

## Geo-Analytics and Eco-Gentrification: A Framework for Equitable Resilience in the Gowanus Canal

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**ABSTRACT:** This conceptual paper examines the intersection of Geospatial Artificial Intelligence (GeoAI), Internet of Things (IoT) technologies, and urban climate resilience through a justice-centered lens. Using the Gowanus Canal Superfund redevelopment in Brooklyn, New York, as a focal case, the paper explores how AI-driven zoning models shape environmental outcomes and socioeconomic equity in post-industrial cities. Drawing on interdisciplinary literature in sustainability, planning, and digital governance, this analysis examines tools such as predictive modeling, equity mapping, and algorithmic audits to evaluate displacement risk, eco-gentrification, and the reproduction of spatial inequality. The paper proposes a conceptual mixed-methods framework that integrates geospatial analytics, participatory zoning, and policy audits, supported by comparative case studies of the New York City High Line and the Atlanta BeltLine. The study presents a novel governance model that emphasizes transparency protocols, affordable housing mandates, and community-led planning that is inclusive. It argues that without inclusive governance mechanisms, GeoAI and IoT innovations may deepen socio-spatial disparities despite their potential to enhance environmental resilience. The framework provides policymakers with measurable equity indicators, including distributional and procedural justice metrics, to assess and mitigate the unintended consequences of smart redevelopment. By synthesizing environmental technology with anti-displacement strategies, this research offers a roadmap for ethical and inclusive climate adaptation in urban planning.

**KEYWORDS:** geospatial artificial intelligence, eco-gentrification, risk management, climate adaptation, algorithmic governance, housing equity

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### **Purpose of Study**

This conceptual study proposes a framework for evaluating how Geospatial Artificial Intelligence (GeoAI) and Internet of Things (IoT) technologies can enhance urban climate resilience while mitigating the risks of eco-gentrification. Focusing on the redevelopment of the Gowanus Canal in Brooklyn, the study examines the technological, environmental, and social implications of AI-driven sustainability initiatives.

The specific objectives of this study are to:

1. Assess the effectiveness of GeoAI and IoT in flood risk management, pollution mitigation, and environmental monitoring within urban redevelopment zones.
2. Analyze the unintended socioeconomic consequences of AI-integrated urban planning, particularly displacement, rising property values, and affordability crises.
3. Identify equity-centered policy interventions that can prevent eco-gentrification while supporting inclusive climate adaptation.
4. Propose a governance framework that aligns AI-driven decision-making with participatory urban planning practices to ensure social inclusion and procedural justice.

By addressing these objectives, this study contributes to the expanding discourse on equitable AI governance and offers practical insights for policymakers, urban planners, and environmental advocates seeking to harmonize technological innovation with environmental justice.

### **Significance of the Study**

The significance of this study lies in its interdisciplinary integration of Geospatial Artificial Intelligence (GeoAI), environmental monitoring, and urban policy. As cities increasingly adopt AI-driven tools to support ecological planning and climate adaptation, it is crucial to ensure that these technologies promote inclusivity and resilience rather than exacerbate existing socioeconomic disparities.

This research contributes to the field in three keyways:

1. **Advancing Environmental Risk Mitigation**
  - Develops a conceptual framework for applying AI-driven predictive models to flood risk assessment, pollution tracking, and climate vulnerability analysis.
  - Offers data-informed strategies to support adaptive urban resilience planning.
2. **Promoting Inclusive Urban Renewal**
  - Identifies equity-centered policy interventions to prevent eco-gentrification in post-industrial redevelopment zones.
  - Recommends tools such as inclusionary zoning mandates and community-driven planning frameworks to ensure AI-informed decisions reflect the needs of marginalized populations.
3. **Strengthening Accountability Through Equity Metrics**
  - Emphasizes the need for measurable frameworks to assess the social and environmental impacts of GeoAI and IoT technologies in urban redevelopment.

- Recommends the development of metrics to evaluate distributional equity (e.g., who benefits from green infrastructure) and procedural justice (e.g., who participates in zoning decisions).
- Supports the use of fairness audits, transparency tools, and policy evaluation frameworks to monitor affordability protections and displacement risks.
- Calls for future research to establish standardized indicators that cities can adopt to evaluate the equity outcomes of AI-integrated planning systems.

By advancing these contributions, the study provides a foundation for ethical, inclusive, and sustainability-oriented AI governance in urban redevelopment.

## Global Relevance

The issues explored in this study, including climate adaptation, algorithmic governance, and housing equity, are not limited to the Gowanus Canal. They reflect broader global trends in urban redevelopment. As cities increasingly incorporate GeoAI and IoT technologies into sustainability planning, there is a critical need to ensure that these tools do not reinforce socioeconomic exclusion or contribute to displacement. This study offers a framework that cities can apply across diverse urban contexts to assess how technological innovation enhances both environmental and social resilience.

The primary research question guiding this inquiry is: How can GeoAI and IoT technologies be used to improve climate resilience and environmental equity in urban redevelopment projects while also preventing displacement and social exclusion?

Supporting this question are the following secondary inquiries:

1. What are the unintended social and economic impacts of AI-driven urban monitoring systems?
2. How do algorithmic zoning models affect housing affordability and displacement trends in post-industrial neighborhoods?
3. What policy mechanisms can align sustainability goals with anti-displacement strategies?
4. How can participatory governance models ensure inclusive decision-making in AI-driven urban planning?

## Hypothesis

**Main Hypothesis (H1):** The integration of Geospatial Artificial Intelligence (GeoAI) and Internet of Things (IoT) technologies will enhance climate resilience and environmental sustainability by enabling real-time pollution tracking, AI-driven flood mitigation, and data-informed planning of green infrastructure. However, in the absence of equity-centered governance

frameworks, such as affordable housing protections, community land trusts, inclusionary zoning policies, and participatory urban planning, these technological advancements may unintentionally exacerbate socioeconomic inequities, resulting in eco-gentrification and residential displacement.

### ***Sub-Hypotheses:***

**H1a:** AI-driven flood modeling and pollution monitoring tools will support improved climate adaptation outcomes but may rely on non-transparent algorithms that obscure public understanding of zoning outcomes if not implemented with transparent oversight and public accountability mechanisms.

