

1 Title: Identifying, projecting, and evaluating informal urban expansion spatial patterns

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3 **Beth Tellman^{1*}, Hallie Eakin², B.L. Turner II^{2,3}**

4 *Corresponding Author, btellman@arizona.edu

5 1. School of Geography, Development, and Environment, University of Arizona. OrcID: 0000-

6 0003-3026-6435. Twitter: @paz.justicia.vida

7 2. School of Sustainability, Arizona State University. OrcID: 0000-0001-8253-1320. Twitter:

8 @HallieEakin

9 3. School of Geographical Sciences and Urban Planning, Arizona State University. OrcID: 0000-

10 0002-6507-521X

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1 **Abstract**

2 Informal urban land expansion is produced through a diversity of social and political
3 transactions, yet “pixelizable” data capturing these transactions is commonly unavailable.
4 Understanding informal urbanization entails differentiating spatial patterns of informal
5 settlement from formal growth, associating such patterns with the social transactions that
6 produce them, and evaluating the social and environmental outcomes of distinct settlement
7 types. Demonstrating causality between distinct urban spatial patterns and social-institutional
8 processes requires both high-resolution spatial temporal time series data of urban change and
9 insights into social transactions giving rise to these patterns. We demonstrate an example of
10 linking distinct spatial patterns of informal urban expansion to the institutional processes each
11 engenders in Mexico City. The approach presented here can be applied across cases, potentially
12 improving land projection models in the rapidly urbanizing Global South, characterized by high
13 informality. We conclude with a research agenda to identify, project, and evaluate informal
14 urban expansion patterns.

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16 **Keywords:** Urbanization, Informality, Mexico City, Politics, Spatial Patterns, Remote Sensing

17

18 **Introduction**

19 Land system science increasingly addresses urban land dynamics and their consequences
20 (Seto & Reenberg, 2014). Planning for more sustainable urban land use demands an accurate
21 understanding of the sociodemographic, institutional and environmental drivers of urban land
22 change. In the Global South, where over 90% of all urban growth is taking place (UN Habitat,
23 2016), the institutional drivers of informal urban land change are particularly salient, yet poorly
24 addressed in land change models. Inhabitants of informal urban settlements have little to no
25 tenure security and lack basic services. Growth of informal urban settlements is often associated
26 with political transactions, because settlements outside the urban zone are commonly

1 prohibited from official access to urban services, titles, and infrastructure, which politicians may
2 facilitate in exchange for electoral support. Spatial-temporal patterns of land change may
3 identify where informal political and economic transactions, such as rent-seeking, clientelism,
4 and corruption, take place. Despite the recognizable differences in informal urban spatial
5 patterns from formal growth (Graesser et al., 2012; Hofmann et al., 2015; Kerins et al., 2021;
6 Kohli et al., 2012; Kuffer et al., 2014; Owen & Wong, 2013), models of urban land change often
7 incorrectly assume spatial signatures of urbanization from the Global North are transferrable to
8 model informal urban growth in the Global South. Urban land change models fail to capture
9 drivers of change in the Global South, including the institutions guiding the process. As a result,
10 models of urban growth projections are likely inaccurate and problematic for urban planning
11 efforts. In this perspective, we argue that a better understanding of urban land systems requires
12 understanding institutional drivers of change and linking them to subsequent spatial patterns of
13 informal settlement growth.

14 Understanding urban land systems can be assisted by systematic, empirical analysis,
15 addressing how much and what kind of land is consumed and the process at play (Solecki et al.,
16 2013). Quantifying land-use change is key to understanding what drives urban form in many
17 cities worldwide, with implications for sustainability. Nevertheless, while most spatial urban
18 growth models include landscape (e.g., slope or land use), location (e.g., distance to city center,
19 road, or amenities), and zoning constraints (Irwin, 2010) as predictive variables, models do not
20 disaggregate formal and informal growth. The conflation of informal and formal urban growth
21 in these analyses display mixed performances for cities in which informal growth plays a
22 dominant role. Models perform reasonably well in Kampala, Uganda, for example (Vermeiren et
23 al., 2012), while modeling patterns in other places of the Global South has proved challenging
24 (Pontius et al., 2008). A recent review of the few existing informal urban expansion growth
25 models (D. Roy et al., 2014) found only one that incorporated political transactions (Patel et al.,
26 2012). No model disaggregates informal urban growth into component types. Most models of

1 informal urban expansion rely on physical or environmental variables only (Augustijn-Beckers
2 et al., 2011; Dubovyk et al., 2011). Failing to incorporate social and political transactions shaping
3 informal urban settings into land change models limits understanding of how to manage and
4 project urban expansion (Nagendra et al., 2018). As a result, existing efforts to simulate and
5 project patterns of urbanization in the developing world often fall short of adequately capturing
6 the dynamics at play (Pontius et al., 2008).

7 We argue that improving monitoring, modeling, and understanding informal urban
8 expansion requires addressing at least three challenges: i) understanding the mechanisms,
9 rules, and payoffs associated with distinct land change patterns, ii) distinguishing dominant
10 spatial patterns of informality associated with dominant institutional mechanisms, and iii)
11 evaluating the implications of the social and environmental outcomes specific patterns of
12 informality. We describe each of these challenges and illustrate them using the case of informal
13 urban expansion in Mexico City. We conclude with a research agenda key to address the
14 challenges identified.

15 **Three challenges to understand drivers and consequences of informal urban** 16 **expansion**

17 The first challenge is to understand the mechanism, rules, and payoffs to actors
18 associated with informal urban expansion. Urban informality is a form of land use that is
19 challenging to study, given its frequent association with undocumented, illegal, or illicit
20 transactions and intentionally hidden exchanges. Despite narratives by urban authorities that
21 urban informality is “chaotic” (Gilbert & De Jong, 2015; Lerner et al., 2018), research shows that
22 informality is shaped, co-produced, and “ordered” by power relations, incentives, and rules of
23 interaction amongst those selling, regulating, and urbanizing land parcels (Duhau & Giglia,
24 2008; Fernandes & Smolka, 2004; McFarlane, 2012; A. Roy, 2005; Tellman et al., 2021).
25 Systematically documenting the contemporary mechanisms defining informal urban expansion
26 across a megalopolis requires social science methods (mainly qualitative) to systematize the

1 dominant rules at play (Banks et al., 2020; Connolly, 2009; Doshi, 2018; Fernandes & Smolka,
2 2004; McFarlane, 2012; Pradilla Cobos, 1995; A. Roy, 2005, 2009; Tellman et al., 2021; Van
3 Gelder, 2013; Varley, 1998). In any complex megalopolis, there will likely be multiple forms of
4 informal urban expansion, each with differing sets of rules and norms (i.e., institutions) (North,
5 1990; Ostrom, 2011). These institutions are often shaped by powerful actors or institutional
6 entrepreneurs (Pacheco et al., 2010; Tellman et al., 2021).

7 Once distinct institutional mechanisms and associated entrepreneurs of urban expansion
8 are identified and categorized, the second challenge is to determine if each category is associated
9 with a distinct spatial pattern. Linking social and institutional processes to land-use patterns is
10 particularly challenging in these cases, but possible, via two approaches (Tellman, Magliocca, et
11 al., 2020). One approach is to use causal inference to link variation in observed spatial patterns
12 to “pixelizable” social data (Belhabib et al., 2020; Jain, 2020; Ordway et al., 2019; Tellman,
13 Sesnie, et al., 2020). The other approach is to use understanding of social mechanisms to
14 identify the process generating distinct spatial patterns (Curtis et al., 2018; Sesnie et al., 2017)
15 or generate spatial patterns with social rules using Agent Based Models (N. R. Magliocca et al.,
16 2019). Agent Based Models are commonly used to link land-use pattern to process to
17 understand what drives change in both forest (Curtis et al., 2018; De Oliveira Filho & Metzger,
18 2006; Manson & Evans, 2007) and urban systems (Irwin et al., 2009; N. Magliocca et al., 2011).
19 The increasing availability of both high-resolution time series satellite data and deep learning
20 classification models that identify contextual spatial patterns (e.g., convolutional neural
21 networks or CNNs) (Dolley et al., 2020; Isikdogan et al., 2017; Jean et al., 2016; Kerins et al.,
22 2021; Wang et al., 2020; Zhang et al., 2018) enables researchers to identify distinct patterns of
23 land change and link them to illicit processes, even in urban settings.

