Assessment 3: Automation, Robotics, and Smart Cities

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As a technology expert for RoboX and SmartX, this paper provides comprehensive analysis and strategic recommendations on automation, robotics, and smart city technologies. The analysis addresses fundamental aspects of automation systems, AI-driven computer vision, robotic technologies, smart city infrastructure, and ethical considerations in technology deployment, with a focus on practical implementation strategies that drive value and innovation.

1 Exercise 1: Automation and Robotics

1.1 Introduction

Jensen Huang, Nvidia's CEO in recent times has positioned physical AI as the next frontier of AI that we ought to pay close attention to (Huang, 2024). Physical AI here making reference to the robotics ecosystem his company looks to support. In fact releasing Nvidia Isaac GR00T N1 as an open source model for humanoid reasoning and skills further cements Nvidia's commitment to collaboratively furthering this robotics narrative (NVIDIA Corporation, 2024).

We stand at an inflection point where AI, robotics, and automation converge to transform industries and urban environments. This transformation is materializing rapidly across sectors from manufacturing to healthcare. Organizations that fail to adapt risk obsolescence in an increasingly automated world. For RoboX, strategic recommendations focus on portfolio optimization, service-based revenue expansion, and leveraging Nvidia Isaac GR00T N1 capabilities. For SmartX, implementing ethical frameworks for smart city technologies while addressing security and privacy concerns will be critical. The future of automation and robotics has arrived, and today's strategic decisions will determine which organizations thrive in this new technological landscape.

1.2 Task 1: Fundamentals of Automation & Testing Methodologies

1.2.1 Automation Foundations

Automation makes reference to the implementation of technologies towards perfomance of tasks with minimal human intervention. Automation could be categorized into three general buckets (Groover, 2019).

- **Fixed Automation**: Systems designed for repetitive tasks with high volume and minimal variation, such as bottling lines or packaging operations
- **Programmable Automation**: Configurable systems that can be reprogrammed for different product variations, like CNC machines or robotic welding in auto manufacturing
- Flexible Automation: Highly adaptable systems capable of producing different products simultaneously with minimal downtime, exemplified by modern electronics assembly lines
- Integrated Automation: Comprehensive systems connecting multiple processes through centralized control networks, such as fully automated warehouses with coordinated picking, sorting and shipping operations

The ecosystem of automation systems comprises a combination of interconnected components including sensors, controllers, software interfaces, communication networks, and human-machine interfaces (HMI), all working together to enable autonomous operation (Johnson & Zhang, 2023). This architecture has two broad primary areas of implementation:

1.2.1.1 Enterprise / Industry

In industrial application, automation boosts overall plant / factory productivity by handling repeatable processes and effectively reducing human error(which could about as a result of natural fatigue (machines do not suffer this)). Machines use sensors to perceive their environment through technologies such as machine vision), actuators perform actions, and controllers act as the brains. These tools work in tandem with traditional power, network and safety systems to enable optimal functioning. Use cases include packagingend products at a product line, installing car doors at an auto manufacturing plant.

1.2.1.2 Consumer

In the consumer realm, automation primarily aims to improve user convenience such as through smart assistants controlling other connected home devices through voice. This segues smoothly into the pivotal role of automation in smart home tech that allows residents to achieve greater energy efficiency(through for example smart bulbs that dim during day) and improved security(security cameras that notify home owners of suspected intrusions, or motion sensor installed security lights that trigger an alarm on sensing movement)

1.2.2 Testing Methodologies

Effective quality assurance for automation systems follows a hierarchical testing framework commonly known as the testing pyramid (Cohn, 2009).

- Unit Tests: Validating individual components at the code level. This tests the most granular aspect of the system possibly at a software programs function level.
- Integration Tests: Verifying interactions between several combined code components or modules
- API (Application Programming Interface) Tests: Validating that different software components communicate and exchange data correctly and securely, could be internal to a software program or between two or more independent software programs
- **UI Tests**: Confirming proper functionality (encompassing performance, visual and functional aspects) of user interfaces

This hierarchical testing framework ensures comprehensive validation while prioritizing the most fundamental elements, enabling detection of issues early in the development lifecycle and a modular approach towards troubleshooting potential issues (Smith, 2023).

