

SYNAPTIC ECO-INTELLIGENT CITIES: A NEURO-INSPIRED FRAMEWORK FOR REGENERATIVE URBAN SYSTEMS IN THE SOCIETY 5.0 ERA

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Abstract

This article introduces the concept of Synaptic Eco-Intelligent Cities as a novel research framework that integrates ecological infrastructure, digital intelligence, and civic participation to support regenerative and inclusive urban transformation. Drawing from biological principles, the city is modeled as a distributed system in which these elements interact through feedback mechanisms analogous to synaptic connections in neural networks. The methodology is structured into three functional layers: environmental sensing embedded in ecological systems, decentralized computation through artificial intelligence and edge technologies, and participatory governance incorporating real-time citizen input. To operationalize the concept, a set of process-based indicators is proposed to assess urban system performance in terms of connectivity, responsiveness, ecological regeneration, civic engagement, and learning capacity. This approach shifts the focus from static sustainability metrics toward dynamic and context-sensitive measures of urban intelligence. The proposed framework is articulated through conceptual reasoning and system logic to clarify an emerging viewpoint in urban landscape design and planning aligned with the principles of Society 5.0.

Keywords: Synaptic urbanism; Eco-intelligent systems; Adaptive infrastructure; Urban regeneration; Society 5.0.

1. Introduction: Beyond Smart and Green Cities

In recent years, urban sustainability agendas have been dominated by two prevailing paradigms: smart cities and green cities. Smart cities have primarily focused on enhancing the efficiency and performance of urban services through digitalization, deploying technologies such as the Internet of Things (IoT), artificial intelligence (AI), and real-time data systems to optimize infrastructure and mobility (Wolniak & Stecula, 2024). Green cities, in contrast, aim to reduce environmental impacts by incorporating nature-based solutions, promoting biodiversity, and implementing measures to reduce carbon emissions and resource consumption (Richter & Behnisch, 2019). Despite important advances, both approaches exhibit limitations in addressing the increasingly complex, dynamic, and systemic nature of urban challenges.

Smart city frameworks have been criticized for prioritizing technological efficiency over social and ecological resilience, often reinforcing centralized control and digital divides (Colding et al., 2024). Similarly, green city initiatives, while ecologically oriented, tend to operate through fixed infrastructures and lack real-time

adaptability, limiting their ability to respond to rapid environmental and socio-economic shifts (Artmann et al., 2019). As cities face compounding pressures from climate change, resource depletion, and urban inequality, there is a need to rethink urban models beyond efficiency or mitigation, and toward dynamic, regenerative systems capable of continuous learning and adaptation (Bibri et al., 2023; Esfandi et al., 2024). This is especially evident in traditional urban governance models, where decision-making is centralized and often disconnected from the various subsystems that constitute the urban environment.



Figure 1. Limitations of the traditional urban model: centralization and systemic disconnection.

As illustrated in **Figure 1**, centralized decision-makers attempt to control domains such as transportation, energy, waste, and water management in isolation, without effective feedback loops or citizen involvement. This fragmented approach can lead to long commuting times, inefficient resource use, and limited adaptation capacity.

The vision of Society 5.0 (originally proposed in Japan) offers a relevant framework to address this transition (Alimohammadlou & Khoshsepehr, 2023). It calls for the integration of advanced technologies with human-centered and sustainable development goals, aiming to harmonize digital innovation, ecological balance, and social well-being. However, practical urban models that embody this convergence remain scarce (Bencekri et al., 2025). In this context, we propose the concept of Synaptic Eco-Intelligent Cities: a novel framework that draws inspiration from the distributed cognition and adaptive behavior observed in biological neural systems. Rather than treating cities as static infrastructures managed from the top down, this perspective conceptualizes them as living, decentralized systems where ecological, technological, and social components interact through dynamic feedback.