24 The third challenge is to understand the social and environmental consequences of
25 distinct patterns of informal urban expansion. Disaggregating different types of urbanization
26 patterns is important for improving modeling of urbanization, and understanding how distinct

1 urban growth patterns generate distinct assemblages of land system architecture (Turner et al.,
2 2013), or the spatial arrangement of heterogeneous land uses in a given landscape. These spatial
3 patterns influence urban climate (Benson-lira et al., 2016), energy budgets and emissions
4 (Frolking et al., 2013), flood risk (Wheater & Evans, 2009), biodiversity (Stuhlmacher et al.,
5 2020), and human health (Ahern, 2011), among other sustainability factors. While informality
6 plays an important role in mega-urban development worldwide, little is known about how it
7 affects urban form and the aforementioned environmental services (Henderson et al., 2016).
8 The spatial patterns produced by urban informality remain unexplored, despite the importance
9 these land patterns represent for urban socio-ecological systems and the implications they have
10 for urban sustainability management and planning.

11 Do distinct orders of social processes driving informality produce similarly ordered
12 spatial morphology? If so, can we identify the different underlying processes by way of detecting
13 the morphological development in satellite imagery? We believe the answer is yes, and
14 demonstrate through an example in Mexico City. While various typologies of informality for
15 Mexico City have been developed (Azuela de la Cueva, 1987; Cymet, 1992; Jones & Ward, 1998;
16 Tellman et al., 2021; Ward, 1976), they have not been linked to specific spatial morphological
17 patterns each may produce on the landscape (e.g. see Bazant, 2001; Nurko et al., 2016) . One
18 study disaggregated spatial morphology of settlement types in Mexico City (Kerins et al., 2021)
19 by training a CNN on satellite data (Sentinel-2) based on the Atlas of Urban Expansion (Lincoln
20 Institute of Land Policy, 2016), but categories of settlement were not linked to distinct social
21 processes (e.g., informal settlements are labeled in this classification as “atomistic” settlements
22 with irregular spatial layouts). Here we address the three aforementioned challenges of i)
23 understanding informal urban mechanisms and payoffs, ii) distinguishing processes by spatial
24 patterns, and iii) evaluating some of their social and environmental outcomes in the case of
25 Mexico City. We extend previous work that examined informal urban institutional types and
26 mechanisms in Mexico City (Tellman et al 2021) by linking each institutional type to a spatial

1 pattern on the landscape. In this case, we explore both the potential and the constraints in
2 analysis posed by available datasets, and from this exploration, propose a way forward for
3 research.

4 **Spatial Patterns of Four Informal Urban Land Institutions for Low- and Middle-** 5 **Income Housing in Mexico City**

6 The Mexico City Metropolitan Area is one of the largest megacities in the world, with nearly
7 30 million people, and has the highest percentage of urban expansion through informal growth
8 (an estimated 65%) in Latin America (Connolly, 2009). Informal urban growth for low and
9 middle income housing persists because of inadequate public housing efforts, a formal housing
10 market inaccessible to the poor, and economic and political gains to politicians, developers, and
11 intermediaries (informal actors who provide access to services like water or electricity) profiting
12 from informal settlement expansion (Tellman et al., 2021). This kind of urban expansion
13 involves an ordered social and institutional process with four distinct types: i) so-called “ant”
14 urbanization (direct sale of one plot to one settler), ii) illegal subdivision (one actor who buys
15 and sells many plots of land), iii) land invasion (a group of settlers illegally squatting on land,
16 usually facilitated by a political group), and, iv) via social or public housing (city or federal
17 subsidized housing for low or middle income populations) (Tellman et al., 2021). The last type is
18 not typically considered informal in the literature (Berglund, 2019; Lambert, 2021). Evidence
19 from Mexico City, however, demonstrates that this development is often deeply embedded in
20 social transactions that deviate from stated legal norms and procedures.

21 The four types of informality are identifiable by common types of land transactions of
22 each and the economic and political payoffs to the actors involved (Fig. 1, see Tellman et al.,
23 2021 for more details). Economic returns include cash earned through bribes and side
24 payments, sale of land or urban services, kickbacks paid by developers to politicians to change
25 zoning or evade regulation, and increased municipal budgets from an expanded tax base.
26 Political returns include opportunities for politicians to advance in their party or win office by

1 gathering loyal clienteles of voters and citizens to participate in mass mobilization. Urban
2 consolidation expands in space and accelerates in time as returns in political and economic
3 capital increase (Fig. 1).

4 The hypothesized spatial pattern of each informal urban type, from initial construction
5 to consolidation, is depicted in Figure 2, with an empirical case example from Google Earth
6 Satellite imagery provided in Figure 3. Cases were identified via GPS points taken in the field,
7 which were uploaded into Google Earth. Screen shots were taken using the Google Earth
8 timelapse feature, but place names are intentionally anonymous due to the sensitive nature of
9 the topic (e.g., potentially making communities vulnerable to eviction). An explanation of the
10 hypothesized pattern and empirical example for each urbanization type follows.

11 Ant urbanization generates a small amount of economic capital to landowners selling
12 their plot on the informal market and to government officials bribed by informal settlers to
13 prevent foreclosure. Politicians engage ant urbanization areas to influence the distribution of
14 services, provide construction materials, or encourage consolidation to garner votes. Ant
15 urbanization appears near the urban fringe on existing agricultural or conservation lands. It is
16 produced by one-off land sales in Southern Mexico City on land zoned as “conservation.” Urban
17 growth is not permitted outside areas zoned urban in local government land-use plans (which
18 began in 1997), but some agricultural activity is allowed. Farmers with land rights who no longer
19 wish to farm or need cash may sell plots (typically 250m² in size) informally to settlers seeking
20 to build a home. Ant urbanization produces a slowly developing, dispersed settlement pattern
21 (Fig. 2). Growth may accelerate or consolidate when intermediaries install urban services,
22 making the land more valuable and increasing demand. Ant urbanization is typified by one
23 community (Fig. 3) that took 15 years to develop and is still slowly growing. In addition to the
24 spatial pattern, ancillary data of land zones that identifies areas not permitted for urban growth
25 are required to identify this urbanization type.

1 Subdivision generates larger economic returns than ant urbanization, because the
2 process is concentrated in one actor who sells dozens to thousands of lots. This actor purchases
3 agricultural land from a member of an agrarian community with communally-held property
4 (*ejidos* in Mexico) and who holds individual land-use rights to farm a small portion of the *ejido*
5 (~ 1 ha). It is either outright illegal to sell this land or expensive to do so legally because of high
6 transaction costs. As such, an *ejido* member illegally sells the land to an entrepreneur known as
7 a “land flipper.” The land flipper buys many *ejido* plots, which are subdivided and re-sold to
8 settlers. The political returns are often higher than ant urbanization due to larger numbers of
9 settlers (and potential votes to be gained). Subdivision occurs on rectangular agricultural plots
10 (Fig 2), and the urbanization pattern is more rapid and spatially consolidated than ant
11 urbanization. One such community reached its development in about 10 years (Fig. 3).
12 Subdivision and ant urbanization together represent at least 10% of all urbanization in the
13 Mexico City Metro Area, occupying at least 3,200 ha in Mexico City’s conservation land. In
14 addition to its characteristic spatial pattern, ancillary data on land tenure is required to identify
15 areas of collective title where subdivision typically occurs.