**H1b:** AI-assisted zoning models are likely to prioritize redevelopment in high-income areas unless equity safeguards and inclusionary land-use policies explicitly guide them.

**H1c:** Municipalities that integrate bias audits and participatory AI governance into their planning processes will experience reduced displacement pressures compared to those that rely exclusively on technocratic, AI-driven land-use planning models.

These hypotheses are visually represented in the conceptual framework (see Figure 1), which outlines the relationships between technological inputs, governance processes, and their potential outcomes on climate resilience and social equity in urban redevelopment.

This logic model proposes a theoretical relationship among technological inputs, governance processes, and urban sustainability outcomes. It highlights how the presence or absence of equity-centered safeguards affects whether AI-driven urban redevelopment fosters social inclusion or leads to eco-gentrification and displacement.

### **Problem Statement**

Urban redevelopment initiatives often emphasize environmental restoration and climate resilience but frequently overlook social equity considerations. As a result, these well-intentioned efforts can produce unintended consequences such as eco-gentrification, housing displacement, and the marginalization of vulnerable populations. While Geospatial Artificial Intelligence (GeoAI) and Internet of Things (IoT) technologies offer advanced capabilities for flood risk management, pollution tracking, and adaptive urban sustainability, their implementation may reinforce economic and spatial inequalities if not accompanied by robust governance structures, including bias audits, transparency mechanisms, and community engagement.

The Gowanus Canal in New York City exemplifies these tensions. Decades of industrial activity, untreated wastewater discharge, and combined sewer overflows (CSOs) have led to profound environmental degradation, prompting a multi-billion-dollar Superfund remediation initiative. Although these

interventions improve climate resilience, they also contribute to speculative real estate investment, which escalates property values and leads to demographic shifts that displace long-term residents and small businesses.

Absent proactive policy safeguards, sustainability efforts risk becoming instruments of exclusion rather than tools for justice. This study presents a novel framework that integrates environmental technology and equity metrics in AI-driven redevelopment. It evaluates policy interventions, equity-focused governance models, and participatory planning strategies that can help ensure AI-driven urban redevelopment supports community-centered, inclusive growth.

## **Literature Review: Theoretical Foundations of Smart Cities, Environmental Gentrification, and AI-Driven Urban Governance**

### ***Smart Cities and AI-Driven Urban Planning***

Smart cities deploy artificial intelligence (AI), Internet of Things (IoT), and big data analytics to optimize transportation networks, environmental monitoring, and infrastructure resilience (Bibri, 2021). While these technologies promise efficiency and sustainability, critics have raised concerns about their implications for democratic governance. Kitchin (2014) argues that AI-assisted urban planning can shift decision-making power away from communities and toward opaque technical systems. A central concern is the rise of "black-box urbanism," where algorithmic systems make planning decisions without public oversight or input (Mattern, 2017). In this context, technological optimization often supersedes participatory governance and social inclusion.

### ***Environmental Gentrification and Housing Displacement***

Many now view environmental improvements, such as flood-resilience infrastructure and green amenities, as "climate luxuries" that can accelerate gentrification. Environmental gentrification refers to the process by which sustainability-focused redevelopment increases property values, often displacing low-income residents and small businesses (Checker, 2011; Gould & Lewis, 2017). Case studies illustrate this dynamic. In Atlanta, the BeltLine project contributed to a 200–400% increase in property values in surrounding areas, resulting in widespread displacement (Anguelovski et al., 2019). Similarly, New York City's High Line sparked a wave of luxury development that priced out existing communities (Gould & Lewis, 2017).

### ***Algorithmic Bias and Digital Exclusion***

The use of AI in urban planning introduces new risks related to algorithmic bias and digital exclusion. Researchers and policymakers are increasingly proposing Explainable AI (XAI) frameworks to enhance transparency in zoning and land-use decisions, thereby mitigating these risks. XAI enables urban planners and residents

to better understand the criteria behind zoning recommendations, thereby facilitating accountability and public trust.

Without such safeguards, AI-assisted planning models tend to favor high-income redevelopment, reinforcing existing socioeconomic divides (Sanchez et al., 2024). Persistent infrastructural inequality in U.S. cities means that AI-driven zoning tools often reinforce patterns of racial exclusion, channeling green infrastructure investment toward affluent, predominantly white neighborhoods while overlooking the risks of displacement and underinvestment in communities of color (Archer and Joshi, 2025). Similarly, the AI Now Institute (Crawford et al., 2019) reported that predictive zoning tools in San Francisco favored commercial developments over affordable housing in historically marginalized communities. Observable trends further illustrate these concerns: in many rezoned areas, rents have risen substantially faster than in comparable neighborhoods, and eviction rates tend to be noticeably higher in districts prioritized for redevelopment through algorithmic tools.

The broader context of algorithmic discrimination in predictive policing and credit scoring (VoPham et al., 2018) underscores the urgency of integrating equity safeguards into AI zoning models. For the Gowanus Canal, these findings suggest that rigorous bias audits must accompany GeoAI tools to prevent speculative zoning from becoming a driver of displacement. As Taylor and Lethabo (2023) argue, community-led AI decision-making is crucial in combating digital exclusion.

### *Participatory AI Governance Models*

In response to these concerns, several cities have begun implementing participatory AI governance frameworks to ensure that technological innovation remains aligned with democratic values. In Barcelona, the Decidim platform allows residents to engage directly in AI-driven zoning decisions, serving as a model for inclusive digital governance. Amsterdam's AI4Cities initiative similarly invites citizens to co-design sustainability policies informed by AI technologies.

These models demonstrate that AI can support, rather than supplant, community participation in urban planning. To operationalize participatory AI, cities should adopt a phased governance model:

1. **Public AI Transparency Portals** – Open-source zoning dashboards that allow residents to review, interrogate, and challenge AI-generated planning decisions.
2. **Community-Led AI Policy Councils** – Local panels that co-design zoning algorithms and conduct ongoing fairness audits.
3. **Digital Participatory Mapping Tools** – Interfaces, such as Decidim, that enable residents to propose zoning alternatives using AI-assisted tools.
4. **Legally Mandated AI Oversight** – Government agencies empowered to enforce fairness audits and pause deployment of biased models.

