

GeoAI and Urban Justice: An Equity-Centered Governance Model for Redeveloping the Gowanus Canal

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ABSTRACT: The rapid redevelopment of New York City's Gowanus Canal neighborhood poses a critical question: Can geospatial artificial intelligence (GeoAI) be governed in a way that promotes environmental justice, or will it exacerbate urban inequities through eco-gentrification? This conceptual paper introduces the Equity-Centered GeoAI Governance Model, which combines algorithmic fairness audits, affordability safeguards, and participatory oversight to address displacement risks associated with AI-driven zoning. Using the Gowanus Canal as an illustrative case, the paper situates local dynamics within broader international examples—including Amsterdam, Helsinki, and Vienna—which demonstrate partial but incomplete advances in transparency, affordability, and community participation. The framework synthesizes lessons from urban planning theory, critical geography, and digital governance to propose a three-phase policy roadmap: (1) transparent auditing of algorithmic zoning tools; (2) mandatory affordability mechanisms, such as Community Land Trusts; and (3) robust resident engagement via open data and digital democracy platforms. While conceptual in scope, this model highlights pathways for cities to align smart urban redevelopment with social equity goals and suggests directions for future empirical testing.

KEYWORDS: geospatial artificial intelligence (GeoAI), Gowanus canal, environmental justice, participatory governance, social equity

1. Introduction

Urban sustainability initiatives are often promoted as solutions for environmental degradation and socioeconomic inequality. Yet, green investments can unintentionally lead to eco-gentrification: environmental upgrades that raise property values and displace long-term, lower-income residents (Anguelovski et al., 2018; Curran & Hamilton, 2022). This risk intensifies when planners deploy GeoAI-driven zoning algorithms that optimize stormwater resilience and fiscal

returns but neglect equity safeguards, steering capital toward already affluent areas and amplifying displacement pressures in historically underserved communities.

This paradox is especially evident in the redevelopment surrounding New York City's Gowanus Canal Superfund. Since 2015, median condominium prices within the Gowanus rezoning footprint have risen sharply, reaching approximately \$1.8 million in mid-2025, well above the Brooklyn borough-wide median of around \$857,000 for the same period. This steep price growth underscores how sustainability investments and rezoning policies can exacerbate speculative real estate pressures if left unchecked. Without algorithmic fairness audits, enforceable affordability mandates, and the expansion of community land trusts (CLTs), GeoAI risks becoming a catalyst for speculative redevelopment rather than a lever for socially just urban renewal (PropertyShark, 2025).

As a conceptual study, this paper introduces the Equity-Centered GeoAI Governance Model, which combines algorithmic fairness audits, affordability-based inclusionary zoning, and community-led oversight committees to help prevent eco-gentrification. We employ comparative policy-scenario illustrations to benchmark Gowanus against equity safeguards adopted in Amsterdam's Eastern Docklands, Vienna's Donaukanal, and Helsinki's Kalasatama. Yet, despite scattered safeguards, such as model audits in Amsterdam, affordability mandates in Vienna, and participatory platforms in Helsinki, no city has integrated these pillars into a single, enforceable governance framework explicitly calibrated for urban AI equity. This gap highlights an urgent need for a replicable model that combines algorithmic transparency, robust affordability mandates, and community-led oversight to prevent eco-gentrification. This paper addresses this gap by proposing the Equity-Centered GeoAI Governance Model and outlining a practical roadmap for its implementation in the Gowanus Canal redevelopment and other cities worldwide.

The sections that follow develop this conceptual framework; they do not present new empirical analyses.

2. The Need for an Equity-Centered GeoAI Governance Model

Current urban-governance frameworks regulate the technical performance of GeoAI systems (accuracy, speed) far more rigorously than their social outcomes. As a result, AI-driven zoning decisions often exacerbate long-standing inequities through three closely linked problem areas.

2.1. Algorithmic Bias & Unequal Resource Allocation

Predictive zoning tools often reproduce historic redlining patterns, channeling infrastructure, green-space upgrades, and public services toward higher-income districts while leaving marginalized neighborhoods underserved (Green, 2019; Leszczynski, 2023). Limited model transparency compounds the problem:

community stakeholders cannot audit, contest, or improve biased recommendations when training data, feature weights, and evaluation metrics are withheld.

