

Protecting Indigenous Agricultural Knowledge in the Age of Artificial Intelligence

*Pruthvi Savaliya¹, Lipsa Dobariya¹, Dhara Vaish¹ and Dr. Harsiddhi Limbani²

¹M. Sc. Scholar, College of Agriculture, JAU, Junagadh, Gujarat

²Assistant Professor, Department of Genetics and Plant Breeding,

College of Agriculture, JAU, Junagadh, Gujarat

Corresponding Author's email: pruthviivsp@gmail.com

Indigenous Agricultural Knowledge (IAK) is a valuable source of traditional farming practices, innovations, and skills developed by local communities over generations. It plays a crucial role in sustainable agriculture, biodiversity conservation, climate resilience and food security. The growing use of Artificial Intelligence (AI) in agriculture has improved precision farming, resource management, and decision-making. However, the use of AI also raises concerns about the ownership, protection, and ethical use of indigenous knowledge. AI systems often rely on data collected from farming communities, creating risks of knowledge misappropriation and commercialization without proper recognition or benefit sharing. This review examines the importance of IAK, the impact of AI-driven agriculture, existing legal and policy frameworks, and strategies to protect traditional knowledge in the digital age. It emphasizes the need for ethical AI governance and stronger intellectual property protections to ensure that indigenous communities benefit from technological advancements while preserving their cultural heritage.

Introduction

Agriculture has evolved through centuries of observation, experimentation, and adaptation by indigenous and local communities. Traditional farming systems have generated a wealth of knowledge regarding crop selection, soil management, water conservation, pest control, and climate adaptation. This indigenous agricultural knowledge (IAK) forms an integral part of rural livelihoods and biodiversity conservation (Altieri, 2004). Unlike modern scientific knowledge, indigenous knowledge is often transmitted orally through generations and is deeply embedded within cultural traditions and local ecological conditions. The growing challenges of climate change, land degradation, biodiversity loss and food insecurity have renewed interest in indigenous agricultural practices. Numerous studies have demonstrated that traditional farming systems contribute significantly to sustainable resource management and resilience against environmental stresses (Berkes, 2018). Indigenous communities have developed sophisticated techniques.

Agriculture is experiencing a technological revolution driven by Artificial Intelligence (AI). AI-based technologies are increasingly used for crop monitoring, disease diagnosis, irrigation scheduling, yield prediction, and precision farming (Wolfert *et al.*, 2017). Machine learning algorithms can process large datasets and generate recommendations that improve productivity and resource-use efficiency.



Despite these benefits, the increasing use of AI raises important concerns regarding the protection of indigenous agricultural knowledge. AI systems require extensive datasets for training and decision-making. In many cases, valuable traditional knowledge is incorporated into digital databases, research projects and commercial applications without the prior informed consent of indigenous communities (Shiva, 2016). This creates a risk of "digital biopiracy," where traditional knowledge is extracted, digitized and commercialized without equitable benefit sharing. The intersection of AI, agriculture and traditional knowledge presents a complex challenge for policymakers, researchers and legal experts. While AI can help preserve and document indigenous knowledge, it can also facilitate its exploitation if adequate safeguards are not established. Therefore, understanding the opportunities and risks associated with AI-driven agriculture is essential for developing effective protection mechanisms.

Indigenous Agricultural Knowledge: Concept and Significance

Indigenous agricultural knowledge refers to the cumulative body of knowledge, practices, beliefs and innovations developed by local communities through long-term interactions with their environment (Warren, 1991). This knowledge encompasses crop cultivation techniques, seed conservation practices, soil fertility management, water harvesting systems, livestock management and ecological monitoring.

Researchers have highlighted that indigenous agricultural knowledge is highly location-specific and adapted to local environmental conditions (Chambers *et al.*, 1989). Farmers often possess detailed understanding of soil characteristics, seasonal weather patterns and plant-animal interactions that may not be captured by formal scientific systems. Such knowledge has enabled communities to sustain agricultural production under challenging ecological conditions for centuries.