### Legal Framework for AI Governance in New York City

To ensure AI zoning adheres to principles of transparency and justice, specific legal and regulatory reforms are required:

1. **Mandated AI Bias Audits** – Independent third-party assessments of zoning models must be required before implementation.
2. **AI Transparency Legislation** – New York City should adopt a localized version of the EU AI Act, ensuring open access to zoning algorithms and training data.
3. **Legislative Review Panels** – Oversight bodies should be established to review and approve AI-generated zoning recommendations.
4. **Public Legal Recourse** – Residents must have formal mechanisms to challenge AI-driven zoning outcomes, building upon existing New York City Open Data laws.

Together, these reforms and models support a more ethical and inclusive integration of AI technologies into urban governance. They also offer a roadmap for safeguarding social equity in climate-resilient redevelopment projects such as the Gowanus Canal.

### Comparative Case Studies: AI, Gentrification, and Affordability Failures

The Atlanta BeltLine and the New York City High Line provide instructive examples of high-profile urban redevelopment initiatives where technological innovation coincided with affordability crises and displacement. Although these cases are not universally representative of all redevelopment projects, they offer valuable insights into the risks associated with AI-driven planning in the absence of safeguards for social equity. Future research should expand the comparative framework to include a broader range of cities and redevelopment contexts, including those without AI integration.

Table 1. Case Study Comparison

Project	Environmental Benefits	Displacement Impact	AI Integration	Policy Effectiveness
Atlanta BeltLine	Flood mitigation, greenway expansion	200-400% property value increase	AI zoning forecasts, Smart Growth models	Weak affordability protections
NYC High Line	Urban reforestation, walkways	Rents ↑ 400%, rapid luxury development	Predictive real estate AI models	No affordability mandates

Note. This table compares two high-profile redevelopment projects on four criteria: environmental benefits, displacement impacts, AI integration, and policy effectiveness. The Atlanta BeltLine and NYC High Line both contributed to green infrastructure goals but lacked sufficient affordability protections, highlighting risks of eco-gentrification when smart technologies are deployed without equity safeguards.

### ***Key Policy Failures***

- Atlanta BeltLine: Although initial plans called for the creation of 5,600 affordable housing units, these goals were largely unmet due to rezoning loopholes and speculative real estate investment. Affordability measures lacked enforcement, and tax incentives favored luxury development.
- NYC High Line: The redevelopment proceeded without Mandatory Inclusionary Housing (MIH) policies. The lack of affordability mandates led to unchecked property value escalation, driving the displacement of long-term residents.
- Gowanus Canal: This site presents a forward-looking opportunity to embed affordability protections before gentrification pressures fully materialize. Lessons from the BeltLine and High Line suggest that without legal mandates, affordability commitments are unlikely to be realized. Inclusionary zoning must be codified early and monitored rigorously.

### ***Lessons Learned***

- Urban climate adaptation projects must be paired with anti-displacement strategies to prevent the reproduction of environmental gentrification.
- Smart city technologies, including predictive zoning models, require democratic oversight and housing equity metrics to ensure inclusive outcomes.
- Mandatory affordability policies and participatory governance models are essential for mitigating the negative externalities of AI-assisted redevelopment.

### ***Conclusion: AI, Sustainability, and Housing Equity***

GeoAI and IoT technologies hold significant promise for enhancing urban resilience, particularly through real-time flood monitoring, pollution control, and adaptive infrastructure design. However, if deployed without inclusive planning mechanisms, these tools risk amplifying displacement and deepening socioeconomic inequality. This study underscores the importance of incorporating bias audits, participatory AI governance, and affordability mandates into AI-driven urban planning. Only through a proactive, equity-centered framework can technological innovation truly support both environmental sustainability and housing justice.

### **Implementation Roadmap for Equity-Centered AI Urban Planning**

To ensure that AI-assisted urban redevelopment promotes both environmental sustainability and social inclusion, cities must adopt a phased implementation strategy:



1. **Short-Term (1–2 Years):** Mandate AI bias audits and zoning fairness evaluations as prerequisites for deploying predictive zoning and land-use models. Independent bodies should conduct these assessments and make them publicly accessible to ensure transparency and accountability.
2. **Medium-Term (3–5 Years):** Establish participatory AI zoning review boards within city planning departments. These boards should include residents, data ethics experts, housing advocates, and urban planners who collectively co-design zoning algorithms and monitor displacement risks in real time.
3. **Long-Term (5–10 Years):** Advocate for the development of federal AI zoning regulations modeled after the European Union's AI Act. These regulations should establish legal safeguards, auditing standards, and equity metrics to guide the ethical use of AI in land-use planning across municipalities.

## Global Implications

While this study focuses on New York City, its findings and proposed framework have broader relevance for cities worldwide that are navigating the integration of AI in urban planning.

- **High-Income Cities (e.g., Amsterdam, London):** These cities face the deployment of advanced AI, as well as worsening affordability crises. Without mandatory affordability protections, AI-assisted zoning risks accelerating exclusionary development and deepening housing inequality.
- **Global South Cities (e.g., Jakarta, Mumbai):** In regions experiencing rapid urbanization and acute climate vulnerability, GeoAI and IoT provide transformative tools for enhancing flood resilience and planning infrastructure. However, policymakers must embed these technologies within equity-centered urban policy frameworks to prevent the recurrence of inequitable development patterns.

## Synthesis and Contribution

By aligning GeoAI, IoT, and participatory AI governance, cities can pursue a redevelopment model that is both sustainable and socially just. Prior research has demonstrated the environmental benefits of AI in urban planning; however, these gains have often come at the expense of housing affordability and community stability. This paper presents a novel, equity-centered approach that addresses a critical gap by integrating geospatial analytics, stakeholder perspectives, and policy analysis to propose a comprehensive, equity-driven framework for AI-assisted urban redevelopment.

## Proposed Methodological Framework for Future Research

This paper proposes a conceptual framework to guide future empirical research rather than presenting original data analysis. The goal is to provide a structured methodology for examining how AI-driven sustainability initiatives impact climate resilience, displacement risk, and equity in urban redevelopment zones, such as the Gowanus Canal. Central to this approach is the conceptual model depicted in Figure 1, which illustrates the relationships between GeoAI and IoT technologies, governance interventions, and socio-environmental outcomes. While this study does not conduct empirical modeling, it identifies data sources and analytical strategies that can operationalize the framework in future studies. For example, indicators such as rent burden, flood zone classification, and access to green infrastructure could be analyzed through spatial equity mapping.

