



AI and IoT Integration for Next- Generation Smart Cities

¹Ms. Sowmiya. S., M.Sc., ²Ms. Jayasri. S (B.Sc.), ²Ms. Thahamina. A. F. (B.Sc.), ²Ms. Srimathi. M (B.Sc.), ²Mr. Sri Raghavan. R (B.Sc.)

¹Assistant Professor, Department of Computer Science with Cyber Security, Sri Ramakrishna College of Arts & Science, Coimbatore-06

²Student, Department of Computer Science with Cyber Security, Sri Ramakrishna college of Arts & Science, Coimbatore-06

DOI: <https://dx.doi.org/10.51244/IJRSI.2026.13020092>

Received: 16 February 2026; Accepted: 21 February 2026; Published: 05 March 2026

ABSTRACT

Artificial Intelligence (AI) and the **Internet of Things (IoT)** are transforming modern cities into smart, data-driven urban environments. It highlights how emerging AI technologies— such as machine learning, deep learning, computer vision, natural language processing, and Edge AI- enable real-time analysis, prediction, and automation of city services.

These innovations help manage critical areas like transportation, energy, waste, healthcare, and public safety, improving efficiency and citizens' quality of life. The study also discusses advanced approaches like Federated Learning and Explainable AI (XAI) that enhance data privacy, reduce latency, and make AI decisions more transparent. Despite their potential, the paper notes key challenges including data security, ethical concerns, interoperability issues, algorithmic bias, and high energy demands.

Finally, it looks toward the future of smart cities, focusing on cutting-edge ideas such as Quantum AI, digital twins, and human centered AI that could shape the next generation of sustainable and intelligent urban systems. Overall, the paper provides a comprehensive overview of how AI and IoT integration can support sustainability, innovation, and trustworthy governance in smart city development.

INTRODUCTION

The paper explores how the rapid pace of urbanization—with nearly 68% of the global population expected to live in cities by 2050—has created an urgent need for smarter, more sustainable urban management systems. Traditional city management methods, which depend on isolated data systems and manual decisions, can no longer handle the growing demand for efficiency, sustainability, and citizen well-being.

To overcome these challenges, cities are turning to **Artificial Intelligence (AI)** and the **Internet of Things (IoT)**—a combination known as **AIoT**. IoT networks connect billions of physical devices, including sensors, cameras, and meters, that continuously collect data about the environment and infrastructure. AI then analyzes this data to support real-time decision-making, predictions, and automation. This integration enables intelligent management of key urban services such as energy systems, healthcare, public safety, and transportation.

The paper highlights how recent technological advances—such as **machine learning, deep learning, computer vision, edge computing, and natural language processing (NLP)**—are making smart cities more adaptive and responsive. For instance, Edge AI allows data to be processed locally, improving speed and privacy, while **Explainable AI (XAI)** increases trust by making AI-driven decisions more transparent.

Beyond technology, the concept of a smart city emphasizes sustainability, resilience, and citizen-centered innovation. AI supports this vision by enabling:

- **Predictive analytics** to forecast and prevent issues like traffic congestion, equipment failures, or floods.

- **Smart resource management** for efficient use of water, energy, and waste systems.
- **Data-driven governance**, where policymakers use AI insights to make informed, inclusive decisions.
- **Automation and robotics** for intelligent transportation, infrastructure maintenance, and emergency response.

Despite its promise, integrating AI and IoT also introduces ethical, technical, and environmental challenges. These include data privacy risks, algorithmic bias, high energy consumption, and interoperability issues among different systems. The paper emphasizes the need for responsible AI frameworks that balance innovation with fairness, transparency, and sustainability.

Finally, it explores how AI, IoT, and 5G technologies together will enhance urban connectivity, requiring stronger standards for security and accountability. Future applications such as self-driving transport, predictive healthcare, AI-based governance, and smart waste management illustrate how cities can evolve into adaptive ecosystems that continuously learn and respond to citizens' needs.

