

The Role of Artificial Intelligence Techniques in Analyzing the Sustainable Development Goals, Practice, Indicators, Values and Environment

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Abstract

This study introduces a framework aimed at enhancing the role of Artificial Intelligence (AI) in achieving the Sustainable Development Goals (SDGs). The primary objective is to address key challenges in AI applications, such as data scarcity, ethical concerns, and cultural diversity, by integrating explainable AI (XAI), simulation environments, and modular customization. The study emphasizes on region-specific datasets, synthetic data generation, and iterative refinement to improve AI solutions in sectors like poverty, healthcare, and climate action. The findings emphasizes on AI's potential to transform theoretical solutions into practical, scalable implementations, driving sustainable development. While addressing challenges like data quality, algorithmic bias, and regulatory issues, the study also highlights the importance of ethical principles and contextual adaptability to achieve long-term, inclusive progress toward the SDGs.

Keywords: Artificial Intelligence, Sustainable Development Goals, Explainable AI, Decision-Making.

1. Introduction

AI is an accelerating progress toward achieving the Sustainable Development Goals (SDGs). The 17 SDGs, which span a wide range of global challenges such as poverty eradication, quality education, good health, climate action, and sustainable urban development, provide a comprehensive framework for building a just, equitable, and sustainable future for all. As these goals demand concerted efforts across diverse sectors, AI is uniquely positioned to analyze colossal datasets, identify hidden patterns, and generate actionable insights that can solve some of the world's most complex problems. For instance, in addressing poverty (SDG 1), AI-powered tools can analyze satellite imagery and other socio-economic data to uncover insights about regions that have historically been underserved or overlooked. These insights allow governments and organizations to better allocate resources, target interventions, and develop more effective poverty alleviation strategies. In healthcare (SDG 3), AI is enhancing diagnostic accuracy, predicting disease outbreaks, and enabling personalized treatments, thus improving both the accessibility and quality of healthcare. This is particularly important in remote or underserved areas, where traditional healthcare infrastructure is limited or lacking. Through AI, healthcare systems can overcome the traditional barriers, ensuring that better healthcare reaches those who need it the most [1].

In the field of education (SDG 4), AI-driven technologies are revolutionizing the learning experience by providing customized and adaptive learning content. AI helps bridge educational gaps, catering to individual learning needs and making education more inclusive and effective. For students worldwide, this technology enables personalized learning experiences that take into account diverse needs, which can drastically improve learning outcomes. AI's role in climate action (SDG 13) is another critical example. By utilizing machine learning algorithms and predictive analytics, AI is improving weather forecasting, optimizing energy consumption, and contributing to the development of renewable energy solutions. AI systems can analyze environmental data to predict potential climate risks, helping policymakers prepare for extreme weather events and mitigate their impacts. Additionally, AI-driven innovations are essential for designing more energy-efficient urban infrastructure, thereby reducing greenhouse gas emissions and promoting sustainability in urban development [2,3].

In terms of urban sustainability (SDG 11), AI technologies are powering smart city solutions that optimize waste management, streamline traffic systems, enhance disaster

preparedness, and improve overall urban planning. These AI-driven initiatives are making cities more sustainable, livable, and resilient to the challenges of rapid urbanization and climate change. Despite AI's vast potential, integrating it into sustainable development efforts presents several challenges, particularly ethical concerns. Issues like data privacy, algorithmic bias, and unequal access to AI technologies must be addressed to ensure that AI benefits everyone. Without careful regulation and inclusive practices, AI could inadvertently deepen existing disparities and inequalities. Therefore, it is essential to establish transparent governance frameworks, inclusive policies, and collaborative approaches to ensure that AI serves as a tool for empowerment, rather than exclusion. This includes prioritizing investments in AI research and technology development in low-income regions, where access to AI resources may be limited. Furthermore, enhancing collaboration among governments, private sectors, and civil society is essential to expand AI's benefits globally. By pooling resources and expertise, these stakeholders can work together to develop AI solutions that are both scalable and culturally relevant, ensuring that no one is left behind in the pursuit of the SDGs. With ethical considerations at the forefront and global partnerships in place, AI can become a driving force in accelerating progress toward the SDGs, enabling a more equitable, sustainable, and prosperous future for all.