1.3 Task 2: AI in Computer Vision for IoT

1.3.1 Al in Computer Vision

Artificial Intelligence has positively transformed the abilities of vision systems in IoT devices by supercharging their ability to interpret and act upon visual data with never-before-seen accuracy. Using techniques such as deep learning, particularly convolutional neural networks (CNNs), machines can now recognize patterns and objects with accuracy that is either on par with or even surpasses human capabilities (LeCun et al., 2015). This advancement enables a wide ranging array of practical applications such as object detection, facial recognition, defect detection amongst others.

While speaking of equivalence with normal human sight, its is noteworthy that it is their very ability to exceed the human vision capabilites that they find their most practical and irreplaceable usecases in industry. For example in the manufacturing domain, for mission critical components such as aircraft engines, the ability to detect defects such as miniscule gaps and cracks that are invisible to the naked eye is paramount (Wang et al., 2018). AI-powered computer vision systems can analyze images and videos in real-time, identifying defects and anomalies with a level of precision and magnification that far exceeds human capabilities. This not only enhances product quality but also reduces the risk of costly recalls and catastrophic safety incidents.

1.3.2 Industry Applications and Real World Examples

AI-powered computer vision delivers transformative capabilities across multiple domains:

- Quality Control: Automated defect detection with classification capabilities that exceed traditional machine vision systems in both accuracy and adaptability, consider the General Electric's use of AI-powered computer vision to inspect turbine blades for defects, significantly reducing inspection time and improving accuracy (Smith, 2023)
- Assembly Operations: Real-time monitoring for process optimization, identifying bottlenecks and ensuring adherence to standard operating procedures (SOPs). For example, Tesla's use of AI-powered computer vision to monitor assembly line operations, ensuring that each step is performed correctly and efficiently
- Worker Safety: Continuous monitoring for hazardous conditions and PPE compliance, enabling proactive intervention before incidents occur
- Intelligent Video Analytics: Advanced capabilities including object detection, behavioral analysis, and predictive insights that extend far beyond traditional video surveil-lance, though consider the data privacy violations implications of this usecase where data subjects have to be duly informed and have to give consent to their data being used in this manner. For example, Amazon's use of AI-powered computer vision in its Go stores to monitor customer behavior and optimize store layouts

1.4 Task 3: Types of Robotic Machines & Mobile Robotics

There are various kinds of robotic machines used in automation, each suited for specific tasks and environments. The choice of robotic system depends on the application requirements, including payload capacity, precision, and operational environment. The following sections provide an overview of some of the different types of robotic systems and their applications.

1.4.1 Robotic System Taxonomy

The robotics landscape encompasses diverse machine categories optimized for specific applications (Siciliano & Khatib, 2016):

- Industrial Robots: Fixed automation systems including articulated arms, SCARA, and delta configurations. An example is the door welding thingamajig at a an automanufacturing plant, optimized for precision and speed in repetitive tasks.
- Collaborative Robots (Cobots): Systems designed to work alongside humans safely. Delta's D-Series robotic arms are examples of such, being a robotic arm used for Pick and Place, Palletizing and Machine Tending optimized to safely work alongside a human.

- Medical Robots: Purpose-built machines used in surgical procedures, rehabilitation, and telemedicine. For example, the da Vinci Surgical System by Intuitive is a robotic surgical platform that enables minimally invasive procedures with enhanced precision and control.
- Humanoid Robots: Anthropomorphic systems designed to navigate human environments and mimic human behavior, appearance and gait. For example, Figure 02, Figure's humanoid robot is touted as the first commercially viable automous human robot (Figure AI, 2024). It boasts a feature set employable across manufacturing, logistics, retail and warehousing.