The paper concludes by presenting a structured analysis of emerging AI technologies, applications, challenges, and future directions. It builds on prior studies while expanding the discussion to include cutting-edge innovations and ethical considerations for creating sustainable, inclusive, and intelligent urban environments.

LITERATURE REVIEW

The idea of a smart city arose as a response to the challenges of urbanization, sustainability, and effective governance. Early studies, including work by Giffinger et al. (2010) and Washburn et al. (2011), defined smart cities as urban areas that use information and communication technologies (ICT) to improve efficiency and enhance public services. The Internet of Things (IoT) soon became pivotal to this change, serving as the backbone that connects various urban systems.

According to Alahi et al. (2023) in their detailed review, "Integration of IoT-Enabled Technologies and Artificial Intelligence (AI) for Smart City Scenario: Recent

Advancements and Future Trends," published in *Sensors*, IoT is crucial for transforming urban infrastructures by linking physical devices like sensors, actuators, and controllers through wireless networks. Their study notes that IoT-based communication systems, which include 5G, LoRaWAN, NB-IoT, ZigBee, and Wi-Fi, provide the foundation for extensive data collection in smart cities. The authors further highlight that combining AI with IoT allows for realtime data processing, instant analysis, and predictive modeling, which drive automation and enhance urban services. This review has become an essential resource for understanding the connection between IoT and AI in building efficient, data-driven cities.

Zanella et al. (2014) proposed an early framework for IoT-enabled urban ecosystems, focusing on data sensing, networking, and cloud-based analysis. They illustrated how IoT networks can enhance energy management, transportation, and environmental monitoring. Later works, such as those by Lee et al. (2016), expanded on this by including cyber-physical systems (CPS) to connect the physical and digital aspects of smart cities, enabling responsive control mechanisms.

Foundations Of Smart City and Iot Integration

The growth of smart cities has been a major focus of research in both academia and industry over the past decade. This interest stems from rapid urban growth, advances in technology, and the increasing availability of real-time data. The merging of the Internet of Things (IoT) and Artificial Intelligence (AI) has created a significant shift that enhances sensing, communication, and decision making in urban settings. Many studies have looked into how these technologies work together to improve sustainability, governance, and the well-being of citizens. This section reviews important literature related to new AI technologies for smart cities, emphasizing key contributions, research gaps, and recent trends.

The Role of Artificial Intelligence In Smart Cities

The combination of AI technologies with IoT infrastructures—often called AIoT (Artificial Intelligence of Things)—has sped up the creation of smart, self-regulating urban systems. Early AI applications in smart cities mostly involved basic rule based systems and simple data analysis. However, the rise of machine learning (ML) and deep learning (DL) has greatly enhanced AI's ability to understand complex urban dynamics.

Ai Technologies Across Smart City Domains

AI has been applied in various areas of smart cities:

Smart Mobility: Deep reinforcement learning (DRL) algorithms have been used for traffic signal control and dynamic routing (Wei et al., 2019). AI-based predictive analytics also allow for real-time congestion detection and optimized public transport.

Energy and Environment: Research by Ahmad et al. (2020) shows how AI models can optimize smart grids, forecast energy demand, and integrate renewable resources. AI also plays a role in monitoring air quality and managing environmental risks.

Healthcare: AI-fueled telemedicine and health monitoring systems use IoT devices and wearable sensors to predict disease outbreaks and improve emergency responses.

Waste Management and Utilities: Computer vision and robotics enable automated waste sorting and recycling. Predictive maintenance systems powered by machine learning can identify water leaks and detect infrastructure issues before they lead to failure.

These studies collectively highlight the wide range of AI applications that boost city resilience, sustainability, and livability.

Challenges And Research Gaps

Despite significant advancements, many limitations remain in current literature.