2.2. Affordability & Displacement Risk

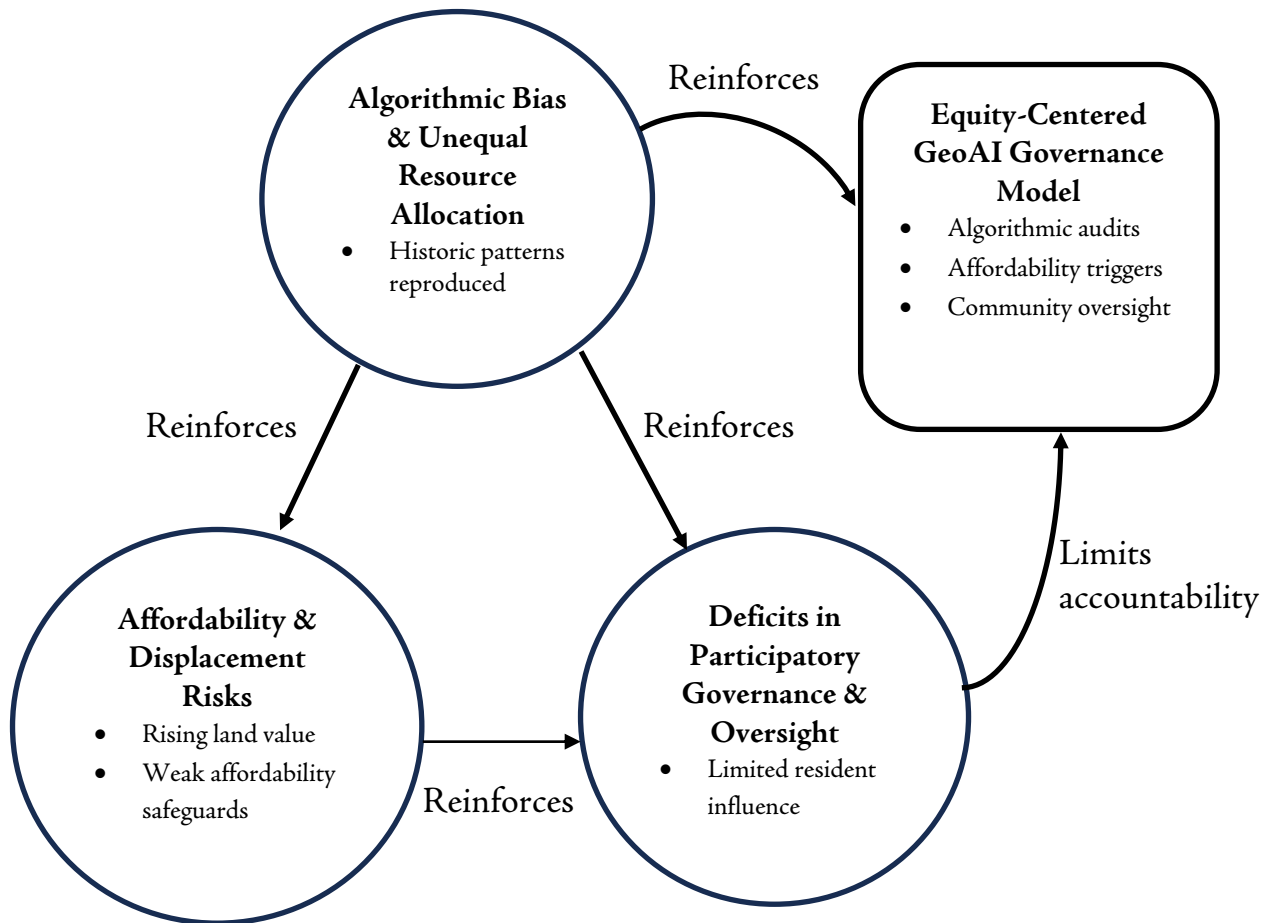
Profit-maximizing real estate algorithms often privilege large-scale developers over the housing stability of existing residents. Unregulated smart-city redevelopment and green upgrades frequently drive substantial property value increases in redeveloped corridors, putting additional pressure on lower-income households and contributing to involuntary displacement (Anguelovski et al., 2018; Curran & Hamilton, 2012). Weak affordability safeguards—such as inclusionary zoning quotas, rent stabilization rules, and Community Land Trust (CLT) expansions—allow speculative capital to dominate post-rezoning markets.

