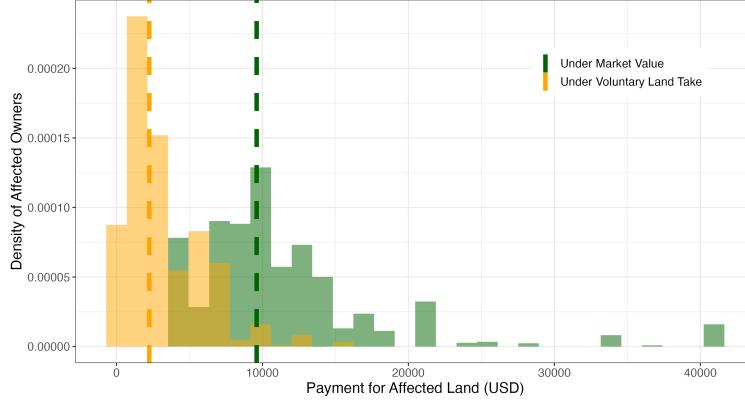
I propose a static general equilibrium model of a city, populated by a fixed number of households who choose where to live, work, and which route to commute on. Commuting happens on a road network, that connects neighboring locations. The road network takes up land, so that in each location, residential use and road infrastructure compete for land. Residential land is owned by private owners. These owners may be compensated if their land is transformed into roads and they do not give it up for free. A city government determines the optimal road improvement to maximize city residents' welfare, tying together the spatially heterogeneous benefits and land costs of road improvements.

### 4.1 Quantitative Spatial Model of Optimal Road Improvements with Land Acquisition

#### Environment

The city is composed of distinct locations, indexed by  $i, j \in \mathcal{J}$ . Locations are connected through a road network (undirected graph), where each location  $i$  is connected to  $\mathcal{C}(i)$  other locations. Workers must use this road network to commute. Locations are endowed with fixed amounts of land  $H_i$  as well as exogenous residential amenities  $B_i$ , productivity  $A_j$ , and landowners property rights  $Z_i$ . Each link between neighboring locations  $kl$  has a baseline amount of road infrastructure of width  $\bar{R}_{kl}$  (meters).

Figure 3: Owner-Level Predicted Land Costs For Realized Road Upgrades



Notes: I plot the distribution of (owner-level) land costs for the realized road upgrades, when payment is made at market value (green) or under the voluntary land take existing system (orange). I assume that each owner is affected by  $75m^2$  (50m long over 1.5m wide), and that they are compensated at the median parish-level residential rate as recovered from the broker survey. To build the orange distribution, I further predict the share of owners negotiating in each parish given this amount of affected land and residential rate, as well as the parish coordinates and property right regimes (col 2, Table 7).

## Land

At baseline, in each location  $i$ , total land  $H_i$  is divided between private residential land  $H_i^r$  and public road land  $H_i^p$ , such that:

$$H_i = H_i^r + H_i^p \quad \forall i. \quad (4)$$

Road land  $H_i^p$  is the sum of the area used by roads of width  $R_{mi}$  and length in  $l_{mi}$ , connecting  $i$  to neighboring locations  $m \in \mathcal{C}_i$  and assuming that the road is split equally across the two locations,

$$H_i^p = \sum_{m \in \mathcal{C}_i} R_{mi} \times \frac{l_{mi}}{2} \quad \forall i. \quad (5)$$

## Land Owners

Residential land  $H_i^r$  is owned by  $N_i^o$  local immobile representative landowners who get utility from consuming the freely tradeable good. Owners in location  $i$  share rental income  $H_i^r \times q_i$ , taxed at rate  $\phi$ , plus a transfer from the government  $\frac{T_i}{N_i^o}$ , such that owners' welfare is given by

$$W_i^o = \frac{H_i^r q_i (1 - \phi) + T_i}{N_i^o}. \quad (6)$$

## Workers

The city is inhabited by a fixed mass  $\bar{L}$  of mobile workers who choose where to live  $i$ , work  $j$ , their commuting route  $r$  as well as how much to consume of the tradeable good  $C_{ij}$  and housing  $H_{ij}$  to solve

their budget constraint:

$$\begin{aligned} \max_{i,j,C_{ij},H_{ij}} U_{ij,r} &= B_i \left( \frac{C_{ij}}{\beta} \right)^\beta \left( \frac{H_{ij}}{1-\beta} \right)^{1-\beta} \epsilon_{ij}. \\ \text{s.t.} \quad \frac{w_j}{\tau_{ij}} &= C_{ij} + H_{ij}q_i. \end{aligned} \quad (7)$$

$q_i$  is the rental rate in location  $i$ ,  $B_i$  the residential amenity in location  $i$ ,  $w_j$  the wage in location  $j$  and  $\tau_{ij}$  the average iceberg commuting costs between  $i$  and  $j$ . Commuting costs  $\tau_{ij}$  between their residence  $i$  and workplace  $j$  decreases workers' available income  $\frac{w_j}{\tau_{ij}}$ .  $\epsilon_{ij}$  is a preference shock over the pair  $ij$  drawn from a Fréchet distribution with shape parameter  $\theta$ . In equilibrium, workers' expected welfare  $W^w$  is equalized over space, so that

$$W^w \equiv \left( \sum_{ij} \left( \frac{B_i w_j}{q_i^{1-\beta} \tau_{ij}} \right)^\theta \right)^{\frac{1}{\theta}}. \quad (8)$$

The larger  $\theta$ , the more responsive are workers to local economic conditions given by  $\frac{B_i w_j}{q_i^{1-\beta} \tau_{ij}}$ , as their idiosyncratic preferences are less dispersed. Consequently, the larger  $\theta$ , the more workers will relocate in response to a change in commuting costs  $\tau_{ij}$ .

These preferences imply that the number of workers commuting from  $i$  to  $j$  in equilibrium is

$$L_{ij} = \bar{L} \frac{\left( \frac{B_i w_j}{q_i^{1-\beta} \tau_{ij}} \right)^\theta}{\sum_{mn} \left( \frac{B_m w_n}{q_m^{1-\beta} \tau_{mn}} \right)^\theta} = \bar{L} \frac{\left( \frac{B_i w_j}{q_i^{1-\beta} \tau_{ij}} \right)^\theta}{(W^w)^\theta}. \quad (9)$$

Local residential count increases in the rent-adjusted amenities  $\left( \frac{B_i}{q_i^{1-\beta}} \right)$  with elasticity  $\theta$ . Similarly, the larger the nominal local wage  $w_j$ , the larger local labor supply. On the other hand, the larger the commuting costs  $\tau_{ij}$ , the less likely are workers to select the  $ij$  option. The attractiveness of a pair  $ij$  therefore depends on the characteristics of its residential location  $i$ , its workplace location  $j$ , the commuting between the two locations. The attractiveness of this pair is relative to the attractiveness of all other available pairs in the city, summarized by the endogenous aggregate variable  $W^w$ .

The number of residents  $L_i^R$  and number of workers  $L_j^F$  in locations  $i$  and  $j$  can be further defined as

$$L_i^R = \sum_j L_{ij}, \quad L_j^F = \sum_i L_{ij}. \quad (10)$$



## Commuting

$\tau_{ij}$  is the average iceberg commuting costs of workers living in  $i$  and working in  $j$ . It is a function of the  $\tau_{ij,r}$  on all possible routes  $r \in \mathcal{R}_{ij}$  between  $i$  and  $j$ , such that

$$\tau_{ij} \equiv \left[ \sum_{r \in \mathcal{R}_{ij}} \tau_{ij,r}^{-\rho} \right]^{-\frac{1}{\rho}}. \quad (11)$$

Average commuting costs between  $i$  and  $j$  are decreasing as route-level commuting costs  $\tau_{ij,r}$  decrease, with elasticity  $\rho$ . The larger  $\rho$ , the larger the decrease in average commuting costs  $\tau_{ij}$  in response to a decrease in the lowest route-level commuting costs  $\tau_{ij,r}$ . As  $\rho \rightarrow \infty$ , all commuters use the least cost path.

The total costs of traveling through route  $r \in \mathcal{R}_{ij}$  is a function of the cost of commuting  $d_{kl}$  on all individual edges  $kl$  on route  $r$ , such that

$$\tau_{ij,r} = \prod_{kl \in r} d_{kl}. \quad (12)$$

Edge-level commuting costs  $d_{kl}, kl \in r$  are an exponential function of travel time, as in [Ahlfeldt et al. \(2015\)](#) and are given by

$$d_{kl} = \exp \left( \kappa \cdot \frac{\bar{t}_{kl}}{(R_{kl})^\xi} \right), \quad (13)$$

where  $\kappa$  is the elasticity of commuting costs on commuting times,  $\bar{t}_{kl}$  is the time it takes to commute on  $kl$  under 1 unit of road infrastructure  $R_{kl}$ , which depends on the length of link  $kl$ .  $\xi$  governs the elasticity of commuting time on the level of road infrastructure. The larger  $\xi$ , the more speed will increase (and commuting time decrease) in response to an increase in road infrastructure (corresponding to road improvements in this model). There is no congestion in this model and commuting costs are symmetrical, s.t.  $d_{kl} = d_{lk} \forall kl$ . I assume that workers must use roads to commute, which is a reasonable assumption in my setting where there is not subway or commuter train in the city.

Given workers' preferences for specific routes, governed by  $\rho$ , the equilibrium expected commuting cost  $\tau_{ij}$  depends on all edge-level commuting costs, represented by a matrix. Following [Allen and Arkolakis \(2022\)](#), I write this relationship as the matrix  $\tau$  of dimensions  $N \times N$ , where the  $(i, j)$  elements is  $\tau_{ij}$ ,

$$\tau = \left( (\mathbf{I} - \mathbf{A})^{-1} \right)^{-\frac{1}{\rho}}, \quad (14)$$

where  $\mathbf{A} \equiv [d_{kl}^{-\rho}]$  is a matrix with  $(k, l)$  element  $d_{kl}^{-\rho}$ .

## Production

Perfectly competitive firms produce a freely traded final good using only labor, such that

$$Y_j = A_j \cdot L_j^F \quad \forall j,$$

where  $L_j^F$  is the number of workers in  $j$  and  $A_j$  the productivity in that location. This production function implies that

$$w_j = A_j \quad \forall j. \quad (15)$$

### Residential Land Market Clearing

Land markets are competitive so the equilibrium rental rate  $q_i$  equates the supply of residential land  $H_i^r$  to the demand for residential land pinned down by equation 7. In equilibrium, all workers spend a fraction  $(1 - \beta)$  of their income on housing due to their Cobb-Douglas utility function, so that

$$q_i = (1 - \beta) \frac{\sum_j L_{ij} \frac{w_j}{\tau_{ij}}}{H_i^r}. \quad (16)$$

Road improvements will impact  $q_i$  directly through a decrease in commuting costs  $\tau_{ij}$  increasing workers' available income and thus the total amount spent on housing and, if happening in location  $i$ , through a decrease in  $H_i^r$ . Road improvements will also impact  $q_i$  indirectly, through equilibrium changes in  $L_{ij}$ .

### Land Payments

Improving road infrastructure on link  $kl$ , from width  $R_{kl}^0$  to width  $R_{kl}$ , requires acquiring  $(R_{kl} - R_{kl}^0) \times \frac{1_{kl}}{2}$  units of land from owners in  $k$ , at price  $p_k^1$ . According to the “*voluntary land take*” framework, the final land price is equal to the market rate under  $R_{kl}^0$ ,  $q_k^0$ , times the probability that owners in  $k$  get paid  $\gamma_k^1$ :

$$p_k^1 = q_k^0 \times \gamma_k^1, \quad (17)$$

where I assume that the fraction of owners who negotiate,  $\gamma_k^1$ , get paid.<sup>20</sup>

As described in Section 3.2.2, owners negotiate if the potential benefits are greater than the costs. The potential benefits are the amount of affected land in location  $k$ , split across all owners in  $k$ ,  $\frac{\Delta H_k^r}{N_k^o}$ , evaluated at the pre-road upgrade market rate  $q_k^o$ . Negotiation costs are decreasing owners' property rights  $Z_k$  at rate  $\mu_z$  and have an idiosyncratic component  $\epsilon_k$  following a logistic distribution of mean 0 and standard deviation 1, such that

$$\gamma_k^1 = P \left( \alpha q_k^o \frac{\Delta H_k^r}{N_k^o} \geq Z_k \mu_z + \epsilon_k \right). \quad (18)$$

The fraction of owners negotiating,  $\gamma_k^1$ , is an equilibrium object which depends on the residential market rate before the policy  $q_k^o$ , road widths  $\{R_{kl}^0\}$ , as well as the change in residential land in location  $k$ ,  $\Delta H_k^r$ , which is itself increasing in the new road area in that location, as  $\Delta H_k^r = \sum_l \frac{l_{kl}(R_{kl} - R_{kl}^0)}{2} \geq 0$ .

<sup>20</sup>I do not observe whether all owners who negotiate actually get paid but focus group discussions with affected owners highlighted the prevalence of formal and informal monetary payments, as well as in kind compensations to replace destroyed fences and gates for example and whose value is likely to be correlated with the local market value. If not all owners get compensated (or at a value below market rate), I would be overestimating  $p_k^1$ . However, negotiation also brings up other costs not modeled here (e.g. delays, processing times and court fees), which would increase the final cost.

These land payments are transferred to the affected owners as a transfer, and so,

$$T_k = p_k^1 \times \Delta H_k^r. \quad (19)$$

### Government's Optimal Road Improvement Problem

Conditional on the road infrastructure at baseline  $\{R_{kl}^0\}$ , the government chooses the new road infrastructure  $R_{kl} \geq R_{kl}^0 \forall kl$  to maximize a weighted average of residents' (workers and owners) expected welfare, subject to two budget constraints. First, the government has access to external funds  $\bar{M}$ , which can be used exclusively for road construction alone, at fixed marginal construction cost  $\bar{p}$ . Second, to fund any additional construction expenditures  $\hat{C}$ , as well as the acquisition of the underlying land from private owners at the equilibrium rate  $p_k^1$  defined by equation 17, the government must levy property taxes at rate  $\phi$ , given a tax revenue wedge  $\eta \geq 0$ , corresponding to tax leakage and corruption. Property taxes contributed to 47% of the Kampala Capital City Authority (KCCA)'s own-source revenues in 2021/2022, making it the top source of local revenues (Regan and Manwaring 2024). These two budget constraints match to the empirical setting of Kampala, where the government has access to external funds from the World Bank and the African Development Bank to fund road construction, but cannot use these funds for land acquisition.

The government's objective function is the weighted average of owners' welfare  $W_k^o$  and workers' expected welfare,  $W^w$ , defined in equations 6 and 8, respectively, and which both depend on the optimal policy. Owners in location  $k$ , and workers are weighted by Pareto weights  $\omega_k^o$  and  $\omega^w$ , respectively, so that the problem of the government is as follows:

$$\max_{\{R_{kl}\}_{\forall kl}, \phi} \mathbf{W} \equiv \sum_k [\omega_k^o W_k^o(\{R_{kl}\}, \phi) + \omega^w W^w(\{R_{kl}\}, \phi)], \quad (20)$$

s.t.

$$\hat{C} \equiv \min \left\{ 0.0, \bar{p} \cdot \sum_{kl} \frac{l_{kl} (R_{kl} - R_{kl}^0)}{2} - \bar{M} \right\}, \quad [\text{External Funds BC}]$$

$$\sum_k p_k^1 \underbrace{\frac{l_{kl} (R_{kl} - R_{kl}^0)}{2}}_{\equiv T_k} + \hat{C} \leq (1 - \eta) \phi \sum_i H_i^r q_i. \quad [\text{Domestic Funds BC}]$$

### Equilibrium

Given the parameters  $\{\beta, \theta, \kappa, \xi, a, \eta, \alpha, \mu_z, \bar{M}, \bar{p}, \bar{L}\}$ , location and link characteristics  $\{B_i\}$ ,  $\{A_i\}$ ,  $\{H_i\}$ ,  $\{Z_i\}$ ,  $\{\bar{t}_{kl}\}$ , welfare weights  $\{\omega_i^o\}$ ,  $\omega^w$  and baseline infrastructure  $\{R_{kl}^0\}$ , the equilibrium of the model is the set of prices  $\{q_i\}$ ,  $\{p_i^1\}$ ,  $\phi$  and quantities  $\{L_{ij}\}$ ,  $\{\gamma_i^N\}$ ,  $\{H_{ij}\}$ ,  $\{C_{ij}\}$ ,  $\{R_{kl}\}$  such that (i) the government chooses  $\{R_{kl}\}$  and the corresponding tax rate  $\phi$  to maximize workers and owners' welfare ; (ii) the government's budget constraints are satisfied ; (iii) workers choose  $\{i, j, r, C_{ij}, H_{ij}\}$  to maximize their utility ; (iv) owners consume all their income on the tradeable good and choose to negotiate with probability  $\gamma_i^N$  ; (v) residential land markets clear ; (vi) the good market clears, and (vii) labor markets clear.