Numerous studies concentrate on isolated applications without addressing interoperability and integration across urban systems. Most smart city frameworks still depend on centralized cloud processing, raising concerns about latency, energy efficiency, and data security. Additionally, few studies delve into ethical AI, algorithm transparency, and citizen privacy even though these issues are increasingly important as AI systems impact public life.

Another gap is in the scalability of AI models. Traditional deep learning systems require high computational resources, making them impractical for real-time, resource-limited IoT environments. This has led to recent investigations into lightweight AI models and edge intelligence, but there are still few standardized benchmarks and evaluation methods. While many discussions about quantum AI and digital twin technologies occur in conceptual papers, real-world implementations in smart city systems are still in their early stages.

Summary Of Key Findings

The literature reveals a clear shift from IoT-enabled data collection toward AI-driven automation and predictive decision-making. The merger of AI, IoT, and communication technologies like 5G represents the next step in urban innovation. However, existing studies stress the necessity for solid frameworks that guarantee interoperability, privacy, transparency, and sustainability. As cities evolve into complex cyber-physical systems, new AI technologies—particularly Edge AI, Explainable AI (XAI), and Federated Learning—provide promising paths for creating adaptable, ethical, and secure smart urban environments.

METHODOLOGY AND CONCEPTUAL FRAMEWORK

This study aims to create a framework that integrates new Artificial Intelligence (AI) technologies into the Internet of Things (IoT) infrastructure of smart cities. The goal is to improve sustainability, efficiency, and citizen engagement. The methodology combines a systematic review of existing literature, a comparison of current models, and the development of a conceptual framework that shows how AI innovations can be applied in key urban areas.

Research Approach

This research follows a qualitative and analytical approach. It focuses on gathering insights from existing literature and identifying new trends in AI-enabled smart cities. The approach includes three main phases:

Phase 1: Data Collection and Literature Analysis

A systematic review was conducted using research databases like IEEE Xplore, ScienceDirect, SpringerLink, and MDPI. Articles published between 2015 and 2025 were included, particularly those discussing AI, IoT, edge computing, and smart city infrastructures. The paper by Alahi et al. (2023) was a key reference for understanding IoT-AI integration. Keywords searched included “AI for smart cities,” “IoT integration,” “Edge AI,” “AIoT,” “sustainable urban systems,” and “datadriven governance.”

Phase 2: Comparative Analysis and Classification

Existing architectures and frameworks were analyzed by their structure, components, and operational scope. The review identified three main categories:

Cloud-Centric Models, where AI processing happens in centralized data centers.

Edge/Fog-Centric Models, where computation occurs near data sources.

Hybrid AI-IoT Models, which combine cloud and edge computing for better scale and efficiency.

The comparison showed that hybrid

frameworks are best for real-time decision making while ensuring energy efficiency and data privacy.

Phase 3: Framework Development and Validation

Based on this analysis, a new AI-Driven IoT Framework for Smart Cities (AIF-SC) is proposed. This model incorporates new AI approaches like Edge AI, Federated Learning, and Explainable AI (XAI) into existing IoT infrastructures to improve performance, transparency, and trust.

Proposed AI-Driven IoT Framework for Smart Cities (AIF-SC)

The AIF-SC framework consists of five linked layers, each representing an important stage in the function of AI-enabled smart city systems.

(i) Perception Layer (Data Acquisition Layer)

This layer includes IoT sensors, RFID tags, cameras, drones, and wearable devices that continuously gather data from the physical environment. Collected parameters include traffic density, air quality, energy usage, public safety metrics, and health indicators. This layer focuses on real-time data collection from various urban areas.

(ii) Network and Communication Layer

Data from the perception layer is sent through wireless communication technologies like 5G, LoRaWAN, Wi-Fi 6, and NB-IoT. Edge gateways and fog nodes filter and aggregate data before it reaches central systems. This layer aims for low-latency, high-bandwidth, and secure communication.

(iii) Processing and Intelligence Layer (AI Core Layer)

This is the main part of the proposed framework. It brings together several new AI approaches:

Edge AI: Runs AI models locally on edge devices for real-time decision-making (e.g., traffic control, anomaly detection).