2.3. Deficits in Participatory Governance & Oversight

Decision authority often remains concentrated in private consultancies or top-down municipal agencies, which limits meaningful resident influence over zoning choices (Bua & Bussu, 2020). Despite the promise of digital-democracy tools such as Barcelona's Decidim platform, practical implementations often reveal gaps in deliberative quality and accessibility, suggesting these platforms remain underutilized in urban planning practice (Aragón et al., 2017). Most GeoAI zoning systems still lack end-to-end audit trails or public transparency mechanisms, making it difficult for residents and policymakers to trace how data inputs (e.g., land value forecasts) translate into outputs (e.g., density bonuses) and neighborhood-level impacts (Green, 2019).

These three deficits reinforce one another: biased algorithms allocate resources inequitably, rising land values exacerbate affordability crises, and limited participatory checks hinder course correction. The cycle reveals a critical policy gap: mainstream planning practice fails to integrate algorithmic transparency, affordability protections, and participatory governance into a single, enforceable framework. A handful of cities—Amsterdam, Vienna, and Helsinki—have piloted discrete safeguards (open-source model audits, mandatory affordability mandates, neighborhood data trusts). Yet no jurisdiction has consolidated these elements into a comprehensive governance architecture explicitly calibrated to social equity goals. Closing this gap requires an innovative, multi-pillar approach that aligns GeoAI deployment with community well-being, rather than speculative market interests—precisely the aim of the Equity-Centered GeoAI Governance Model outlined in the sections that follow.

Figure 1. Equity-Centered GeoAI Governance Model – Conceptual Diagram



Cycle of Reinforcing Deficits in GeoAI-Driven Urban Planning and the Disruptive Role of Equity-Centered Governance

Note. This conceptual diagram illustrates how algorithmic bias, affordability risk, and governance deficits form a self-reinforcing cycle that drives eco-gentrification. The Equity-Centered GeoAI Governance Model disrupts this cycle through algorithmic audits, affordability mandates, and participatory oversight.

3. Research Questions

Building on the Gowanus Canal case, this conceptual study asks:

1. **Displacement Dynamics.** How do GeoAI-driven zoning decisions affect housing affordability, measured through rent-to-income ratios, sales prices, and eviction filings, among marginalized residents during redevelopment?
2. **Transferable Safeguards.** Which international practices—Amsterdam's model audits, Vienna's affordability mandates, and Helsinki's co-governance platforms—offer the most relevant design principles for equity-focused GeoAI?
3. **Model Operationalization.** How can the Equity-Centered GeoAI Governance Model be implemented to prioritize housing stability, community co-governance, and environmental justice in complex redevelopment contexts?

4. Policy Efficacy. Among candidate interventions, bias audits, transparency dashboards, and Community Land Trust (CLT) expansion, which prove most scalable for mitigating algorithmic exclusion in zoning workflows?

Answering these questions will underpin a governance framework that steers AI-enabled planning toward inclusive, justice-oriented urban transformation rather than reinforcing socioeconomic exclusion.

4. Framework-Development Process

This study employed a three-step conceptual design protocol aimed at generating, rather than empirically testing, an equity-oriented GeoAI governance model. Each step is described below to ensure methodological transparency without implying new data collection or statistical analysis.

4.1. Structured Literature Review (2015–2025)

Using Scopus, Web of Science, and Google Scholar, we identified approximately 70 peer-reviewed sources organized into thematic clusters (see Appendix A). Sources were included if they met three criteria: (a) an explicit focus on eco-gentrification, AI-enabled urban planning, or digital governance; (b) a clear discussion of distributive or procedural justice; and (c) publication between 2015 and 2025.

4.2. Policy Scan of Benchmark Cities

We reviewed official zoning ordinances, housing statutes, and open-data portals from Amsterdam, Vienna, and Helsinki to catalog existing safeguards, including model audit mandates, inclusionary zoning thresholds, and neighborhood data trusts. Documents were triangulated with municipal white papers and NGO reports to capture both formal rules and practical lessons on implementation.