## 4.2 Main Model Mechanisms: A Simple Economy on a Grid

To highlight the key forces of the model, I simulate a city of 25 locations, arranged on a  $5 \times 5$  grid, represented in Figure 4. Productivity is high in the central location and close to zero everywhere else (panel a). All locations have similar residential amenities (panel b), total land supply and road infrastructure at baseline. The distributions of equilibrium residential population (panel d), employment (panel e) and residential rental rate (panel f) are intuitive: most residents want to work in the central location to access the highest productivity and wages, and, as a result, their demand for land is decreasing in distance to the center, as they must incur a longer commute. This geography is a simplification of Kampala's, where most jobs are located at the center (Figure 4). In addition, property right regimes  $Z_k$  are highest at the central locations (panel c), to mirror the location of leasehold land, the strongest property right in the city (Figure 2).

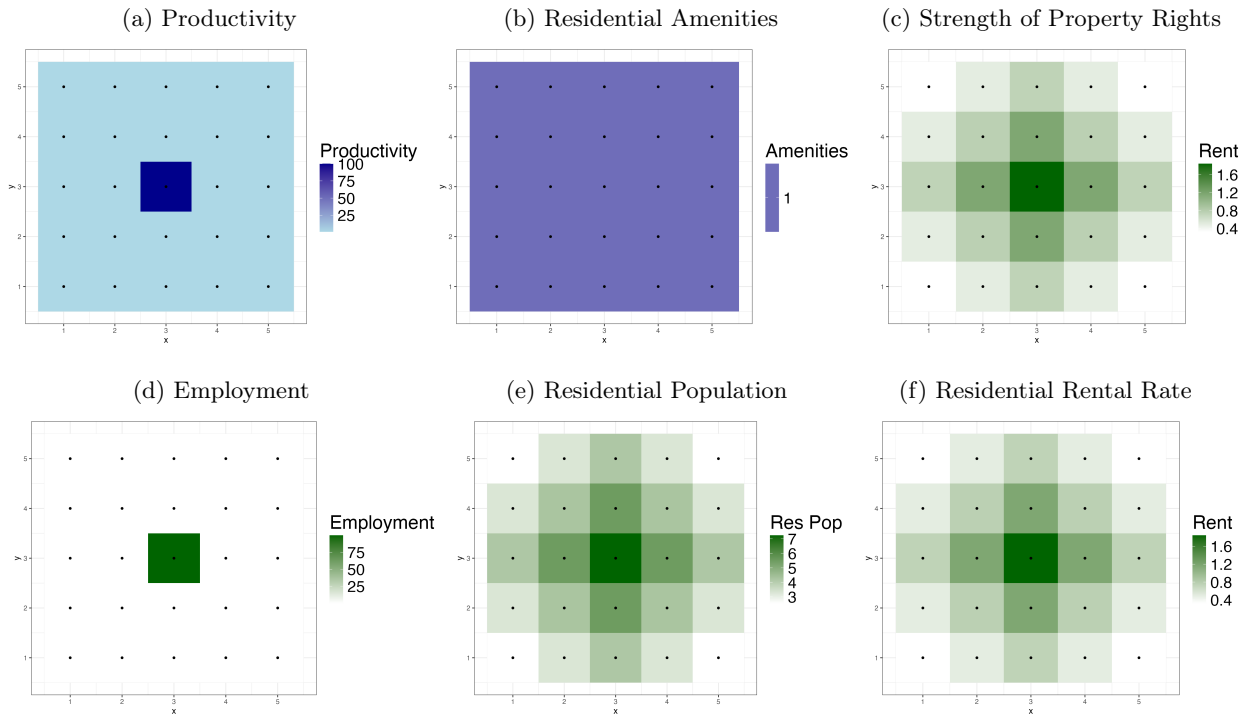
Road improvements have benefits and costs and I illustrate how their joint distribution affects the net welfare gains of different road improvements. In practice, for each road link, I increase width by one unit fixing width at baseline for all other links, solve for the new spatial equilibrium and compute the corresponding welfare change for upgrading this specific link. This exercise highlights how the level and relative welfare gains from different road upgrades vary in response to the different forces of the model and land acquisition rule. By looking at road upgrades one by one, I abstract from complementarity effects across road upgrades.

**Net welfare gains of road improvements in the absence of land payments:** I first assume that road improvements do not use land and have no other costs than construction costs, which is the dominant approach in the literature. Given the homogeneous distribution of road infrastructure at baseline, the roads connecting the central location host the well paying jobs and thus have the largest marginal benefits (Figure 5, panel a). Yet, these central locations also have the largest opportunity cost of land use, as workers would like to live as close as possible to the majority of jobs. The welfare-maximizing planner accounts for this opportunity cost of land use, decreasing the net welfare gains from road improvements in all locations, and especially by the city center (panel b). The larger the supply of residential land, the less impactful is this channel (not shown).

**A large tax revenue wedge decreases net welfare gains of road improvements, especially at the center:** Under the eminent domain legal framework, landowners must be compensated at market value for their affected land. This compensation is funded through property taxes  $\phi$ . In the absence of a tax revenue wedge  $\eta = 0$ , this transfer is welfare neutral for a utilitarian planner, as owners are assumed to have linear utility in consumption. Consequently, as displayed on panel c, the road improvements and corresponding welfare at the same than in the absence of land payments. However, under a large tax revenue wedge (panel d), the high land payments at the center of the city have a high welfare costs, both their decreasing their net welfare gains and making these central improvements less attractive than improvements at the periphery, which are lower benefit but also lower cost. Note that if landowners had decreasing marginal utility of income, compensating them at market value would be welfare improving compared to the no payment counterfactual, but the tax revenue wedge would still shift optimal improvements away from the high benefit central locations.

**Weak property rights and *voluntary land take* can lead to higher welfare gains but change the hierarchy of returns across locations:** Under the voluntary land take approach, adopted by the government, land acquisition prices  $p_k^l$  in location  $k$  are tied to owners' property right  $Z_k$ , so that for a given amount of land being taken, the stronger  $Z_k$ , the larger  $p_k^l$ . In Kampala, leasehold land features the strongest property rights and largest negotiation rate (Section 3.2.2), and is also dominant at the city center as displayed on Figure 2. As displayed on panel *e*, the net welfare gains of road improvements under voluntary land take are higher than under market value (panel d) because of the lower associated fiscal burden (not all owners get compensated). In addition, the gradient of relative land costs is steeper: central locations do not only have relatively high land market value, but also relatively high rates of owner negotiations, compared to peripheral locations. Consequently, in this simulation, road improvements with the largest net welfare gains under voluntary land take are displaced further from the city center compared to market value.

Figure 4: Simulated Economy - Baseline Equilibrium



## 5 Model Estimation and Net Returns to Realized Road Improvements

### 5.1 Estimating The Parameters of the Model

To estimate the model parameters, I use the data and variation from the reduced-form analysis. I start by describing my estimation of key elasticities. I then calibrate the remaining parameters using estimates from other studies and public data on Kampala.

### 5.1.1 The elasticity of local speed on road infrastructure $\xi$

$\xi$  governs the changes in speed in response to an increase in road infrastructure and thus is a key parameter in my model to determine the benefits of new roads. The larger  $\xi$ , the larger the increase in speed for any additional meter of road width. As in Section 3.1.1, I use the Google Maps API queries to study the relationship with road speed and road infrastructure. I compare speed on roads upgraded early in the rollout of the policy, to speed on roads upgraded later or to be upgraded. An important change compared to the results presented in Section 3.1.1 is that in the model, the measure of road upgrades is continuous. Thus, here, I map the road upgrade (dummy) to a change in the amount of road infrastructure, or road width. Non-upgraded road width is defined as per the KCCA road inventory<sup>21</sup> and upgraded roads add 3.2 meters (one lane) to that initial width. Indeed, as most road improvements involved *dualling* and/or *paving*, I assume to this corresponds to adding one lane to the original road. I exclude trips on roads under construction.

The identification assumptions are analogous to the ones in Section 3.1.1, and I present the results in columns 3 and 4 of Table 4. I estimate an elasticity  $\xi$  between 0.31\*\*\* and 0.390\*\*\*. This number is 2 to 3 times larger than existing estimates for this parameter, all in middle or high income countries. For example, Fajgelbaum and Schaal (2020) translate estimates by Couture et al. (2018) on U.S. data into  $\xi = 0.10$ . In Santiago, Chile, Bordeu (2024) estimates  $\xi = 0.13$ . These papers use either OLS specifications, or instrument for local road infrastructure with institutional features, but do not leverage changes in road infrastructure from a policy, as I do here. In addition, the discrepancies in these coefficients should not be surprising, given the dramatically different levels of road infrastructure at baseline, illustrated by the average speed in Couture et al. (2018) being 38.5 km/h, or 60% higher than in Kampala.

### 5.1.2 The elasticity of commuting flows on commuting costs

Now, turning to the parameters governing how workers' location choice responds to changes in commuting times using data from a local ride-sharing company. This increase in local road speed will induce changes in commuting flows in the city. Taking workers' residence and work locations as given, workers will be more likely to choose a route using these faster roads. I provide evidence that this re-routing is happening in Appendix D.2, by showing that over time, Google Maps trips between parish centroids are increasingly likely to use roads in the second wave of the policy, which are progressively upgraded over the course of the sample. Workers will then reallocate in response to these local changes, changing the number of residents and workers in each location, thereby affecting commuting flows  $L_{ij}$ . The magnitude of this response is governed by the model parameters  $\theta$ , the elasticity of commuting flows on commuting costs, and  $\kappa$ , the elasticity of commuting costs on commuting times. These elasticities are important: the larger  $\theta\kappa$ , the larger the impacts of road improvements on workers' welfare.

To estimate this elasticity, I rely on the ride hailing data discussed in Section 2.2.2. To capture work-related trips, I restrict my sample to morning and evening trips during days of the week.<sup>22</sup> I aggregate

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<sup>21</sup>For the 27% of roads with missing width, I assign the average width of roads in their OSM category in that sample.

<sup>22</sup>This includes trips from 6am to 10am, and 4pm to 8pm. Motorcycles (bodas) is one of the most common mode of transportations in urban Uganda, and 90% of the public transportation providers in Kampala are motorcycle taxis (KCCA 2020).

the data to the origin-destination  $\times$  year level (or  $\times$  month), where  $t_{kl,t} = \frac{1}{N_{o,kl,t}} \sum_o t_{o,kl,t}$ , is the average commuting time for user  $o$  on  $ij$  in period  $t$ . I assume that drivers used the shortest path, so I combine equations (11), (12) and (13) and plug in the observed time  $t_{kl,t}$ , such that  $\tau_{kl,t} = \exp(\kappa \times t_{kl,t})$ . I take the log and first difference of the gravity equation (9), so

$$\log L_{ij,ym} = \alpha + \gamma_{i,y} + \gamma_{j,y} - \theta\kappa t_{ij,ym} + \eta l_{ij} + \gamma_m + \epsilon_{ij,ym}.$$

where the dependent variable is the log number of trips from parish  $i$  to  $j$ , in month  $m$  year  $y$ . Year-origin  $\gamma_{i,y}$  and year-destination  $\gamma_{j,ty}$  fixed effects absorb annual changes in location unobservables and month fixed effects  $\gamma_m$  capture time patterns. I do not have enough power to include pair fixed effects, but by controlling for the straight-line distance  $l_{ij}$ , I effectively use variation in  $t_{ij,ym}$  across trips of the same straight-line distance to identify the parameters  $\theta\kappa$ . Standard errors are clustered by origin-destination pairs. The identification assumption to causally identify  $\theta\kappa$  is that difference in  $t_{ij,ym}$  across pairs and over time are uncorrelated with unobservable differences captured by  $\epsilon_{ij,ym}$ . This condition would be violated if changes in commute times were correlated with unobserved pair-specific characteristics also affecting trends in commute flows, after conditioning on origin-month and destination-month. The results are presented in Table 8. In column 1, where observations are defined at the pair  $\times$  month level, I estimate  $\hat{\theta}\kappa = 0.033^{***}$ . This result is robust to restricting the sample to evening commuting alone (column 2,  $0.023^{***}$ ) and defining observations at the pair  $\times$  year level (Appendix Table A6). It is worth acknowledging two potential issues with these data. First, there is a potential selection issue as users of this motorcycle taxi company may be different than the average commuter in Kampala. Second, using data on actual trips (rather than predicted) prevents me from including an observation (pair  $\times$  period) for which no trip was recorded in a given time period.<sup>23</sup> Nevertheless, my estimate  $\hat{\theta}\kappa$  is in line with estimates in other contexts, including Bogota (Tsivanidis 2023,  $\hat{\theta}\kappa \in [0.028, 0.046]$ ). As  $\theta\kappa$  governs workers' response to a change in commuting times, a large  $\theta\kappa$ , as estimated for Santiago (Bordeu 2024,  $\hat{\theta}\kappa = 0.0656$ ) and Berlin (Ahlfeldt et al. 2015,  $\hat{\theta}\kappa = 0.0683$ ), would imply larger benefits from road improvements.

Given  $\hat{\theta}\kappa = 0.033$ , I follow the literature (Ahlfeldt et al. 2015) and I choose  $\kappa$ , the elasticity of commuting costs on commuting times = 0.01, resulting in  $\theta = 3.33$ .

### 5.1.3 Parameters Governing Owners' Negotiation

There is a direct mapping between the parameters government owners' decision to negotiate in the model and the empirical specification in Section 3.2.  $\alpha$  governs the elasticity of negotiation on the market value of affected land, and  $\mu_z$  governs the elasticity of negotiation on owners' property right regimes, as well as other observable characteristics. Given the stability of the coefficients across the different specifications presented in Table 7, my preferred specification to take to the model is in column 2, where I include parish coordinates (recentered longitude, longitude<sup>2</sup>, latitude, latitude<sup>2</sup>), parish-level socio-economic controls, and share of property right regimes per location. When solving the model, I also assume that the length of each owner's property is 50 meters (48m average in the owner survey), so that the total amount of affected land is equal to  $50m$  times  $\frac{1}{2}$  the additional road width ( $R_{kl} - R_{kl}^0$ ), so that land is taken equally

<sup>23</sup>Going forward, I will investigate how to address these issues by relying a travel habit survey conducted in 2016/2017 in Kampala. which includes approximately 600 respondents.