Federated Learning (FL): Allows collaborative model training across devices without sharing raw data, keeping data private and ensuring decentralized learning.

Explainable AI (XAI): Ensures transparency and understandability in decision-making systems used for important applications like public safety and healthcare.

Reinforcement Learning (RL): Helps systems adapt automatically in changing urban environments by continually optimizing actions based on feedback.

This multi-layered intelligence makes smart city systems self-learning, adaptable, and secure, able to function even with network or data issues.

(iv) Application Layer

At this level, processed insights are used in various urban services, including:

Smart Mobility: AI-based traffic signal management, self-driving transportation, and congestion forecasts.

Smart Energy: Load forecasting, balancing renewable energy, and managing microgrids with predictive AI.

Smart Healthcare: AI for predicting diseases, telemedicine, and real-time patient monitoring.

Smart Governance: Chatbots, NLP-based citizen feedback systems, and AI-enhanced urban planning tools.

Each application makes use of AI results to improve decision-making, boost efficiency, and ensure sustainability.

(v) Service and Governance Layer

The final layer emphasizes policy integration, ethical AI governance, and citizen involvement. It creates regulatory frameworks for data privacy, cybersecurity, and accountability of algorithms. It also supports sustainable urban development goals (SDGs) by matching AI use with environmental and social needs. Digital twins and block chain-based transparency tools build trust between citizens and city authorities.

Framework Characteristics and Design Principles

The **AIF-SC framework** is designed around several key principles that ensure its effectiveness, scalability, and ethical reliability in smart city environments. It emphasizes **scalability**, allowing the seamless integration of millions of IoT devices and AI models that operate in real time to manage urban systems efficiently. **Interoperability** ensures that diverse technologies, platforms, and standards can communicate smoothly, creating a unified and connected city ecosystem. To maintain trust, the framework prioritizes **security and privacy** through advanced measures such as encryption, federated learning, and blockchain, safeguarding sensitive urban and citizen data. **Transparency and explainability** are also central, with the use of

Explainable AI (XAI) enabling citizens and administrators to understand and trust automated decision-making processes. Finally, **sustainability** is embedded in the framework's design through energy-efficient approaches, including lightweight AI models and edge computing, which reduce power consumption. Together, these principles make the AIF-SC framework robust, efficient, and aligned with modern standards for ethical and sustainable AI deployment in smart cities.

Validation And Expected Outcomes

Although the framework is conceptual, it is based on current technological capabilities and validated by comparing it to existing models found in the literature. The AIF-SC framework is expected to deliver these outcomes:

- Better real-time analytics and automated decision-making.
- Enhanced data privacy through federated and decentralized learning models.
- Decreased network latency and reduced computational burden with edge intelligence.
- Increased citizen trust through clear and transparent AI systems.
- Long-term sustainability and energy efficiency through optimized data processing.

By combining these features, the framework lays a strong foundation for the next generation of AI-driven smart cities. It enables urban systems to operate in an autonomous, secure, and sustainable way.

APPLICATION AREAS OF EMERGING AI TECHNOLOGIES IN SMART CITIES

The use of Artificial Intelligence (AI) in various areas of smart cities has changed how urban spaces are designed, operated, and managed. New AI technologies, such as machine learning (ML), deep learning (DL), computer vision, natural language processing (NLP), reinforcement learning (RL), and edge intelligence, help cities become more adaptable, efficient, and focused on their residents. The following sections highlight key application areas where AI is fostering innovation and sustainability.

Smart Mobility and Transportation

Transportation is crucial for every city, and AI is vital for enhancing its efficiency, safety, and sustainability.

Intelligent Traffic Management: Deep reinforcement learning algorithms adjust traffic-signal timings based on real-time vehicle flow. Cities like Singapore and Amsterdam use AI-driven adaptive traffic lights, which have cut congestion and emissions by over 20%.