4.3. Comparative Synthesis

Findings from the literature review and policy scan were cross-tabulated to isolate standard safeguards across contexts: (i) algorithm-transparency audits, (ii) affordability mandates, and (iii) co-governance mechanisms. This matrix served as the evidentiary backbone for the proposed three-pillar governance architecture. Throughout this process, no primary data were collected, and no statistical models were estimated. The resulting framework is therefore conceptual and agenda-setting, synthesizing existing knowledge and best practices to guide future empirical validation and policy experimentation.

5. Theoretical Foundations and Framework Integration

Equitable urban renewal necessitates a multifaceted strategy that integrates economic, political, and technological frameworks to effectively address systemic inequalities in urban development. To understand and address the socioeconomic

consequences of AI-driven urban planning, this study draws on theoretical perspectives from critical urban geography, political philosophy, and emerging literature on algorithmic governance.

David Harvey's theory of spatial dialectics (Harvey, 2003) offers crucial insights into how urban development often prioritizes capital accumulation and economic interests over social equity. This dynamic is exacerbated within the context of AI-driven zoning models, where predictive GeoAI algorithms, primarily designed for optimizing real estate investments and infrastructure planning, can reinforce historical exclusionary patterns. Such algorithms tend to disproportionately direct urban improvements and resources toward affluent neighborhoods, marginalizing lower-income districts. This digital manifestation of Harvey's spatial dialectics highlights how contemporary urban spaces are increasingly shaped not just by capital directly, but by algorithmically mediated decision-making processes that obscure political accountability and transparency.

Complementing Harvey's economic critique, Giorgio Agamben's concept of "bare life" (Agamben, 1998) critiques the role governance structures play in marginalizing vulnerable populations by stripping them of political agency and subjecting them to structural exclusion. In the context of AI-driven urban planning, Agamben's theory becomes particularly salient. Zoning algorithms operating as opaque, "black-box" systems diminish residents' capacities to contest, influence, or even fully understand urban policy decisions, effectively relegating marginalized communities to passive subjects of redevelopment rather than empowered participants. Consequently, algorithmic governance has the potential to either exacerbate displacement and exclusion or actively challenge such dynamics, depending on how resource allocation, affordability, and participatory decision-making are prioritized.

The combination of Harvey's economic restructuring lens and Agamben's governance critique provides a robust theoretical framework for evaluating the socioeconomic implications of GeoAI. Contemporary literature on GeoAI and algorithmic governance (Green, 2019; Leszczynski, 2023) reinforces these concerns, emphasizing that unregulated AI zoning models risk amplifying existing socioeconomic disparities rather than mitigating them. Addressing these risks necessitates the development of comprehensive governance frameworks that mandate algorithmic transparency, equity-driven zoning policies, and meaningful participatory oversight, explicitly designed to ensure digital technologies contribute positively to spatial justice rather than exacerbating systemic inequities. Thus, by integrating these critical theoretical frameworks, this study provides the conceptual foundation necessary for developing the Equity-Centered GeoAI Governance Model, designed explicitly to counteract algorithmically mediated spatial injustices.

6. Equity-Centered GeoAI Governance Model

The Equity-Centered GeoAI Governance Model offers an interdisciplinary framework specifically designed to address eco-gentrification by integrating advanced technological tools, robust affordability protections, and meaningful participatory governance. By bridging technological innovations with equitable urban policy, the model ensures that sustainability initiatives actively prevent displacement and reduce socioeconomic inequalities.

6.1. GeoAI-Driven Environmental Monitoring

IoT sensor networks and AI-driven spatial analytics systematically track real-time environmental conditions, guiding infrastructure investments that promote environmental justice and mitigate spatial inequities (Alotaibi & Nassif, 2024). Open-access environmental dashboards offer communities transparent, real-time access to AI-guided zoning and infrastructure decisions, empowering residents to challenge inequitable resource allocation. AI-powered predictive analytics support the equitable distribution of environmental remediation efforts, counteracting patterns that often favor affluent neighborhoods. Community-informed impact assessments provide residents with direct, timely feedback on proposed developments, ensuring responsiveness and local relevance.

6.2. Affordability Protections

Mandatory inclusionary zoning policies ensure that a defined share of new housing remains permanently affordable, explicitly safeguarding housing stability for marginalized communities. Community Land Trusts (CLTs) and similar inclusionary housing schemes secure long-term affordability by removing key land parcels from speculative markets and embedding stewardship in community governance (Grounded Solutions Network, 2023). These models have gained momentum both globally and domestically. In the U.S., the 2022 CLT Census reported 113 responding entities with a 62% response rate, and by early 2024, over 300 active CLTs were recorded nationwide. AI-driven rent stabilization frameworks that dynamically adjust affordability requirements based on predictive displacement risk offer a promising complement to fixed quotas, enhancing flexibility and local responsiveness (Salazar-Miranda & Talen, 2025).