Table 8: Estimated  $\theta\kappa$ 

	<i>Dependent variable:</i>	
	(1)	(2)
Time (in Min)	-0.033*** (0.001)	-0.023*** (0.001)
Period Def	Month	Month
Sample	All	Evening
Fixed Effects		
- origin x year	Y	Y
- destination x year	Y	Y
- month	Y	Y
SE Clustered	o-d	o-d
Observations	59,300	45,077
R <sup>2</sup>	0.437	0.687

Notes: Standard errors are displayed in parentheses and clustered at the o-d pair level, with significance levels \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ . All specifications include origin-year, destination-year and month fixed effect. I control for log distance in all specifications. In column 1, I include all trips from 6am to 10am, and from 4pm to 8pm. In column 2, I restrict the sample to the evening rush hour from 4pm to 8pm. Observations are at the pair-month level. Trips are motorcycles (bodas) trips.

on both sides of the road, belonging to two different owners.

#### 5.1.4 Additional public data and calibrated parameters on Kampala

**Data on residential population  $\{L_i^r\}$  and employment  $\{L_i^F\}$ :** I get data on workers' residence and workplace at baseline from the Population and Housing Census (2014) and Census of Business Establishments (2011) from the Uganda Bureau of Statistics (UBOS), respectively. I aggregate this data to the location level, where one location corresponds either to one parish, or to a group of small neighboring parishes.

**Residential Land  $H_i^r$ :**  $H_i^r$ , the quality-adjusted quantity of residential land in the model, is pinned down by the land market clearing condition in equation 16, given data on  $\{q_i\}$  from the real estate broker survey.

**Share of workers' expenditure on housing  $1 - \beta$ :** I recover the share of workers' expenditure on housing,  $1 - \beta$  from the Uganda National Panel Survey (2019), restricted to households in the Kampala region. I recover  $(1 - \beta) = 0.2355$  as the ratio of the monthly rental expenditure (or monthly-equivalent value of housing for owners) and the household's total income, excluding respondents who report 0 income.

**Elasticity of commuting route on commuting costs  $\rho$ :** I set  $\rho = 50$ , which is significantly larger than in Allen and Arkolakis (2022) ( $\rho = 6.83$ ), corresponding to workers' choice route being closely aligned with the shortest route. Importantly this value of  $\rho$  ensures that the commuting cost matrix  $I - A$  is invertible.



**Weight on owners  $\{\omega_i^o\}$  and workers  $\omega^w$ :** I assume a utilitarian planner, putting equal weights on worker and owner households. I recover the total number of residents in age of working residing in each location from the 2014 Uganda Population Census, and I assume that 20.15% of them are owners, as per the 2019 Uganda National Panel Survey, which asks detailed characteristics to a subset of households, restricting the sample to Kampala.  $\omega^w$  is the total welfare weight on all workers in the city. To compare workers' utility to owners' utility, and be able to interpret these utilities as dollars, I scale the number of workers so that the welfare gains associated with making each worker a lump-sum transfer is equal to the total monetary sum of these transfers. This scaling factor is not constant given workers' non-linear expected utility in income, so I estimate it so that if the realized road construction expenditures were instead transferred lump sum to workers, the net welfare gains would be 0. Owners' utility is linear in income, no such scaling is needed and  $\omega_i^o$  is the number of owners in location  $i$ .

**$\{\tau_{ij}\}$  and Mapping Roads to the Model:** To estimate  $\{\tau_{ij}\}$ , I need to recover link-level commuting times in the absence of road infrastructure  $\bar{t}_{kl}$ , as well as the road width on each link at baseline  $R_{kl}^0$ . To do so, I use data from Open Street Map and from Google Maps. Open Street Map categorizes all roads in Kampala into five categories associated with different speeds as summarized in Appendix Table A4. I recover the average uncongested speed of each road type from the Google Maps trips between neighboring grid cells. In practice, I regress a trip's average speed on the share of the trip on each road category, controlling for the trip's distance, its hour, and second degree polynomials for the origin and destination coordinates, to control for local heterogeneity in road speed. The results are presented in Appendix Table A2, along with additional details on the estimation strategy.

To map these estimates to the model, I then aggregate roads by location, such that the average speed in location  $k$  is the weighted average of category-specific free flow speed on all roads in that location, weighted by the share of each road category. I recover the time between two neighboring locations  $t_{kl}$  as the distance between these locations times this average speed. I further recover  $\bar{t}_{kl} = t_{kl} \times (R_{kl}^0)^\xi$  from equation (13), where  $R_{kl}^0$  is the average width of roads in locations  $k$  and  $l$  at baseline. Given  $\{\bar{t}_{kl}\}, \{R_{kl}^0\}, \xi, \kappa$  and  $\rho$  I can compute  $\{\tau_{ij}\}$ .

I then map the improvements into the model's aggregated road network by summing over all improvements in a location. Assuming that the average road improvements increased road width by 3.2 meters (one additional lane, as discussed in Section), I compute the total amount of new road infrastructure by location and split it equally across all links connecting this location to its neighbors. The resulting map of road improvements is displayed on Figure 7.

**Fundamental Location Productive  $\{A_i\}$  and Residential  $\{B_i\}$  Amenities:** I follow the standard inversion procedure in the literature to recover the exogenous productivities  $\{A_i\}$  and residential amenities  $\{B_i\}$ . As data on property rates  $q_i$  is in USD, I scale  $\{A_i\}$  up to the median local wage (KCCA 2019). From the gravity equation (9), I then recover flows  $L_{ij}$  given  $\left\{ \tau_{ij}, w_j, q_i^{\beta-1} B_i, L_i^R, L_j^F \right\}_{\forall i \forall j}$  and  $\theta$ . Last, I solve for the quality-adjusted supply of land  $\{H_i^T\}$  given  $\{L_{ij}, w_j, q_i\}$  using the equilibrium condition for the market of residential land (16).<sup>24</sup>

<sup>24</sup>Note that as  $H_i^T$  is the quality-adjusted supply of residential land, it cannot be directly compared to, or recovered from satellite data on residential land use from the Global Human Settlement Layer. This data further does not differentiate

The data on  $L_i^R, L_j^F, q_i$  and the recovered  $A_i, B_i, H_i^r$  are displayed on Figure 6. Most jobs are concentrated in the city center (panel a), and both residential population (panel b) and residential rental rate (panel c) are high in the areas surrounding the city center. The relatively low population density on the immediate north east of the city center is explained by the prevalence of administrative and official buildings in these locations, rationalized by relatively low residential amenities (panel e) and quantity of quality-adjusted residential land  $H_i^r$  (panel f) in the model.

**tax revenue wedge  $\eta$ :** I calibrate  $\eta = 0.61$  from Regan and Manwaring (2024) who, in an experiment in collaboration with the Kampala city government, estimate that “*over the years 2019-2022, the KCCA has collected 0.39% of the potential revenue from property taxes.*”

## 5.2 Net Returns to Road Improvements in Kampala

Equipped with the estimated model, I solve for the new spatial equilibrium taking the realized road upgrades as given.

**Impacts Commuting Time and Property Values:** First, I fix workers’ locations at baseline and compute changes in commuting times across any pair of location and rental rates coming from the road upgrades  $\Delta R_{kl}$ . I find that the realized road upgrades led to a decrease in workers’ average commuting time by 8.0% (1min 49 sec), and an increase in total property values by 1.44%.<sup>25</sup> Second, I allow workers to relocate, leading to new equilibrium location and affecting rental prices. Facing lower commuting costs, workers increase the distance between their residence and workplace, so that the average commuting time decreased by only 6.75% compared to baseline. In turn, the average increase in property values is 1.45%.

**Welfare Impacts Under Voluntary Land Take:** I compute the compensating differential lump-sum transfer  $s$  that the government would have to make to every resident (owners and workers) to reach the same total expected utility as the road improvements policy. In other words, I solve for  $s$  such that:

$$\sum_k [\omega_k^o \times W_k^o (\{R_{kl}\}, \phi, 0) + \omega^w \times W^w (\{R_{kl}\}, \phi, 0)] = \sum_k [\omega_k^o \times W_k^o (\{\bar{R}_{kl}\}, 0, s) + \omega^w \times W^w (\{\bar{R}_{kl}\}, 0, s)]$$

I find that under a 3% discount rate, the Net Present Discounted Value (NPDV) of the lump-sum transfer that would have led to the same city-level welfare as the road improvements is 181 USD per resident. This number accounts for the land costs of the policy, funded through a property tax. Once I further subtract the one time construction expenditures, equivalent to 83 USD per resident, I recover net welfare gains equivalent to a one time transfer of 98 USD per resident, or 11% of the median annual wage for Kampala workers.<sup>26</sup>

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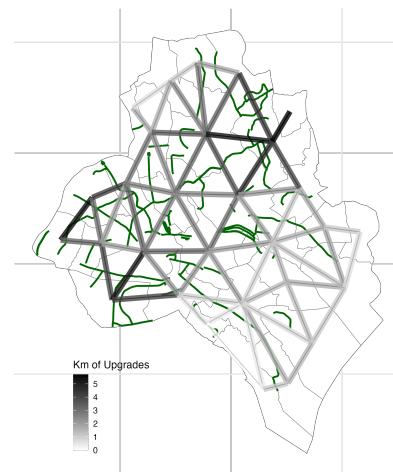
between residential and non-industrial business land use.

<sup>25</sup>Computing expected workers’ welfare while fixing locations at baseline is not informative, as expected welfare is defined over workers’ idiosyncratic preferences over pairs of locations, leading them to choose their residence and workplace.

<sup>26</sup>The 2014 Uganda Population Census counts about 1.6 million inhabitants, but only about 866,000 inhabitants about 18 years old. When talking about the residents in this paper, I focus on those inhabitants about 18 years old, who are of working age and thus agents in my model.

To interpret these numbers and compare them to the reduced-form results from Section 3, it is important to keep in mind what elements are captured or not by the model and by the reduced-form results. The model accounts for spatial spillovers through the road network and for the general equilibrium impacts of workers relocating over space in response to a change in commuting costs, increasing overall economic activity in the city. I do not explicitly model firms' land use (captured into the exogenous local productivity), so any increase in the property value of business premises would not be captured by my model. However, more than 71% of Kampala businesses count at most 2 employees (UBOS 2010), implying that most businesses are small and likely ran on the side of residential dwellings, so capitalization of the road improvements into residential property values should capture most of the gains. Last, my model has exogenous residential amenities  $B_i$ , but, given the empirical evidence that road upgrades improved local amenities (Appendix Figure A10) including a decrease in traffic accidents, my estimate is likely a lower bound on total net benefits.

Figure 7: Realized Road Improvements



**Welfare Impacts Under Alternative Land Acquisition Rules:** I estimate that the net welfare gains would have been 16% larger in the absence of land payments, but would have been 59% lower if land had been acquired at market value, as mandated under eminent domain, which is consistent with about 25% of owners only getting compensated at market value. This result is driven by the large tax revenue wedge, implying that land payments decrease the net welfare gains of the policy.

The government claimed that acquiring land at market value would threaten the viability of the project (World-Bank 2023). I assess this claim through the lens of my model. In practice, I do a grid search for the wedge  $\eta$  such that the net welfare gains of the realized policy are equal to 0. I find that for any  $\eta$  greater than 0.74, the net welfare gains of the realized road improvements would be negative if land was acquired at market value. This threshold goes up to 0.78 if I do not subtract the construction expenditures. The welfare gains for  $\eta \in [0.0, 0.9]$  is displayed on Appendix Figure A15. These estimates are consistent with the  $\eta = 0.61$  calibrated from Regan and Manwaring (2024).

**Sensitivity to Alternative Parameters:** Two key forces in the model are driving these results: the welfare impacts of land payments, governed by the tax revenue wedge  $\eta^\phi$  and workers' benefits from the road improvements, governed by workers' response to changes in local speed  $\theta\kappa$  and the impact of road improvements on local speed  $\xi$ . In Table 10, I investigate the sensitivity of the results with respect to each of these parameters.

### 5.3 Link-Level Upgrades and Model-Predicted Net Returns

To understand whether these benefits come from upgrading any road, or upgrading the right roads, I investigate the spatial distribution of improvements' marginal benefits, costs and net welfare gains. I

compare the length of roads upgraded along each link to the model-predicted link-level net welfare gains, benefits and costs.

### 5.3.1 Link-level Costs, Benefits and Welfare Gains

I start by computing the marginal costs, benefits and net returns of upgrading each parish-to-parish link one by one as in Section 4.2. In practice, for each link, I increase road width by 1 additional lane and solve for the resulting equilibrium. I compute the total change in commuting times, total land costs, and NPDV of the welfare gains (net of costs). The results are displayed on Figure 8. The color of each link indicates the impact of upgrading this specific link on city-level outcomes of interest. This exercise abstracts from complementarities across road upgrades, discussed in the next section.

Starting with a direct measure of benefits, the decrease in commuting times (panel a), I show that the highest marginal benefit links are concentrated at the center of the city, and that there is substantial heterogeneity across links. Moving on to land costs (panels b and c), I show that there is heterogeneity in the predicted share of owners bargaining (dark red), as 21.6% of owners would negotiate for roads at the 25<sup>th</sup> percentile of marginal costs, against 31.7% on roads at the 75<sup>th</sup> percentile. I plot the distribution of net marginal welfare gains on panel d. The first takeaway is that there is a lot of heterogeneity and that not all roads have positive net marginal welfare gains, as expensive construction and land costs are not always offsets by gains in commuting times and property values. The second takeaway is that the highest marginal net welfare gains are at the center of the city, consistent with these locations being also high marginal benefits (panel a).

### 5.3.2 Predicting Realized Road Improvements from Link-Level Net Welfare Changes

Did the government upgrade the highest net return (benefits - cost) roads, as predicted by the model? I answer this question in two steps. First, I look at the correlation between the length of upgraded roads along each link  $kl$ , and predicted link-level welfare gains. On Figure 9, I plot the distribution of predicted link-level welfare gains, split between above (green) and below (pink) median road upgrades along that link. This exercise highlights two patterns. First, there is a positive relationship between realized road upgrades and link-level welfare gains, as the green distribution is to the right of the pink one. Second, however, there is substantial overlap between the two distributions, suggesting some amount of misallocation. In Appendix Figure A16, I further investigate the relationship between model-predicted link-level net welfare gains (from a 1 lane upgrade) and the length of road upgrades around that link. This figure shows a positive relationship between both.

the picture is a little more nuanced when looking at the scatter plot of road upgrades (in km) on predicted net welfare gains (Figure 9, panel (b)).

Second, to unpack this relationship, I regress the length of upgraded roads along each link  $kl$ ,  $D_{kl}$ , on these predicted link-level impacts, controlling for the length of link  $kl$ ,  $l_{kl}$ :

$$D_{kl} = \alpha + \beta \times \Delta W_{kl} + l_{kl} + e_{kl}$$

Table 9: High Marginal Benefits and Low Land Costs Predict Realized Road Upgrades

	<i>Dependent variable:</i>		
	Length of Upgraded Roads (All upgrades)		
	(1)	(2)	(3)
Net Welfare Gains (std)	0.370** (0.184)		
Gain in Commuting Time (std)		0.360** (0.177)	0.320* (0.174)
Land Costs Under Negotiation (std)		-0.515** (0.222)	
Land Costs at Market Value (std)			-0.018 (0.248)
Share Owners Negotiating (std)			-0.528*** (0.175)
Observations	58	60	60
R <sup>2</sup>	0.069	0.135	0.188

Notes: Standard errors are displayed in parentheses, with significance levels \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ . Observations are at the link-level. Covariates are standardized (mean 0, sd 1).

where  $\Delta W_{kl}$  is the standardized change in city-level welfare from upgrading link  $kl$ . The results are displayed in Table 9. In column 1, I unveil a positive correlation between the length of roads upgraded along a link between neighboring locations and the road's predicted welfare gains from upgrading that link. In column 2, I split net welfare gains into benefits (gains in commuting time) and land costs. I unveil a positive correlation between road upgrades and predicted gain in commuting time, and a negative correlation between road upgrades and land costs. In column 3, I further split these land costs between land market value and the share of owners negotiating, and find that most of the negative correlation between upgrades and land costs seem to come from negotiations. Overall, I conclude that the selection process of roads to upgrade in Kampala was relatively good.