Table 1 Literature Review

S.no	Title	Technique used	Advantages	Limitations
1	A panoramic view and swot analysis of artificial intelligence for achieving the sustainable development goals by 2030: progress and prospects [4]	SWOT analysis	Provides a comprehensive view of AI technologies in relation to SDGs. Identifies strengths, weaknesses, opportunities, and threats for AI in the context of SDGs.	Focuses primarily on literature review and SWOT analysis. Limited scope in providing detailed solutions for specific SDGs. Challenges in addressing the complex and multifaceted nature of AI's impact on SDGs.
	Evaluating the role of	Bibliometric analysis of AI'	Highlights the potential of AI, particularly	Limited understanding of AI's long-term

2	Artificial Intelligence in sustainable development goals with an emphasis on “quality education”[5]	contributions to SDGs, particularly SDG4 (Quality Education)	Generative AI, to revolutionize education through personalized learning. Identifies key research trends and contributors in AI's role in sustainable development.	societal and ethical impacts. Gaps in literature on AI’s comprehensive role across all SDGs.
3	A systematic review of current AI techniques used in the context of the SDGs [6]	Supervised Learning (65% of studies) Unsupervised Learning (18%) Reinforcement Learning (17%)	Offers a clear taxonomy of AI techniques in sustainability applications. Identifies trends in the field and proposes new research directions. Highlights ANN dominance in solving environmental challenges.	Limited exploration of broader SDG applications beyond environmental studies. The reliance on existing literature may not reflect the newest AI advancements. Challenges in translating findings into real- world, large-scale solutions.
4	How artificial intelligence plays a role in achieving sustainable development goals [7]	Machine Learning (ML): For predictive modeling, resource optimization, and decision- making in sectors like healthcare and climate change.	Enhances efficiency and decision-making in critical sectors like agriculture, healthcare, and energy. Empowers developing countries by offering cost- effective, scalable solutions. Supports inclusive growth by reducing inequalities and improving resilience to environmental challenges.	Concerns over data privacy, bias, and ethical issues in AI deployment. Potential for uneven adoption due to resource constraints in low- income regions. Requires strong collaboration among stakeholders for responsible and equitable deployment.
	Roles and challenges of	Natural Language	Enhances educational access	The very goal of bringing clean water

5	ChatGPT and similar generative artificial intelligence for achieving the Sustainable Development Goals [8]	<p>Processing (NLP):</p> <p>Enables personalized education (SDG 4), disseminates agricultural knowledge (SDG 2), and supports healthcare communication (SDG 3).</p> <p>Knowledge Generation and Accessibility:</p> <p>Simplifies complex information for widespread use.</p> <p>Adaptive Learning Models:</p> <p>Provides tailored educational content to learners across regions.</p>	<p>and quality, reducing barriers to lifelong learning (SDG 4).</p> <p>Promotes dissemination of vital information in agriculture and healthcare.</p> <p>Facilitates global collaboration by connecting stakeholders with diverse knowledge bases.</p>	<p>to everyone in the world revolves around achieving SDG 6 and SDG 7 of the UN.</p> <p>Among the primary concerns are misrepresentation, lack of impartiality, and data privacy violations, which all contribute to inefficiency when Pursuing</p>
6	Application of AI/ML techniques in achieving SDGs: a	<p>Machine Learning (ML)</p> <p>Techniques:</p> <p>Explored for</p>	Provides a comprehensive overview of AI/ML trends in SDG research.	Overemphasis on AI's positive outcomes, with limited focus on potential adverse impacts.