Suggested tools include hotspot analysis to detect inequality in infrastructure distribution and AI-based flood prediction models that utilize climate and demographic overlays. These examples are intended to demonstrate how the framework could be adapted for real-world testing. Future researchers can employ such techniques to evaluate how algorithmic zoning models influence displacement, affordability, and environmental justice across varied urban settings.

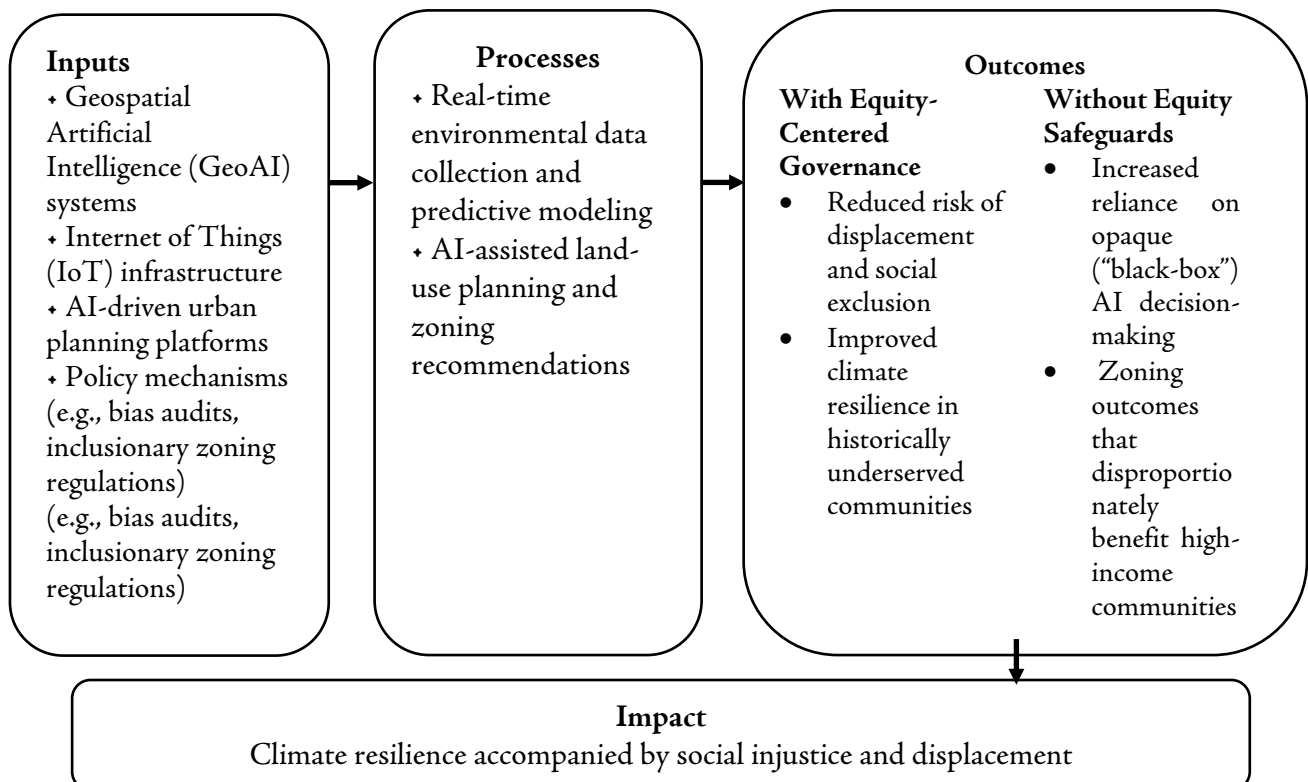


Figure 1. Conceptual framework for evaluating equity impacts of GeoAI and IoT in urban redevelopment

Note. Figure 1 illustrates how the combination of GeoAI and IoT technologies, when integrated with inclusive policy mechanisms, can lead to enhanced climate resilience and social inclusion. In contrast, the absence of equity safeguards may reinforce exclusionary

zoning outcomes and increase reliance on opaque AI systems, contributing to displacement and spatial inequality.

### 1. Geospatial Analytics and IoT-Based Environmental Monitoring

This component triangulates environmental, technological, and socioeconomic indicators to assess the impacts of AI-driven zoning. By combining remote sensing, IoT sensor data, and real estate analytics, the study captures the spatial dynamics of green infrastructure investments and their socioeconomic effects.

#### 1.1 Proposed Data Sources and Metrics

Table 2. Environmental and socioeconomic data sources

Data Type	Source	Metrics	Purpose
Satellite Imagery	NASA Landsat-8, Sentinel-2	NDVI (Vegetation Index), Land Use Changes, Urban Heat Islands	Assess environmental shifts and impacts of green infrastructure
IoT Water Sensors	NYC Open Data, EPA Monitoring Stations	pH, Heavy Metals, Nitrogen Levels	Evaluate water quality post-remediation
IoT Air Sensors	NYC Environmental Monitoring	PM2.5, CO2, VOCs	Track pollution and air quality trends
GeoAI Flood Models	NYC Climate Projections, NOAA	Flood Risk Index, Stormwater Mapping	Predict climate-related vulnerabilities
Real Estate & Zoning Data	NYC Open Data, Zillow, Tax Parcel Records	Property Values, Demographic Shifts	Identify gentrification and displacement patterns

*Note.* This table summarizes key data inputs and metrics that may be used to operationalize the conceptual framework. Environmental indicators such as NDVI, pollutant levels, and flood risk scores, along with real estate and zoning data, can support spatial analysis of green infrastructure outcomes, climate vulnerability, and displacement risk

#### 1.2 Proposed Analytical Methods

- **AI-Driven Spatial Equity Mapping**
  - Objective: Identify disparities in infrastructure benefits by income level.
  - Method: Machine learning models with hot spot analysis (Getis-Ord  $G_i^*$ ) to identify correlations between flood mitigation and rent increases.
- **Predictive AI Flood Modeling**
  - Objective: Model future flood risks under various climate scenarios.