#### 5.4 Optimal Road Improvements in Kampala under the Status-Quo Land Costs

The above exercise recovers the net returns of the realized road improvements, but does not allow me to study how land acquisition affects the optimal set of road upgrades. To do that, I solve the model for the road improvements that maximize the government's objective function (equation 20). In Section 6, I then compare these optimal upgrades under the existing land acquisition framework to the ones under alternative property right regimes, alternative land acquisition rule and restrictions on the use of external funds.

I fix the maximum area of roads that can be upgraded at the level of the implemented policy and I solve for the welfare maximizing (optimal) road improvements in Kampala under the voluntary land take approach, the status quo. When solving for the optimal road improvement  $\{R_{kl} - R_{kl}^0\}$ , the government must keep track of several equilibrium objects: land payments  $\{p_k^l\}$  depend on  $\{R_{kl}\}$  through the probability owners

negotiate, which is increasing in the amount of land taken ; rental rates  $\{q_i\}$  depend not only on commuting costs  $\tau_{ij}(\{R_{kl}\})$  but also on workers' residence and workplace, which are also determined in equilibrium as a function of  $\{R_{kl}\}$ . In addition, unlike in Section 5.3, solving for the optimal upgrades accounts for the complementarities across road upgrades.

To solve for the optimal improvements, I follow the following procedure. First, I start from an arbitrary policy  $\{R_{kl}\}$  and solve for the corresponding spatial equilibrium locations  $\mathbf{x}_1 = \{L_i^R, L_j^F\}$ . Second, given these location choices  $\mathbf{x}$ , the planner solves for the optimal policy  $\mathbf{z} = \{\{R_{kl}\}, \phi\}$  that maximizes residents' expected welfare subject to its budget constraints. In that step, the planner accounts for the impact of the policy on all equilibrium objects  $\{W^w, W_i^o, q_i, \gamma_i^l, p_i^l\}$  but takes workers' equilibrium locations as given from the first step. Third, given the optimal policy from step 2  $\mathbf{z}_2 = \{\{R_{kl}\}, \phi\}$ , I solve for the new spatial equilibrium  $\mathbf{x}_3 = \{L_i^R, L_j^F\}$ . I iterate over steps 2 and 3 until convergence.

The map of optimal road improvements is presented in Figure 10, and corresponding welfare gains and other outcomes of interest in Table 11. I estimate that the NPDV of the net welfare gains in this benchmark is about 344 million USD under a 3% annual discount rate. This corresponds to 397 USD per Kampala adult (Table 11, row 1).

This number is four times larger than the net returns from the realized road improvements, despite the realized upgrades being more prevalent around links with high net welfare gains from a one lane upgrade, as discussed in Section 5.3.2. The discrepancy between the two numbers has two main explanations. First, when solving for optimal road improvements  $\{R_{kl}\}$ , I allow for a continuous solution in terms of additional width, rather than a discrete solution in terms of additional lanes. This simplification, motivated by computational considerations as solving for the optimal combination of additional (discrete) lanes is a high dimensionality discrete-choice combinatorial problem (Arkolakis et al. 2023), de-facto ignores this discreteness constraint faced by the government. While common (Fajgelbaum and Schaal 2020, Bordeu 2024), it may drive part of the observed wedge. Second, the government may face other constraints that are beyond the scope of this paper, but which may include corruption (Olken 2007), procurement constraints (Wolfram et al. 2024) and, more generally political economy considerations (Brueckner and Selod 2006, Glaeser and Ponzetto 2018, Bordeu 2024, Fajgelbaum et al. 2024).

## 6 Does the Existing Land Acquisition Framework Lead to Misallocation?

In this final section, I leverage the model to understand how the specific structure of land payments are affecting the net welfare gains of road improvements in Kampala. The structure of this section mirrors the mechanisms from Section 4.2. In a first part, to study the role of the tax revenue wedge, I compare the welfare gains of the optimal road improvements under the status quo voluntary land take approach (Section 5.4) to a counterfactual where land is acquired at market value. I then investigate a potential solution to this high tax revenue wedge: removing restrictions on the use of external funds. In a second part, I study the allocative impact of property rights heterogeneity in Kampala, by comparing the status quo to a counterfactual with homogeneous property rights.

## 6.1 Welfare Impacts of a High tax revenue wedge for Optimal Road Improvements in Kampala

### 6.1.1 Optimal Road Improvements Under Alternative Compensation Rules

**Market Value Compensation:** Under both the Ugandan constitution and the World-Bank guidelines, owners should receive a fair compensation against land take.<sup>27</sup> I solve for the corresponding counterfactual, where owners are compensated at market value. The locations of optimal upgrades under market value compensation, compared to the optimal upgrades under voluntary land take, are displayed on Figure 11. Under market value acquisition, less roads are upgraded in the city center (orange) and instead upgrades are displaced towards a ring around the city center, to relatively lower land cost locations (blue). Net welfare gains of the policy are equivalent to a 115 USD one-time transfer per resident (Table 10, row 2), or 29% of the net welfare gains under the status quo voluntary land take approach. Under both comparisons, less roads are upgraded without the restriction on fund use than with it, but net welfare gains are larger. This result is driven by the high fiscal cost, as the marginal dollar is best used to fund land payments, replacing levying property taxes subject to a tax revenue wedge. Importantly, these differences across payment rules would be lower if landowners' utility was concave in income, as opposed to linear. In this case, compensating owners at market value would yield not only additional fiscal costs, but also additional owner welfare gains, somewhat mitigating decrease in overall welfare gains.

**No financial compensation:** Consistent with this result, a counterfactual where owners would not be compensated for their land (Table 10, row 3) would yield net welfare gains equivalent to a 568 USD per resident, or 43% more than under the existing voluntary land take rule. The difference between the two optimal allocations is displayed on Figure 12. This result is driven both by the optimal upgrades leading to a lower decrease in commuting time ( $-13.9\%$  against  $-13.2\%$ ) and a larger increase in property values ( $+3.2\%$  against  $+2.9\%$ ) compared to the voluntary land take optimal upgrades, and by the absence of a fiscal cost.

**Decomposition of the difference between tax revenue wedge and improved allocation:** When imposing an analogous fiscal burden as under the voluntary land take optimal upgrades (Table 10, row 4), the net welfare gains drop to 419 USD per resident, implying that 13% of the gains come from the improved allocation, and the remaining 87% from the tax revenue wedge.

### 6.1.2 Optimal Road Improvements Relaxing Restrictions on the Use of External Funds

One potential avenue to remove distortion imposed by the fiscal cost under market value payments would be to relax the existing restrictions on the use of external funds. Currently, the city government can only use external funds (from the World Bank and the African Development Bank) for road construction but not for land acquisition. This status quo restriction is likely driven by concerns of corruption (World Bank 2011). However, in the presence of high fiscal leakage on property value taxes used to fund land acquisition, it may significantly hinder the net welfare gains of road improvements.

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<sup>27</sup>The question of whether market value at baseline is a fair compensation rate is beyond the scope of this paper.

In practice, I calculate the optimal policy when the government faces a consolidated budget constraint given by

$$\sum_k \underbrace{p_k^1 \times [R_{kl} - R_{kl}^0]}_{\text{land payment expenditures}} + \bar{p} \cdot \underbrace{\sum_{kl} [R_{kl} - R_{kl}^0]}_{\text{construction expenditures}} \leq (1 - \eta) \phi \cdot \sum_i \underbrace{(H_i - H_k^r(R_{kl}))}_{H_i^{\text{res}}} \cdot q_i + \bar{M}$$

I find that under this relaxed constraint and voluntary land take approach, the net welfare gains are equivalent to a 425 USD transfer pre resident, or 17% more than under the existing restrictions (Table 10, row 5). The difference in optimal improvements between these two approaches is displayed on Figure 13. Less roads are upgraded everywhere except at the city center, as some of the funds are instead used for land acquisition on these same roads.

Furthermore, relaxing this restriction may help to reach other World Bank’s stated objectives, like ensuring that displaced owners are “*provided prompt and effective compensation [...] for losses of assets attributable directly to the project.*” (World Bank 4.12 Involuntary Resettlement). Removing the fund use restrictions leads to net welfare gains of the policy equivalent to a 261 USD transfer per resident, or 126% more than under the existing restrictions (Table 10, row 6). This increase in welfare, larger than under the voluntary land take approach, comes from the high calibrated fiscal cost.

While corruption costs are outside the scope of this model, these numbers can be used to benchmark whether the anticorruption benefits outweigh the potential benefits of alternate payment and funding rules.

## 6.2 Welfare Impacts of a Heterogeneous Property Right Regimes in Kampala

While the resort to voluntary land take leads to higher net welfare gains than under the eminent domain market value compensation rule because of the large tax revenue wedge, this approach also ties land costs and optimal road improvements to the distribution of property right regimes in the city.

Under voluntary land take, a lower share of owners gets compensated on freehold and mailo land, where property rights are less clear and land disputes are widespread.<sup>28</sup> To assess the welfare impact of this heterogeneity, on Figure 14, I compare the optimal road improvements under the existing property right regimes (Figure 10), to the optimal road improvements if all land was owned under the freehold property right regime, the weakest in the city. Under these hypothetical uniform weak property rights, upgrades would be reallocated towards the central leasehold locations and commuting times would decrease by 0.2 percentage point more than under the existing property right regimes. These higher commuting time benefits, together with the lower fiscal burden from lower total payments, would lead to net welfare gains equivalent to a 461 USD transfer, or 16% more than under the existing rights (Table 10, row 7a). To isolate the allocative effect coming from higher benefits, I then enforce a total fiscal cost equal to the one under the existing rights (Table 10, row 7b). Increasing the fiscal burden up to the status quo, I find that

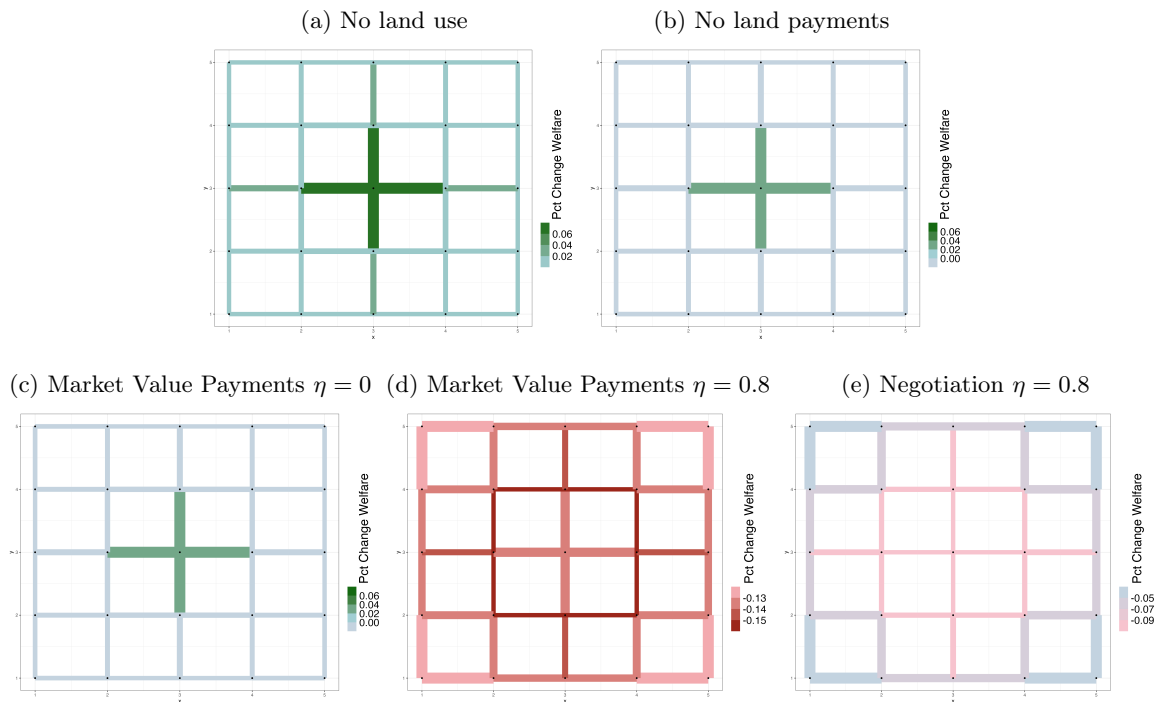
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<sup>28</sup>These issues are well known by the Uganda government, who started to digitize the national land registry to decrease the number of land fraud in an effort to decrease transaction costs and frictions on the land market.



the net welfare gains remain equivalent to a 438 USD, transfer, meaning that 64% of the gains come from a better allocation, and 36% from a lower fiscal burden.

Figure 5: Net Welfare Gains of Road Improvements in a Simple Economy



Notes: These figures display the city-wide welfare effects of improving each link, solving for equilibrium location choices, routing, and accounting for the fiscal cost of funds. In panels (b) to (e), I account for the opportunity cost of land use by assuming that additional roads compete with residential land.

Figure 6: Kampala at Baseline

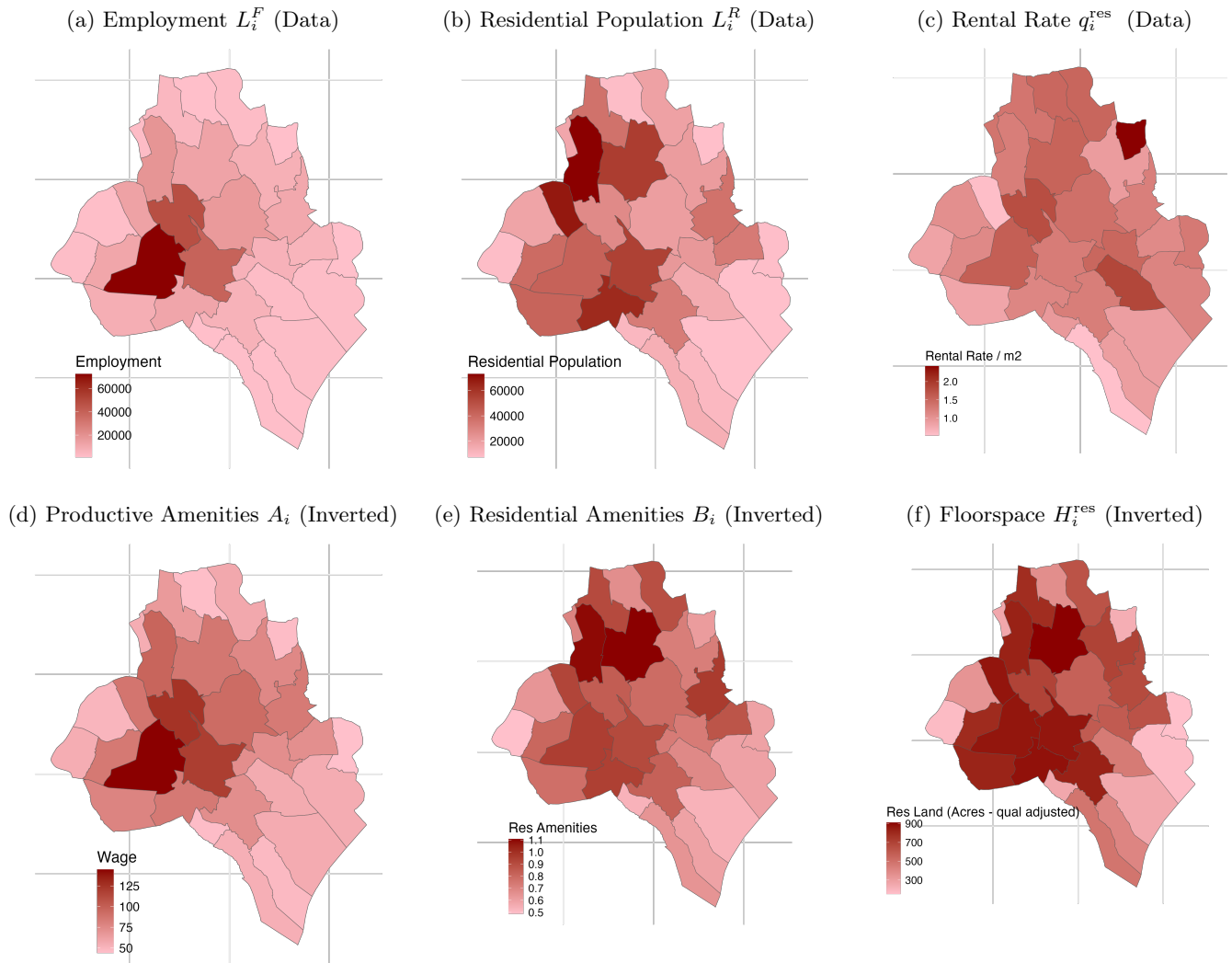
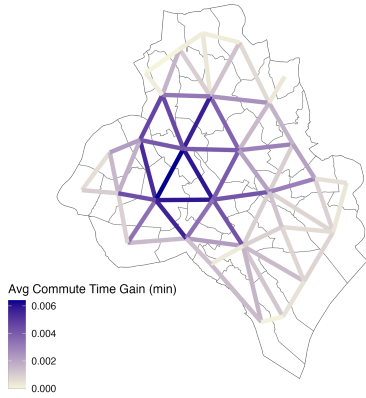
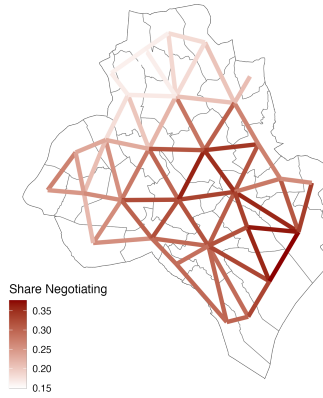