While related frameworks such as the 15-minute city emphasize proximity-based planning and resilient city models focus on infrastructure robustness, the synaptic approach extends these perspectives by embedding continuous feedback, distributed intelligence, and ecological co-regulation into the urban fabric. This comparative positioning highlights its originality and reinforces its relevance within current debates on sustainable urban futures.

This article introduces the core principles, architecture, and potential applications of Synaptic Eco-Intelligent Cities, providing a scientifically grounded, scalable, and realistic foundation for advancing research in regenerative urbanism. Rather than replacing existing urban models, this framework seeks to enhance them by embedding adaptive intelligence and ecological responsiveness into the structural and operational logic of cities. It outlines a biologically inspired approach in which ecological infrastructure, decentralized digital technologies, and participatory governance interact through dynamic feedback systems. This integrated model supports the development of context-aware, learning-oriented urban environments aligned with the vision of inclusive, sustainable, and intelligent cities.

2. Defining Synaptic Eco-Intelligent Cities

The concept of Synaptic Eco-Intelligent Cities emerges from the need to develop urban systems capable of continuous adaptation, learning, and regeneration in response to ecological and social signals. Unlike conventional models of smart or green cities, which often treat technology and nature as separate domains, this framework proposes an integrated vision that treats ecological infrastructure and digital intelligence as co-evolving components of the urban fabric. It draws inspiration from the functioning of biological neural systems (particularly the synapse) as a model for distributed information processing, coordination, and response.

In a biological brain, synapses are junctions through which neurons transmit information. They are dynamic, able to strengthen or weaken over time, depending on the frequency and nature of the signals they receive. This plasticity is fundamental to learning and memory (Ma & Zuo, 2022; Martella, 2023). Translating this to an urban context, a “synapse” can be understood as a functional interface between components of the city (such as a park, a sensor array, a renewable energy node, or a participatory platform) where flows of data, energy,

material, and decisions are exchanged, and through which adaptive behavior emerges.

In Synaptic Eco-Intelligent Cities, these interfaces are not centralized or rigidly hierarchical. Instead, they are distributed across multiple layers of the urban system and operate through decentralized mechanisms. Each urban node, whether ecological (e.g., a wetland, urban forest), technological (e.g., sensor networks, local energy grids), or social (e.g., a neighborhood council, co-design platform), can independently perceive its local context, process information, and interact with other nodes. This structure allows for more resilient and context-sensitive decision-making, as responses are not dictated solely by a central authority, but negotiated through real-time feedback across the system.

A practical example of this can be found in microclimate-responsive urban green spaces. In many cities, green areas suffer from underutilization or ecological stress due to insufficient integration with urban management systems (Feltynowski et al., 2018). However, by embedding sensors that monitor temperature, humidity, soil conditions, and human presence, these spaces can become active participants in urban metabolism. Using edge AI, such nodes can

trigger adjustments in irrigation schedules, control shading devices, or send alerts about invasive species, all without requiring centralized processing (Ahmed et al., 2021). Moreover, when aggregated at the district level, such data can inform broader strategies for heat island mitigation, water use optimization, or biodiversity enhancement.

What sets Synaptic Eco-Intelligent Cities apart is the eco-intelligent logic that underlies these interactions. Rather than viewing green infrastructure as a static amenity or backdrop to urban activity, it is treated as an intelligent and sensing layer of the city. Trees, soils, waterways, and green corridors are seen as systems capable of storing, processing, and responding to information, when coupled with digital interfaces and feedback mechanisms. For example, bioswales designed to filter stormwater can be monitored in real time to assess infiltration performance, water quality, and flow patterns. This information can then be used to optimize upstream land-use decisions, infrastructure maintenance schedules, or even engage community groups in ecological stewardship.

This approach aligns with emerging concepts in urban ecology and systems thinking, particularly the notion of cities as “complex

adaptive systems” (Nel et al., 2018; Nel, 2009). In such systems, the behavior of the whole emerges from the interactions among its parts, rather than being imposed externally. Synaptic Eco-Intelligent Cities operationalize this idea by enabling feedback-rich environments in which infrastructure, ecosystems, and citizens co-regulate urban functions. These functions include not only basic services like transportation or waste management, but also more subtle processes such as thermal comfort, mental health, cultural vitality, and environmental justice.