- Method: AI-enhanced hydrodynamic models using sea-level rise and density projections overlaid with socioeconomic data.
- **Real Estate Displacement Analysis**
  - Objective: Determine if flood adaptation investments correlate with rising rents.
  - Method: Regression analysis on housing price changes and environmental project locations.

## 2. Comparative Case Study Analysis

This component assesses the displacement impacts and policy responses in two high-profile redevelopment projects: the Atlanta BeltLine and the New York City High Line.

### 2.1 Case Selection

Table 3. Comparative case studies of ai-integrated urban redevelopment projects.

Case Study	Environmental Innovation	AI Use in Urban Planning	Gentrification Impact
Gowanus Canal	Superfund remediation, smart flood-barriers	GeoAI zoning models, IoT pollution tracking	Displacement risks increasing
Atlanta BeltLine	Greenway expansion, stormwater retention	AI-based zoning forecasts, Smart Growth plans	Housing costs ↑ 200-400%
NYC High Line	Urban reforestation, smart walkways	Predictive real estate AI models	Rents ↑ 400%, low-income residents displaced

*Note.* This table compares three redevelopment sites, Gowanus Canal, Atlanta BeltLine, and NYC High Line, based on environmental innovation, AI application in planning, and observed gentrification impacts. The cases highlight how smart technologies, if not paired with equity safeguards, can contribute to rising housing costs and displacement pressures.

### 2.2 Analysis Methods

- **Historical Trend Analysis:** Evaluates pre- and post-redevelopment affordability changes.
- **Spatial Mapping of Displacement:** Maps socio-demographic changes in affected neighborhoods.
- **Policy Evaluation:** Assesses enforcement of affordability protections and zoning regulations.

### 3. Policy Document Analysis

With stakeholder interviews removed, the study compensates by analyzing zoning, environmental, and AI governance documents at the federal, local, and international levels.

#### 3.1 Proposed Data Sources

Table 4. Policy document sources and relevance to AI-driven zoning in Gowanus

Policy Level	Document Analyzed	Relevance to Gowanus Redevelopment
Federal	EPA Superfund Guidelines	Determines clean-up priorities but lacks affordability protections
Federal	Fair Housing Act	Evaluate AI-driven zoning discrimination risks
NYC Local	Mandatory Inclusionary Housing (MIH)	Examines if housing mandates prevent displacement
NYC Local	Green Infrastructure Plan	Assesses whether green investments are equitably distributed
International	EU AI Act	Compares AI zoning regulations globally

*Note.* This table identifies key policy documents at the federal, local, and international levels relevant to the Gowanus Canal redevelopment. It highlights how each policy either supports or lacks mechanisms to address equity concerns in AI-assisted zoning and environmental planning.

#### 3.2 Policy Evaluation Framework

Table 5. Policy evaluation framework

Policy Component	Key Questions	Assessment Criteria
AI & Zoning Regulations	Does the policy prevent algorithmic bias in zoning decisions?	AI transparency, data ethics
Climate Resilience vs. Displacement	Are green investments benefiting existing communities?	Equity mapping, policy enforcement
Affordable Housing Mandates	Do zoning laws require affordability protections in redevelopment?	Inclusionary zoning compliance

*Note.* This table outlines a framework for evaluating policies that guide the ethical deployment of AI in zoning and urban redevelopment. It focuses on three core components: algorithmic fairness in zoning regulations, equitable climate resilience investments, and the enforcement of affordability mandates in redevelopment policies.

#### 4. Ethical Considerations: Algorithmic Redlining in AI-Driven Zoning

Modern AI zoning tools risk replicating 20th-century redlining patterns by deprioritizing investments in historically marginalized neighborhoods. Case evidence from California, New York, and Virginia demonstrates how AI models reinforce inequalities when not paired with fairness auditing or equity-driven algorithm design.

##### 4.1 Real-World Examples

- **NYC Predictive Zoning (AI Now Institute, 2021):** AI undervalued historically Black and Latino areas, resulting in reduced resilience investment and commercial upzoning.
- **Amazon HQ2 (Arlington, VA):** AI-driven land valuation prioritized high-growth areas without considering socioeconomic vulnerabilities, resulting in the displacement of over 10,000 renters.
- **Chicago SDSC Predictive Policing:** AI tools have reinforced over-policing in communities of color and under-investment in lower-income areas, highlighting how urban planning data sources contribute to zoning bias.

##### 4.2 Proposed Safeguards for Gowanus AI Planning

Table 6. Real-world cases of algorithmic bias in urban planning

Intervention	Purpose	Implementation Strategy
Bias Audits for AI Zoning Models	Ensure that predictive zoning does not reinforce past redlining	Mandate NYC to conduct independent AI fairness assessments
Participatory AI Decision-Making	Allow local communities to co-design zoning algorithms	Require community input sessions on AI zoning models
Equity Weighting in AI Models	Prioritize investment in historically marginalized neighborhoods	Modify GeoAI algorithms to account for past disinvestment
Open Data & Transparency in AI Decisions	Ensure public scrutiny of AI-generated zoning changes	NYC must provide public access to AI-generated zoning forecasts

Notes. This table summarizes proposed strategies to mitigate algorithmic bias in AI-assisted zoning and planning. Each intervention targets a specific equity concern such as redlining, exclusion, or opacity, through implementation mechanisms like fairness audits, participatory design, and open data mandates.

#### Conclusion: AI, Sustainability, and Housing Equity

This study presents a conceptual framework suggesting that while Geospatial Artificial Intelligence (GeoAI) and Internet of Things (IoT) technologies offer considerable potential to enhance urban climate resilience, they also carry significant

risks. If not carefully governed, these innovations risk reinforcing spatial and economic exclusion. As emphasized in the conceptual framework (see Figure 1), incorporating equity-centered governance strategies such as bias audits, participatory AI governance, and enforceable affordability mandates is critical to mitigating these risks.