16 Land invasion generates larger political returns, because settlers “pay” for rights to land
17 and services via political participation with the group engineering the invasion. The economic
18 returns can be higher than subdivision, depending on the size of the invasion. Land invasion
19 generates the most rapid and consolidated urban pattern. These communities construct many
20 homes rapidly in interstitial urban areas (empty lots, or public spaces such as garbage dumps) or
21 open lands immediately adjacent to the urban fringe (Fig. 2). Urban plots are built next to one
22 another with little undeveloped space, and the community (~100-1,000 homes) is completed
23 within a few months (Fig. 3). Although often receiving significant media attention, this process
24 represents the smallest proportion of informal growth (in terms of spatial extent) in Mexico
25 City. The two most common invasion groups in Mexico City, Antorcha Campesina and the
26 Pancho Villas, have together urbanized 600 ha of land, representing 2% of the city’s growth

1 since 2000. In addition to its distinctive pattern, media reports can be used to identify
2 approximate neighborhoods where an invasion process may have taken place.
3 Social housing generates the highest economic returns, concentrated in one developer. Such
4 housing refers to large, new construction funded by INFONAVIT (National Housing Fund for
5 Workers) on the urban fringe of the metropolitan area, not to the renovation of existing buildings
6 in Mexico City's urban core related to the INVI program (The National Institute of Housing).
7 Developers may purchase cheap agricultural land far from the urban fringe and convince the
8 accompanied municipalities to rezone the land as urban through bribes, offering to aid in zoning
9 plans in under-resourced municipalities, or demonstrating the potential of increased tax
10 returns. Municipalities are required to install urban services, the government facilitates the sale
11 of these homes via a social housing program, and the developer captures the capital gains. Social
12 housing areas are large, and develop within two to four years (Fig. 2). This pattern represents at
13 least 11,000 ha of urban growth, around 30% of new urban land in the Mexico City area from
14 2000-2010. Social housing has slightly smaller rates of consolidation compared to land invasion
15 (Fig. 2), but the area urbanized is much larger. Ancillary data on permits for social housing
16 construction could be used to identify the names and municipality of construction, but specific
17 data on the polygons associated with these permits are not publicly available. Sometimes, the
18 name of the developer appears in Google Earth, which can help identify a social housing
19 complex.

20 **Outstanding challenges in Mexico City**

21 Previous work has addressed the first challenge of systematically examining the
22 mechanisms, rules, and payoffs associated with types of informal urban expansion (Tellman et
23 al., 2021), however, the remaining two identified challenges (distinguishing spatial patterns
24 across the metropolis and estimating social and environmental impacts of patterns) have only
25 been partially addressed. While Google Earth offers exemplary visual trajectories of informal

1 settlement types and spatial patterns, classifying these patterns across the city requires an
2 annual time series of >15 years at <5m spatial resolution. Attempts to identify ant urbanization
3 patterns with a time series of urban land cover with Landsat (at 30m resolution) (Goldblatt et
4 al., 2018) failed because dispersed single family homes fell into mixed pixels covered with forest
5 on the urban fringe. Likewise, in other urbanizing cities, such as Wuhan China, informal
6 urbanization from migrant farmers can only be resolved with high resolution commercial
7 imagery (<3m spatial resolution) (Dolley et al., 2020).

8 While the social consequences of different types of urban informality have been
9 documented (Tellman et al., 2021), understanding how the spatial patterns of each degrade (or
10 retain) environmental services has not been examined. In addition to socio-environmental
11 consequences associated with informal land-use patterns, the underlying institutions that
12 produce these patterns may have social costs. For example, politicians may withhold urban
13 services such as electricity, water, or public housing in exchange for votes or political support
14 (Castro, 2006; de Alba, 2016; Hilgers, 2008), or intermediaries may set up informal electricity
15 or water grids and charge residents substantial fees for access (Tellman et al., 2021;
16 Venkatachalam, 2015). Below we offer a research agenda to address remaining challenges in
17 Mexico City, with lessons for efforts to capture such patterns in other urbanizing regions.

18 **Identifying informal land change patterns in urban areas requires higher** 19 **resolution time series of >15 years**

20 Mapping informal settlements requires high resolution (<10 m) satellite data in an
21 annual time series of enough length to detect settlement evolution (Kuffer et al., 2016). Public
22 satellite data available to map annual urban expansion (MODIS and Landsat, or 250m and 30
23 m² respectively), in this case, from 2000-2015, are not at a resolution sufficient to capture
24 growth patterns, such as ant urbanization in which a typical abode has a roof of no more than
25 60m². Plausible options for high-resolution time series data for urban classification include
26 commercial optical data from PlanetScope (3-5m, from 2017, and Rapideye, from 2009), the

1 recently publicly available (as of August 2020) SPOT archive (2.5-6m for Spot 2, 6 and 7 from
2 2002-2016), or publicly available Sentinel-2 data (10m resolution, available starting 2015).

3 High-resolution urban time series data would facilitate pattern analysis to address
4 several key questions to understand informal urban land systems and aid their governance: Do
5 formal versus informal urban development produce distinct and identifiable patterns? Can
6 distinct types of informal growth be differentiated across a metropolitan area? If so, do these
7 differentiated patterns indicate distinct environmental and social consequences? Answers to
8 these question through the use of high-resolution data could aid impact evaluation of policies
9 such as land titling and regularization, eviction, payments for environmental services, and other
10 interventions designed to curb urban growth.

11 The paucity of such data for Mexico City impeded empirical testing of the hypothesized
12 spatial-temporal patterns presented. The average land invasion in Mexico City from 2000-2016
13 was 1.9 ha, which is equivalent to only 20 Landsat pixels. Subdivisions and ant urbanization are
14 even more dispersed and are unlikely to be distinguishable from land invasions at such a low
15 spatial resolution. Google Earth Imagery tiles can be visualized, but are not available for
16 download for quality empirical analysis, as in the data can only be captured using screen shots
17 in low resolution. Efforts to map informal growth at 5m in Mexico City for one point in time
18 have been successful (Rodriguez Lopez et al., 2017), but no time series are available. Time series
19 of at least 15 years are required to differentiate, for example, a consolidated ant urbanization
20 community from a subdivision (Fig.3). Other studies that have examined patterns in land
21 system architecture rely on 1m resolution images (Li et al., 2017). While we were unable to take
22 advantage of the data for the analysis here, the availability of Sentinel-2 data, r opening of the
23 SPOT archive, or advances in deep learning to classify Google Earth Imagery for informal per-
24 urban development (see Dolley et al., 2020) could help identify distinct informal urban land
25 patterns in Mexico City and elsewhere in the future.