Traditional agricultural practices also contribute significantly to biodiversity conservation. According to Altieri and Toledo (2011), indigenous farming systems maintain diverse crop varieties and genetic resources that enhance ecosystem stability and resilience. These systems often utilize mixed cropping and agroecological approaches that reduce dependence on synthetic inputs while promoting sustainable resource use. Furthermore, indigenous knowledge plays a critical role in climate adaptation. Studies conducted by FAO (2021) indicate that traditional farming communities have developed innovative strategies for coping with droughts, floods and changing climatic conditions. Such practices offer valuable insights for developing climate-resilient agricultural systems in the twenty-first century.

Artificial Intelligence and Modern Agriculture

Artificial Intelligence has emerged as a transformative technology in modern agriculture. AI applications include machine learning, computer vision, robotics, natural language processing and predictive analytics. These technologies facilitate real-time monitoring of crops, automated irrigation management, pest detection and yield forecasting (Liakos *et al.*, 2018).

Precision agriculture represents one of the most significant applications of AI. Through the integration of sensors, drones, satellite imagery and machine learning algorithms, farmers can make informed decisions regarding nutrient management, irrigation scheduling and crop protection. Research suggests that AI-driven precision agriculture can significantly improve productivity while reducing environmental impacts (Zhang *et al.*, 2021).

AI also supports agricultural extension services by providing personalized recommendations to farmers. Mobile-based advisory systems use weather forecasts, soil information and crop models to guide farm management decisions. Such systems are particularly beneficial in developing countries where access to



conventional extension services may be limited. However, AI systems depend heavily on large datasets. Data collection often involves recording local farming practices, indigenous innovations and traditional ecological knowledge. Without appropriate governance mechanisms, valuable community knowledge may become vulnerable to unauthorized use and commercialization.

Intellectual Property Rights and Traditional Knowledge Protection

Intellectual Property Rights (IPR) play a crucial role in protecting Indigenous Traditional Knowledge (ITK), which encompasses the knowledge, innovations, practices and cultural traditions developed by indigenous and local communities over generations. Traditional agricultural knowledge, medicinal practices, seed conservation methods and ecological management systems are valuable resources that contribute to biodiversity conservation and sustainable development (Berkes, 2018).

One of the primary objectives of IPR protection is to prevent biopiracy, where traditional knowledge is used or commercialized without the consent of the knowledge holders. Conventional patent systems often fail to adequately protect ITK because traditional knowledge is collectively owned, transmitted orally and developed over long periods rather than created by a single inventor (Shiva, 2016). As a result, alternative mechanisms such as Traditional Knowledge Digital Libraries (TKDLs), geographical indications, and sui generis protection systems have been developed to safeguard indigenous knowledge and prevent unauthorized patent claims (Carroll, 2020).

The Convention on Biological Diversity (CBD) and the Nagoya Protocol emphasize the principles of Prior Informed Consent (PIC) and Access and Benefit Sharing (ABS). These frameworks require researchers and companies to obtain permission from indigenous communities before accessing their knowledge and ensure that benefits derived from its use are shared fairly (CBD, 2011). In the era of artificial intelligence, the protection of ITK has become increasingly important. AI systems rely on large datasets for training and decision-making and traditional knowledge may be incorporated into these datasets without community authorization. This creates risks of digital biopiracy and loss of control over indigenous intellectual resources. Therefore, ethical AI governance, indigenous data sovereignty, and the application of the CARE Principles (Collective Benefit, Authority to Control, Responsibility and Ethics) are essential for ensuring that AI technologies respect community rights and cultural heritage (Carroll *et al.*, 2020).

Opportunities for AI in Preserving Indigenous Knowledge

Artificial Intelligence (AI) can help preserve indigenous knowledge by creating digital records of traditional farming practices, making sure valuable knowledge is not lost over time (UNESCO, 2021). AI can also translate indigenous languages and organize information in digital databases, making it easier for people to access and learn from traditional knowledge (WIPO, 2023).

AI helps combine indigenous knowledge with modern scientific research to develop sustainable farming practices and improve climate resilience (FAO, 2021). It can also support the transfer of traditional knowledge to younger generations through mobile applications and educational platforms (Ryan, 2022). In addition, AI can help identify and conserve traditional crop varieties that are important for food security and biodiversity conservation (FAO, 2021).