	bibliometric study [9]	<p>sustainability solutions across various SDGs.</p> <p>Collaboration Networks: Identified country and university partnerships for AI/ML applications in SDGs.</p>	<p>Identifies key publications and sources advancing AI/ML for sustainability</p> <p>Highlights collaborative efforts across countries and institutions, offering direction for global research integration.</p>	<p>Regulatory and ethical concerns such as lack of transparency, trust, and universal standards in AI applications.</p> <p>Limited exploration of practical implementation challenges in low-resource settings.</p>
7	Modeling the effects of artificial intelligence (AI)- based innovation on sustainable development goals (SDGs): Applying a system dynamics perspective in a cross- country setting [10]	<p>Artificial Intelligence (AI) System Dynamics Modeling- based Simulation</p> <p>Institutional Theory (Technology Enactment Framework)</p>	<p>Provides forecasts of AI's impact on Sustainable Development Goals (SDGs). Covers both developed and developing countries across continents.</p> <p>Facilitates a systematic, data- driven approach to understanding global challenges.</p>	<p>Predictions are inherently uncertain due to variability in global and local contexts.</p> <p>Limited to 22 countries and specific timeframes (2022–2030).</p> <p>Outcomes may vary as new AI technologies emerge beyond the study's assumptions.</p> <p>Institutional and cultural differences might not be fully captured</p>

				in the model.
8	Impact of renewable energy utilization and artificial intelligence in achieving sustainable development goals [11]	Renewable Energy (RE) Artificial Intelligence (AI) Expert Elicitation Method	Supports 75 SDG targets; AI aids in achieving 42 targets. Promotes sustainability in environment, society, and economy. Highlights potential for future RE-AI synergies. Encourages regulatory insights for progress.	Negatively impacts 27 SDG targets. Limited focus on regional/local challenges. Lacks robust regulatory frameworks. Assumes exponential RE- AI growth without addressing barriers.
9	A Review of Artificial Intelligence Applications to Achieve Water-related Sustainable Development Goals [12]	AI applications in water sector: predictive maintenance, water demand forecasting, water quality monitoring, disaster prediction.	Enhances efficiency in water management. Supports SDG targets 3, 6, 11, 15. Enables real-time monitoring and accurate forecasting. Facilitates disaster prediction with better lead times.	Adoption challenges include affordability and accessibility in some regions. Dependence on data availability and quality.
10	AI for monitoring the Sustainable Development Goals and supporting action and policy	Machine Learning (ML) for modeling SDI measurements.	Addresses SDG measurement gaps with AI-driven insights. Enhances evidence- based	Challenges in data availability and quality for accurate AI modeling. Governance issues around evidence collection and

	development [13]	Explainable AI (XAI) for presenting insights in a human- friendly manner.	<p>polymaking.</p> <p>Facilitates effective communication of results to policymakers</p> <p>Interdisciplinary approach improves policy relevance and applicability.</p>	<p>communicatio n</p> <p>Potential resistance to AI adoption in policymaking</p>
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The integration of AI into the achievement of SSDGs) faces significant technical and systemic challenges, which limit its transformative potential. One of the major issues is the lack of high- quality, comprehensive, and globally representative datasets. This is especially true for low- income and underdeveloped regions, where the data collection infrastructure is either weak or does not exist at all, leaving big gaps that prevent AI from making accurate and fair predictions. Furthermore, ethical concerns such as data privacy, algorithmic bias, lack of transparency, and accountability severely impact trust in AI solutions. Without robust frameworks to address these issues, AI systems risk reinforcing existing inequalities or introducing new ones, especially in vulnerable communities. Resource disparities exacerbate the uneven adoption of AI technologies, where wealthier regions benefit from advanced AI solutions while low-resource settings struggle with basic implementation. Similarly, there is a limited exploration of practical, localized applications of AI that consider unique regional, cultural, and economic contexts, making many AI-driven solutions are unsuitable for diverse settings. This limitation is compounded by inadequate regulatory frameworks that fail to provide guidance on the ethical and responsible use of AI across different jurisdictions, leading to inconsistencies in implementation and enforcement. Another critical issue is the over-reliance on bibliometric analysis and theoretical studies, which often prioritize academic insights over actionable, field-tested solutions. As a result, these approaches fail to address real-world challenges, such as operational scalability, technological accessibility, and societal acceptance.