Crucially, the synaptic metaphor also extends to the temporal dimension of urban behavior. In biological systems, synaptic pathways that are frequently used become stronger over time, while unused ones weaken, a principle known as Hebbian learning (Lansner et al., 2023; Remme et al., 2021). Applied to cities, this suggests that patterns of interaction (such as citizen engagement in local energy decisions, or the frequency of biodiversity monitoring) should influence how resources are allocated, how priorities are set, and how infrastructure is adapted. In practice, this could involve systems that learn from usage patterns to reconfigure spatial layouts, modify traffic flows, or repurpose underutilized spaces.

The technological foundation for implementing synaptic systems in cities is already emerging. The proliferation of low-cost environmental sensors, improvements in wireless connectivity (e.g., 5G and LoRaWAN), and advances in edge computing enable data to be collected, analyzed, and acted upon at the local level (NotionAge, 2024; Schulthess et al., 2023). For instance, a decentralized air quality monitoring network can inform both citizens and urban managers about pollution hotspots in real time. When combined with machine learning, such systems can also predict future conditions and suggest interventions, such as temporary street closures, increased tree planting, or behavior change campaigns.

However, the eco-intelligent model is not solely a technical one. It fundamentally requires the inclusion of human agency through participatory and deliberative processes. Synaptic Eco-Intelligent Cities must embed what could be called civic synapses, mechanisms through which community input, lived experience, and local knowledge are integrated into the adaptive cycles of the city. This may include digital co-design platforms where residents propose and vote on green infrastructure projects, mobile apps that allow real-time reporting of

environmental issues, or citizen science initiatives that generate ecological data at a high spatial resolution.

An essential dimension of Synaptic Eco-Intelligent Cities is ensuring that eco-digital benefits are equitably distributed, particularly among marginalized communities that often experience disproportionate environmental risks and limited access to digital infrastructures. Research on inclusive smart cities highlights that without deliberate policies, technological deployments can deepen existing inequalities by privileging well-connected districts while leaving peripheral or low-income areas behind (Cardullo & Kitchin, 2019). To prevent this, synaptic frameworks must incorporate equity-based design principles, such as prioritizing resource allocation to vulnerable neighborhoods, supporting community-led data initiatives, and ensuring affordable access to digital platforms and green infrastructures (Heeks & Shekhar, 2019). Embedding these practices strengthens not only the social legitimacy of the model but also its long-term resilience by engaging diverse actors in co-creation and stewardship.

■ ■ The ethical implications of this model are significant. Without careful design, distributed intelligence systems risk

reinforcing existing inequalities, creating new forms of exclusion, or enabling surveillance-based governance (Filgueiras & Silva, 2022; Pereira et al., 2024). Therefore, a central tenet of the synaptic framework is just eco-intelligence, a commitment to ensuring that both ecological and digital benefits are equitably distributed, and that systems are transparent, accountable, and responsive to diverse communities. This involves not only technical safeguards (e.g., open-source algorithms, privacy-respecting data practices) but also institutional arrangements that promote shared ownership and stewardship.

To support implementation, the framework must be operationalized through indicators and performance metrics. These should go beyond traditional urban sustainability metrics to capture the degree of ecological responsiveness, system adaptability, and inclusivity. For example, cities could measure the “synaptic density” of their networks (i.e., the number and quality of connections between ecological, technological, and social nodes). Other indicators might include the response time of adaptive systems to environmental events, the diversity of actors engaged in co-governance, or the regenerative capacity of green infrastructure

as measured by soil health, biodiversity levels, or carbon uptake.

Synaptic Eco-Intelligent Cities are not a speculative vision, but a real and actionable paradigm shift in urban development. They build upon existing technologies and practices (such as sensor networks, AI, nature-based solutions, and participatory governance) but reframe them through a systems-based, biologically inspired lens. By doing so, they offer a way to transition from cities that merely sustain current functions toward cities that actively regenerate ecological and social systems through intelligence, feedback, and collective learning.