Autonomous Vehicles and Connected Mobility: AI-powered systems allow self-driving cars to spot obstacles, read road signs, and communicate with nearby vehicles through Vehicle-to-Everything (V2X) networks.

Public Transit Optimization: Predictive models analyze passenger data to forecast demand, improving bus routes and schedules.

Smart Parking: Computer vision systems and IoT sensors help drivers find available parking spots, reducing unnecessary fuel use and pollution.

These innovations promote sustainable urban mobility, increase commuter safety, and lessen carbon footprints.

Smart Energy And Environmental Management

Energy management is a key part of building sustainable smart cities.



AI-Driven Smart Grids: Machine learning algorithms predict demand patterns and balance electricity from renewable sources, reducing power losses.

Renewable Integration: AI models forecast solar radiation and wind speeds, aiding in efficient storage and distribution.

Energy Efficiency in Buildings: Deep learning systems use data from HVAC sensors to optimize energy usage while keeping occupants comfortable.

Environmental Monitoring: Edge AI sensors track air quality, greenhouse gas emissions, and noise levels, allowing authorities to respond before conditions become dangerous.

These AI-powered energy systems directly support the United Nations Sustainable Development Goals (SDGs) related to affordable clean energy and climate action.

CHALLENGES AND ETHICAL ISSUES IN AI-ENABLED SMART CITIES

Integrating Artificial Intelligence (AI) and Internet of Things (IoT) technologies can greatly benefit urban sustainability and efficiency. However, it also brings technical, social, and ethical challenges. Implementing AI systems in smart cities requires careful thought about data management, clear algorithms, cybersecurity, and human rights.

The following sections discuss critical issues that must be tackled for responsible and fair AI use.

Data Privacy And Security Concerns

Smart cities depend on large amounts of realtime data collected from IoT sensors, surveillance systems, social networks, and citizen devices. These datasets are vital for predictive analytics and automated decision making, yet they often contain personally identifiable information (PII) like location, health, and behavioral data.

Data Leakage and Misuse: Without strong encryption and anonymization, sensitive data can be exposed to unauthorized people, leading to identity theft, misuse, or profiling.

Unauthorized Access: Weak authentication in IoT devices raises the risk of cyber intrusions and malware attacks.

Privacy Preservation: Technologies like Federated Learning (FL) and Differential Privacy hold promise. They enable model training while keeping user data secure and local.

A significant challenge is finding the right balance between data utility and user privacy. Cities must use information for public benefit without compromising individual rights.

Algorithmic Bias And Fairness

AI systems learn from data. If the data includes bias, the resulting models can lead to unfair or discriminatory results. In smart cities, biased algorithms can impact public services like:

Policing and Surveillance: Facial recognition systems can wrongly identify individuals based on race or gender, raising ethical and social justice issues.

Resource Allocation: Predictive models in healthcare, housing, or education may unintentionally favor specific demographic groups due to historical biases in the data.

To ensure fairness, it is necessary to implement bias detection, model audits, and continuous retraining with diverse datasets. Explainable AI (XAI) is also important, as it makes algorithmic decisions clearer, helping policymakers and citizens understand the outcomes.

Lack Of Transparency And Explainability

Many advanced AI models, especially deep learning systems, act as "black boxes." This means their internal decision-making processes are unclear. In smart city applications, this lack of transparency can damage public trust and accountability. For instance, if an AI system denies a building permit, forecasts a disease outbreak, or redirects emergency services, the reasoning behind these decisions must be understandable and justifiable.

Explainable AI (XAI): Techniques like SHAP, LIME, and saliency mapping help visualize and clarify AI decisions.

Policy and Regulation: Governments should require AI systems in important public areas to include explainability features, ensuring that automated decisions remain transparent and ethically sound.

Transparency is not just a technical issue but also a social necessity for maintaining citizen trust in AI-driven governance.