2. Analyzing the SDG, Practice, Indicators, Values and Environment

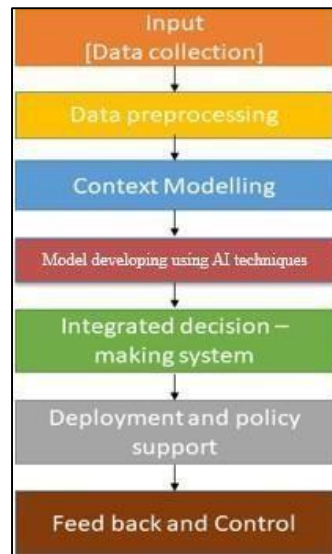


Figure 1. AI System Development Framework

It brings together a unique combination of high-quality, region-specific datasets in environmental, social, and economic indicators in the development of an AI-driven framework (Figure 1) to address challenges related to sustainable development at a local level. The first stage involves extensive data collection exercises ensuring that datasets are complete and relevant to the needs of a particular region. Next comes the step of pre-processing of the data to weed out inconsistencies, biases, and gaps using techniques like normalization and standardization to ensure that the data is clean, consistent, and thus amenable to analysis. In the context modelling phase, the AI system integrates critical regional factors such as cultural norms, resource availability, infrastructure, and local policy frameworks [14]. By analysing these context-specific elements, the model becomes more attuned to the unique challenges and opportunities faced by different regions. This contextual knowledge is then woven into the AI model to ensure that its outputs are not only technically sound but also relevant and actionable for local needs. For the model design, two key components are integrated. First, an XAI component ensures the transparency and interpretability of the decision-making process of AI, for human stakeholders like policymakers and community leaders. It might involve using attention mechanisms, decision trees, or feature attribution methods that allow users to understand the reasoning behind AI predictions. The model is optimized for transparency and accuracy using metrics such as interpretability scores, predictive accuracy, and precision. This

guarantees that the AI is not only effective but also fair and aligned with the ethical considerations of the target communities. The Deployment and Decision Support phase focuses on how to translate the outputs of the AI model into actionable insights for policymakers. Recommendations made by the system will be interpretable, directly applicable in local policy development, and supportive of community involvement in the policy development process through participatory decision-making tools, enabled by XAI, through which communities can interact with the AI's reasoning and contribute to the decision-making process. More importantly, feedback mechanisms are implemented to continuously collect input from both policymakers and community members to make sure the model is constantly being refined based on real-world feedback. The approach contains key strategies to overcome challenges in integrating AI to achieve Sustainable Development Goals (SDGs). Region-specific high-quality datasets are considered to overcome the data gaps of low-income regions, thus giving comprehensive and representative inputs. Integration of explainable AI (XAI) provides answers to ethical issues of algorithmic bias, transparency, and accountability to establish trust and assurance with ethical norms. In all, the approach guarantees region-specific, pragmatic solutions that are applicable and scalable by infusing regional and cultural context into AI models. Clear regulatory frameworks and collaboration among the involved stakeholders provide stability and the necessary guidance for responsible AI use. On the other hand, participatory decision-making tools enable community engagement and inclusivity, particularly for addressing SDGs such as gender equality and reduced inequalities. Bridging theory and practice through field-tested, actionable solutions, this is further supported by dynamic feedback mechanisms that fine-tune the system with real-world input. In addition, the local capacity-building initiatives and equitable distribution of AI resources are meant to address the disparities so that the AI-driven solutions are effective, ethical, and sustainable.