The analysis underscores that AI-driven zoning and predictive land-use models are not inherently objective. Instead, historical inequalities embedded in housing policies, land-use decisions, and capital investment flows shape them. In the absence of transparency, oversight, and community participation, such tools may replicate and exacerbate the very injustices that sustainable redevelopment efforts aim to address. To address these concerns, this study proposes a governance framework centered on equity-focused interventions, including AI bias audits, participatory AI governance, and enforceable affordability mandates. These mechanisms are essential to ensuring that AI-integrated urban redevelopment projects promote inclusion rather than exclusion. While this paper does not present original empirical findings, it outlines a structured methodology for future validation. As the Gowanus Canal redevelopment incorporates AI-based flood resilience planning and zoning reform, it offers a critical opportunity to implement and test the proposed framework in practice. Future empirical research should rigorously evaluate the effectiveness of the framework's components, particularly the predictive modeling, spatial equity mapping, and policy interventions, to determine whether AI-assisted urban sustainability initiatives can truly support both environmental progress and social justice.

## **Discussion**

This study reinforces the dual role of GeoAI and IoT technologies in urban redevelopment: while they enhance environmental resilience and support data-informed planning, they may unintentionally heighten affordability crises and deepen community vulnerability when equity safeguards are absent. These insights align with the conceptual relationships detailed earlier, underscoring the critical role of equity-centered governance in mitigating adverse outcomes.

### ***1. AI-Driven Urban Planning: Opportunities and Risks***

GeoAI-enabled flood models and predictive analytics significantly improve urban climate adaptation, particularly in stormwater management and infrastructure resilience. However, as governments increasingly rely on algorithmic decision-making, they face new challenges in governance. "Black box" AI systems often obscure the process by which zoning decisions are made, making it difficult for communities and policymakers to understand or challenge them. This opacity risks excluding marginalized populations from the urban planning process and weakens democratic oversight.

## ***2. Eco-Gentrification: Insights from Comparative Case Studies***

The case studies of the Atlanta BeltLine and the NYC High Line reveal how environmental improvements, when decoupled from affordability protections, can accelerate gentrification. In both cases, AI-informed planning contributed to rising property values and displacement of long-term residents. Underscore the importance of embedding mandatory inclusionary zoning and affordability mandates within AI-assisted urban planning to avoid replicating these outcomes in future projects like the Gowanus Canal.

## ***3. Toward Equity-Centered AI Governance***

Policymakers must restructure urban policy around equity-centered governance models to ensure that AI-driven sustainability efforts promote inclusive outcomes. Key recommendations include:

- Mandating affordable housing quotas in AI-generated zoning scenarios
- Supporting the development of community land trusts (CLTs) to safeguard local housing
- Requiring bias audits for AI-driven real estate and zoning algorithms to identify and mitigate exclusionary outputs

These strategies can help reorient urban AI systems toward fairness and distributive justice.

## ***4. The Future of AI in Sustainable Urban Planning***

Policymakers and planners must embed GeoAI within participatory governance frameworks that foster inclusive decision-making to ensure it serves all communities equitably. Public engagement in zoning decisions, algorithm transparency, and the co-creation of planning tools are essential. Without these components, AI-enhanced planning risks prioritizing investment returns over social inclusion and equity. This study highlights the urgent need for ethical, regulatory, and participatory reforms to shape the role of AI in the built environment.

### ***Limitations***

Despite its conceptual depth, this study acknowledges several limitations related to AI data governance, transparency, and the feasibility of implementation.

#### ***The "Black Box" Problem in AI Zoning***

Many AI systems used in urban planning operate as "black boxes," lacking explainability and transparency. This opacity undermines public accountability and impedes regulatory oversight. For example, a 2023 report by the Algorithmic Justice League revealed that AI-driven zoning systems in Chicago failed to provide



justifications for land-use changes, effectively preventing residents from participating in the planning process.

#### *Data Bias and Algorithmic Discrimination*

AI zoning models trained on historical land use and real estate datasets frequently replicate embedded structural discrimination. These systems often favor affluent districts for green infrastructure while prioritizing commercial and luxury development in historically disinvested areas. A 2022 MIT study found that predictive zoning tools in California encouraged commercial expansion in high-poverty areas based on outdated risk assumptions despite strong community opposition.

*Recommended Mitigation:* Implement algorithmic fairness audits and integrate "equity weighting" within zoning algorithms to actively reverse, rather than reproduce, past inequalities.

#### *Limited Access to GeoAI and Predictive Zoning Data*

While GeoAI and IoT technologies produce valuable insights, much of the data remains inaccessible to the public due to proprietary restrictions or limited municipal transparency. For example, while New York City's Open Data Law mandates public access to specific zoning layers, predictive zoning algorithms used in rezoning proposals are often held by private vendors and not publicly disclosed.

*Recommended Mitigation:*

- Mandate full public access to AI-generated zoning proposals
- Ensure open-source GeoAI data is available for community analysis and civic engagement

#### *Challenges in Policy Enforcement and Implementation*

Even when strong policies exist, enforcement remains a barrier. Developers often resist inclusionary zoning mandates, and AI-generated projections may favor profit-maximizing models over equitable development. Regulatory fragmentation across multiple stakeholders, planners, tech firms, and local governments can weaken consistency and accountability.

Example: In San Francisco, AI-driven proposals to increase housing density were delayed and weakened due to legal challenges from developers and neighborhood associations.

*Recommended Mitigation:*

- Establish a centralized oversight mechanism for AI zoning compliance
- Foster cross-sector collaboration among city planners, equity advocates, and AI ethics experts

## Future Research Directions

This study provides a comprehensive conceptual framework for understanding the intersection of AI-driven urban planning, environmental justice, and housing equity. However, several critical areas remain underexplored. Future research should prioritize empirical investigations into how GeoAI technologies, zoning algorithms, and regulatory frameworks influence urban redevelopment across diverse contexts.

### 1. Key Research Questions

Future inquiry should consider the following foundational questions:

1. How does AI-driven gentrification manifest across cities with differing social, political, and economic conditions?
2. What are the most effective policy interventions to mitigate the risks of AI-induced displacement?
3. How can communities be empowered to participate meaningfully in the development and governance of AI-driven urban planning tools?