Figure 8: Link-Level Impacts of Road Improvements

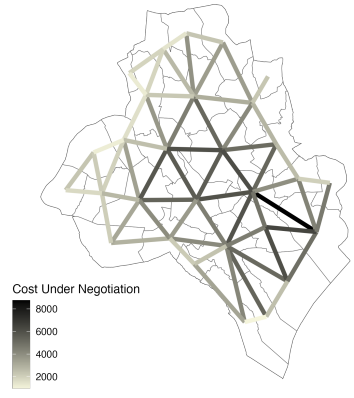
(a) Change in Avg Commuting Times (minutes)



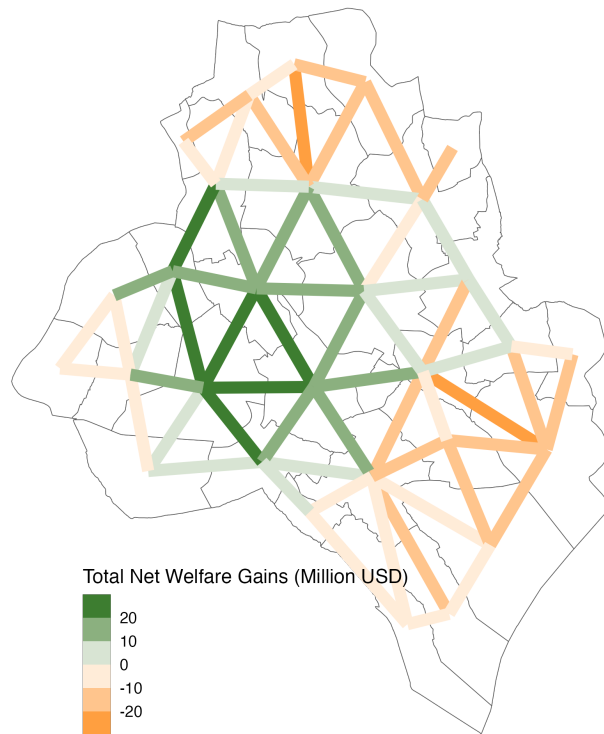
(b) Predicted Share of Owners Negotiating



(c) Land Costs

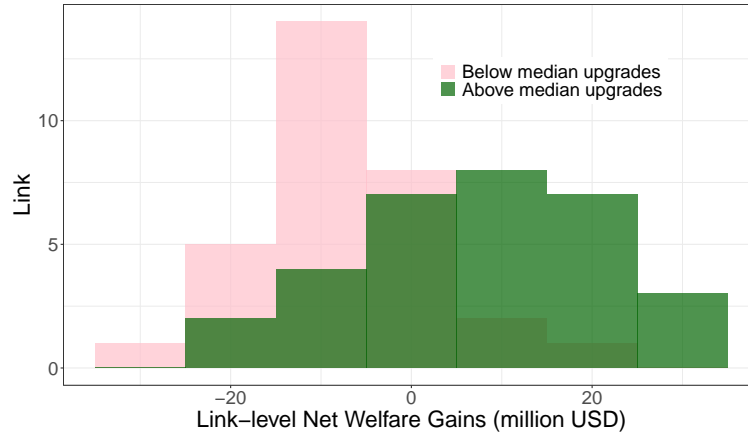


(d) NPDV Net Welfare Gain



Notes: These figures display the impact of adding one lane on each link on city-level average commuting time (panel a), predicted share of owners negotiating (panel b), land costs under the existing voluntary land take approach (panel c) and net welfare gains (panel d, NPDV 3%). The color of each link corresponds to this impact of this link's upgrade of the city-level outcome of interest.

Figure 9: Net Welfare Gains of Link-Level Upgrades and Realized Upgrades



Notes: I display the relationship between link-level net welfare gains as predicted by the model, and the length of realized road upgrades along that link. I split links according to whether the total length of road upgrades along that link is above median (green) or below median (pink)

Figure 10: Optimal Road Improvements Under “Voluntary Land Take”

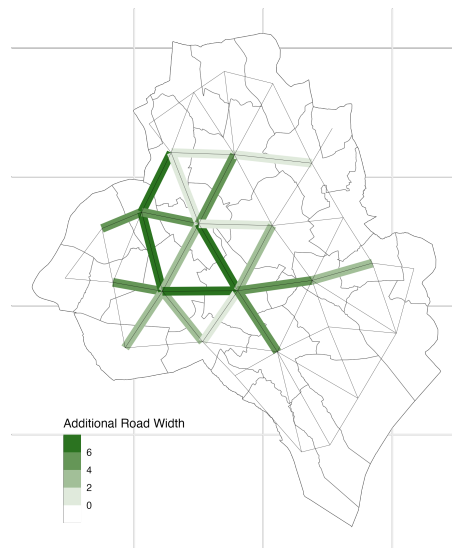
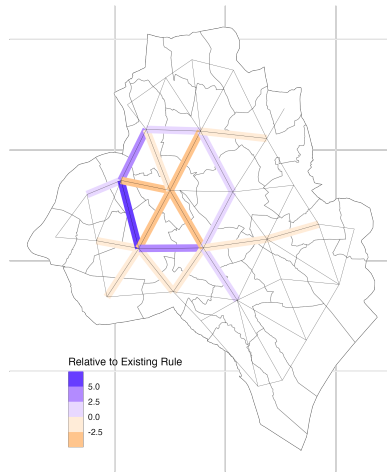
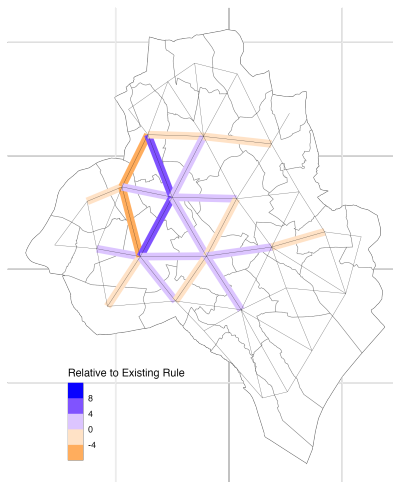


Figure 11: Payment at Market Value



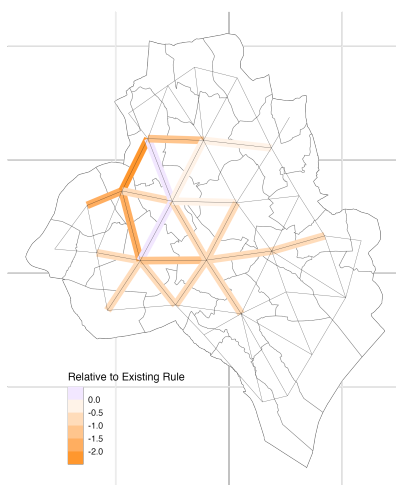
Notes: I display the difference in optimal road improvements between the existing rule and if land was acquired at market value. Orange links correspond to less improvements under the latter counterfactual than under the status quo rule. Purple links correspond to more improvements under the latter counterfactual than under the status quo rule.

Figure 12: No Land Payment



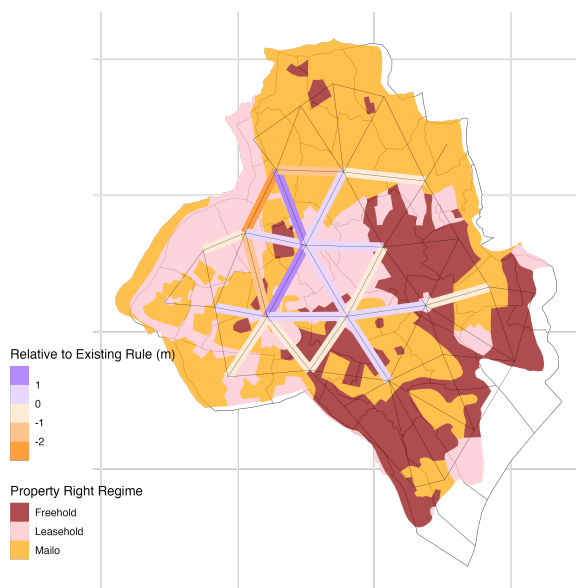
Notes: I display the difference in optimal road improvements between the existing rule and if land was acquired without compensation. Orange links correspond to less improvements under the latter counterfactual than under the status quo rule. Purple links correspond to more improvements under the latter counterfactual than under the status quo rule.

Figure 13: Voluntary Land Take & No Restriction of the Use of External Funds



Notes: I display the difference in optimal road improvements between the existing rule and if land was acquired under voluntary land take but in the absence of restrictions on the use of external funds. Orange links correspond to less improvements under the latter counterfactual than under the status quo rule. Purple links correspond to more improvements under the latter counterfactual than under the status quo rule.

Figure 14: All Freehold Property Rights



Notes: I display the difference in optimal road improvements between the existing rule and if land was acquired under voluntary land take but all land was freehold. Orange links correspond to less improvements under the latter counterfactual than under the status quo rule. Purple links correspond to more improvements under the latter counterfactual than under the status quo rule. I overlap the property rights map.

Table 10: Welfare Impacts of Realized Road Improvements Under Alternative Model Parameters

	Parameters			Outcomes		USD per resident-equivalent transfer (NPDV 3%)	USD net welfare gains (NPDV 3%)	USD net welfare gains (NPDV 6%)
	$\eta^\phi$	$\kappa$	$\xi$	avg $\Delta$ commuting times (%)	$\Delta$ property values (%)			
Main	0.61	0.01	0.39	-7.1	+1.3	+99	+85,640,225	+1,327,000
Robustness	0.3	-	-	-7.1	+1.3	+139	+119,973,517	+19,520,809
	-	0.015	-	-14.9	-6.6	+341	+294,983,996	+107,026,048
	-	-	0.29	-8.4	-2.6	+5	+4,284,928	-3,832,349

Table 11: Optimal Road Improvements

Counterfactual					Outcomes			Net Welfare Gain (equivalent transfer)		
	Land acquisition	Property Right	Land	Restriction	total area	$\Delta$ commuting	$\Delta$ property	USD / resident	USD total	USD total
	Rule	Regimes	Use	Ext. Funds	upgraded (m <sup>2</sup> )	time (%)	values (%)	(NPDV 3%)	(NPDV 3%)	(NPDV 6%)
1	“voluntary land take” (VLT)	Existing	Y	Y	452,765	-13.2%	+2.9%	+397	343,767,916	131,418,009
2	payment at market value	-	Y	Y	441,168	-12.4%	+2.7%	+115	99,865,592	9,127,330
3	no payment	Existing	Y	Y	452,765	-13.9%	+3.2%	+568	491,780,377	205,424,239
4	equally split land payments	-	Y	Y	452,765	-13.9%	+3.2%	+419	363,283,998	141,176,050
5	“voluntary land take”	Existing	Y	N	359,771	-11.2%	+2.5%	+425	367,852,167	151,771,522
6	payment at market value	-	Y	N	252,032	-8.3%	+1.9%	+261	225,827,807	90,388,550
7a	VLT	All freehold	Y	Y	452,765	-13.4%	+3.0%	+461	399,570,320	159,319,210
7b	VLT & fixing total payments at existing	All freehold	Y	Y	452,765	-13.4%	+3.0%	+438	379,676,761	149,372,431



## 7 Conclusion

Improving road infrastructure in developing cities is an investment priority, but little is known about the net returns to these investments in practice, neither about their levels nor their drivers. In this paper, I fill this gap by collecting new data and studying the returns to road improvements in Kampala. I show that the benefits of road improvements are large but so are the costs. These costs hinder the net returns of road improvements in two ways. First, for a given set of improved roads, land costs decrease net returns, as there is a high cost for domestic governments to compensate owners. Compensating owners at market value - mandated under eminent domain - would yield less than 50% of the realized welfare gains. The realized welfare gains are higher because not all owners get compensated, especially if they have weak property rights. In the context of public good provisions, clear property rights can have large negative welfare gains, by preventing budget constrained governments from seizing land that could be used more efficiently for the public good (Acemoglu and Robinson 2012), because there are positive externalities of road improvements on the whole city that owners do not internalize. In counterfactual scenarios, I further explore how land acquisition under more uniform property right regimes could improve welfare outcomes by shifting investments towards higher benefit locations.

Ultimately, my study highlights the need to account for existing property rights and land acquisition conditions when designing infrastructure investments in rapidly growing cities. As international donors weigh in to enforce that owners are compensated at market value are ‘*provided prompt and effective compensation [...] for losses of assets attributable directly to the project.*’ (World Bank 4.12 Involuntary Resettlement), adjustments of the restrictions on the use of external funds for land acquisitions should be considered. More generally, land acquisition in Kampala is a textbook example of the theory of the second best where correcting a market imperfection (e.g. enforcing compensation at market value to prevent abusive expropriation) does not necessarily improve overall welfare in the presence of other market imperfections (e.g. cost of levying domestic funds for compensation).

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## Appendix

### A Context: Property Right Regimes in Kampala

The different property right regimes (or *land tenure systems*) are inherited from the 1900 Buganda Agreement between the British Protectorate and the Buganda Kingdom, which split used land into Mailo and Crown land. On the one hand, crown land, owned by the British protectorate, was leased to private owners for 49 or 99 year leases and are still managed by local land boards. These leases can be renewed. On the other hand, Mailo land (square mile plots) was owned by the Kabaka, head of the Kingdom or by local chiefs and notables. Over time, legal protections were put in place to protect the peasant tenants occupying this land (*kibanja*) from being evicted by the landowners, giving them ownership over developments on this land and creating de-facto double ownership claims over mailo land. These land titles and occupancy certificates are managed by the Buganda Land Board, and issues of multiple land claimants per plot are common. In addition, previously unoccupied land was titled over the course of the 20<sup>th</sup> century and became freehold land. A small number of leases were also replaced by permanent titles. While freehold land titles are the least restrictive (single ownership and unlimited term), the clarity of the owners’ property rights is conditional on the titles being correctly accounted for, which is seldom the case.<sup>29</sup> [Bird and Venables \(2020\)](#) compare Mailo land to all other property right regimes in Kampala, using information from the Census to recover the share of Mailo land in each parish. They find that the larger the share of Mailo land in a parish, the higher its population density, the larger the prevalence of informal housing and the lower economic activity.