Multi-layered policy approach

- 1. Statutory Reforms and Sui Generis IP Law:** Establish sui generis laws that recognize collective Indigenous ownership of ITK, require AI developers to disclose its use in datasets, and enforce penalties for unauthorized exploitation.
- 2. CARE Principles in AI Governance:** Integrate the CARE Principles to ensure community benefits, Indigenous control over data, developer accountability, and ethical AI development.

3. Data Sovereignty and FPIC: Mandate Free, Prior and Informed Consent (FPIC), enable flexible consent management and support Indigenous-controlled data repositories.

4. Technical Safeguards: Use blockchain for provenance tracking, AI tools for monitoring misuse and benefit-sharing mechanisms to ensure equitable returns from ITK-based innovations.

Conclusion

Indigenous agricultural knowledge represents a critical resource for sustainable agriculture, biodiversity conservation, and climate resilience. The emergence of Artificial Intelligence has created both opportunities and challenges for the preservation and utilization of this knowledge. While AI can facilitate documentation, dissemination and innovation, it also increases the risk of digital biopiracy and unauthorized commercialization. Protecting indigenous agricultural knowledge in the age of AI requires a balanced approach that promotes technological innovation while safeguarding community rights. Strengthening intellectual property frameworks, ensuring equitable benefit sharing, and adopting ethical AI governance mechanisms will be essential for achieving this balance. Future agricultural development should recognize indigenous communities not merely as data providers but as knowledge holders and partners in innovation.

References

1. Altieri, M. A. (2004). *Agroecology: The Science of Sustainable Agriculture* (2nd ed.). CRC Press, Boca Raton, Florida, USA.
2. Altieri, M. A. and Toledo, V. M. (2011). The agroecological revolution in Latin America: Rescuing nature, ensuring food sovereignty and empowering peasants. *The Journal of Peasant Studies*, 38(3): 587-612.
3. Berkes, F. (2018). *Sacred Ecology* (4th ed.). Routledge, New York, USA.
4. Carroll, S. R., Garba, I., Figueroa-Rodríguez, O. L., Holbrook, J., Lovett, R., Materechera, S., Parsons, M., Raseroka, K., Rodriguez-Lonebear, D., Rowe, R., Sara, R., Walker, J., Anderson, J. and Hudson, M. (2020). The CARE Principles for Indigenous Data Governance. *Data Science Journal*, 19(1): 43. <https://doi.org/10.5334/dsj-2020-043>
5. Chambers, R., Pacey, A. and Thrupp, L. A. (1989). *Farmer First: Farmer Innovation and Agricultural Research*. Intermediate Technology Publications, London, UK.
6. Convention on Biological Diversity (CBD). (2011). Nagoya Protocol on Access to Genetic Resources and the Fair and Equitable Sharing of Benefits Arising from Their Utilization. Montreal: Secretariat of the Convention on Biological Diversity.
7. FAO. (2021). *The State of the World's Biodiversity for Food and Agriculture*. FAO Commission on Genetic Resources for Food and Agriculture Assessments.
8. Liakos, K. G., Busato, P., Moshou, D., Pearson, S. and Bochtis, D. (2018). Machine Learning in Agriculture: A Review. *Sensors*, 18(8): 1-29.
9. Ryan, J. C. (2022). *Digital Technologies and Indigenous Knowledge Preservation*. Springer.
10. Shiva, V. (2016). *Biopiracy: The Plunder of Nature and Knowledge*. North Atlantic Books, Berkeley, California, USA.
11. UNESCO. (2021). *Recommendation on the Ethics of Artificial Intelligence*. United Nations Educational, Scientific and Cultural Organization.
12. Warren, D. M. (1991). *Using Indigenous Knowledge in Agricultural Development*. World Bank Discussion Paper No. 127, World Bank, Washington, DC, USA.
13. WIPO. (2023). *Traditional Knowledge and Intellectual Property: Background brief*. WIPO.
14. Wolfert, S., Ge, L., Verdouw, C. and Bogaardt, M. J. (2017). Big Data in Smart Farming: A Review. *Agricultural Systems*, 153: 69-80.
15. Zhang, Y., Wang, G. and Wang, J. (2021). Artificial Intelligence Applications in Precision Agriculture. *IEEE Access*, 9: 125078-125090.