While each domain asks for a bespoke set of capabilities, the current push towards AGI(Artificial General Intelligence) is driving towards a unified approach to adaptable humanoids that can perform a wide range of tasks across varied environments. With the open source release of N-Nvidia Isaac GR00T N1, Nvidia is positioning itself as a key player in this space, enabling developers to create versatile humanoid robots capable of learning and adapting to diverse tasks and environments

1.4.2 Mobile Robotics

Mobile robots represent a rapidly evolving sector characterized by autonomous navigation capabilities (Siegwart et al., 2011). These systems integrate sophisticated sensor arrays, positioning systems, and decision-making algorithms to operate effectively in dynamic environments. Some of the most notable and defining features of mobile robots include:

- 1. **Autonomous Navigation**: The ability to navigate complex environments without human intervention, using technologies such as LiDAR, computer vision, and GPS. For example, Boston Dynamics' Spot robot can autonomously navigate construction sites and perform inspections.
- 2. **Obstacle Detection and Avoidance**: Advanced sensor systems enabling real-time detection and avoidance of obstacles, ensuring safe operation in dynamic environments. For example, the Clearpath Robotics Husky A200 is a mobile robot designed for outdoor applications, equipped with advanced sensors for obstacle detection and navigation.
- 3. **Path Planning**: Algorithms that enable robots to determine optimal routes while avoiding obstacles and adapting to changing conditions. For example, the Fetch Robotics Freight robot uses advanced path planning algorithms to navigate warehouses and deliver goods autonomously.
- 4. Multi-Modal Operation: The ability to operate in various environments, including indoor and outdoor settings, and adapt to different tasks. For example, the Amazon Scout is a delivery robot designed for last-mile delivery, capable of navigating sidewalks and pedestrian areas.

1.5 Recommendations: Future Impact and Implementation Strategy

1.5.1 Industry Transformation Forecast

What used to be a futuristic pipe dream, a reserve of scifi movies, is now a reality. The unprecedented advancements in AI and inherently in robotics are set to transform the landscape of automation across multiple industries. As the biggest players in the space in terms of sheer innovation and investment, e.g Nvidia and Tesla, push the boundaries of what is possible and look to democratize access to these technologies(as recently seen with the open source release of N-Nvidia Isaac GR00T N1), the tech will become accessible to a wider audience. What will set apart organizations that thrive in this new era will be their ability to adapt and integrate these technologies into their operations and more strategically and more importantly those who will innovate around productizing this revolution will win even bigger.

Some of the most notable transformations expected in the next 5-10 years include:

- Retail: Fully automated stores with AI-driven inventory management and checkout systems
- Manufacturing: Evolution toward lights-out facilities with adaptive production capabilities and required reskilling for floor staff to facilitate cobot ecosystems
- Healthcare: Expansion of surgical robotics and teleoperation systems, enabling remote surgeries a boon for low-resource settings where access to such specialized care is limited though this will require a robust and reliable connectivity infrastructure to enable real-time remote surgeries
- Agriculture: Widespread adoption of autonomous equipment for farm management, including planting, harvesting, and monitoring crop health. For example, the John Deere See & Spray technology uses computer vision to identify and target weeds for herbicide application, reducing chemical usage and improving crop yields.
- Logistics: Fully automated warehousing and last-mile delivery solutions (already happening with Amazon Drone and Scout)

While these impressive advancements will undoutedly create unprecedented gains in productivity and efficiency, the potential for negative disruptions of the labor market linger in the backburner. Organizations and workers alike will need to quickly adapt to this new reality, with a focus on reskilling and upskilling to remain relevant in an increasingly automated world.

1.5.2 Implementation Best Practices

To maximize return on automation investments, RoboX should adopt a structured approach with these four key strategies:

- 1. Product Portfolio Optimization and Market-Driven Enhancement: Conduct a comprehensive analysis of the current product portfolio to identify high-margin offerings and untapped market opportunities. Map customer usage patterns and pain points to guide feature development priorities, enabling targeted enhancements that directly address validated market needs. This data-driven approach ensures development resources focus on capabilities that customers demonstrably value and are willing to pay premium prices for, maximizing return on R&D investment.
- 2. Service-Based Revenue Expansion and Ecosystem Development: Transform existing products into platforms by developing value-added services that generate recurring revenue streams beyond initial hardware sales. Implement subscription-based analytics offerings, predictive maintenance packages, and performance optimization services that leverage the installed base of RoboX products. This approach increases customer lifetime value while creating a more predictable revenue stream that enhances company valuation, all while utilizing existing technological assets more effectively.
- 3. Cross-Industry Application Scaling and Modular Architecture: Redesign core robotic technologies using modular architectures that enable rapid adaptation to adjacent markets with minimal additional development. Identify high-value components within current products that could be repurposed for new applications, creating economies of scale across previously disconnected market segments. This strategic repurposing of intellectual property and technical capabilities accelerates time-to-market for new offerings while significantly reducing development costs, enhancing overall portfolio ROI.
- 4. Strategic Developer Talent Investment for Next-Generation Technologies: Establish a dedicated talent acquisition and development program focused on Nvidia Isaac GR00T N1, synthetic data generation, and digital twin technologies. Build specialized teams that can leverage Nvidia Isaac GR00T N1's open-source framework to accelerate robotics development cycles while reducing physical prototyping costs. Invest in synthetic data capabilities to train AI models more efficiently than with real-world data alone, and develop digital twin infrastructure that enables virtual testing and validation of product improvements before physical deployment. This targeted talent investment creates a competitive advantage in development speed and simulation accuracy, significantly reducing time-to-market and physical testing costs while enabling more ambitious product innovations.

2 Exercise 2: Smart Cities

2.1 Introduction

Smart cities represent the cumulative intersection of modern digital technologies, forward looking urban planning, and sustainability initiatives designed to enhance quality of life for these cities' dwellers (Albino et al., 2015). As a technical advisor for SmartX, this section outlines strategies for developing AI-driven smart city solutions that create sustainable value.

2.2 Task 1: Smart City Features & Essential Technologies

2.2.1 Core Smart City Elements

Smart cities integrate multiple technological systems to create comprehensive urban management capabilities, enhanced governance and provision of citizen facing services. These systems encompass a wide range of applications, including:

- Smart Energy: Intelligent power distribution networks (Smart Grids), renewable energy integration, and power demand management
- Smart Mobility: Intelligent transportation multi-modal transit integrations that optimize last-mile traveller connections(air-train-tram), traffic optimization using computer vision
- Smart Safety: Enhanced emergency response, predictive policing and public security systems (though data privacy and the legality of biometric profiling would need to be addressed here), and public security systems
- Smart Infrastructure: Connected utilities, intelligent buildings, and adaptive public spaces

These systems deliver measurable improvements in resource efficiency, service quality, and citizen experience when effectively integrated into cohesive urban management frameworks.

2.2.2 Foundational Technologies

The technological foundation of smart cities comprises several interdependent systems (International Standards Organization, 2021):

• Internet of Things (IoT): Distributed sensor networks providing real-time data collection - consider these as the eyes and ears of the smart city, collecting data on everything from traffic patterns to energy consumption

- Artificial Intelligence (AI): Advanced algorithms enabling data analysis, predictive modeling, and decision-making automation these are the brains of the smart city, processing the data collected by the IoT sensors and making decisions based on it
- Cloud and Edge Computing: Distributed processing architectures balancing centralized management with local responsiveness by placing computing resources in close proximity to data sources
- Connectivity Infrastructure: High-capacity and high-speed networks such as 5G enabling system-wide communication these are essential for powering realtime data transfers from the IoT sensors

The integration of these technologies enables the pervasive intelligence that characterizes truly smart urban environments.

2.3 Task 2: Challenges & Concerns in Smart Cities

2.3.1 Implementation Barriers

Despite their potential benefits, smart city initiatives face substantial challenges that can hinder successful implementation and operation: * Data Security: Ensuring the integrity and confidentiality of the massive volume of sensitive data in interconnected systems is no small feat and requires huge investments in cybersecurity measures.