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<sup>29</sup>The Uganda government started to digitize the national land registry to decrease the number of land fraud cases in 2013, but, as of 2023, this process had not been finalized yet.

## B Broker survey: Comparison with Scraped Online Property Ads

To increase confidence in the reliability of the broker survey data, I compare it with data scrapped online from [RealEstateDatabase.net](https://www.realestate.com.ug/), which includes posts of properties to rent or sale in Kampala and other parts of Uganda. This online database includes some geographical information (at the *neighborhood* level, which is less precise than parish and very noisy), as well as property characteristics. In addition, the date at which the property is posted can be recovered from the file name of the photos accompanying the post (but not the transaction date).

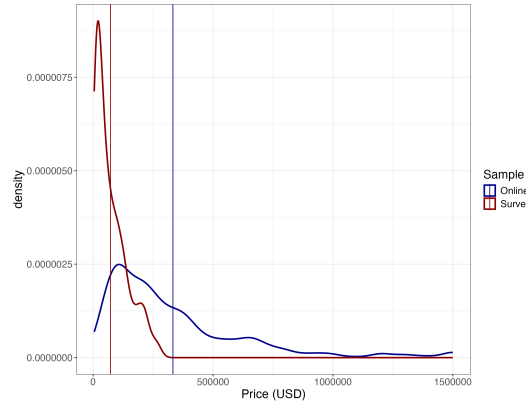
On Figure A1, I plot the distributions of property-level prices for sold residential properties in my broker survey data (red) and the online data (blue). The distribution of online property prices is skewed to the right, with an average price list at 332,694 USD (against 70,261 USD in the survey data). Given that the average monthly net salary in Kampala is \$188 USD (KCCA 2019 Statistical Abstract) and 18.4% of Kampala households own a laptop (UBOS 2014), property posted online are upscale properties only, not within reach of the average household, and not representative.

Reassuringly, however, while price levels are different across the databases, they are positively correlated over space. To check this, I run the following regression

$$\ln p_{pt}^b = \alpha + \beta \times \ln p_{pt}^o + \gamma_t + \gamma_p + e_{pt}$$

where  $p_{pt}^b$  is the average property price in parish  $p$ , year  $t$  in the broker data, and analogously for the online database  $\ln p_{pt}^o$ .  $\gamma_t$  and  $\gamma_p$  are year and parish fixed effects, respectively. The results are displayed on Table A1 are confirm the robust positive correlation between the online and broker survey databases at the parish and parish-year levels.

Figure A1: Comparison of Property Prices in the Broker Survey and the Online Database



Notes: This figure displays property-level log sale prices for residential properties in the real estate broker survey and the online database. The vertical lines correspond to the average prices in each database

Table A1: Comparison of Online Data and Broker Data

	<i>Dependent variable:</i>				
	Log Parish-Level Mean Broker Survey Price (USD)				
	(1)	(2)	(3)	(4)	(5)
Log Parish-Level Mean Online Price (USD)	0.396*** (0.099)	0.400*** (0.100)	0.497** (0.194)	0.494** (0.198)	0.272*** (0.080)
Year FE		Y		Y	
Parish FE			Y	Y	
Level of Observation	Parish-Year	Parish-Year	Parish-Year	Parish-Year	Parish
Observations	181	181	181	181	309
R <sup>2</sup>	0.083	0.108	0.644	0.659	0.036
Adjusted R <sup>2</sup>	0.078	0.077	0.433	0.431	0.033

Notes: standard errors are displayed in parentheses, s.t. \*p< 0.1; \*\*p< 0.05; \*\*\*p< 0.01. Price variables are trimmed at the top and bottom 1%. The sample is restricted to residential properties sold out (or posted) in 2018 onwards.

## C Owner survey

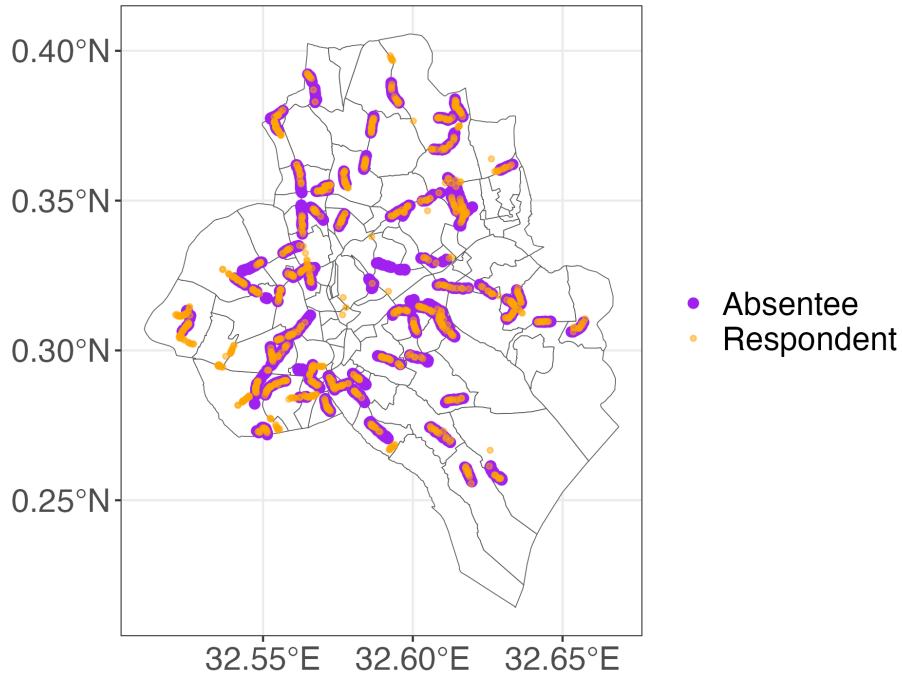
### C.1 Sampling Strategy

The sampling frame for the owner survey was at the road segment level. More precisely, roads within Kampala were split into categories based on their policy status: road targeted by the policy with completed upgrade (pre-2020 or 2020-2024), road targeted by the policy but upgrade ongoing (2024 onwards), potential road candidate for upgrade (either mentioned by policy documents but not upgraded, or unpaved roads). To ensure coverage of the different land tenure systems in the city, the main tenure system within a 50 meters buffer of each road was identified. Then, for each triplet {division, land tenure, road category}, road segments were randomly ordered and the first 97 valid segments were selected.<sup>30</sup>

Enumerators conducted a census of properties over a pre-specified road segment of approximately  $\frac{1}{2}$  to 1 mile, between two intersections, starting from a randomly selected point over the road segment. In addition to the property census, enumerators were instructed to interview all available potential respondents, moving from the (randomly selected) starting point to the end of the road segment. Figure A2 shows a map of the survey areas, displaying both interviewed respondents (in orange) and listed plots (in purple).

<sup>30</sup>Segments within approximately 1km from a previously selected segment were excluded, as well as segments where data collection was impossible (based on on-the-ground observations) or where road characteristics were different than in the database (for example a large paved road vs an unpaved road).

Figure A2: Spatial Coverage of the Owner Survey



Notes: This map displays the approximate location of respondents and potential respondents (but not interviewed) of the owner survey. Roads covered include roads targeted by the policy with completed upgrade (pre-2020 or 2020-2024), roads targeted by the policy but upgrade ongoing (2024 onwards) and potential roads candidate for upgrade (either mentioned by policy documents but not upgraded, or unpaved roads). Respondents on the latter roads are not used in the main analysis because their property was not affected by road upgrades.

Eligible respondents were property (land or building) owners above 18 years old, their spouse or an individual with knowledge and decision power over land and property related issues. The targeted property must be on the side of one of the targeted roads. If the respondent was not available at that time, but willing to participate, the survey team offered to make an appointment and come back at a later time.

## C.2 Overview of Survey Instruments

1. Eligibility and consent
2. Road quality and road upgrade
  - Name and characteristics of the closest road and road upgrade characteristics.
  - Road quality on a scale of 1 to 10, today and before the road upgrade (conditional on upgrade).
3. Land acquisition (conditional on land or property affected by the road upgrade)<sup>31</sup>
  - What was affected: how much land was affected, whether developments in the road reserve were affected.

<sup>31</sup>In addition, if respondents' road was not part of the policy, or if their property had not been affected by the road upgrade, respondents were asked analogous questions about their decision and reasoning, in the hypothetical case their property had been affected by a road upgrade.



- Compensation, offer and negotiations: timeline and accept / negotiate / court and outcomes, reasons.
- Perceptions: pivotality, potential returns from negotiating, affected neighborhood
- Interactions: with neighbors, family, local leaders
- Knowledge of the legal system
- Suggestions

#### 4. Property value and property characteristics

- Self-assessed property value if sold and rented today, as well as either before the road upgrade and today if the road had not been upgraded.
- Ownership status: timeline and characteristics of the acquisition and ownership, ownership documents.
- Property characteristics, including recent improvements.

#### 5. Household and neighborhood characteristics

- religion, ethnicity, local leader popularity
- basic socio-economic characteristics on the model of the Uganda National Panel Survey

## D Additional Analyses

### D.1 Estimating Speed by Road Category

To recover the average free-flow speed on each type of roads as per the Open Street Map (OSM) classification  $\in \{\text{highway, primary, secondary, tertiary and other}\}$ , I leverage information from my Google Maps trips. From the Direction API, I recover each trip’s approximate path from a list of coordinates roughly corresponding to turns and intersections. I draw a buffer of 50 meters around the straight lines between these points to account for measurement error, and map them to the OSM underlying road network. I compute the share of the total length of OSM intersected roads belonging to each road type,  $c^{RT}$ , and regress the trip’s average speed  $s_{h,d,od}$  on road categories and trip controls:

$$s_{h,d,od} = \alpha + \sum_{RT} \beta^{RT} \times c_{h,d,od}^{RT} + \mathbf{X}_{od}' \beta^{od} + \gamma_h + \gamma_d + \epsilon_{h,d,od}.$$

$\mathbf{X}_{od}$  is a vector of origin and destination characteristics, including longitude, latitude, longitude<sup>2</sup>, latitude<sup>2</sup> at origin and destination and  $\gamma_h$  is an hour fixed effect. The sample of trips is restricted to grid-to-grid trips, to ensure a detailed coverage of Kampala as a whole.  $\alpha + \beta^{RT}$  can be interpreted as the average speed on a trip where all the trip happens on a road of type  $RT$  (the share of the trip on a road of type  $RT$  is equal to 1). Standard errors are clustered at the day level to reflect the query sampling strategy. The results are presented in Table A2. In columns 1 to 4,  $s_{h,d,od}$  is defined as the speed at time of query and the sample is restricted to non-rush hour speed to capture free-flow speed. In columns 5 to 8, speed is defined as the average speed reported by Google Maps on each query for that trip. The speed patterns are robust across specifications: compared to a trip exclusively on primary roads (the omitted category), the larger the share of the trip on motorway, the faster the trip. On the other hand, the larger the share of the trip on secondary, tertiary or other roads, the slower. Given the stability of the coefficients across specifications, I pick the third specification to map to OSM, as it is the most comprehensive in terms of road characteristics, but does not include geographical characteristics and I am assuming that local speed is constant over space within a road category.

### D.2 Evidence of Traffic Reallocation Towards Upgraded Roads

A road being improved may also affect the number of commuters using this road, coming from both (i) a change in path for existing trips in both the short and long term and (ii) new trips. The data allows me to study (i) by conducting the following thought experiment: take a trip between the centroids of two parishes  $o$  and  $d$ . Over time, as a given road  $r$  gets improved (late wave), does the probability the trip takes road  $r$  increase?

To answer this question, I leverage my Google Maps data between all pairs of parish centroids in Kampala. Over the period covered by the data (March 2023 - September 2024), upgrades of some roads in the late policy wave were finalized, allowing me to study whether trips queried later were more likely to take roads part of the late policy wave, as these roads were increasingly improved over time. Google Maps API itinerary is a series of coordinates, corresponding to turns and intersections. I construct straight lines between these points, calculate a buffer of 50 meters around these straight lines to account for measurement

Table A2: Speed by Road Category

	<i>Dependent variable:</i>							
	Speed at time of Query (km/h)				Average Speed (km/h)			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Distance (km)		0.187*** (0.006)	0.177*** (0.004)	0.170*** (0.005)		0.192*** (0.004)	0.188*** (0.004)	0.178*** (0.004)
Paved (share)			8.843*** (0.457)	9.026*** (0.388)			7.853*** (0.086)	8.128*** (0.096)
Motorway (share)	10.890*** (0.812)	11.790*** (0.713)	7.501*** (0.590)	7.573*** (0.538)	11.360*** (0.444)	12.510*** (0.548)	8.620*** (0.435)	8.661*** (0.462)
Secondary (share)	-4.218*** (0.196)	-1.976*** (0.156)	-3.766*** (0.138)	-3.572*** (0.125)	-4.559*** (0.113)	-2.155*** (0.144)	-3.968*** (0.102)	-3.726*** (0.112)
Tertiary (share)	-4.210*** (0.203)	-1.797*** (0.212)	-2.931*** (0.154)	-2.667*** (0.212)	-4.161*** (0.172)	-1.748*** (0.145)	-2.653*** (0.135)	-2.577*** (0.137)
Other (share)	-7.198*** (0.360)	-4.961*** (0.354)	-1.094*** (0.233)	-1.162*** (0.233)	-6.378*** (0.129)	-3.983*** (0.090)	-0.445*** (0.080)	-0.547*** (0.080)
Sample	Grid	Grid	Grid	Grid	Grid	Grid	Grid	Grid
Ref Road Cat: Primary								
Flexible Geo Controls		Y		Y		Y		Y
Time	Non-rush	Non-rush	Non-rush	Non-rush	All	All	All	All
Hour FE	Y	Y	Y	Y	Y	Y	Y	Y
Dep Var Mean (km/h)	20.2	20.2	20.2	20.2	19.1	19.1	19.1	19.1
Observations	36,267	36,267	36,267	36,267	58,783	58,783	58,783	58,783
R <sup>2</sup>	0.142	0.269	0.380	0.381	0.087	0.245	0.339	0.344

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Notes: Standard errors are clustered at the day of query level and displayed in parentheses with significance levels \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ .

error. I recover the corresponding roads by intersecting (buffered) straight lines between these points and the OSM underlying road network. I run the following regression:

$$\frac{\text{dist policy roads late wave}}{\text{dist total}}_{h,d,ij} \times 100 = \beta \times t_d + \gamma_h + \gamma_{ij} + e_{h,d,ij}$$

where  $\frac{\text{dist policy roads late wave}}{\text{dist total}}_{h,d,ij} \times 100$  is the share of trip between  $i$  and  $j$  on day  $d$ , hour  $h$  that is mapped to roads in the late policy waves,  $t_d$  corresponds to the month of query, while  $\gamma_h$  and  $\gamma_{ij}$  are hour and origin-destination fixed effects, respectively. Standard errors are clustered at the day of query level.<sup>32</sup>

Regression results are presented in Table A3. In columns 1 to 4, I show that over time, the share of trips happening on roads upgraded in the late wave of the policy increases, as more and more of these roads get upgraded. This finding is robust to including origin-destination fixed effects (column 4), so that the identifying variation comes from changes in the path of the same  $o - d$  trip over time. I reproduce the analysis but change the dependent variable to be the share of trips happening on roads upgraded in the early wave of the policy. These roads' upgrades were already completed by the beginning of the period covered by my Google Maps data, so they can be used as a placebo test. In columns 5 and 6, I find that indeed, the share of trips on these roads did not increase since the beginning of the Google Maps data.

