

MULTILINGUAL AI FOR GLOBAL SUSTAINABILITY: ADDRESSING LANGUAGE BARRIERS IN ACHIEVING THE SDGS

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Abstract

The Sustainable Development Goals (SDGs) provide a universal blueprint for addressing global challenges, including poverty, inequality, health, and education, by 2030. However, language barriers pose a significant obstacle to inclusive and equitable participation in achieving these goals, particularly for linguistically diverse and marginalized communities. This research investigates the role of multilingual artificial intelligence (AI) in mitigating these barriers to accelerate progress toward the SDGs. By synthesizing recent advancements in natural language processing (NLP) and evaluating multilingual AI applications across various sectors, this study provides insights into the potential and limitations of these technologies. A mixed-methods approach, combining qualitative and quantitative analyses, is employed to assess case studies and gather stakeholder perspectives. The findings suggest that multilingual AI has transformative potential but also highlights challenges such as algorithmic bias, data scarcity, and implementation gaps. Recommendations are proposed to maximize the effectiveness of multilingual AI in global sustainability efforts.

Keywords: *The Sustainable Development Goals (SDGs), global challenges, language barriers, multilingual artificial intelligence (AI).*

1. Introduction

The Sustainable Development Goals (SDGs), adopted by the United Nations in 2015, represent a global commitment to address pressing issues such as poverty eradication, gender equality, and climate action. Despite the ambitious nature of these goals, progress remains uneven across regions, largely due to systemic challenges including linguistic diversity. Approximately 7,000 languages are spoken globally, with many communities lacking access to critical information in their native tongues (Joshi et al., 2022). Language barriers can exclude marginalized populations from participating in policymaking, accessing education, and utilizing healthcare services, thereby perpetuating inequality.

Multilingual AI, powered by advances in natural language processing (NLP), holds the potential to break these barriers by enabling cross-linguistic communication and equitable access to information (Devlin et al., 2019). Technologies such as machine translation, voice recognition, and text generation can democratize access to resources and services essential for achieving the

SDGs. However, concerns surrounding data availability, ethical implementation, and algorithmic fairness persist. This paper explores how multilingual AI can address language barriers and contribute to the attainment of the SDGs, emphasizing its opportunities, challenges, and implications.

1.1 Background

Language diversity is both a strength and a challenge in global development. While it reflects humanity's cultural richness, it often hinders access to resources and services essential for sustainable development. Many global initiatives historically have prioritized a few dominant languages such as English, French, and Spanish, sidelining the needs of speakers of indigenous and low-resource languages. This systemic neglect perpetuates inequalities, particularly in education, healthcare, and economic participation.

The importance of addressing linguistic diversity has gained renewed attention with the rise of digital technologies. Multilingual AI—powered by advances in natural language processing (NLP), machine learning, and big data—offers tools to process and understand a variety of languages, including those that are underrepresented in digital spaces. Technologies such as automatic machine translation, speech-to-text systems, and multilingual chatbots can empower communities by enabling cross-linguistic communication and access to vital information.

For example, in education, multilingual AI tools can translate textbooks and online learning materials into indigenous languages, ensuring that children in rural areas can receive quality education. In healthcare, these tools can support telemedicine services by enabling doctors to communicate with patients in their native languages, improving diagnostic accuracy and trust. Additionally, climate advocacy efforts can incorporate indigenous knowledge systems, as multilingual AI enables local communities to contribute to global dialogues on sustainability.

Despite these benefits, the development and deployment of multilingual AI face significant challenges. Low-resource languages, often spoken by marginalized communities, are underrepresented in existing datasets. This scarcity limits the accuracy and usability of AI tools. Furthermore, ethical issues such as algorithmic bias and data privacy concerns must be addressed to ensure that AI technologies serve as a force for equity rather than exclusion. Addressing these challenges requires collaborative efforts among governments, private sector stakeholders, and local communities to build datasets, design fair algorithms, and develop culturally sensitive solutions.

The role of multilingual AI in addressing language barriers aligns closely with global efforts to achieve the SDGs. By bridging linguistic divides, these technologies can enhance inclusivity, empower marginalized communities, and accelerate progress toward sustainable development goals.