- Security Vulnerabilities: The inherent interconnectivity of these systems creates a larger attack surface for potential cyber threats
- Privacy Concerns: Ubiquitous data collection raising citizen privacy issues, particularly regarding unfettered biometric data collection and unconsented surveillance additional concerns abound regarding how that data is used, who has access to it, and how it is protected potential violations of GDPR and other data protection regulations

 Infrastucture Costs: High initial investment requirements for infrastructure development and maintenance, particularly in low-resource settings where the cost of building out the necessary infrastructure is prohibitive and is weighed in less favourably considering competing priorities for basics such as food, water and shelter
- Interoperability: Lack of standardization across systems and platforms, leading to integration challenges and vendor lock-in worth noting here is artificially created or politically motivated vendor lock-in where a government or city administration chooses to work with a specific vendor for political reasons, rather than based on the vendor's technical capabilities or suitability for the project. This can lead to suboptimal solutions and increased costs, as the chosen vendor may not be the best fit for the city's needs.

These challenges require comprehensive mitigation strategies to ensure successful implementation and sustainable operation.

2.3.2 Risk Management Approach

Risk mitigation in smart city implementations involves systematically identifying, assessing, and addressing potential threats to the smart city project success. A structured approach like the NIST Risk Management Framework would provide a comprehensive methodology for managing technological and operational risks through continuous assessment and adaptation. To address the specific challenges of smart city deployments, the following targeted strategies should be implemented:

- Data Security & Encryption: Implement end-to-end encryption, regular security audits, and advanced threat detection systems to safeguard sensitive data assets
- Security Architecture: Adopt security-by-design principles with defense-in-depth strategies and regular vulnerability assessments to reduce the attack surface of interconnected systems
- Privacy Protection: Establish transparent data governance frameworks with clear consent mechanisms, data minimization practices, and compliance with regulations like GDPR
- Financial Sustainability: Develop public-private partnership models with phased implementation approaches to distribute costs and demonstrate ROI through pilot projects
- Standards Adoption: Implement open standards and API-first architectures to ensure interoperability across vendors and preserve future flexibility

This comprehensive approach addresses both technical vulnerabilities and organizational challenges while maintaining focus on citizen-centric outcomes.

2.4 Task 3: Connectivity Methods in Smart Cities

2.4.1 Wireless Technology Ecosystem

Smart cities leverage multiple wireless technologies optimized for different use cases (Zanella et al., 2014).

2.4.2 Wireless Technologies

- **5G**: Fifth-generation cellular technology operating in sub-6 GHz and mmWave (24-100 GHz) bands, delivering up to 20 Gbps throughput, <5ms latency, and massive connectivity capacity (1M devices/km²) with network slicing capabilities to support diverse service requirements.
- LoRaWAN: Low Power Wide Area Network protocol utilizing unlicensed ISM bands with Chirp Spread Spectrum modulation, providing 10+ km range with minimal power

Wireless Technologies Comparison

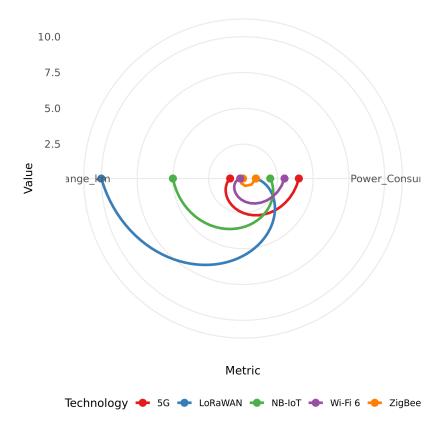


Figure 1: Wireless Technologies Comparison for Smart City Applications

- consumption (10-year battery life). Operates at 0.3-50 kbps data rates optimized for infrequent, small payload transmissions.
- NB-IoT: Narrowband IoT cellular technology operating in licensed spectrum with 180 kHz bandwidth channels, delivering 250 kbps downlink/20 kbps uplink speeds with enhanced coverage (+20dB link budget) for deep penetration and power efficiency supporting 10-year battery life.
- **ZigBee**: IEEE 802.15.4-based mesh protocol operating in 2.4 GHz band with 250 kbps data rate, implementing 128-bit AES encryption and self-healing mesh architecture supporting up to 65,000 nodes with 100m range per node.
- Wi-Fi 6: IEEE 802.11ax standard utilizing 2.4/5/6 GHz bands with 1024-QAM modulation, OFDMA, and MU-MIMO, delivering 9.6 Gbps theoretical throughput and 4x capacity improvement in dense environments through efficient spectrum utilization.