<sup>32</sup>We do not cluster at the road level as we include the universe of parish-to-parish centroids in the city of Kampala.

Table A3: More Likely to Use a Road As It Gets Upgraded

	<i>Dependent variable:</i>					
	Share of Total Trip Length on Late Policy Roads * 100			Share of Total Trip Length on Early Policy Roads * 100		
	(1)	(2)	(3)	(4)	(5)	(6)
Time (months)	0.024** (0.010)	0.022** (0.010)	0.022** (0.011)	0.013* (0.007)	-0.001 (0.006)	-0.005 (0.004)
Straight line o-d length controls	Y					
Trip length controls				Y		
Spatial FE		o and d	od pair	od pair	od pair	od pair
Observations	104,404	104,404	104,404	104,404	104,404	104,404
R <sup>2</sup>	0.247	0.563	0.912	0.914	0.877	0.878
Adjusted R <sup>2</sup>	0.247	0.562	0.904	0.906	0.865	0.867

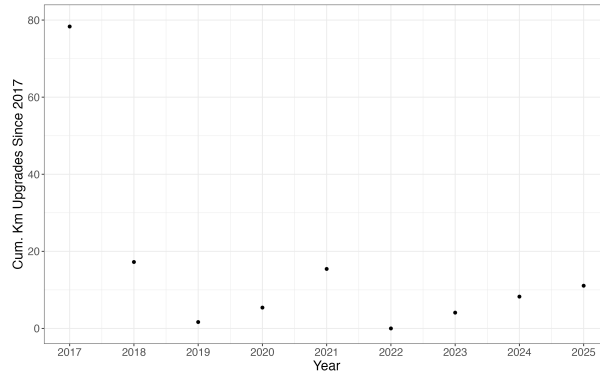
*Note:*

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Notes: Standard errors are displayed in parentheses and clustered at the day of query level, with significance levels \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ . I do not cluster at the road level as I include the universe of parish-to-parish centroids in the city of Kampala.

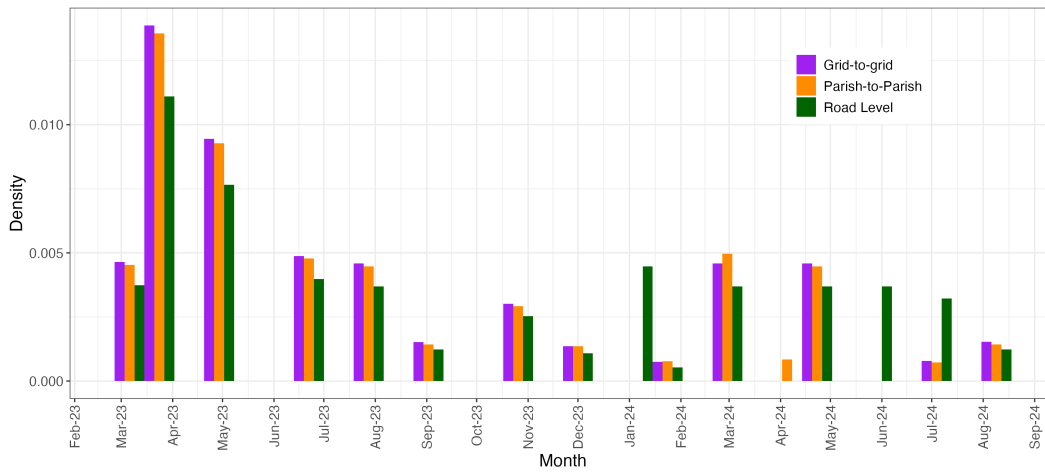
## E Additional Figures and Tables

Figure A3: Cumulative Length of Road Upgrades in Kampala since 2017



Notes: This figures displays the length of road upgrades between 2017 and 2024 in Kampala, 2017 being the first year KIIDP2 policy roads are upgraded. Source: KCCA, WB, AfDB.

Figure A4: Timing of Google Maps API Queries By Type of Queries



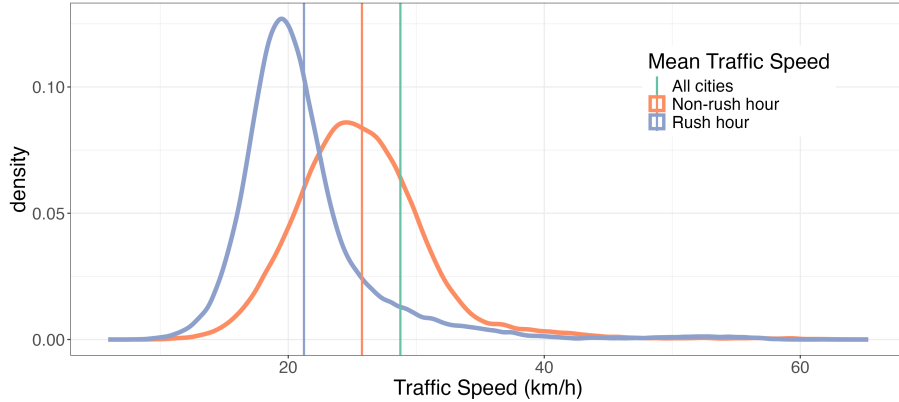
Notes: This figures plots the distribution of Google Maps Direction API queries over time by type of queries  $\in$  {between neighboring grid cells, between all pairs of parish centroids, road level}.

Table A4: Length of Roads by Category

Road Class		Length (km)	share
Motorway	1	125	0.03
Primary	3	129	0.03
Secondary	4	186	0.04
Tertiary	5	309	0.07
Other	2	3873	0.84

Source: Open Street Maps

Figure A5: Distribution of Traffic Speed



Notes: This figure plots the distribution of traffic speed from Google Maps Direction API queries for parish-to-parish (unweighted) queries. The data spans March 2023 - August 2024, as displayed on Figure A4. The average non-rush hour speed is 25.76 km/h and the average rush-hour speed is 21.23 km/h. There are 57,662 non-rush hour trips and 55,535 rush-hour trips.

Table A5: Impact of Road Improvement on Rush-Hour Traffic Speed

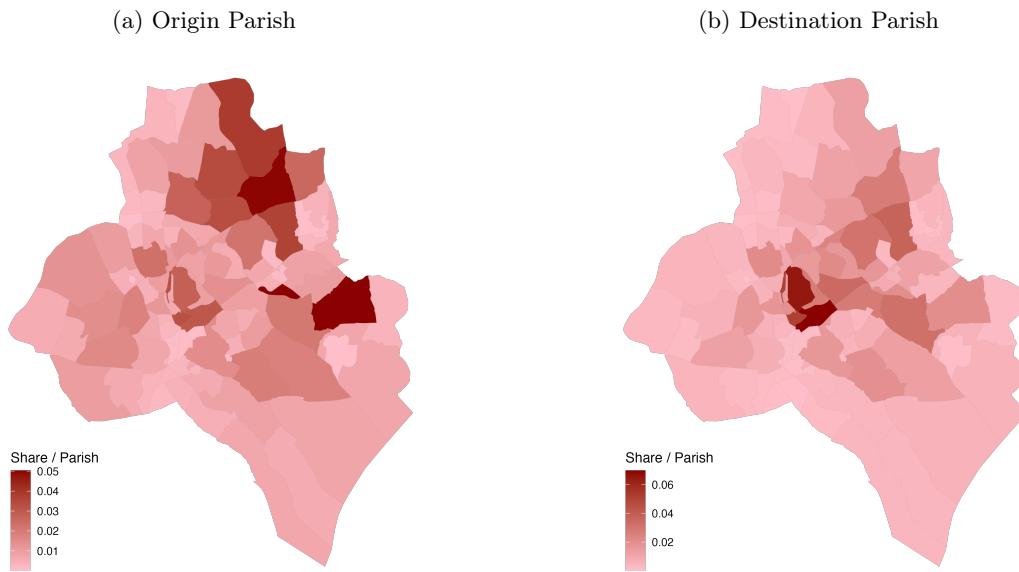
	<i>Dependent variable:</i>						
	Traffic Speed at time of query in km/h						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Upgraded (dummy)	4.804*** (1.151)	4.504*** (1.105)	4.449*** (0.859)	4.351** (1.598)	1.735** (0.884)	2.616*** (0.612)	1.957 (2.167)
Construction (dummy)	0.771 (1.368)	0.395 (1.585)	-1.319 (1.264)	-1.112 (1.267)	-1.173 (0.888)	-1.283 (0.907)	-1.523* (0.751)
Mean dep var (km/h)	22.39	22.39	22.39	22.39	22.85	20.21	20.21
Time of Day	Non-rush	Non-rush	Non-rush	Non-rush	Non-rush	Rush	Rush
Hour + Day FE	Y	Y	Y	Y	Y	Y	Y
Control for road class		Y	Y	Y	Y	Y	Y
Flexible geo controls			Y		Y	Y	
Road FE				Y			Y
Sample	Policy	Policy	Policy	Policy	Policy + Mentioned	Policy	Policy
Observations	682	682	682	682	1,663	426	426
R <sup>2</sup>	0.311	0.323	0.473	0.641	0.331	0.407	0.613
Adjusted R <sup>2</sup>	0.267	0.276	0.434	0.592	0.310	0.362	0.544

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

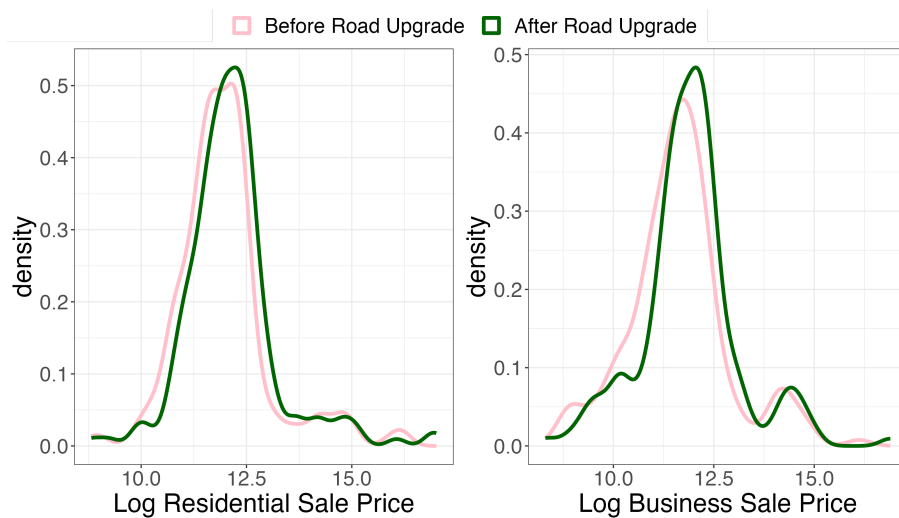
Notes: Standard errors are displayed in parentheses and clustered at the road and day levels, with significance levels \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ . Observations are at the road level (trips covering a single road) from Google Maps API queried between 2023/03/17 and 2024/06/30. Rush hour is defined as 6 to 9am and 4 to 7pm.

Figure A6: Ride-Hailing Trips: Origin and Destination of Morning Rush Hour Trips



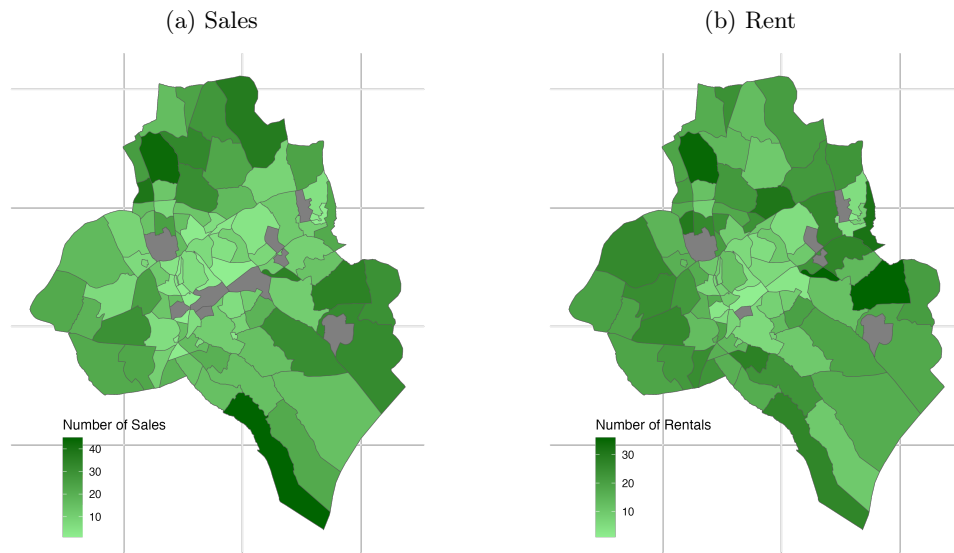
Notes: These figures show the share of morning rush-hour trips by parish of origin (panel a) and destination (panel b) from the ride-hailing data.

Figure A7: Broker Appraisal: Impact of Road Upgrades on Local Property Values



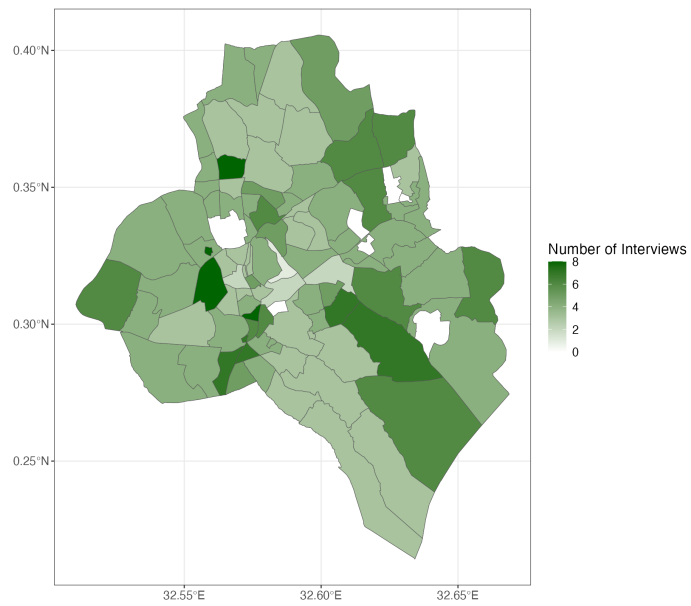
Notes: I plot the distribution of log sales price for residential (left) and business (right) properties, as assessed by brokers during the appraisal of a standardized property if the road was/wasn't upgraded. Source: own survey.

Figure A8: Number of Transactions By Parish



Notes: These maps display the number of transactions by parish from the broker survey.

Figure A9: Number of Real Estate Broker Interviews By Parish



Notes: This map displays the number of real estate brokers interviewed by parish.