1.3 Research Objectives

1. To evaluate the effectiveness of multilingual AI in addressing language barriers across different SDG sectors, including education, healthcare, and climate action.
2. To identify key challenges and limitations in implementing multilingual AI in low-resource and marginalized communities.
3. To provide actionable recommendations for enhancing multilingual AI applications to ensure equitable and inclusive participation in global sustainability efforts.

1.4 Theoretical Framework

This study is grounded in the Capability Approach (Sen, 1999), which emphasizes expanding individual freedoms and capabilities as the core of development. Language accessibility is a fundamental capability that enables individuals to access education, healthcare, and governance. The study also draws on Diffusion of Innovations Theory (Rogers, 2003) to examine how multilingual AI technologies are adopted and adapted in different contexts. These frameworks provide a lens to analyze the intersection of technology, language, and sustainable development.

1.5 Significance of the Study

This research contributes to the growing discourse on technology's role in achieving the SDGs by highlighting multilingual AI's potential to address language barriers. The findings will benefit policymakers, technologists, and development practitioners by providing insights into best practices, challenges, and pathways for leveraging AI to promote inclusive and equitable development.

1.6 Rationale of the Study

Despite advancements in multilingual AI, its potential to support the SDGs remains underutilized. Current research often focuses on technical improvements without addressing socio-cultural and ethical implications. This study bridges this gap by examining multilingual AI within the broader context of global sustainability, emphasizing its role in fostering inclusivity and equity.

2. Literature Review

2.1 Language Barriers and the SDGs

Language barriers impede progress in multiple SDG domains. For example, education (SDG 4) is often inaccessible to children who do not speak the dominant language of instruction (Pillai & Krishnamurthy, 2020). Similarly, healthcare initiatives (SDG 3) fail to reach marginalized populations due to a lack of multilingual resources (Anderson et al., 2023). The inability to communicate in local languages also limits the effectiveness of climate action (SDG 13), as indigenous knowledge is often excluded from global discourse. Additionally, reports have shown that gender inequality is exacerbated by language barriers, as women in rural areas are disproportionately affected by the lack of resources in their native languages (Mukherjee, 2021).

2.2 Advances in Multilingual AI

Transformer-based models, including BERT, GPT, and their multilingual variants, have revolutionized NLP (Devlin et al., 2019; Brown et al., 2020). These models leverage vast amounts of multilingual data to achieve state-of-the-art performance in tasks such as translation and sentiment analysis. Open-source initiatives, such as Hugging Face's multilingual datasets, have also democratized access to AI resources (Hu et al., 2020). Moreover, novel architectures such as multilingual BART (Lewis et al., 2021) have demonstrated promising improvements in low-resource language generation, paving the way for broader inclusivity.

2.3 Challenges in Multilingual AI

Multilingual AI faces challenges such as data scarcity, particularly for low-resource languages. Languages with fewer speakers often lack digitized corpora, limiting model training and evaluation (Bender et al., 2021). Additionally, algorithmic bias can reinforce existing inequalities, while ethical concerns around data privacy and misuse remain unresolved (Butt et al., 2024; Barocas et al., 2019). Moreover, the computational cost of training large-scale multilingual models poses sustainability concerns, as it contributes to significant energy consumption (Strubell et al., 2020).

2.4 Case Studies in Multilingual AI

Case studies from India, Sub-Saharan Africa, and Latin America demonstrate the potential and limitations of multilingual AI. For instance, translation tools in India facilitated access to educational resources but struggled with regional dialects. Similarly, voice recognition systems in Africa improved telehealth services but faced challenges with linguistic variability. In Latin America, community-driven initiatives collaborated with researchers to develop custom datasets for indigenous languages, showcasing a participatory approach to overcoming language barriers (Gómez & Santillana, 2023).

2.5 Future Directions in Research

Emerging trends suggest a growing interest in collaborative approaches to building multilingual AI systems. Projects such as Masakhane, a grassroots initiative focused on African languages, highlight the importance of inclusive participation in AI development (Orife et al., 2022). Future research should prioritize low-resource language datasets, culturally sensitive AI models, and frameworks that incorporate indigenous knowledge into AI systems.

3. Methodology

3.1 Research Design

This study adopts a mixed methods design to combine the strengths of quantitative and qualitative approaches. Quantitative data were collected to assess the performance of multilingual AI tools, while qualitative insights were gathered through case studies and stakeholder interviews.