2.4.3 Wired Technologies

- Fiber Optic: Light-based transmission medium utilizing single-mode (1310/1550nm) or multi-mode (850nm) optical wavelengths, providing terabit-scale bandwidth (100+Tbps), sub-microsecond latency, and immunity to electromagnetic interference over distances exceeding 100km without amplification.
- Ethernet: IEEE 802.3 standard implementing multiple speed grades (1/10/40/100 GbE) over twisted pair or fiber, with Power over Ethernet (IEEE 802.3bt) delivering up to 100W for connected devices and Time-Sensitive Networking (TSN) capabilities for deterministic communication.
- Ultraethernet: Emerging standard targeting 800 Gbps to 1.6 Tbps using PAM4/PAM6 signaling and coherent optical technologies, optimized for backbone infrastructure with enhanced forward error correction and reduced power consumption per bit transmitted.
- PLC (Power Line Communication): Data transmission over electrical infrastructure utilizing frequencies from 50 kHz to 100 MHz, employing OFDM modulation with adaptive bit loading to achieve up to 500 Mbps throughput while navigating variable impedance and noise conditions.
- **DOCSIS 4.0**: Latest Data Over Cable Service Interface Specification enabling symmetrical 10 Gbps data transfer over existing coaxial infrastructure through Full Duplex DOCSIS and Extended Spectrum DOCSIS technologies, supporting low-latency applications and backward compatibility.

Each technology addresses specific requirements related to bandwidth, power consumption, range, and reliability. The choice of technology depends on the specific use case and deployment environment, with many smart city implementations leveraging a combination of wired and wireless solutions to achieve optimal performance.

2.5 Recommendations: Ethical Framework and Data Governance

2.5.1 Ethical Principles for Smart City Implementation

• Transparency:

- Clear disclosure of data collection practices and algorithmic decision-making
- Example: Amsterdam's Open City initiative publishing sensor documentation
- Aligns with ISO 37106 guidelines (Section 2.3.1)

• Equity and Inclusion:

- Ensure benefits reach all demographic groups to prevent digital divide
- Example: Barcelona's Digital Inclusion program with community tech hubs
- Follows ISO 37122 principles for sustainable communities (Section 2.3.2)

• Security Architecture:

- Defense-in-depth strategies protecting data ecosystems (Section 2.2.2)
- Example: Singapore's Smart Nation security assessments
- Implements NIST Cybersecurity Framework core functions

• Regulatory Compliance:

- Adherence to privacy standards for citizen data protection
- Example: Helsinki's MyData initiative implementing GDPR principles
- Addresses privacy concerns from Section 2.3.1

• Data Access and Control:

- Mechanisms for citizens to view, correct and delete personal data
- Example: Estonia's e-governance platform with access transparency
- Implements GDPR Article 15 rights (Section 2.3.2)

3 Conclusion

The convergence of automation, robotics, and smart city technologies presents unprecedented opportunities to enhance operational efficiency, sustainability, and quality of life. For RoboX, implementing the four-pronged strategy of portfolio optimization, service-based revenue expansion, modular cross-industry scaling, and strategic talent investment in Nvidia Isaac GR00T N1 capabilities will position the company at the forefront of the physical AI revolution that Nvidia's Jensen Huang has highlighted (Johnson & Zhang, 2023). Meanwhile, SmartX should embrace the ethical framework outlined—particularly focusing on transparency and equity while implementing rigorous security architecture that follows the NIST Risk Management Framework (International Standards Organization, 2021). Both organizations must balance

technological advancement with careful consideration of societal impacts, including privacy concerns and workforce transitions (Chen, 2023). By addressing the technical, ethical, and organizational dimensions of these technologies, organizations can create sustainable value while managing associated risks and maintaining citizen trust in increasingly connected urban environments.

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