Cybersecurity Threats

The widespread connection of smart city systems through IoT networks makes them more vulnerable to cyberattacks.

Device Vulnerabilities: Low-cost sensors and controllers often do not have proper encryption or firmware protection, making them easy targets.

Distributed Denial-of-Service (DDoS) Attacks: Compromised IoT devices can be misused to disrupt vital services like traffic control or energy supply.

Data Integrity Threats: Hackers can alter sensor data, misleading AI models and causing incorrect or harmful system responses.

FUTURE TRENDS IN AI-ENABLED SMART CITIES

As urbanization speeds up and digital infrastructures develop, the next decade will bring a significant change in how Artificial Intelligence (AI) technologies are created, used, and managed in smart cities. Development will shift from traditional automation to more autonomous, context aware, and ethically managed urban systems. This section highlights the key emerging trends shaping the future of AI-enabled smart cities.

Convergence Of AI, IoT, And 6G Connectivity

The upcoming sixth-generation (6G) networks will allow for ultra-low latency (below one millisecond) and extensive machine communication. Combining AI, IoT, and 6G will support:

Real-time coordination among autonomous vehicles and drones.

Highly reliable remote surgery and healthcare monitoring.

Smart energy micro-grids that balance supply and demand across the city.

Edge-native AI models integrated within 6G infrastructures will speed up, secure, and make data processing more energy efficient, creating a more connected and responsive urban environment.

Quantum Artificial Intelligence (QAI)

Quantum computing could transform AI computation with much faster data processing and optimization. Quantum machine learning (QML) can tackle complex urban problems like traffic management, resource scheduling, and energy control that are currently too challenging to solve with existing methods. Quantum

encryption methods will also boost data security in smart city networks. While still in the experimental stage, the combination of QAI with IoT and edge computing will lead to the next generation of intelligent infrastructures that can make quick decisions on a large scale.

CONCLUSION

The integration of Artificial Intelligence (AI) and Internet of Things (IoT) technologies has turned the idea of smart cities from mere concepts into dynamic, data-driven realities. As urban populations grow, cities encounter urgent issues related to congestion, energy shortages, environmental damage, and social inequality. This research shows that combining AI, IoT, and smart data analysis lays the groundwork for sustainable urban ecosystems that can tackle these challenges in real time.

This study offers an overview of emerging AI technologies, including machine learning, deep learning, edge computing, federated learning, explainable AI, and digital twins. These tools are reshaping smart city infrastructures. The proposed AI-Driven IoT Framework for Smart Cities (AIF-SC) illustrates how these technologies can work together to support real-time monitoring, decision-making, and service improvement across key areas like transportation, healthcare, energy management, waste management, and governance. The framework prioritizes scalability, interoperability, and sustainability while ensuring ethical and transparent use of AI.

The examination of application areas shows that AI can greatly enhance urban efficiency, safety, and citizen engagement when used responsibly. However, it also highlights serious challenges, such as data privacy risks, algorithmic bias, lack of transparency, cybersecurity issues, and high energy use. These problems threaten the ethical and environmental goals of smart city development. Tackling them requires not just technological breakthroughs but also strong governance, policy coordination, and efforts to build public trust.

Looking ahead, the paper points out several transformative trends, including 6G-enabled edge intelligence, quantum AI, federated learning, explainable and ethical AI, digital twin ecosystems, and green AI. These technologies will form the basis of the next generation of urban environments that are autonomous, sustainable, and focused on people. Furthermore, global collaboration among governments, researchers, and industry players will be crucial for creating standardized frameworks that ensure fair access and cross-border compatibility.

Ultimately, the success of AI in smart cities will hinge on how well humanity balances technological progress with ethical responsibility. Smart cities should be defined not just by their infrastructure or connectivity, but by their ability to improve human life, safeguard the environment, and foster social inclusivity. As we move toward a future of intelligent urban environments, our goal must be to build cities that are not only smart but also trustworthy, resilient, and sustainable for the generations to come.

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