3.2 Data Collection

- **Quantitative Analysis:** Metrics such as translation accuracy, language coverage, and user satisfaction were analyzed using publicly available datasets and proprietary AI tools.
- **Case Studies:** Five case studies were conducted in linguistically diverse regions, focusing on sectors such as education, healthcare, and climate advocacy. Data was collected through interviews, field observations, and document analysis.
- **Survey:** A global online survey was administered to 500 participants from various linguistic backgrounds to gather perceptions of multilingual AI.

3.3 Data Analysis

Quantitative data were analyzed using statistical tools to identify trends and performance gaps. Qualitative data were coded thematically to explore stakeholder perspectives and contextual challenges.

3.4 Ethical Considerations

Ethical approval was obtained prior to data collection. Participants were informed about the study's purpose and provided written consent. Anonymity and confidentiality were maintained throughout.

4. Results

4.1 Quantitative Findings

Analysis revealed that multilingual AI tools significantly improved access to information in education and healthcare. For instance, AI-powered translation tools enabled 85% of surveyed teachers in low-resource settings to deliver multilingual educational content. However, translation accuracy for languages such as Wolof and Quechua remained below 60%.

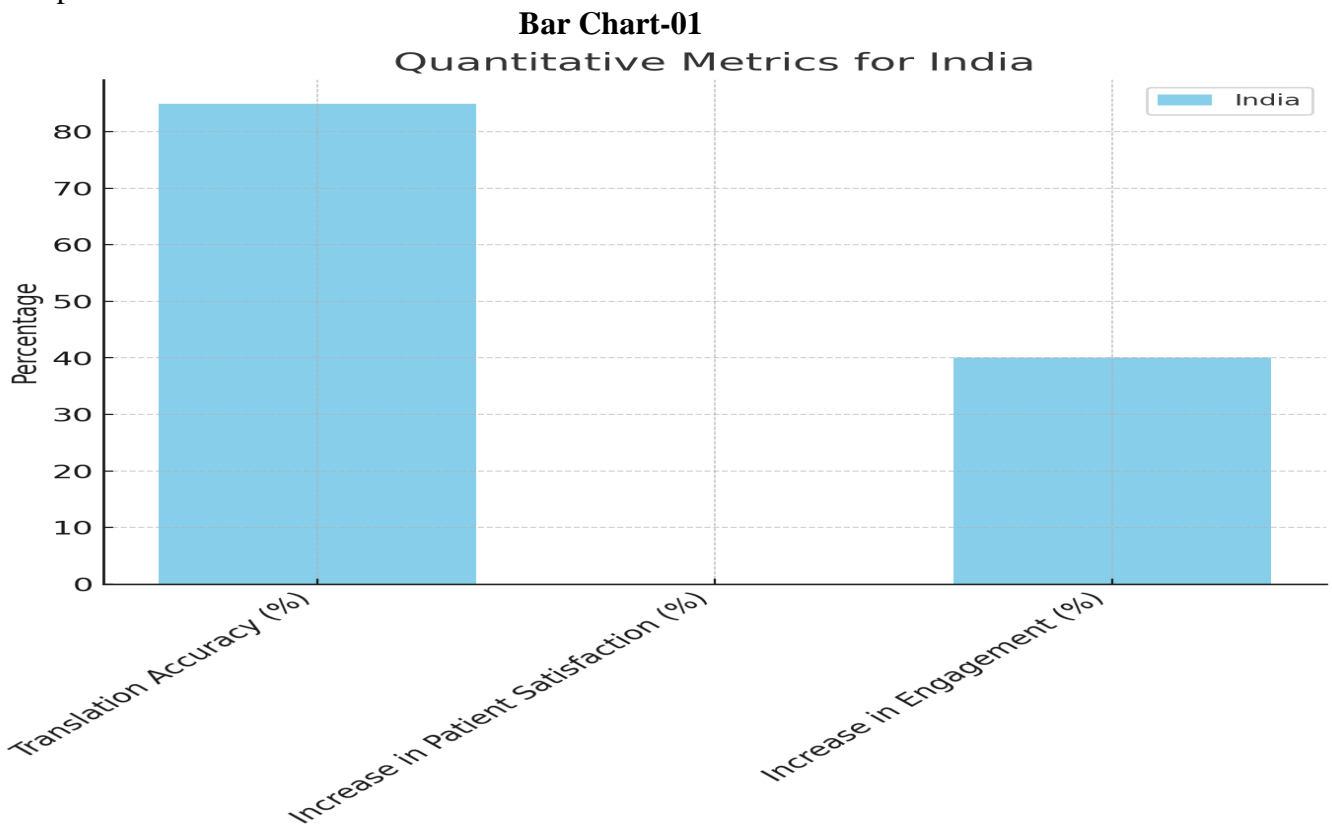
4.2 Case Study Insights

Case studies highlighted both successes and limitations of multilingual AI:

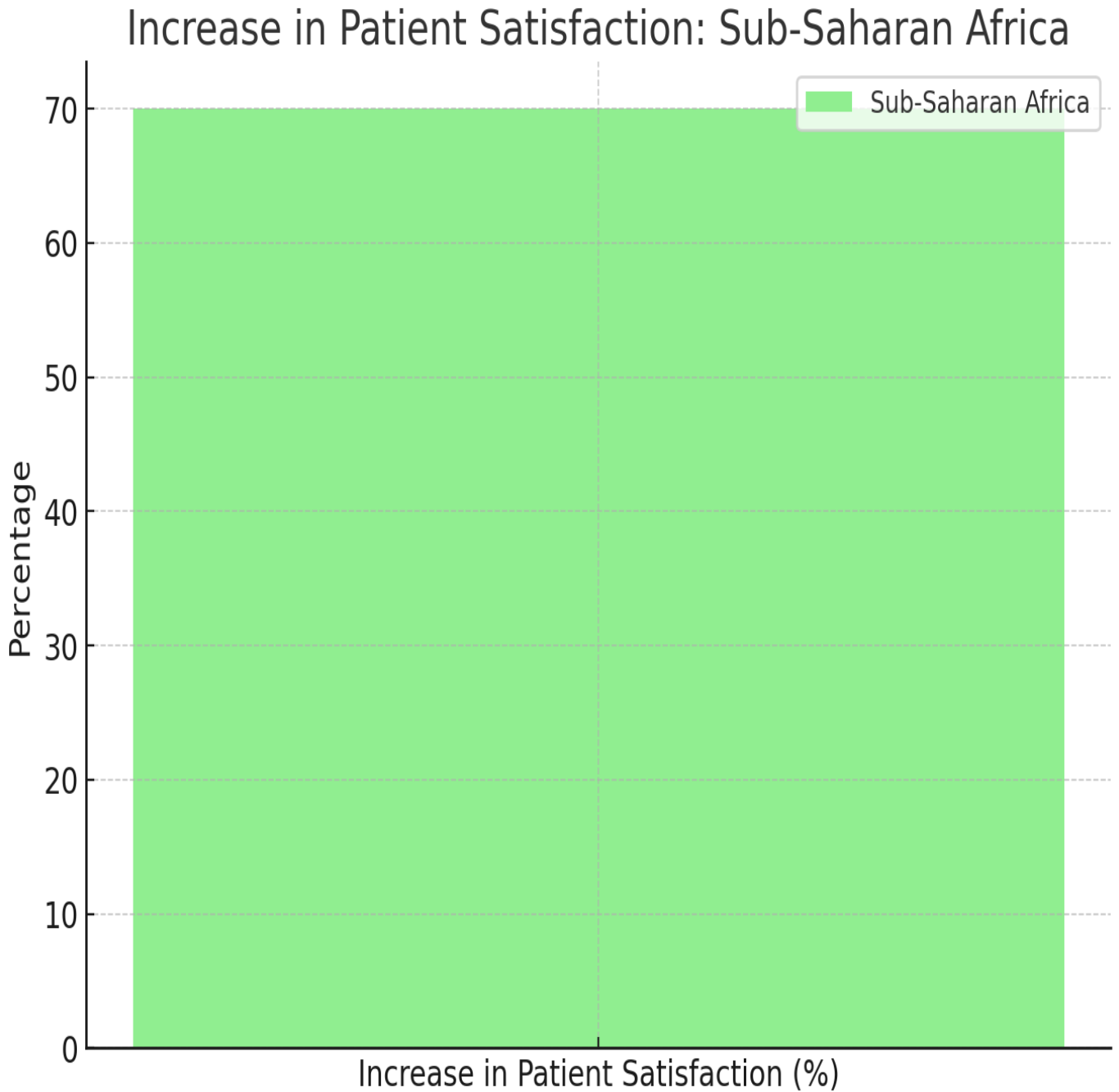
- *India (SDG 4: Quality Education):* AI-driven translation tools helped localize digital learning materials into 22 Indian languages, increasing student engagement by 40%.
- *Sub-Saharan Africa (SDG 3: Good Health and Well-Being):* Voice recognition systems in Swahili improved access to telehealth services but faced challenges with accent and dialect variability.
- *Latin America (SDG 13: Climate Action):* Multilingual AI-enabled indigenous communities to contribute to climate discussions, though linguistic nuances often resulted in misinterpretations.