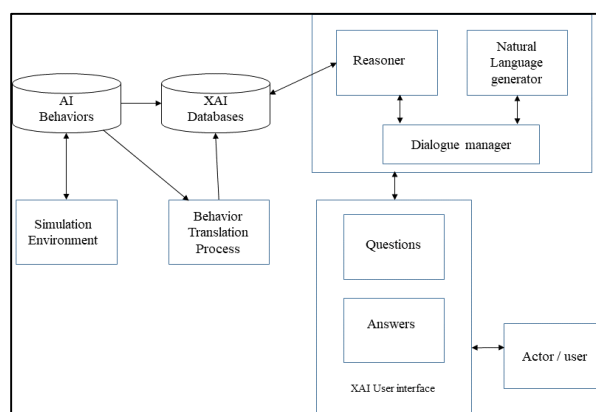


Figure 2. Architecture of explainable AI [15]

The architecture (Figure 2) addresses several limitations of AI in achieving Sustainable Development Goals (SDGs) by emphasizing explainability, inclusivity, and adaptability. The integration of simulation environments with explainable AI (XAI) databases compensates for the lack of high-quality datasets in underdeveloped regions, enabling the generation of synthetic data and iterative refinement of models. Ethical concerns like bias, transparency, and accountability are mitigated through the "Reasoner," which ensures decision-making processes are clear and justifiable, and the "Natural Language Generator," which provides users with comprehensible explanations. By utilizing an accessible XAI user interface, the system democratizes AI adoption, making it suitable for low-resource settings while supporting localized applications through modular customization of AI behaviors [15]. Furthermore, the transparent feedback loop facilitated by the dialogue manager empowers stakeholders to address regulatory gaps and evaluate compliance with ethical principles. The simulation environment and iterative behavior translation process enable the transition of theoretical AI solutions into practical, scalable, and field-tested implementations, reducing predictive uncertainty and enhancing societal acceptance. Finally, the architecture enhances collaboration among diverse stakeholders by providing interactive interfaces for real-time dialogue, ensuring that AI solutions are inclusive, culturally sensitive, and aligned with long-term societal and ethical considerations.

AI offers transformative advancements in sustainable development, including precision in resource management, climate modeling, and optimization of energy use. Machine learning facilitates accurate predictions and region-specific interventions, addressing local challenges effectively. However, limitations persist, such as data biases, lack of transparency in AI models, and ethical concerns regarding accountability. Low-resource regions face challenges with data availability and technical capacity. Additionally, over-reliance on AI without human oversight risks misaligned outcomes. Bridging these gaps requires integrating explainable AI (XAI), ethical frameworks, and capacity-building measures to ensure AI-driven solutions are equitable, transparent, and culturally adaptable. AI offers transformative advancements in sustainable development, including precision in resource management, climate modeling, and optimization of energy use. Machine learning facilitates accurate predictions and region-specific interventions, addressing local challenges effectively. However, limitations persist, such as data biases, lack of transparency in AI models, and ethical concerns regarding accountability. Low-resource regions face challenges with data availability and technical

capacity. Additionally, over-reliance on AI without human oversight risks misaligned outcomes. Bridging these gaps requires integrating explainable AI (XAI), ethical frameworks, and capacity-building measures to ensure AI-driven solutions are equitable, transparent, and culturally adaptable. Challenges in implementing AI for sustainability include data scarcity in underdeveloped regions, algorithmic biases, lack of interpretability, and limited stakeholder trust. Ethical and regulatory gaps further hinder adoption. Solutions involve developing high-quality, region-specific datasets and using synthetic data generation in simulation environments. Incorporating explainable AI ensures transparency, while participatory decision-making tools enhance inclusivity. Regulatory frameworks and international collaboration can establish accountability and trust. Local capacity-building initiatives and equitable access to AI resources address disparities. Continuous feedback loops between stakeholders and AI systems ensure adaptive, practical, and community-aligned solutions for sustainable development goals.

3. Conclusion and Future Work

The findings emphasize on AI's transformative ability to translate theoretical solutions into practical, impactful applications for sustainable and inclusive development. Future efforts will focus on expanding region-specific datasets, enhancing strategies to mitigate algorithmic bias, and scaling field testing in diverse, resource-constrained settings. Strengthening stakeholder engagement and addressing regulatory complexities will be critical to ensuring AI solutions are ethical, scalable, and responsive to local contexts, driving tangible progress toward achieving the Sustainable Development Goals (SDGs).

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