### 2. Research Prioritization Framework

To structure inquiry over the next decade, this study identifies three tiers of research based on urgency and feasibility:

- **Immediate-Term Focus:** Develop Explainable AI (XAI) models to increase zoning transparency, conduct real-time AI fairness audits, and ensure legislative compliance with open data mandates.
- **Mid-Term Comparative Studies:** Evaluate how AI zoning outcomes differ between high-income and low-income cities, with particular attention to affordable housing impacts.
- **Long-Term Global Research:** Investigate how AI-based land-use models function in the Global South, especially in contexts marked by informal settlements and limited regulatory infrastructure.

By organizing research according to implementation urgency, this framework provides a roadmap for advancing ethical and inclusive AI governance in urban planning.

### 3. Critical Areas for Future Study

#### 3.1 Explainable AI (XAI) and Algorithmic Transparency in Zoning

AI-based zoning systems often operate as "black boxes," obscuring the decision-making process and leaving policymakers and residents unable to effectively interrogate the outcomes. Future research should investigate:

- How an Explainable AI (XAI) can be embedded in zoning tools to improve public accountability.

- The design of algorithmic fairness auditing frameworks for municipal planning agencies.
- The effect of AI transparency laws (e.g., NYC Open Data Law, EU AI Act) on zoning outcomes and civic participation.

**Example Research Question:** Can XAI models increase public trust and policy accountability in AI-driven land-use planning?

### 3.2 *Community-Led AI Governance in Smart Cities*

Currently, the control of AI zoning tools rests largely with municipal agencies, private tech firms, and real estate developers. This centralized governance model excludes communities most affected by redevelopment. Future research should explore:

- Participatory AI platforms that allow residents to co-create or challenge zoning proposals.
- Governance models that decentralize AI decision-making (e.g., blockchain-based zoning feedback systems).
- Metrics to evaluate how participatory AI improves housing equity and planning legitimacy.

**Example Research Question:** How can participatory AI governance ensure community involvement and reduce exclusion in urban planning?

### 3.3 *Comparative Studies Across Diverse Urban Contexts*

Most existing literature focuses on high-income cities such as New York, London, and Amsterdam. However, there is an urgent need for research in low- and middle-income cities (LMICs), where informal housing, limited digital infrastructure, and weak regulatory frameworks pose unique challenges.

- How do AI-driven urban analytics function in cities like Lagos, Mumbai, or São Paulo?
- What are the implications of AI-led flood adaptation planning in climate-vulnerable cities such as Dhaka and Jakarta?
- How does zoning automation affect housing affordability and spatial justice in informal settlements?

**Example Research Question:** What are the impacts of AI-assisted zoning on housing rights and affordability in Global South cities?

## 4. *Synthesis: Building an Inclusive Future for AI Urban Planning*

To guide future scholarship, the table below summarizes the priority focus areas, corresponding questions, and potential impacts:

## Towards an Equitable Future for AI-Driven Urban Planning

### *Summary of Key Future Research Areas:*

Table 7. Summary of priority future research areas in AI-driven urban planning

Future Research Focus	Key Question	Potential Impact
Explainable AI (XAI) in Zoning	How can transparent AI models improve accountability in urban planning?	Improves public trust, prevents hidden biases
Community-Led AI Governance	Can participatory AI models ensure community involvement in zoning decisions?	Enhances equity, prevents top-down gentrification
Comparative Global Studies	How does AI zoning affect affordability in high-income vs. low-income cities?	Expands AI governance research beyond the Global North

Notes. This table outlines three strategic areas for future research: explainable AI (XAI), participatory AI governance, and comparative global studies. Each area is paired with a guiding research question and highlights the potential impact on urban equity, accountability, and global knowledge exchange.

### *Final Takeaway*

The future of AI-driven urban planning must prioritize equity, transparency, and democratic governance. As GeoAI and IoT technologies become more integral to redevelopment strategies, it is essential that their deployment actively works to undo historical inequities rather than reinforce them. Future research should ensure that algorithmic tools support inclusive, just, and climate-resilient urban futures for all communities.

### *Policy Document Analysis: AI Governance, Zoning, and Sustainability Regulations*

To assess the regulatory landscape shaping AI-driven urban redevelopment, this study conducts a policy document analysis focused on the intersection of climate resilience, data governance, affordable housing mandates, and community engagement. Given the removal of stakeholder interviews, this component serves as a crucial source of qualitative insight into how laws and frameworks either support or hinder the equitable integration of AI in urban planning.

### *Data Collection: Policy Selection Criteria*

The study selected policy documents based on their relevance to four key dimensions of equitable redevelopment:

- Direct influence on AI-driven zoning and sustainability planning in New York City

- Inclusion of environmental justice, housing equity, or data governance provisions
- Applicability to federal, local, and international urban policy contexts
- Demonstrated impact on climate adaptation, land-use planning, or community participation

The documents reviewed include:

- **Federal:**
  - *EPA Superfund Remediation Guidelines* (Environmental restoration vs. gentrification risks)
  - *Fair Housing Act* and *AI Ethics Guidelines* (Housing discrimination prevention and algorithmic fairness)
- **Local (NYC):**
  - *Mandatory Inclusionary Housing (MIH)* (Affordability zoning mandates)
  - *NYC Green Infrastructure Plan* (Equitable distribution of climate adaptation resources)
  - *NYC Open Data Law* (AI zoning transparency)
- **International:**
  - *European Union AI Act* (Algorithmic governance in planning contexts)
  - *United Nations Sustainable Development Goals (SDG 11)* (Inclusive and sustainable urbanization)

*Analytical Framework: Evaluating Policy Intent and Equity Impact*

To guide the analysis, this study applies a comparative policy evaluation framework organized around four key components. Each component is tied to a central research question and linked to a representative policy document. This framework assesses whether policies promote transparent, equitable, and community-responsive AI governance in urban redevelopment.

*Policy Evaluation Framework for AI-Driven Urban Redevelopment*

Table 8. Policy evaluation framework for AI-driven urban redevelopment

Policy Component	Key Questions for Analysis	Example Policy
Sustainability & Climate Resilience	Does the policy ensure equitable access to flood mitigation, green spaces, and pollution control?	NYC Green Infrastructure Plan
Housing & Displacement Protections	Does the policy require affordable housing mandates in gentrifying areas?	NYC Mandatory Inclusionary Housing
AI & Data Governance	Does the policy regulate how AI-driven zoning models impact urban development?	NYC Open Data Law, EU AI Act

Community Participation	Does the policy ensure local communities have a voice in AI-driven planning?	UN SDG 11, Participatory AI Models
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*Note.* This table presents a framework for analyzing urban redevelopment policies based on four components: climate resilience, housing protections, AI governance, and community participation. Each policy component is linked to a guiding question and a real-world example to assess whether equity, transparency, and inclusivity are embedded in AI-based zoning initiatives.