1 **Disaggregating urban informality links institutions to landscape patterns and** 2 **social and environmental consequences**

3 Not all urbanization is driven by the same mechanisms or produces the same patterns. At
4 first glance, heterogeneity in informal urban pattern appears seemingly chaotic, and has
5 overwhelmed attempts to model the pattern of urban growth in the Global South robustly
6 (Pontius et al., 2008). In Mexico City, we identified explicit and distinct sets of actors and rules
7 characterizing different types of informal urbanization, each with its unique spatial footprint,
8 demonstrating the potential links between processes and patterns for one metropolitan area.
9 The general approach of linking urban social process to spatial pattern could be applied to other
10 cities, always starting with the specific informal settlement typology and spatial pattern. For
11 example, analyses of Caracas would include vertical slums (Schmaeler, 2016), whereas analysis
12 of Wuhan would include peri-urban migrant farmers(Dolley et al., 2020), while analysis for
13 Nairobi would describe communities of individual “shacks” versus denser tenements, each with
14 a distinct spatial pattern (Mwau & Sverdlik, 2020). On this basis, we posit that codifying and
15 translating rule sets into urban growth models could improve their accuracy. To improve urban
16 growth models, however, a series of such studies should be made across the Global South to
17 enable meta-analyses (e.g (Kuras et al., 2020; Narayanan et al., 2017; Seto et al., 2011; Teo et al.,
18 2021; Tuholske et al., 2021). The outcomes may point to common patterns associated with the
19 general type of processes involved or by regional context. Connecting institutional and social
20 patterns to spatial footprints, as demonstrated here for Mexico City, provides a framework that
21 could be replicated in other cases and lead to informal urban pattern meta-analyses.
22 Understanding the spatial patterns of informality could lead to more robust modeling
23 approaches and assessment of consequences for the environment.

24 Informal urbanization is a social-ecological process that produces diverse but potentially
25 consistent patterns for their projections with measurable environmental outcomes, and with
26 future advancements, perhaps predictable outcomes. Each pattern is driven by institutional

1 arrangements that produce a distinct land system architecture with implications for both social
2 outcomes and environmental services (Turner , 2017). Social outcomes of urban informal types
3 can be assessed by their ability to provide land and housing, the location and quality of housing
4 and services, and the economic and social cost (e.g. exclusion, violence, or control) to residents.
5 For example, for residents in Mexico City, participating in invasions is the cheapest way to
6 obtain land for a home, but such participation then requires political participation in protests
7 and electoral allegiance (Tellman et al., 2021).

8 Stocks and flows of environmental services in and for cities (also known as urban ecosystem
9 or environmental services (Tan et al., 2020)) are largely dependent on spatial land-use patterns
10 (Andersson et al., 2015). In Mexico City, for example, land system architecture of small areas of
11 slow, distributed ant urbanization on conservation land versus the large areas of rapid,
12 consolidated social housing on agricultural land could have different consequences for
13 hydrologic ecosystem services, such as water filtration, aquifer recharge, and flood mitigation.
14 Yet, public discourse in Mexico City assumes ant urbanization and invasion as the primary cause
15 of reducing environmental services and increasing water scarcity and flood risk (Eakin et al.,
16 2019; Lerner et al., 2018). This assumption leaves other types of informal urbanization and their
17 spatial and environmental consequences unexamined. Disaggregating informal urbanization
18 based on institutional arrangements and landscape patterns could provide a new avenue to
19 analyze tradeoffs based on environmental costs of informal growth in Mexico City and
20 elsewhere. Open source modeling software such as InVEST (Integrated Valuation of Ecosystem
21 Services and Tradeoffs) can be used to translate land-use patterns into quantified biophysical
22 and economic value of a large suite of urban environmental services (Hamel et al., 2021). Open
23 source web-tools leverage deep learning algorithms to classify Google Earth imagery and
24 quantify tradeoffs in informal housing and agriculture with urban food security, and other
25 sustainable development goals (Dolley et al., 2020). Trade-offs between environmental services
26 for identified typologies of urban growth could be examined in three ways: i) comparing past

1 changes in urban land-cover patterns using extant maps or remote sensing (e.g., (Xiang et al.,
2 2022)) across urban informality types; ii) using land-use models to project future patterns of
3 urban growth scenarios assuming different percentages or types of urban informality and
4 quantifying the resulting services, or iii) stakeholder scenario mapping of potential urban land-
5 use plans and quantified trade-offs of social benefits and environmental services (Dolley et al.,
6 2020) . Methods exist to examine tradeoffs in future land change projections and their
7 environmental services (Lang & Song, 2019), but have not been applied to distinct types of
8 urban development. A better understanding of environmental service impacts generated by
9 distinct types of informal growth could aid urban planning efforts to improve urban
10 sustainability. Yet differentiated environmental service assessments are predicated on the ability
11 to identify the process, spatial pattern, and local of informal settlement types, which is currently
12 a challenge for the land change community.

13 **Conclusion**

14 Overall, land system science has yet to fully engage with the range and nuances of
15 institutional contexts that shape informal or illicit transactions. These transactions may result in
16 unique landscape signatures, urban or non-urban in kind. Linking institutional rules and social
17 relations to the distinct spatial footprints they engender requires bringing together disparate
18 knowledge communities which have focused almost exclusively on either informal urban
19 processes or urban land cover patterns. Linking spatial patterns to social process represents an
20 exciting frontier for land systems scientists, and a necessary step to understanding urbanization
21 in the Global South. Identifying, projecting, and evaluating informal urban expansion patterns
22 requires meta-analyses of socio-institutional informal development across cities, developing
23 higher spatial and temporal resolution time series data, and methods to assess socio-
24 environmental consequences that avoid further marginalizing vulnerable communities.

25

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8 **References**

- 9 Ahern, J. (2011). From fail-safe to safe-to-fail: Sustainability and resilience in the new urban
10 world. *Landscape and Urban Planning*, 100(4), 341–343.
11 <https://doi.org/10.1016/j.landurbplan.2011.02.021>
- 12 Andersson, E., McPhearson, T., Kremer, P., Gomez-Baggethun, E., Haase, D., Tuvendal, M., &
13 Wurster, D. (2015). Scale and context dependence of ecosystem service providing units.
14 *Ecosystem Services*, 12(2003), 157–164. <https://doi.org/10.1016/j.ecoser.2014.08.001>
- 15 Augustijn-Beckers, E. W., Flacke, J., & Retsios, B. (2011). Simulating informal settlement
16 growth in Dar es Salaam, Tanzania: An agent-based housing model. *Computers,
17 Environment and Urban Systems*, 35(2), 93–103.
18 <https://doi.org/10.1016/j.compenvurbsys.2011.01.001>
- 19 Azuela de la Cueva, A. (1987). Low income settlements and the law in Mexico City. *International
20 Journal of Urban & Regional Research*, 11(4), 522–541.
- 21 Banks, N., Lombard, M., & Mitlin, D. (2020). Urban Informality as a Site of Critical Analysis.
22 *Journal of Development Studies*, 56(2), 223–238.
23 <https://doi.org/10.1080/00220388.2019.1577384>
- 24 Bazant, J. (2001). Periferias urbanas. In *México. Trillas*.
- 25 Belhabib, D., Le, P., & David, B. (2020). *Narco-Fish : Global fisheries and drug trafficking.*
26 *February*, 1–16. <https://doi.org/10.1111/faf.12483>
- 27 Benson-lira, V., Georgescu, M., Kaplan, S., & Vivoni, E. R. (2016). Loss of a lake system in a
28 megacity: The impact of urban expansion on seasonal meteorology in Mexico City. *Journal
29 of Geophysical Research : Atmospheres*, 1–19.
30 <https://doi.org/10.1002/2014JD022994>.Received
- 31 Berglund, L. (2019). Excluded by design: informality versus tactical urbanism in the
32 redevelopment of Detroit neighborhoods. *Journal of Cultural Geography*, 36(2), 144–181.
33 <https://doi.org/10.1080/08873631.2018.1516600>
- 34 Castro, J. E. (2006). *Water, Power, and Citizenship*. Palgrave Macmillan.
- 35 Connolly, P. (2009). Observing the evolution of irregular settlements: Mexico City's colonias
36 populares, 1990 to 2005. *International Development Planning Review*, 31(1), 1–35.
- 37 Curtis, P. G., Slay, C. M., Harris, N. L., Tyukavina, A., & Hansen, M. C. (2018). Classifying
38 drivers of global forest loss. *Science*, 361(6407), 1108–1111.
39 <https://doi.org/10.1126/science.aau3445>
- 40 Cymet, D. (1992). *From ejido to metropolis, another path: an evaluation on ejido property
41 rights and informal land development in Mexico City* (pp. xiv, 275 p.). P. Lang.
42 <file://catalog.hathitrust.org/Record/002884210>
- 43 de Alba, F. (2016). Challenging state modernity: Governmental adaptation and informal water
44 politics in Mexico City. *Current Sociology*, 001139211665728.
45 <https://doi.org/10.1177/0011392116657288>
- 46 De Oliveira Filho, F. J. B., & Metzger, J. P. (2006). Thresholds in landscape structure for three
47 common deforestation patterns in the Brazilian Amazon. *Landscape Ecology*, 21(7), 1061–
48 1073. <https://doi.org/10.1007/s10980-006-6913-0>
- 49 Dolley, J., Marshall, F., Butcher, B., Reffin, J., Robinson, J. A., Eray, B., & Quadrianto, N.
50 (2020). Analysing trade-offs and synergies between SDGs for urban development, food