6.3. Participatory Governance

AI-assisted citizen engagement platforms democratize urban planning, enabling direct and accessible participation from diverse community members in zoning and redevelopment decisions. Algorithmic transparency mandates legally require public disclosure and routine fairness audits of AI zoning models (Bua & Bussu, 2020), empowering communities with critical information to effectively challenge or support development plans. Community-led oversight boards, composed of diverse resident stakeholders, oversee redevelopment initiatives, ensuring planning

priorities reflect genuine local needs rather than exclusively corporate or investor interests.

Interaction and Integration: Together, these components create a reinforcing cycle where technological transparency supports effective community governance, affordability protections anchor equitable outcomes, and active participation ensures local responsiveness. Implementing this model requires addressing critical barriers such as initial financial investment, fostering digital literacy among community members, and overcoming political resistance from vested interests. Nonetheless, clearly defined success metrics, such as reductions in displacement rates, enhanced community engagement levels, and demonstrable improvements in equitable resource allocation, provide a robust evaluative framework for measuring effectiveness. By systematically integrating these elements, the Equity-Centered GeoAI Governance Model provides a comprehensive, scalable, and practical approach for cities worldwide seeking to align AI-driven urban planning innovations with enduring social equity and environmental justice.

7. Implementation Roadmap: Gowanus Canal

The accelerated clean-up and rezoning of the canal have raised property values, local news outlets estimate increases of 25–40 percent, and intensified displacement pressures for long-standing renters and small-business owners. The ongoing redevelopment of the Gowanus Canal in New York City presents an ideal case study for examining the real-world implications of AI-driven zoning on affordability, displacement, and community agency. As one of New York's largest and most high-profile urban renewal projects, the Gowanus Canal transformation has been publicly presented as a pioneering model for eco-friendly urban planning, utilizing advanced technologies including AI-enhanced land-use decision-making, real estate forecasting, and predictive analytics. Despite these promising objectives, the lack of explicit equity safeguards has inadvertently reinforced speculative investment cycles, significantly raised property values (by an estimated 25–40%), and intensified displacement pressures for lower-income residents and local small businesses. To confront and mitigate these urgent challenges, the Equity-Centered GeoAI Governance Model proposes a structured, multi-phase intervention strategy tailored explicitly to Gowanus's complex redevelopment landscape:

- **Phase 1: Algorithmic Transparency**
The first step establishes real-time AI fairness audits and transparency dashboards to empower residents to scrutinize and shape zoning decisions.
- **Phase 2: Affordability Mandates**
The second step strengthens housing security by enforcing mandatory inclusionary zoning and expanding Community Land Trusts (CLTs) to protect vulnerable parcels from speculative development.

- **Phase 3: Participatory Governance**
Finally, local oversight boards and digital participation platforms ensure that redevelopment priorities reflect community needs rather than market interests.

By explicitly prioritizing affordability, meaningful participation, and technological transparency, this intervention strategy actively prevents AI systems from merely facilitating speculative real estate investments. Instead, it positions GeoAI technologies as integral tools for achieving genuine housing justice, environmental resilience, and sustainable, equitable community development.

7.1. Implementation Considerations

The successful implementation of this multi-phase strategy requires addressing anticipated barriers, such as political resistance, developer opposition to affordability mandates, and potential community skepticism regarding technological interventions. Addressing these concerns proactively, through clear policy incentives (e.g., tax credits, density bonuses for developers), robust legal mandates for transparency, and extensive community education initiatives, will be critical for ensuring practical feasibility and broad stakeholder support.

7.2. Evaluating Outcomes

Clearly defined success metrics, such as reduced displacement rates, increased affordable housing stock, demonstrably higher community engagement levels, and transparent accountability in zoning decisions, will provide critical evaluative benchmarks, ensuring continuous improvement and accountability in Gowanus's redevelopment process.