This model aligns with the transformative ambitions of Society 5.0, not by proposing futuristic technologies, but by reconfiguring relationships (between people, machines, and nature) so that cities become more than the sum of their parts (Deguchi, 2020). They become living systems, capable of evolving in harmony with the planet and the people who inhabit them.

3. Architecture and Functionality of Synaptic Eco-Intelligent Cities

The architecture of a Synaptic Eco-Intelligent City is grounded in the integration of three

interdependent subsystems: ecological infrastructure, digital-technological infrastructure, and participatory governance mechanisms. Each subsystem performs a distinct role, yet they interact through dynamic, decentralized feedback loops that allow the urban system to learn, adapt, and respond to internal and external changes. This architecture does not require speculative technologies; rather, it relies on the purposeful configuration of existing capabilities into a cohesive, responsive whole.

At the foundation of the synaptic architecture lies the ecological layer, composed of green and blue infrastructures such as urban forests, green roofs, rain gardens, restored wetlands, river systems, and permeable surfaces. These elements perform essential functions for climate regulation, air and water purification, urban biodiversity, and psychological well-being. However, in conventional urban planning, these assets are often under-monitored and disconnected from broader decision-making systems (Du et al., 2018). In a synaptic city, ecological infrastructure is enhanced with environmental sensing technologies that collect continuous data on temperature, humidity, soil moisture, vegetation health, hydrological cycles, and

carbon sequestration. These data streams feed into local or distributed processors (such as edge computing nodes) that analyze environmental conditions and generate targeted recommendations or automated responses.

Overlaying the ecological layer is the digital-technological layer, which includes networks of sensors, data platforms, AI models, and distributed control systems. These tools enable real-time analysis of the data collected from ecological systems, energy use, mobility patterns, and social interactions (Alahi et al., 2023; Bibri et al., 2024). Importantly, processing is not confined to centralized data centers; instead, intelligent functions are embedded at the local level (e.g., in park management systems, district energy controllers, or building automation systems) to enable timely, context-sensitive responses. For example, if a localized increase in temperature and air pollution is detected, the system might trigger the release of fine mist in public spaces, adjust public transit frequency to reduce emissions, or send alerts to nearby schools or health centers. Such responses are not based on static rules but on learning models that adjust over time as conditions and behavior evolve.

The third layer is participatory governance, which connects human actors (citizens, planners, researchers, and local governments) to the eco-intelligent system. Through mobile applications, civic platforms, and urban dashboards, citizens can report issues, contribute localized data (e.g., observations of wildlife, urban heat discomfort, or waterlogging), or participate in collaborative planning processes. These tools facilitate two-way communication, allowing the system to not only inform and educate but also receive feedback and preferences from users. This participatory architecture enhances transparency and ensures that technological decisions are aligned with local needs and values. For example, a neighborhood that experiences frequent flash flooding could collectively propose the installation of bioswales and monitor their effectiveness, with data and outcomes accessible to all stakeholders.

The interactions between these three layers form what can be understood as urban synapses, points of exchange where information, resources, and decisions are coordinated (see **Figure 2**). Unlike traditional urban systems, which rely on siloed departments and long planning cycles, synaptic cities emphasize functional

connectivity and response fluidity. For instance, when green infrastructure and energy systems are coupled through shared data flows, surplus energy generated by solar installations could be redirected to power water retention or irrigation systems during peak heat events, reducing strain on public utilities.