- 1 security and poverty alleviation in rapidly changing peri-urban areas: a tool to support
2 inclusive urban planning. *Sustainability Science*, 15(6), 1601–1619.
3 <https://doi.org/10.1007/s11625-020-00802-0>
- 4 Doshi, S. (2018). Greening Displacements, Displacing Green: Environmental Subjectivity, Slum
5 Clearance, and the Embodied Political Ecologies of Dispossession in Mumbai.
6 *International Journal of Urban and Regional Research*, 112–132.
7 <https://doi.org/10.1111/1468-2427.12699>
- 8 Dubovyk, O., Sliuzas, R., & Flacke, J. (2011). Spatio-temporal modelling of informal settlement
9 development in Sancaktepe district, Istanbul, Turkey. *ISPRS Journal of Photogrammetry*
10 *and Remote Sensing*, 66(2), 235–246. <https://doi.org/10.1016/j.isprsjprs.2010.10.002>
- 11 Duhau, E., & Giglia, Á. (2008). *Las reglas del desorden: habitar la metrópoli*. Siglo XXI.
- 12 Eakin, H., Siqueiros-García, J. M., Hernández-Aguilar, B., Shelton, R., & Bojórquez-Tapia, L. A.
13 (2019). Mental Models, Meta-Narratives, and Solution Pathways Associated With Socio-
14 Hydrological Risk and Response in Mexico City. *Frontiers in Sustainable Cities*,
15 1(November), 1–13. <https://doi.org/10.3389/frsc.2019.00004>
- 16 Fernandes, E., & Smolka, M. (2004). Regularización de la tierra y programas de mejoramiento:
17 nuevas consideraciones. *Land Lines*, 16(3).
- 18 Froking, S., Milliman, T., Seto, K. C., & Friedl, M. a. (2013). A global fingerprint of macro-scale
19 changes in urban structure from 1999 to 2009. *Environmental Research Letters*, 8,
20 024004. <https://doi.org/http://dx.doi.org/10.1088/1748-9326/8/2/024004>
- 21 Gilbert, L., & De Jong, F. (2015). Entanglements of Periphery and Informality in Mexico City.
22 *International Journal of Urban and Regional Research*, 39(3), 518–532.
23 <https://doi.org/10.1111/1468-2427.12249>
- 24 Goldblatt, R., Stuhlmacher, M. F., Tellman, B., Clinton, N., Hanson, G., Georgescu, M., Wang,
25 C., Serrano-Candela, F., Khandelwal, A. K., Cheng, W.-H., & Balling, R. C. (2018). Using
26 Landsat and nighttime lights for supervised pixel-based image classification of urban land
27 cover. *Remote Sensing of Environment*, 205(November 2017), 253–275.
28 <https://doi.org/10.1016/j.rse.2017.11.026>
- 29 Graesser, J., Cheriyyadat, A., Vatsavai, R. R., Chandola, V., Long, J., & Bright, E. (2012). Image
30 based characterization of formal and informal neighborhoods in an urban landscape. *IEEE*
31 *Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 5(4),
32 1164–1176. <https://doi.org/10.1109/JSTARS.2012.2190383>
- 33 Hamel, P., Guerry, A. D., Polasky, S., Han, B., Douglass, J. A., Hamann, M., Janke, B., Kuiper, J.
34 J., Levrel, H., Liu, H., Lonsdorf, E., McDonald, R. I., Nootenboom, C., Ouyang, Z., Remme,
35 R. P., Sharp, R. P., Tardieu, L., Vigié, V., Xu, D., ... Daily, G. C. (2021). Mapping the
36 benefits of nature in cities with the InVEST software. *Npj Urban Sustainability*, 1(1).
37 <https://doi.org/10.1038/s42949-021-00027-9>
- 38 Henderson, J. V., Venables, A. J., Regan, T., & Samsonov, I. (2016). Building functional cities.
39 *Science*, 352(6288).
- 40 Hilgers, T. (2008). *Perspective of Political Clientelism : Mexico ' s PRD in Comparative*
41 *Perspective*. 50(4), 123–153.
- 42 Hofmann, P., Taubenböck, H., & Werthmann, C. (2015). Monitoring and modelling of informal
43 settlements - A review on recent developments and challenges. *2015 Joint Urban Remote*
44 *Sensing Event, JURSE 2015*, 27–30. <https://doi.org/10.1109/JURSE.2015.7120513>
- 45 Irwin, E. G. (2010). New directions for urban economic models of land use change:
46 Incorporating spatial dynamics and heterogeneity. *Journal of Regional Science*, 50(1), 65–
47 91. <https://doi.org/10.1111/j.1467-9787.2009.00655.x>
- 48 Irwin, E. G., Jayaprakash, C., & Munroe, D. K. (2009). Towards a comprehensive framework for
49 modeling urban spatial dynamics. *Landscape Ecology*, 24(9), 1223–1236.
50 <https://doi.org/10.1007/s10980-009-9353-9>
- 51 Isikdogan, F., Bovik, A. C., & Passalacqua, P. (2017). Surface water mapping by deep learning.