Figure A10: Change in Road Quality and Local Amenities Following Road Upgrades

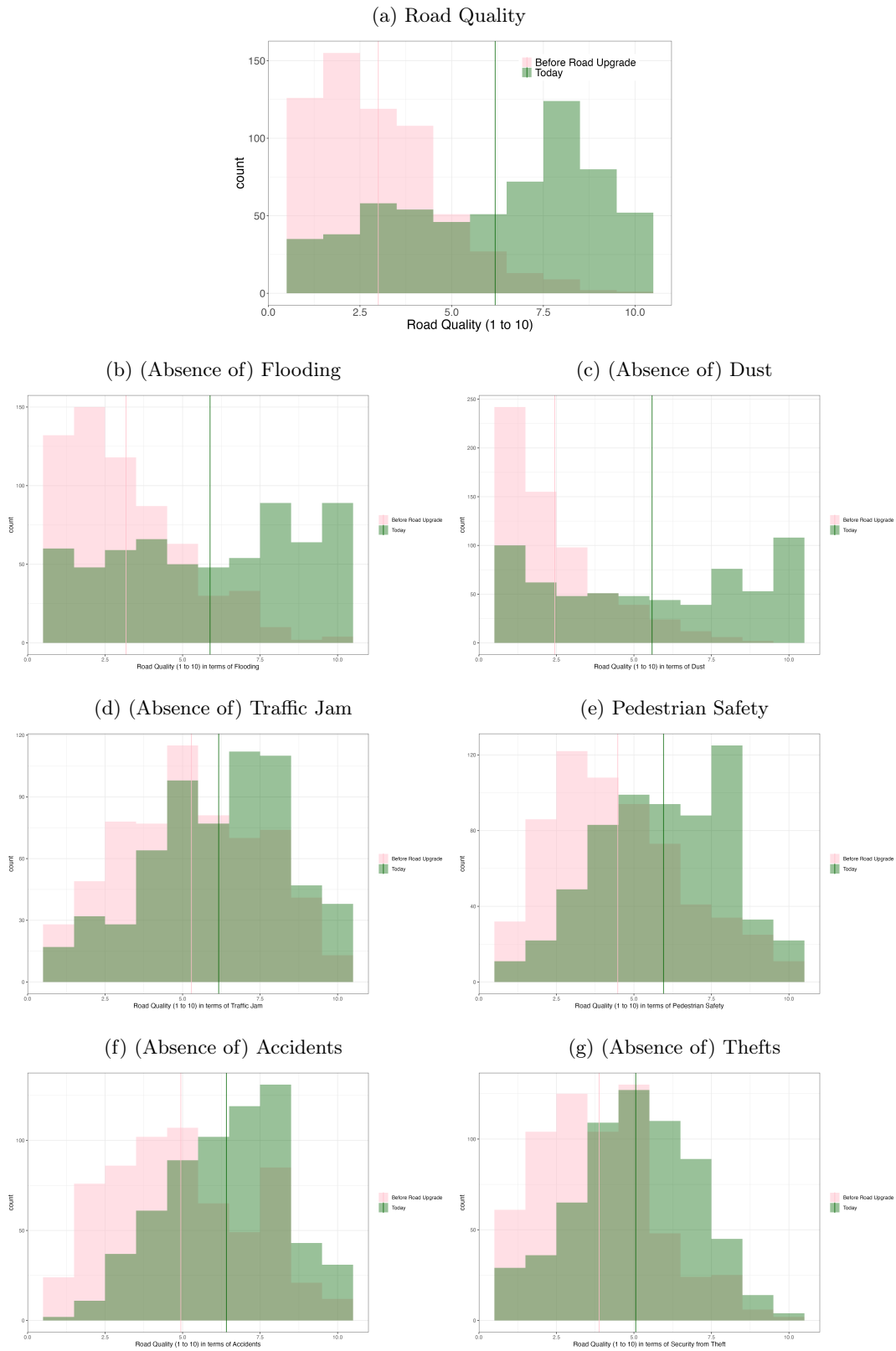


Figure A11: Owners' Stated Reasons to Accept Not to Get Compensated

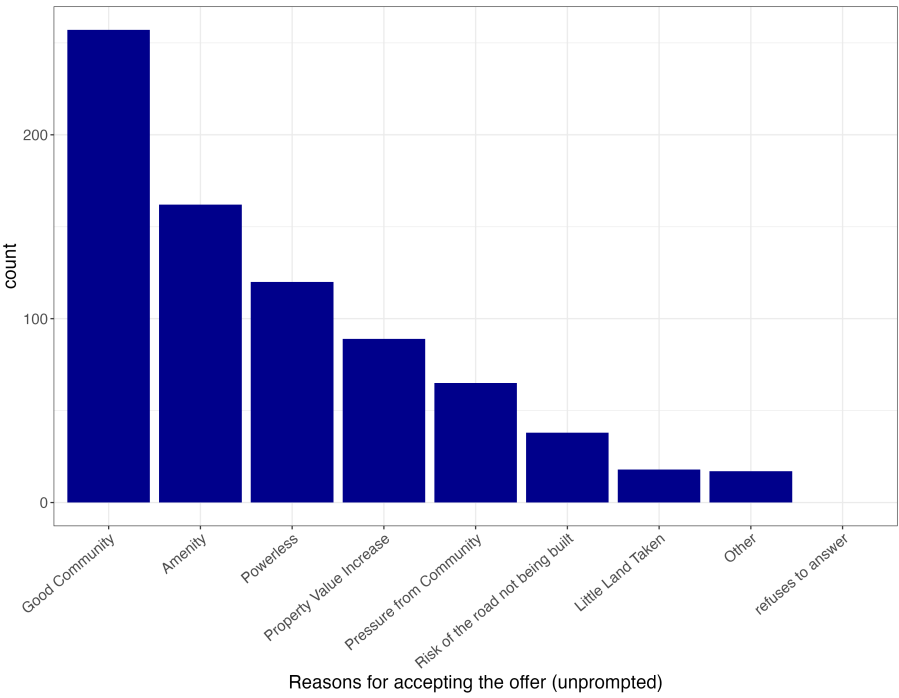


Figure A12: Owners' Stated Reasons to Negotiate to Get Compensated

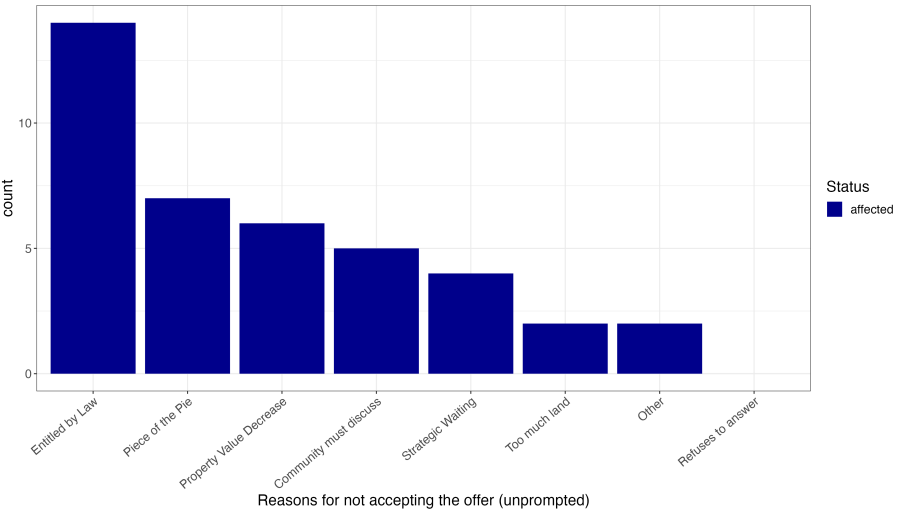


Figure A13: Interviewed Owners Do Not Believe that their Decision has an Impact on The Probability of the Road Being Built

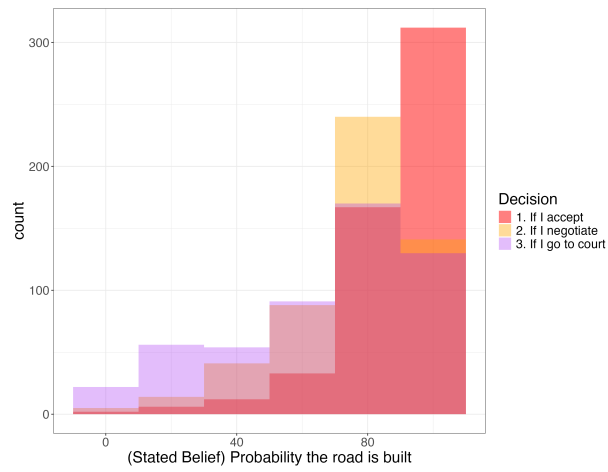


Figure A14: Predicted Market Value of Affected Land by Parish for 75m<sup>2</sup>

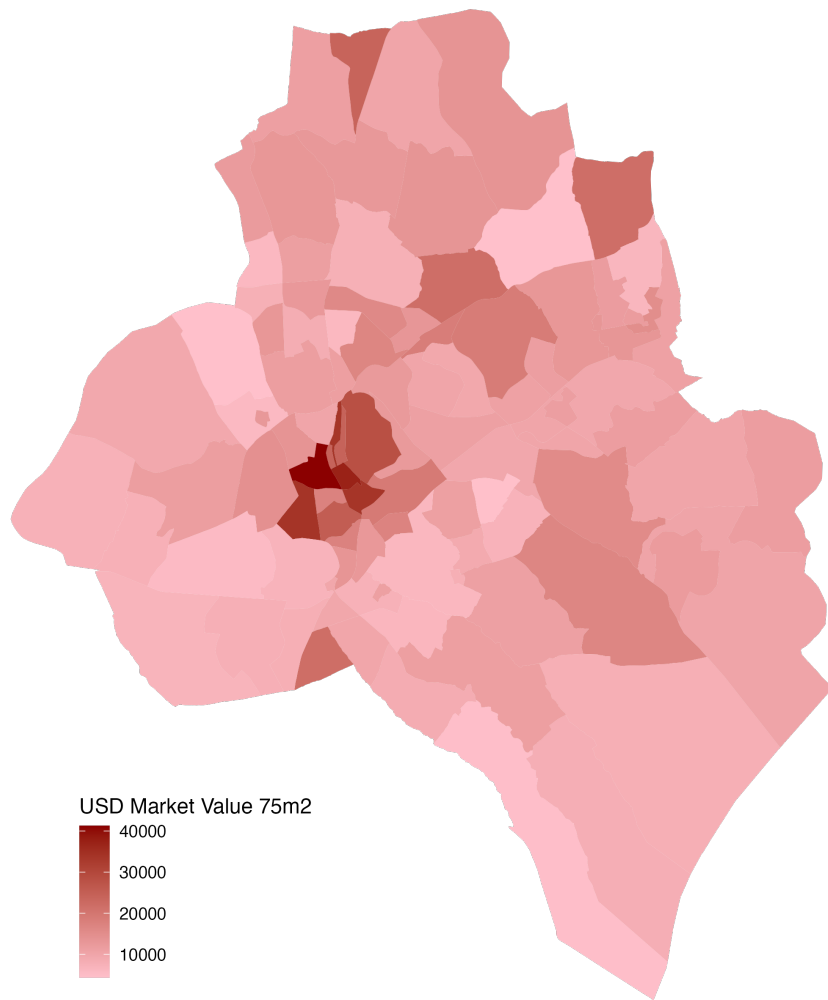
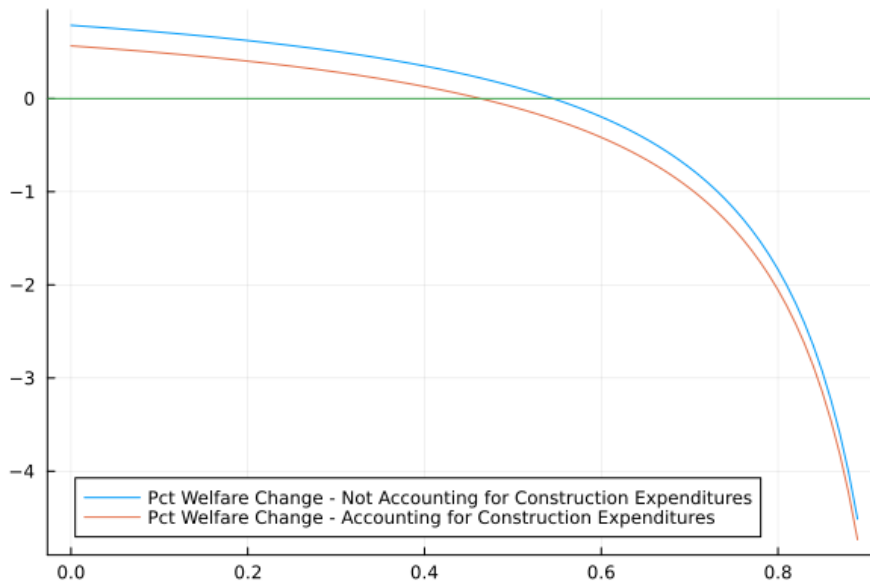


Table A6: Estimated  $\theta\kappa$  - Alternative Specifications

	<i>Dependent variable:</i>			
	(1)	(2)	(3)	(4)
Time (in Min)	-0.033*** (0.001)	-0.023*** (0.001)	-0.045*** (0.001)	-0.035*** (0.001)
Period Def	Month	Month	Year	Year
Sample	Evening	All	Evening	All
Fixed Effects				
- origin x year	Y	Y	Y	Y
- destination x year	Y	Y	Y	Y
- month			Y	Y
SE Clustered	o-d	o-d	o-d	o-d
Observations	59,300	45,077	17,084	15,009
R <sup>2</sup>	0.437	0.687	0.513	0.813

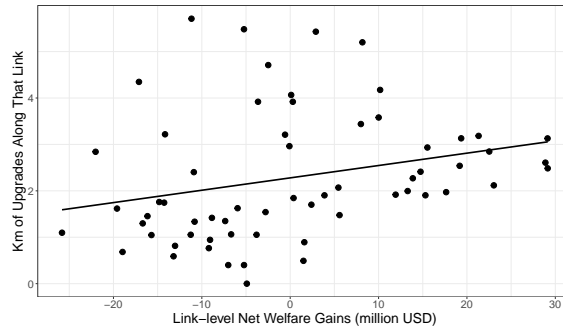
Notes: Standard errors are displayed in parentheses and clustered at the o-d pair level, with significance levels \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ . All specifications include origin-year, destination-year and month fixed effect. I control for log distance in all specifications. In column 1, I include all trips from 6am to 10am, and from 4pm to 8pm. In column 2, I restrict the sample to the evening rush hour from 4pm to 8pm. Observations are at the pair-month level. Trips are motorcycles (bodas) trips.

Figure A15: Welfare Gains of the Realized Improvements Under Market Value Acquisition as a Function of the tax revenue wedge  $\eta$



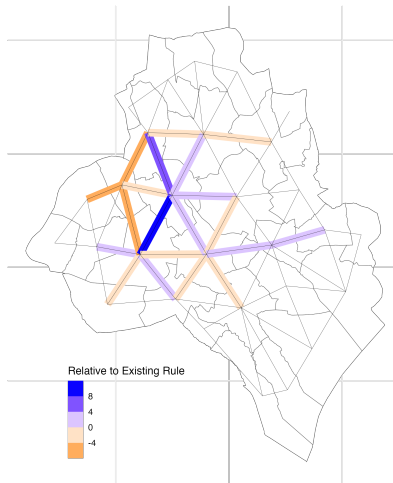
Notes: This figure displays the percentage welfare gains from the realized road improvements if land is acquired at market value, under various values for the tax revenue wedge  $\eta \in [0.0, 0.9]$ . The red curve accounts for the construction costs to compute the net welfare gains, but assume that the tax revenue wedge on these funds is 0 because they are funded through international agencies. The blue curve takes the construction costs as free.

Figure A16: Net Welfare Gains of Link-Level Upgrades and Realized Upgrades



Notes: I display the relationship between link-level net welfare gains as predicted by the model, and the length of realized road upgrades along that link.

Figure A17: No Land Payment & No Land Use



Notes: I display the difference in optimal road improvements between the existing rule in the absence of land payments (Figure 12). Orange links correspond to less improvement under the no land payment counterfactual than under the status quo rule. Purple links correspond to more improvements under the no land payment counterfactual than under the status quo rule. In addition to removing land payments, I remove the opportunity cost of land use, so that road improvements do not take up any residential land, and so their optimal location is only driven by the relative benefits.