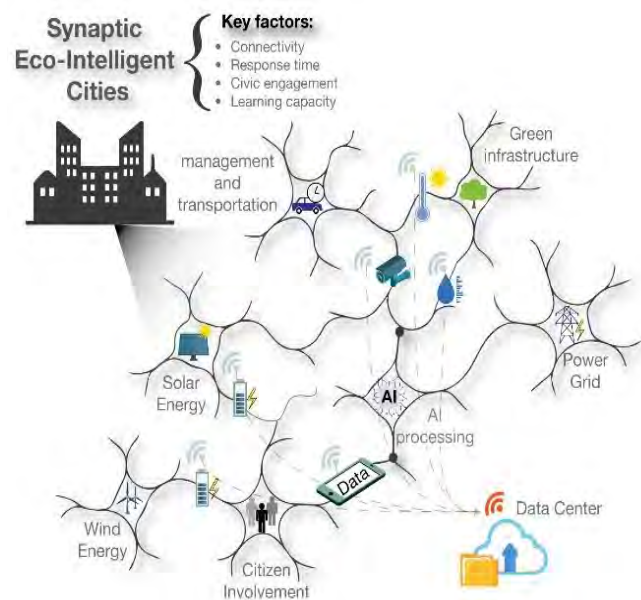


Figure 2. Urban synapses in eco-smart cities: connectivity and distributed learning.

Functionality in Synaptic Eco-Intelligent Cities is therefore not about automation for its own sake, but about co-regulation; a process through which nature, technology, and society modulate each other's behavior through continuous feedback. This co-regulation makes cities more resilient, not only in terms of infrastructure robustness, but in their ability to adapt governance,

redistribute resources, and recalibrate priorities in the face of uncertainty.

Furthermore, this architecture is inherently scalable and modular. It can be implemented incrementally in specific urban districts or infrastructure projects and expanded through interoperable systems. For instance, a single green corridor equipped with sensors and public feedback tools can evolve into a district-wide ecological monitoring network, which then informs broader planning decisions. This flexibility allows cities with limited resources to experiment and scale based on capacity, without requiring full-system overhauls.

The architecture of Synaptic Eco-Intelligent Cities reimagines the city as an interconnected set of ecological, technological, and social nodes that collaborate to maintain and enhance urban life. The core functionality is not driven by central command, but by the quality and density of the synaptic links that allow each part of the city to sense, decide, and act, responsively, locally, and regeneratively. This systemic perspective represents a shift from efficiency-driven automation to adaptive intelligence rooted in ecological integrity and social inclusion.

A critical challenge for Synaptic Eco-Intelligent Cities lies in the interoperability of technological systems. The integration of heterogeneous sensors, data platforms, and communication protocols often encounters barriers of standardization, leading to fragmented infrastructures that cannot seamlessly exchange information. Studies on smart city deployments show that without interoperable frameworks, local innovations risk remaining isolated “islands of technology” rather than functioning as part of a cohesive urban system (Gil-Garcia et al., 2020). Ensuring connectivity across ecological, technological, and social nodes, therefore, requires common standards, open data architectures, and interoperable platforms that allow scaling while preserving inclusiveness and transparency (Anthopoulos, 2017).

4. Application Outlook and Indicators of Synaptic Eco-Intelligent Cities

The practical implementation of Synaptic Eco-Intelligent Cities does not require starting from scratch; rather, it involves reconfiguring existing urban systems to operate in a more integrated, adaptive, and ecologically responsive manner. This section explores concrete applications of the synaptic model and proposes a preliminary framework

of indicators that could guide its evaluation and future refinement.

One of the most promising domains for synaptic applications is urban climate resilience. Increasing temperatures, extreme precipitation, and variable wind patterns pose severe risks to cities, particularly in vulnerable communities (Dharmarathne et al., 2024). In a synaptic city, distributed networks of environmental sensors embedded in green infrastructure (such as tree canopies, bioswales, and green roofs) can monitor microclimatic changes in real time. These data streams, when analyzed locally, can inform targeted interventions such as the activation of shade mechanisms, fogging systems, or localized cooling through green-water-soil interaction. At the same time, community-based platforms can receive alerts, suggest behavioral adjustments (e.g., reducing outdoor activity or avoiding heat-prone routes), and solicit input in areas where environmental stress is most acutely felt.