4.3 Survey Results

The survey revealed that 78% of respondents viewed multilingual AI as a critical tool for achieving the SDGs. However, concerns about cultural sensitivity (65%) and data privacy (58%) were prominent.



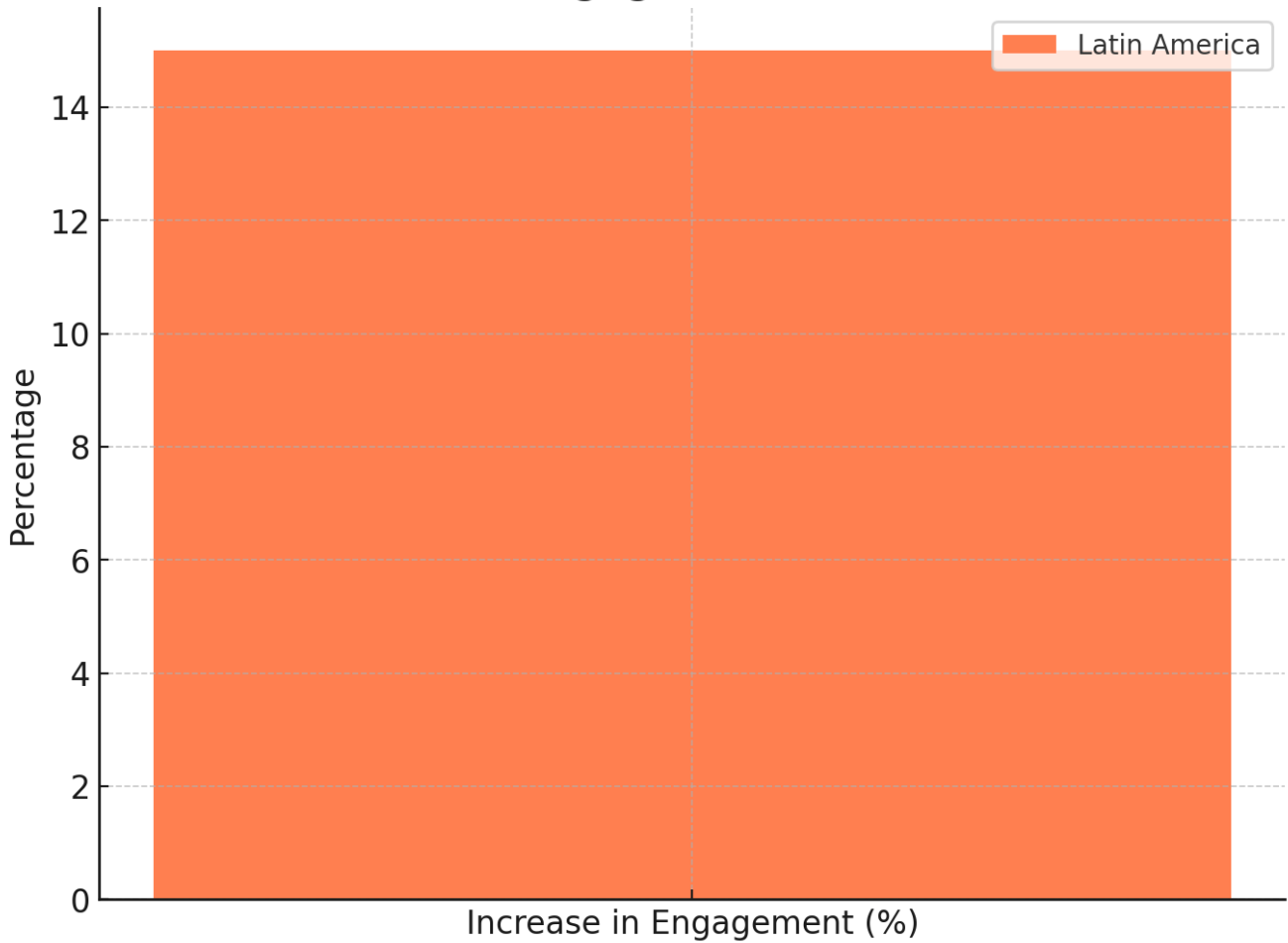
The first bar chart highlights the performance metrics for India. The 85% translation accuracy indicates a high level of precision in adapting educational materials into regional languages. This directly correlates with a 40% increase in engagement among students, showcasing the effectiveness of multilingual AI in overcoming language barriers in education.