8. Literature Review

Recent scholarship exploring eco-gentrification, urban renewal, and spatial justice has deepened our understanding of how sustainability-driven urban redevelopment intersects with socioeconomic inequalities and technological advancements. Foundational theories, such as David Harvey's (2006) spatial dialectics, critically highlight tensions between capital-driven urban development and displacement pressures, emphasizing how prioritizing economic growth can often result in adverse outcomes for marginalized communities. Complementing this perspective, Giorgio Agamben's (1998) theory of "bare life" further critiques governance structures, revealing the systematic marginalization and exclusion faced by vulnerable populations during urban transformations. These seminal theories underpin contemporary debates on algorithmic governance, AI-driven urban planning, and zoning policies, particularly concerning biases and equity concerns (Leszczynski, 2023).

8.1. AI-Driven Urban Governance: Opportunities and Risks

Emerging studies illuminate the complex dualities inherent in AI-enhanced urban governance. Leszczynski (2023) and Green (2019) highlight the potential benefits of AI-driven zoning, including optimizing resource allocation and infrastructure planning, while simultaneously underscoring significant risks related to algorithmic bias. Specifically, AI models trained on historically biased economic and housing data can unintentionally perpetuate patterns of urban exclusion. Empirical research on automated decision systems shows that algorithmically driven infrastructure investments and predictive service targeting often reinforce wealthier districts while marginalizing lower-income areas (Eubanks, 2018). Similarly, predictive algorithms in real estate valuation and pricing can embed racially biased assumptions, producing unfair assessments and heightened eviction risks for historically marginalized neighborhoods (Noble, 2018; Barocas et al., 2019). Illustrating these risks concretely, AI-driven real estate analytics in New York City have categorized historically marginalized communities as investment risks, intensifying displacement pressures on residents (Anguelovski et al., 2018). Collectively, these findings underscore a pressing need for robust equity safeguards in AI-driven urban planning processes to prevent the exacerbation of socioeconomic inequalities.

8.2. International Approaches to AI Governance in Urban Planning

International case studies offer valuable insights into how cities are addressing the governance challenges presented by AI-driven urban redevelopment. Amsterdam and Barcelona, for example, have strengthened transparency by implementing digital democracy platforms, such as Decidim, and maintaining public AI registers that list the algorithms used for urban decision-making, thereby improving accountability and enabling citizen oversight (Aragón et al., 2017; GPAI, 2024). Singapore's Smart Nation 2.0 initiative continues to prioritize rapid economic growth through AI-enabled zoning and smart infrastructure. Still, it has raised questions about potential displacement and affordability risks if stronger social safeguards are not enacted (GovInsider, 2023). Meanwhile, Helsinki's AGAPP Project has advanced a model for algorithmic fairness audits and "human-in-the-loop" oversight to enhance the legitimacy of urban AI systems (University of Helsinki AGAPP Project, 2023; OECD, 2024). Vienna's robust social housing regime, widely regarded as the "Vienna Model," remains an influential example of how affordability-centered frameworks can counteract speculative real estate pressures, providing a template for developing AI-guided rent stabilization strategies (Amann & Mundt, 2015). Despite these promising directions, no city yet integrates algorithmic transparency, strong affordability mandates, and participatory governance into a unified, enforceable AI urban planning framework, highlighting a key gap that future governance models must address.

9. Research Gaps and the Need for an Integrated Governance Framework

This study directly addresses this critical gap by proposing the Equity-Centered GeoAI Governance Model. Designed to mitigate risks associated with AI-driven zoning, this comprehensive framework incorporates three essential components:

- (1) AI transparency and accountability mechanisms, such as algorithmic oversight committees and bias audits,
- (2) Robust housing justice measures, including mandatory inclusionary zoning, community land trusts, and adaptive rent stabilization algorithms, and
- (3) Digitally enabled participatory governance platforms, ensuring residents actively engage in and influence redevelopment decisions. By integrating these elements, the Equity-Centered GeoAI Governance Model aims to ensure that technological advances in urban planning align closely with principles of social equity, housing justice, and sustainable development, thus positioning AI as a positive force rather than a potential amplifier of urban inequality.

10. Addressing Methodological Limitations

Acknowledging potential limitations, the following mitigation strategies are implemented (see Table 1).