- 1 *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*,
2 10(11), 4909–4918. <https://doi.org/10.1109/JSTARS.2017.2735443>
- 3 Jain, M. (2020). The Benefits and Pitfalls of Using Satellite Data for Causal Inference. *Review of*
4 *Environmental Economics and Policy*, 14(1), 157–169.
5 <https://doi.org/10.1093/reep/rez023>
- 6 Jean, N., Burke, M., Xie, M., Davis, W. M., Lobell, D. B., & Ermon, S. (2016). Combining satellite
7 imagery and machine learning to predict poverty. *Science (New York, N.Y.)*, 353(6301),
8 790–794. <https://doi.org/10.1126/science.aaf7894>
- 9 Jones, G. A., & Ward, P. M. (1998). Privatizing the commons: reforming the ejido and urban
10 development in Mexico. *International Journal of Urban and Regional Research*, 22(1),
11 76–93. <https://doi.org/10.1111/1468-2427.00124>
- 12 Kerins, P., Guzder-Williams, B., Mackres, E., Rashid, T., & PIETRASZKIEWICZ, E. (2021).
13 *MAPPING URBAN LAND USE IN INDIA AND MEXICO USING REMOTE SENSING AND*
14 *MACHINE LEARNING* (Issue January).
15 <https://wrirosscities.org/?fbclid=IwAR1J4FJjTPk1PmIZVJFsRhg2C9hPFO1PpUdbUjs96Y>
16 [ByHka6PNuiNraY_84](https://wrirosscities.org/?fbclid=IwAR1J4FJjTPk1PmIZVJFsRhg2C9hPFO1PpUdbUjs96Y)
- 17 Kohli, D., Sliuzas, R., Kerle, N., & Stein, A. (2012). An ontology of slums for image-based
18 classification. *Computers, Environment and Urban Systems*, 36(2), 154–163.
19 <https://doi.org/10.1016/j.compenvurbsys.2011.11.001>
- 20 Kuffer, M., Barros, J., & Sliuzas, R. V. (2014). The development of a morphological unplanned
21 settlement index using very-high-resolution (VHR) imagery. *Computers, Environment and*
22 *Urban Systems*, 48, 138–152. <https://doi.org/10.1016/j.compenvurbsys.2014.07.012>
- 23 Kuffer, M., Pfeffer, K., & Sliuzas, R. (2016). Slums from space-15 years of slum mapping using
24 remote sensing. *Remote Sensing*, 8(6). <https://doi.org/10.3390/rs8060455>
- 25 Kuras, E. R., Warren, P. S., Zinda, J. A., Aronson, M. F. J., Cilliers, S., Goddard, M. A., Nilon, C.
26 H., & Winkler, R. (2020). Urban socioeconomic inequality and biodiversity often converge,
27 but not always: A global meta-analysis. *Landscape and Urban Planning*, 198(March),
28 103799. <https://doi.org/10.1016/j.landurbplan.2020.103799>
- 29 Lambert, R. (2021). Land Trafficking and the Fertile Spaces of Legality. *International Journal*
30 *of Urban and Regional Research*, 45(1), 21–38. <https://doi.org/10.1111/1468-2427.12975>
- 31 Lang, Y., & Song, W. (2019). Quantifying and mapping the responses of selected ecosystem
32 services to projected land use changes. *Ecological Indicators*, 102(February), 186–198.
33 <https://doi.org/10.1016/j.ecolind.2019.02.019>
- 34 Lerner, A. M., Eakin, H. C., Tellman, E., Bausch, J. C., & Hernández Aguilar, B. (2018).
35 Governing the gaps in water governance and land-use planning in a megacity: The example
36 of hydrological risk in Mexico City. *Cities*, 83(June), 61–70.
37 <https://doi.org/10.1016/j.cities.2018.06.009>
- 38 Li, X., Kamarianakis, Y., Ouyang, Y., Turner, B. L., & Brazel, A. (2017). On the association
39 between land system architecture and land surface temperatures: Evidence from a Desert
40 Metropolis—Phoenix, Arizona, U.S.A. *Landscape and Urban Planning*, 163, 107–120.
41 <https://doi.org/10.1016/j.landurbplan.2017.02.009>
- 42 Lincoln Institute of Land Policy. (2016). *Atlas of Urban Expansion*. NYU, UNHABITAT.
- 43 Magliocca, N. R., McSweeney, K., Sennie, S. E., Tellman, E., Devine, J. A., Nielsen, E. A.,
44 Pearson, Z., & Wrathall, D. J. (2019). Modeling cocaine traffickers and counterdrug
45 interdiction forces as a complex adaptive system. *Proceedings of the National Academy of*
46 *Sciences of the United States of America*, 116(16), 7784–7792.
47 <https://doi.org/10.1073/pnas.1812459116>
- 48 Magliocca, N., Safirova, E., McConnell, V., & Walls, M. (2011). An economic agent-based model
49 of coupled housing and land markets (CHALMS). *Computers, Environment and Urban*
50 *Systems*, 35(3), 183–191. <https://doi.org/10.1016/j.compenvurbsys.2011.01.002>
- 51 Manson, S. M., & Evans, T. (2007). Agent-based modeling of deforestation in southern Yucatan,

1 Mexico, and reforestation in the Midwest United States. *Proceedings of the National*
2 *Academy of Sciences of the United States of America*, 104, 20678–20683.
3 <https://doi.org/10.1073/pnas.0705802104>

4 McFarlane, C. (2012). Rethinking Informality: Politics, Crisis, and the City. *Planning Theory &*
5 *Practice*, 13(1), 89–108. <https://doi.org/10.1080/14649357.2012.649951>

6 Mwau, B., & Sverdlík, A. (2020). High rises and low-quality shelter: rental housing dynamics in
7 Mathare Valley, Nairobi. *Environment and Urbanization*, 32(2), 481–502.
8 <https://doi.org/10.1177/0956247820942166>

9 Nagendra, H., Bai, X., Brondizio, E. S., & Lwasa, S. (2018). The urban south and the
10 predicament of global sustainability. *Nature Sustainability*, 1(7), 341–349.
11 <https://doi.org/10.1038/s41893-018-0101-5>

12 Narayanan, S., Rajan, A. T., Jebaraj, P., & Elayaraja, M. S. (2017). Delivering basic infrastructure
13 services to the urban poor: a meta-analysis of the effectiveness of bottom-up approaches.
14 *Utilities Policy*, 44, 50–62. <https://doi.org/10.1016/j.jup.2017.01.002>

15 North, D. C. (1990). *Institutions, institutional change and economic performance*. Cambridge
16 university press.

17 Nurko, R., Ruiz Durazo, J. E., & Gonzalez Rodriguez, J. L. (2016). Morfología y dinámica
18 familiar de la autoconstrucción en Ciudad Nezahualcóyotl: de casa unifamiliar a “vecindad”
19 familiar. In F. de Alba (Ed.), *Megalopolis, Un Debate a Voces Diferentes* (pp. 111–141).
20 CESOP.

21 Ordway, E. M., Naylor, R. L., Nkongho, R. N., & Lambin, E. F. (2019). Oil palm expansion at the
22 expense of forests in Southwest Cameroon associated with proliferation of informal mills -
23 Supplementary Material. *Nature Communications*, 1, 1–8.
24 <https://doi.org/10.1038/s41467-018-07915-2>

25 Ostrom, E. (2011). Background on the Institutional Analysis and Development Framework.
26 *Policy Studies Journal*, 39(1), 7–27. <https://doi.org/10.1111/j.1541-0072.2010.00394.x>

27 Owen, K. K., & Wong, D. W. (2013). An approach to differentiate informal settlements using
28 spectral, texture, geomorphology and road accessibility metrics. *Applied Geography*, 38(1),
29 107–118. <https://doi.org/10.1016/j.apgeog.2012.11.016>

30 Pacheco, D. F., York, J. G., Dean, T. J., & Sarasvathy, S. D. (2010). The Coevolution of
31 Institutional Entrepreneurship: A Tale of Two Theories. *Journal of Management*, 36(4),
32 974–1010. <https://doi.org/10.1177/0149206309360280>

33 Patel, A., Crooks, A., & Koizumi, N. (2012). Slumulation : An Agent-Based Modeling Approach
34 to Slum Formations. *Journal of Artificial Society and Simulation*, 15(2012), 1–15.