Another area of application is integrated water management, particularly relevant for regions facing both scarcity and flooding (Chen et al., 2025). Synaptic logic enables coordination between ecological systems like permeable pavements, retention ponds, green corridors, and technical infrastructure such as

storm drains and pumping stations. Sensors measuring soil saturation and water flow can autonomously adjust the operation of pumps or valves, while AI models predict runoff patterns and optimize stormwater retention. These systems can interact with citizen interfaces that report incidents of waterlogging or infrastructure failure, allowing for joint monitoring and rapid response.

In the energy domain, local energy loops exemplify synaptic dynamics. Rooftop solar panels, community batteries, and demand-responsive devices can interact through real-time energy trading platforms. These nodes exchange data about generation, consumption, and environmental conditions to balance loads at the neighborhood level (Mediwaththe & Blackhall, 2021). A green corridor equipped with solar panels, for instance, might divert excess energy to an adjacent community center or power irrigation systems. Feedback from residents can inform of energy priorities during peak hours or emergencies, contributing to social equity and adaptive planning.

Public space governance also benefits from the synaptic model. Data on pedestrian flows, user preferences, and environmental quality can be combined with adjusting lighting,

scheduling maintenance, or reprogramming spaces dynamically (e.g., pop-up shade zones, temporary gardens, or mobile service hubs). These actions are not dictated top-down, but co-designed through platforms that allow residents to express needs, validate system performance, and contribute to the continuous improvement of public infrastructure.

To assess the performance of Synaptic Eco-Intelligent Cities, traditional indicators of urban sustainability (such as greenhouse gas emissions, energy efficiency, or green space per capita) remain relevant but insufficient. What is needed is a new class of process-based indicators that capture the system's capacity to sense, adapt, and regenerate through distributed feedback. **Table 1** proposes a set of preliminary indicators, categorized into five functional domains.

1 **Table 1.** Proposed indicators for evaluating the performance of Synaptic Eco-Intelligent Cities.

Domain	Indicator	Definition / Description	Measurement Unit
Synaptic Connectivity	Functional node integration rate	Proportion of ecological and technological nodes with active data exchange	Percentage of total nodes (%)
Environmental Responsiveness	Adaptive response time	Average time between detection of environmental change and system-level response	Minutes (min)
Regenerative Performance	Ecological Surplus Index (ESI)	Composite score based on normalized carbon sequestration (kg CO ₂ /m ² ·year), water retention (L/m ²), and biodiversity	Normalized index (0 to 1, dimensionless)
Civic Engagement	Participatory feedback loop density	Average number of citizen inputs integrated into adaptive system responses, disaggregated by demographic group	Validated inputs per 1000 residents/month
Learning Capacity	System adjustment rate	Number of rule-based or automated system modifications based on new environmental or social data	Adjustments per system node per month (n/unit)

2

These indicators are not prescriptive but exploratory. They offer a way to initiate data collection and system reflection in urban environments, transitioning toward eco-intelligent governance. Over time, cities can

refine these metrics to fit local conditions, capacities, and goals.

Potential technical limitations must also be considered, including sensor malfunctions, cybersecurity vulnerabilities, and

maintenance challenges, which could disrupt feedback loops and temporarily weaken the resilience of synaptic systems if not properly anticipated.

Beyond technical performance, the long-term success of the synaptic model depends on governance innovation. Institutional frameworks must evolve to support polycentric decision-making, data interoperability, and participatory regulation. Urban laboratories and pilot districts offer promising sites to test these systems at manageable scales, document outcomes, and develop standards for broader adoption.

Synaptic Eco-Intelligent Cities are not just a vision but a methodology for transforming the function and evaluation of urban systems. By embedding intelligence into the interactions between ecology, technology, and society, cities can shift from reactive, siloed models to proactive, adaptive systems capable of facing the uncertainties of the 21st century. The indicators proposed here represent a starting point for this transformation, one grounded in evidence, transparency, and co-creation.