## Global Relevance: Urban Climate Risks in High- and Low-Resource Settings

### *Climate Adaptation Case Study: Jakarta, Indonesia*

Jakarta, one of the world's most flood-prone megacities, faces compounding threats from land subsidence, rising sea levels, and climate-driven displacement. In response, the city has implemented AI-enhanced flood prediction models that integrate IoT-based rainfall sensors and machine learning algorithms to optimize stormwater drainage and provide early warnings. However, the uneven distribution of resilience infrastructure, focusing on wealthier districts, has exacerbated social inequalities. Jakarta underscores the urgent need for equity-centered AI governance to ensure that climate adaptation efforts do not exacerbate existing disparities.

### *Key Urban Climate Risks in LMICs*

Low- and Middle-Income Countries (LMICs) face climate vulnerabilities with limited adaptation resources. Key challenges include:

- **Urban Heat Island (UHI) Intensification:** Rising temperatures disproportionately affect informal settlements that lack cooling infrastructure, thereby compounding health risks.
- **Stormwater Management Failures:** Outdated drainage systems contribute to chronic flooding and infrastructure breakdowns.
- **Water Scarcity:** Rapid urban expansion and inefficient management exacerbate the stress on already limited water supplies.

Weak governance structures, limited environmental monitoring, and inequitable investment exacerbate these vulnerabilities.

## GeoAI and Urban Resilience

### *Technological Promise*

GeoAI integrates satellite imagery, machine learning, and IoT systems to enable:

- Real-time environmental monitoring
- Predictive modeling of flood risk and pollution
- Data-driven land-use and infrastructure planning

### ***Challenges to Equitable Deployment***

1. **Data Governance:** Marginalized communities often lack access to critical environmental data.
2. **Algorithmic Bias:** Inadequate training datasets risk perpetuating socioeconomic disparities.
3. **Infrastructure Gaps:** High implementation costs limit access in low- and middle-income countries (LMICs).

Effective deployment requires coupling technological innovation with inclusive governance, community participation, and transparency.

### ***Case Study: The Gowanus Canal***

Once a biodiverse tidal marshland, Brooklyn's Gowanus Canal suffered decades of industrial pollution. Designated a Superfund site in 2010, it is now undergoing extensive remediation efforts, including:

- Sediment dredging and capping
- Wastewater infrastructure upgrades
- Green infrastructure for stormwater adaptation

While these efforts represent environmental progress, they also drive property value increases, raising concerns over displacement and eco-gentrification.

### ***GeoAI, IoT, and Smart Redevelopment in Gowanus***

Technological interventions offer the potential for just redevelopment, including:

- IoT-enabled water quality monitoring
- AI-driven flood modeling and storm surge prediction
- Machine learning for green infrastructure placement
- Spatial analytics for predicting displacement risks

However, without affordability mandates, rent stabilization, and community land trusts, the redevelopment may replicate exclusionary trends observed in the High Line and Atlanta's BeltLine projects.

### ***Policy Considerations for Equitable Urban Renewal***

Key Questions for Gowanus

1. Who benefits from ecological improvements?
2. How can AI tools support equitable redevelopment?
3. What safeguards can prevent displacement?

### ***Recommendations:***

- Mandatory inclusionary zoning and rent stabilization
- Participatory data governance
- Bias audits for AI-driven zoning and property valuation tools

The conceptual framework proposed in this study offers planners, policymakers, and environmental justice advocates a replicable model for aligning AI technologies with equitable redevelopment goals. City planning departments, environmental NGOs, and AI oversight boards can use this framework to assess zoning decisions, evaluate algorithmic transparency, and implement procedural equity safeguards. As climate adaptation efforts intensify, applying this model in real-time zoning, infrastructure funding, and green redevelopment initiatives will be essential to ensuring that smart city transformations do not replicate historical patterns of exclusion but instead promote truly inclusive urban resilience.

### ***Global Applications: Scaling GeoAI in LMICs***

Cities in LMICs face resource and governance constraints but can still leverage scalable GeoAI tools:

- Open-source satellite imagery (e.g., Google Earth Engine, SERVIR)
- Cloud-based predictive modeling
- Low-cost IoT sensor deployment

Case studies, such as those in Jakarta, illustrate how machine learning and remote sensing support flood risk management, even in resource-constrained environments.

### ***SDG Alignment***

The Gowanus redevelopment aligns with multiple UN Sustainable Development Goals (SDGs), including:

- SDG 6 (Clean Water): IoT-enhanced monitoring improves wastewater management.
- SDG 11 (Sustainable Cities): AI supports inclusive urban planning.
- SDG 13 (Climate Action): Predictive models inform flood resilience.
- SDG 9 (Innovation & Infrastructure): GeoAI guides sustainable green infrastructure.
- SDG 10 (Reduced Inequalities): Equity-focused policies address displacement risks.

### ***Mitigating Eco-Gentrification***

Environmental improvements often result in rising housing costs and demographic shifts. To ensure inclusive urban sustainability, policymakers should implement:

1. Permanently affordable housing via zoning mandates
2. Rent stabilization and anti-displacement protections
3. Community land trusts to preserve long-term affordability

### ***Community-Led Smart City Initiatives***

Participatory approaches enhance the legitimacy and effectiveness of GeoAI:

- Citizen science for pollution monitoring



- Community-led mapping to track affordability
- Participatory AI training to audit planning models

These efforts foster transparency and community ownership in smart city planning.

### ***Toward a Global Framework for Inclusive Urban Resilience***

This research proposes a framework that integrates:

- GeoAI and IoT for real-time environmental insight
- Equity-based policy to prevent displacement
- Community engagement to ensure inclusive governance

The Gowanus Canal redevelopment offers a replicable model for global cities seeking to align environmental sustainability with social justice in the era of climate change.

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