Table 1. Methodological Limitations and Mitigation Strategies

Limitation	Mitigation Strategy
Limited access to proprietary AI zoning models	Use open-source GeoAI datasets and publicly available municipal zoning records
Variability in AI governance across case studies	Conduct weighted policy comparisons to standardize governance variations
Potential bias in case study selection	Select diverse urban governance models, comparing top-down (Vienna) and participatory (Helsinki) approaches

Note: This study rigorously adheres to ethical standards regarding AI utilization in urban governance by ensuring:

1. Transparency in the use of AI-driven urban data analytics.
2. Equity-focused assessments addressing affordability implications.
3. Comprehensive community impact assessments are conducted to ensure the viability and ethical appropriateness of proposed policies and programs.

Scope, Limitations, and Directions for Future Research

This paper is conceptual; empirical testing of the model, e.g., using parcel-level datasets, eviction records, or agent-based simulations, remains future work. Future empirical work should employ GIS-based price mapping, eviction record analysis, and stakeholder surveys to test the hypotheses generated by this framework. Data access challenges, including limited transparency and accessibility to proprietary AI zoning models, restrict comprehensive evaluation. Future research should investigate open-source GeoAI models to examine algorithmic biases and their

effects. Regulatory variability in AI governance also influences effectiveness, so comparative studies between regulated and unregulated contexts can help clarify the optimal conditions for equitable AI deployment. Cross-sectional data provide limited insights into long-term impacts, so longitudinal research assessing fairness audits, affordability interventions, and participatory governance is recommended.

To advance this agenda, future empirical research should include:

- Parcel-level GIS simulations to test how GeoAI zoning decisions shift affordability and displacement patterns over time.
- Resident household surveys and interviews to assess perceived impacts of AI-driven redevelopment on housing security.
- Pilot implementation of algorithmic fairness audits, measuring zoning outcomes before and after audit integration.
- Comparative case studies across diverse cities to validate the framework's transferability in different governance contexts.
- Longitudinal tracking of displacement trends, rent stabilization effectiveness, and community trust in AI governance systems.

11. Conclusion

Unchecked AI-driven urban planning amplifies economic displacement and exacerbates spatial inequalities. Effective mitigation requires real-time AI fairness audits, enforceable affordability protections, and robust participatory governance structures. To this end, this paper proposes the Equity-Centered GeoAI Governance Model, an integrated framework designed to align next-generation urban analytics with housing justice and inclusive redevelopment.

Key policy recommendations include:

- Mandating transparency in AI zoning decisions to prevent algorithm-driven exclusion.
- Implementing and enforcing affordability mandates within AI-driven urban renewal projects.
- Establishing real-time algorithmic bias audits to address spatial inequities proactively.
- Strengthening participatory governance to empower community stakeholders in decision-making processes.

As a conceptual contribution, this framework sets a clear agenda for future empirical testing, including parcel-level GIS simulations, resident surveys, fairness audit pilots, comparative studies, and longitudinal tracking. By combining innovative policy safeguards, ethical AI governance, and collaborative stakeholder engagement, cities can harness AI technology as a tool for equitable, sustainable urban development, rather than a driver of displacement and exclusion.

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Appendix A

Conceptual Framework Source Clusters

Cluster	Sample Sources	Focus
Eco-Gentrification & Justice	Curran & Hamilton (2012); Anguelovski et al. (2018)	How green infrastructure upgrades can trigger displacement: balancing sustainability and equity
GeoAI & Algorithmic Zoning	Leszczynski (2023); GPAI (2024); Salazar-Miranda & Talen (2025) [Preprint]	How AI and algorithmic tools affect zoning, land valuation, and bias audits
Digital Democracy & Participation	Aragón et al. (2017); Bua & Bussu (2020)	Participatory platforms (Decidim), co-design workshops, governance-driven democratization
IoT & Real-Time Monitoring	Alotaibi & Nassif (2024); PropertyShark (2025)	Sensor networks, open data, and real-time local environmental monitoring for accountability
Housing & Affordability Tools	Amann & Mundt (2015); Grounded Solutions Network (2023)	Inclusionary zoning, Community Land Trusts (CLTs), and rent stabilization
Theoretical Foundations	Harvey (1989, 1996, 2006, 2019); Agamben (1998)	Spatial dialectics, urban inequality, bare life, and governance theory
Algorithmic Bias & Ethics	Eubanks (2018); Noble (2018); Barocas et al. (2019)	Structural bias in algorithms, fairness audits, and accountability frameworks