4.1 Practical Applicability and Pilot Experiences

Several cities worldwide have begun to experiment with initiatives that reflect the principles of Synaptic Eco-Intelligent Cities, even if not under this terminology. For instance, Amsterdam Smart City has developed a collaborative platform that integrates citizen participation, sensor-based monitoring, and data-driven decision-making for energy, mobility, and air quality management. This approach demonstrates the feasibility of distributed feedback loops between ecological infrastructure, technology, and citizens, which improves energy efficiency and strengthens collective governance of public space (Van Winden & Van den Buuse, 2017; Haarstad, 2017).

In Singapore, the Smart Nation initiative has combined green-blue infrastructure with advanced sensing networks to address challenges such as flooding and the urban heat island effect. Through real-time monitoring of rain gardens, drainage systems, and microclimates, the city applies predictive models to activate adaptive responses, including automated pumping, cooling mechanisms, and targeted community alerts.

These projects exemplify how ecological and technological nodes can operate as urban

synapses to sustain functionality under climate variability (Mell, 2024).

Another relevant case is Barcelona's superblocks (superilles), which serve as pilot districts for integrated mobility, environmental quality, and citizen participation. By reorganizing street layouts, embedding green corridors, and using digital platforms for co-design, the city has successfully reduced traffic, improved air quality, and increased the usability of public spaces. These interventions show that decentralized governance, ecological responsiveness, and participatory mechanisms are already being tested in practice, validating the applicability and scalability of the synaptic framework (Rueda, 2019).

5. Conclusion and Research Agenda

The concept of Synaptic Eco-Intelligent Cities introduces a biologically inspired, systems-oriented framework that challenges linear, centralized approaches to urban sustainability. Rather than proposing a new technological layer or isolated innovation, it emphasizes reconfiguring the functional relationships between nature, technology, and society in ways that allow cities to become reflexive, adaptive, and regenerative systems.

This perspective opens a space for methodological experimentation and critical inquiry that has so far remained peripheral in mainstream smart or green city literature.

A key strength of the synaptic approach lies in its emphasis on distributed agency and ecological co-intelligence. However, several open questions remain regarding its implementation, governance, and evaluation. First, there is a need to understand the limits and trade-offs of decentralized decision-making in urban systems. While local autonomy can enhance responsiveness, it may also create coordination challenges or inefficiencies without proper protocols and standards for interoperability across nodes.

The implementation of Synaptic Eco-Intelligent Cities also faces significant institutional limitations that must be acknowledged. Governance structures in many cities remain highly centralized and sectoral, which can hinder the coordination required for decentralized, feedback-based systems. Financial constraints are another barrier, as pilot projects often rely on temporary funding without long-term investment mechanisms, making it difficult to scale initiatives beyond experimental districts. Regulatory frameworks can also be misaligned, particularly when data-sharing

protocols, environmental standards, and procurement rules are not adapted to support polycentric governance and cross-sector integration. Addressing these institutional obstacles requires not only technological innovation but also reforms in financing models, urban governance, and regulatory flexibility to enable adaptive and participatory infrastructures.

Second, the model invites a deeper examination of data ethics, accessibility, and ownership. As ecological systems become instrumented and citizens engage in real-time feedback, issues of surveillance, exclusion, and algorithmic opacity must be addressed. Future research should focus on mechanisms to ensure that eco-digital infrastructures are inclusive, accountable, and co-governed.

Third, there is an opportunity to develop simulation tools and digital twins adapted specifically to synaptic logic, capable of modeling nonlinear feedback among ecological processes, infrastructure behavior, and citizen responses. These tools could support scenario analysis and inform design interventions that are both context-sensitive and scalable.

From an institutional perspective, further work is needed to explore hybrid governance

models that combine municipal leadership, civic engagement, and algorithmic coordination in decision-making processes. Research can also investigate how regulatory frameworks can evolve to support adaptive infrastructure without compromising transparency or democratic oversight.