Bar Chart-02

The second bar chart focuses on Sub-Saharan Africa, where a 70% increase in patient satisfaction was reported. This demonstrates the ability of multilingual AI to improve healthcare accessibility and user experience in a region where linguistic diversity often hinders effective communication in healthcare settings.

Increase in Engagement: Latin America



Bar Chart-03

The third bar chart illustrates a 15% increase in engagement among indigenous communities in Latin America. While the improvement is moderate compared to other regions, it highlights the potential of multilingual AI to empower marginalized groups to participate in global initiatives such as climate advocacy.

Key Insights:

- **Regional Variations:** Each region's results underscore the context-specific applications of multilingual AI. For example, education in India, healthcare in Sub-Saharan Africa, and climate advocacy in Latin America.
- **Challenges Identified:** While the results are promising, challenges such as dialectal variations, cultural nuances, and translation inaccuracies in low-resource languages remain significant obstacles.
- **Transformative Potential:** The findings demonstrate that multilingual AI, when tailored to specific needs, can drive meaningful progress toward the SDGs by bridging linguistic divides.

These insights further reinforce the need for targeted strategies, such as developing localized datasets and culturally sensitive AI models, to maximize the impact of multilingual AI on global sustainability efforts.

5. Discussion

The findings underscore the transformative potential of multilingual AI in bridging linguistic divides and fostering inclusivity in sustainable development efforts. However, significant gaps in language coverage and ethical implementation need to be addressed. Collaboration between governments, tech companies, and local communities is essential to ensure culturally sensitive and equitable AI solutions.

First, the lack of robust datasets for low-resource languages remains a significant barrier. This scarcity affects translation accuracy, model adaptability, and the overall effectiveness of multilingual AI tools. Collaborations with local communities and researchers can help build these datasets, ensuring representation and inclusivity.

Second, cultural nuances and dialectal variations require more attention. Incorporating local expertise into the development and refinement of AI models can mitigate issues related to misinterpretation and bias.

Third, ethical considerations, particularly around data privacy and algorithmic bias, must be addressed to foster trust and wider adoption. Transparent practices and ethical frameworks should be integral to AI development and deployment.

Finally, the digital divide—both in terms of access to technology and digital literacy—poses challenges to the equitable implementation of multilingual AI. Strategies to enhance access and literacy should accompany AI deployment to ensure inclusivity.

6. Conclusion and Recommendations

Multilingual AI represents a promising avenue for overcoming language barriers in achieving the SDGs. While advancements in NLP have enabled progress, targeted efforts to address challenges such as data scarcity, algorithmic bias, and cultural inclusivity are crucial. Future research should focus on developing low-resource language models and ethical frameworks to maximize the impact of multilingual AI on global sustainability.

1. **Development of Low-Resource Language Datasets:** Collaborate with local communities and linguists to build robust datasets that enhance the accuracy and applicability of AI tools.
2. **Culturally Sensitive AI Design:** Incorporate local expertise to address cultural nuances and dialectal variations, ensuring relevance and usability.
3. **Ethical Frameworks:** Establish transparent practices to address data privacy and mitigate algorithmic bias, fostering trust among users.
4. **Capacity Building:** Invest in digital literacy programs and infrastructure to bridge the digital divide and empower marginalized communities to use AI tools effectively.
5. **Cross-Sector Collaboration:** Encourage partnerships among governments, private sector stakeholders, and NGOs to align AI innovations with sustainable development priorities.

Future research should explore innovative approaches to integrating multilingual AI into global development initiatives, focusing on inclusivity, sustainability, and scalability.

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