35 Pontius, R. G., Boersma, W., Castella, J. C., Clarke, K., Nijs, T., Dietzel, C., Duan, Z., Fotsing, E.,
36 Goldstein, N., Kok, K., Koomen, E., Lippitt, C. D., McConnell, W., Mohd Sood, A.,
37 Pijanowski, B., Pithadia, S., Sweeney, S., Trung, T. N., Veldkamp, A. T., & Verburg, P. H.
38 (2008). Comparing the input, output, and validation maps for several models of land
39 change. *Annals of Regional Science*, 42(1), 11–37. [https://doi.org/10.1007/s00168-007-](https://doi.org/10.1007/s00168-007-0138-2)
40 [0138-2](https://doi.org/10.1007/s00168-007-0138-2)

41 Pradilla Cobos, E. (1995). El mito neoliberal de la informalidad urbana. *Ciudad*.

42 Rodriguez Lopez, J. M., Heider, K., & Scheffran, J. (2017). Frontiers of urbanization: Identifying
43 and explaining urbanization hot spots in the south of Mexico City using human and remote
44 sensing. *Applied Geography*, 79, 1–10. <https://doi.org/10.1016/j.apgeog.2016.12.001>

45 Roy, A. (2005). Urban Informality: Toward an Epistemology of Planning. *Journal of the*
46 *American Planning Association*, 71(2), 147–158.
47 <https://doi.org/10.1080/01944360508976689>

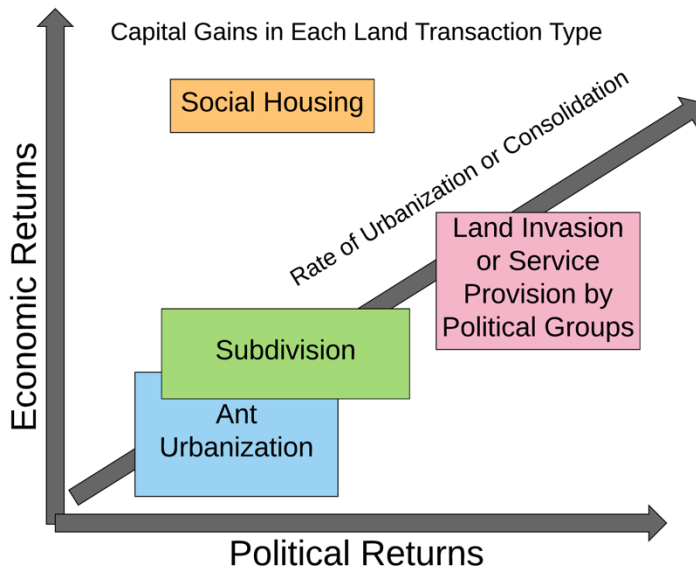
48 Roy, A. (2009). Why India Cannot Plan Its Cities: Informality, Insurgence and the Idiom of
49 Urbanization. *Planning Theory*, 8(1), 76–87. <https://doi.org/10.1177/1473095208099299>

50 Roy, D., Lees, M. H., Palavalli, B., Pfeffer, K., & Sloot, M. A. P. (2014). The emergence of slums:
51 A contemporary view on simulation models. *Environmental Modelling and Software*,

- 1 59(2014), 76–90. <https://doi.org/10.1016/j.envsoft.2014.05.004>
- 2 Santos, C. (2013). Interacciones y tensiones entre la expansión urbana y el Suelo de
3 Conservación. *Adrián Aguilar e Irma Escamilla (Coords.), La Sustentabilidad En La*
4 *Ciudad de México. El Suelo de Conservación En El Distrito Federal. México:*
5 *UNAM/Miguel Ángel Porrúa Ediciones.*
- 6 Schmaeler, L. (2016). More People More Slum: Venezuela’s Struggle with Urbanization.
7 *International Affairs Review Journal.* [https://www.usfca.edu/journal/international-](https://www.usfca.edu/journal/international-affairs-review/spring-2015/venezuelas-struggle-with-urbanization)
8 [affairs-review/spring-2015/venezuelas-struggle-with-urbanization](https://www.usfca.edu/journal/international-affairs-review/spring-2015/venezuelas-struggle-with-urbanization)
- 9 Sesnie, S. E., Tellman, B., Wrathall, D., McSweeney, K., Nielsen, E., Benessaiah, K., Wang, O., &
10 Rey, L. (2017). A spatio-temporal analysis of forest loss related to cocaine trafficking in
11 Central America. *Environmental Research Letters*, 12(5), 054015.
12 <https://doi.org/10.1088/1748-9326/aa6fff>
- 13 Seto, K. C., Fragkias, M., Güneralp, B., & Reilly, M. K. (2011). A Meta-Analysis of Global Urban
14 Land Expansion. *PLOS ONE*, 6(8), e23777. <https://doi.org/10.1371/journal.pone.0023777>
- 15 Seto, K. C., & Reenberg, A. (2014). *Rethinking global land use in an urban era.* MIT Press.
- 16 Solecki, W., Seto, K. C., & Marcotullio, P. J. (2013). It’s Time for an Urbanization Science.
17 *Environment: Science and Policy for Sustainable Development*, 55(1), 12–17.
18 <https://doi.org/10.1080/00139157.2013.748387>
- 19 Stuhlmacher, M., Andrade, R., Turner, B. L., Frazier, A., & Li, W. (2020). Environmental
20 Outcomes of Urban Land System Change: Comparing Riparian Design Approaches in the
21 Phoenix Metropolitan Area. *Land Use Policy*, 99(July 2019), 104615.
22 <https://doi.org/10.1016/j.landusepol.2020.104615>
- 23 Tan, P. Y., Zhang, J., Masoudi, M., Alemu, J. B., Edwards, P. J., Grêt-Regamey, A., Richards, D.
24 R., Saunders, J., Song, X. P., & Wong, L. W. (2020). A conceptual framework to untangle
25 the concept of urban ecosystem services. *Landscape and Urban Planning*, 200(September
26 2019), 103837. <https://doi.org/10.1016/j.landurbplan.2020.103837>
- 27 Tellman, B., Eakin, H., Janssen, M. A., Alba, F. De, & Ii, B. L. T. (2021). The Role of Institutional
28 Entrepreneurs and Informal Land Transactions in Mexico City’s Urban Expansion. *World*
29 *Development*, 140, 1–44. <https://doi.org/10.1016/j.worlddev.2020.105374>
- 30 Tellman, B., Magliocca, N. R., Turner, B. L., & Verburg, P. H. (2020). Understanding the role of
31 illicit transactions in land-change dynamics. *Nature Sustainability*, 3(3), 175–181.
32 <https://doi.org/10.1038/s41893-019-0457-1>
- 33 Tellman, B., Sesnie, S. E., Magliocca, N. R., Nielsen, E. A., Devine, J. A., McSweeney, K., Jain,
34 M., Wrathall, D. J., Dávila, A., Benessaiah, K., & Aguilar-Gonzalez, B. (2020). Illicit Drivers
35 of Land Use Change: Narcotrafficking and Forest Loss in Central America. *Global*
36 *Environmental Change*, 63. <https://doi.org/10.1016/j.gloenvcha.2020.102092>
- 37 Teo, H. C., Zeng, Y., Sarira, T. V., Fung, T. K., Zheng, Q., Song, X. P., Chong, K. Y., & Koh, L. P.
38 (2021). Global urban reforestation can be an important natural climate solution.
39 *Environmental Research Letters*, 16(3). <https://doi.org/10.1088/1748-9326/abe783>
- 40 Tuholske, C., Caylor, K., Funk, C., Verdin, A., Sweeney, S., Grace, K., Peterson, P., & Evans, T.
41 (2021). Global urban population exposure to extreme heat. *Proceedings of the National*
42 *Academy of Sciences of the United States of America*, 118(41), 1–9.
43 <https://doi.org/10.1073/pnas.2024792118>
- 44 Turner, B. L., Janetos, A. C., Verburg, P. H., & Murray, A. T. (2013). Land system architecture:
45 Using land systems to adapt and mitigate global environmental change. *Global*
46 *Environmental Change*, 23(2), 395–397. <https://doi.org/10.1016/j.gloenvcha.2012.12.009>
- 47 Turner II, B. L. (2017). Land system architecture for urban sustainability: new directions for
48 land system science illustrated by application to the urban heat island problem. *Journal of*
49 *Land Use Science*, 12.
- 50 UN Habitat. (2016). *INFORMAL SETTLEMENTS* (Issue .).
51 <https://doi.org/10.3402/gha.v5i0.19065>