On the empirical side, urban testbeds and pilot programs will be critical to assess the viability of synaptic configurations in different contexts. Comparative studies across cities (particularly those experimenting with participatory energy systems, climate-resilient green corridors, or distributed sensing networks) can help identify success factors, bottlenecks, and context-specific adaptations.

The synaptic framework also provides valuable opportunities for public policy by offering a more integrated lens for urban planning and sustainability strategies. At the local level, it can guide municipalities to design adaptive regulations that link ecological infrastructure with digital systems and community participation, fostering more context-sensitive responses to climate, mobility, and resource challenges. At the national scale, the framework can inform policies that encourage decentralized innovation, promote data interoperability,

and allocate funding to pilot projects that test regenerative urban practices. By embedding the principles of connectivity, responsiveness, and equity into planning instruments, Synaptic Eco-Intelligent Cities can support long-term transitions toward more inclusive and resilient urban policies.

Finally, there is a broader theoretical task: to advance interdisciplinary models that bridge urban ecology, systems engineering, human-computer interaction, and political theory. This requires moving beyond disciplinary silos and co-creating knowledge with planners, scientists, community organizations, and policymakers.

Advancing Synaptic Eco-Intelligent Cities requires the active collaboration of interdisciplinary teams that bridge technical, ecological, and social domains. Engineers contribute to the design of sensing networks, data platforms, and adaptive infrastructures; urban planners align these tools with spatial design and mobility systems; ecologists ensure that regenerative principles and biodiversity are embedded in decision-making; and social actors, including community organizations and policy stakeholders, provide the contextual knowledge and legitimacy necessary for implementation. Strengthening these

synergies transforms the synaptic model from a conceptual framework into a practical pathway, ensuring that cities evolve as truly adaptive and inclusive living systems.

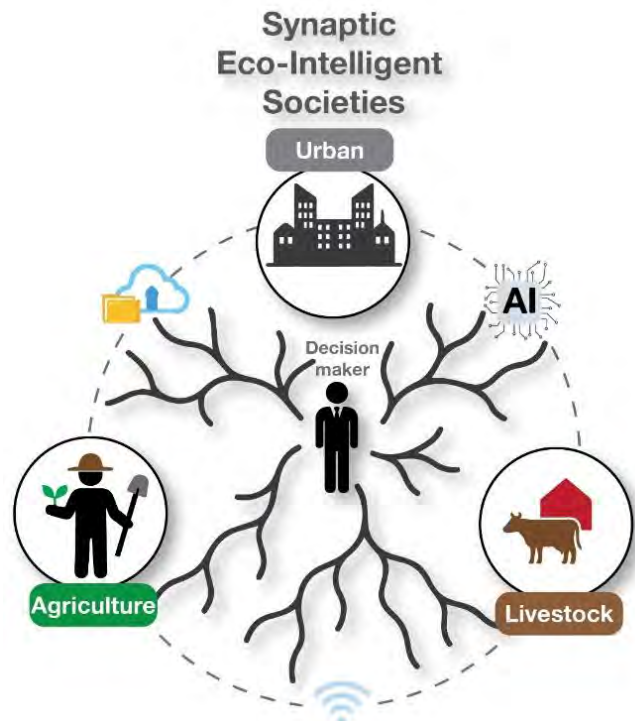


Figure 3. Integrated Decision-Making in Synaptic Eco-Intelligent Societies.

Synaptic Eco-Intelligent Cities are not merely a destination but a framework for inquiry; inviting cities to think, adapt, and evolve as living systems embedded in dynamic ecological and social realities. The path forward is not predetermined, but it is increasingly necessary. Extending this vision beyond urban boundaries, Synaptic Eco-Intelligent Societies propose a holistic integration of urban, agricultural, and livestock systems through interconnected

decision-making supported by AI and digital infrastructure. As illustrated in **Figure 3**, this model places the decision-maker at the center of a dynamic network that continuously senses, learns, and responds across diverse sectors, enabling coordinated strategies for sustainability and resilience.

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