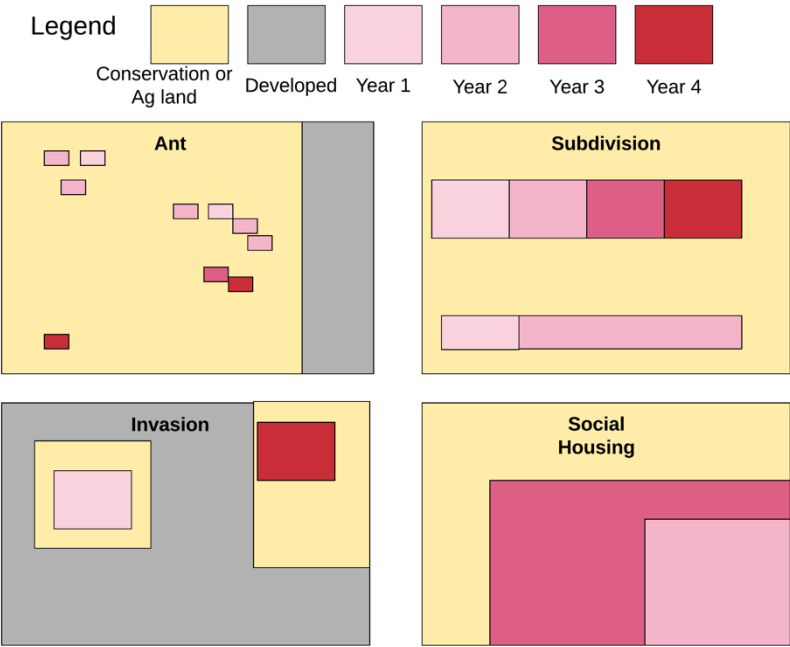
1 Van Gelder, J. L. (2013). Paradoxes of Urban Housing Informality in the Developing World.
2 *Law and Society Review*, 47(3), 493–522. <https://doi.org/10.1111/lasr.12030>
3 Varley, A. (1998). *The political uses of illegality: evidence from urban Mexico*. Zed Books.
4 Venkatachalam, L. (2015). Informal water markets and willingness to pay for water: a case study
5 of the urban poor in Chennai City, India. *International Journal of Water Resources*
6 *Development*, 31(1), 134–145. <https://doi.org/10.1080/07900627.2014.920680>
7 Vermeiren, K., Van Rompaey, A., Loopmans, M., Serwajja, E., & Mukwaya, P. (2012). Urban
8 growth of Kampala, Uganda: Pattern analysis and scenario development. *Landscape and*
9 *Urban Planning*, 106(2), 199–206. <https://doi.org/10.1016/j.landurbplan.2012.03.006>
10 Wang, S., Chen, W., Xie, S. M., Azzari, G., & Lobell, D. B. (2020). Weakly supervised deep
11 learning for segmentation of remote sensing imagery. *Remote Sensing*, 12(2), 1–25.
12 <https://doi.org/10.3390/rs12020207>
13 Ward, P. M. (1976). The Squatter Settlement as Slum or Housing Solution: Evidence from
14 Mexico City. *Land Economics*, 52(3), 330–346. <https://doi.org/10.2307/3145530>
15 Wheeler, H., & Evans, E. (2009). Land use , water management and future flood risk. *Land Use*
16 *Policy*, 251–264. <https://doi.org/10.1016/j.landusepol.2009.08.019>
17 Xiang, X., Qiu, C., Hu, J., Shi, Y., Wang, Y., Schmitt, M., & Taubenb, H. (2022). *Remote Sensing*
18 *of Environment The urban morphology on our planet – Global perspectives from space.*
19 269(November 2021), 1–11. <https://doi.org/10.1016/j.rse.2021.112794>
20 Zhang, Z., Liu, Q., & Wang, Y. (2018). Road Extraction by Deep Residual U-Net. *IEEE*
21 *Geoscience and Remote Sensing Letters*, 15(5), 749–753.
22 <https://doi.org/10.1109/LGRS.2018.2802944>

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24 Figures:
25 1.



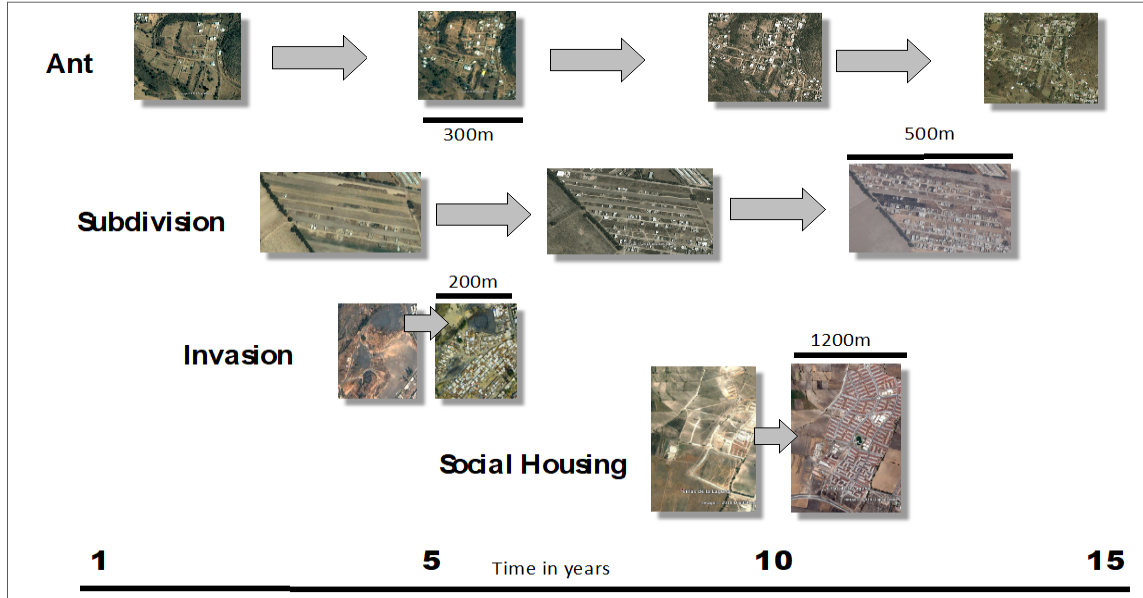
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6 **Figure 1.** Economic and political capital returns in each informal land transaction type in the
7 Mexico City metropolis, and a hypothesized vector of how rate of change is orthogonally and
8 positively correlated, adapted from Tellman et al 2021. Social housing refers to large housing
9 complexes funded by INFONAVIT on the fringe of the metropolitan area.

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12 **Figure 2.** Hypothesized spatial patterns of informal urban development, illustrated from
13 textual interview analysis (Tellman et al., 2021). Yellow represents existing undeveloped land (in
14 Mexico City, often conservation land or ejido agricultural land, except for invasions that may
15 “infill” empty lots or public spaces such as garbage dumps in the urban core), grey is existing
16 urban land cover. New informal urban settlement is displayed in pink to red, with shades
17 representing year of development.

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20 **Figure 3.** Empirical examples of each informal urbanization type by hypothesized spatial
21 pattern and speed of development, starting from the first image date, when initial signs of urban
22 land cover change take place (e.g. visible homes, roads, or other impervious surface) to the final
23 image date, when consolidation has finished and thereafter no additional visible roads and
24 homes were added.