Smart Seeds: How Digital Technologies Are Revolutionizing Plant Breeding G. Anusha¹ and G. Usha rani²

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Abstract

The world faces an urgent need to produce more food with fewer resources and under increasingly unpredictable climate conditions. Traditional plant breeding, while powerful, is time-consuming and limited by human capacity for data analysis. Enter the digital era: the fusion of genomics, artificial intelligence (AI), big data and advanced phenotyping tools is transforming plant breeding into a faster, more precise and scalable process. Digital technologies enable the breeders to analyze vast datasets, simulate outcomes and make data-driven decisions that enhance crop resilience, yield and nutritional quality. High-throughput phenotyping using drones, sensors and imaging tools allows for real-time, non-invasive assessment of thousands of plants. Genomic selection leverages machine learning to predict the performance of untested plant lines. Cloud computing facilitates the storage and processing of massive breeding datasets across global networks, improving collaboration and speeding up research cycles. Digital twins of plantsvirtual replicas that model growth and stress responses-offer insights previously impossible to capture. This article explores how digital transformation is reshaping plant breeding, highlighting current innovations, case studies from global breeding programs and future prospects. It also discusses the importance of digital literacy and data infrastructure in ensuring equitable access to these technologies, especially in developing countries. By integrating traditional knowledge with modern tools, digital plant breeding is set to help feed the future sustainably.

Introduction

Plant breeding has been a cornerstone of agricultural innovation, playing a vital role in shaping the crops that sustain global food systems. From Gregor Mendel's early genetic discoveries with pea plants to the high-yield varieties of the Green Revolution, generations of breeders have advanced crop development through careful observation, genetic insight and iterative selection. These conventional methods have served humanity well, but the growing complexity of global challenges now calls for a more advanced and responsive approach.

As the global population approaches 10 billion, agriculture is under increasing pressure to produce more food with fewer resources while adapting to the intensifying impacts of climate change, land degradation and emerging pests and diseases. Traditional breeding cycles, which often take 10 to 15 years to deliver improved varieties, struggle to meet the urgent demands of modern agriculture. There is a growing need for faster, more precise breeding systems that can deliver high-performing, climate-resilient crops tailored to specific environments and consumer preferences (Araus and Cairns, 2014; FAO, 2021).

Digital technologies are rising to meet this challenge by transforming the way plant breeding is conducted. Through the integration of genomics, data science, artificial intelligence, and field-based monitoring systems, breeders now have powerful tools to accelerate and optimize the development of new crop varieties. Machine learning enables the prediction of trait performance with high accuracy, while drones and sensor networks rapidly capture phenotypic data at scale, providing real-time insights into plant health and development.

This transformation goes beyond speed, it enhances decision-making, reduces breeding costs and democratizes innovation. Digital breeding opens doors for researchers, seed companies and smallholder farmers alike, paving the way for a more resilient, inclusive and data-driven future in agriculture (Bayer Crop Science, 2022 and CIMMYT, 2023).

Key Technologies in Digital Breeding

1. Genomic Selection and AI-Based Predictions:

Genomic selection uses DNA markers across the entire genome to predict the performance of breeding lines without extensive field trials. Combined with artificial intelligence (AI) and machine learning, it enables breeders to identify superior genotypes faster and with greater precision. These algorithms learn from past breeding data and continuously improve predictions, optimizing the selection of parent lines (Crossa *et al.* 2017)

2. High Throughput Phenotyping (HTP)

HTP technologies—such as drones, groundbased robots and sensor platforms—collect detailed phenotypic data rapidly across large field plots. Using tools like multispectral cameras, thermal sensors and LiDAR, breeders can measure traits like plant height, chlorophyll content, water stress and disease symptoms—non-invasively and at scale ICRISAT, 2022).

3. Digital Twins and Crop Modeling

Digital twins are virtual models of plants that simulate their growth, development and response to environmental conditions. These simulations help breeders test multiple scenarios (like drought or high heat) without real-world trials, offering predictive insights into trait performance under stress (Varshney, *et al.* 2021).

4. Cloud Computing and Data Platforms

Managing the massive data generated from breeding programs requires robust cloud infrastructure. Platforms like BreedBase, Germinate and KDDart provide centralized databases for storing, analyzing and sharing genetic and phenotypic data among breeding teams worldwide (Hickey, *et al.* 2019).

Case Studies

1. CIMMYT (International Maize and Wheat Improvement Center)

CIMMYT has adopted genomic selection and HTP to develop climate-resilient maize and wheat varieties for smallholder farmers. Their integrated breeding platform uses drones for field scouting and cloud databases for real-time data analytics, reducing the breeding cycle by 30–40% (Reynolds, *et al.* 2020).

2. Bayer Crop Science

Bayer's breeding programs utilize Al-driven genomic prediction models and automated greenhouses where robots assess traits 24/7. These technologies have significantly increased breeding efficiency and product development speed (Roorkiwal, *et al.* 2020).

ICRISAT (International Crops Research Institute for the Semi-Arid Tropics)

ICRISAT leverages mobile phenotyping tools and open-access data platforms to assist national breeding programs in Africa and Asia. Their digital tools help identify drought-tolerant and early-maturing varieties of sorghum and millet.

Challenges and Opportunities

Challenges

- **Digital Divide:** Many breeding programs, especially in low-income countries, lack access to infrastructure, training and high-speed internet needed to adopt digital tools.
- Data Standardization: Inconsistent data formats and protocols across breeding programs can hinder collaboration and interoperability.
- **Cost and Complexity:** Initial investment in sensors, software and training can be a barrier for smaller institutions.

Opportunities:

- **Open-Source Platforms:** Free tools and open databases can democratize access to digital breeding resources.
- **Capacity Building:** Investing in digital literacy and technical training for breeders and technicians is key to inclusive innovation.
- **Public-Private Partnerships:** Collaboration between governments, research institutions, and tech companies can foster scalable solutions and reduce costs.

Conclusion

Digital plant breeding is the natural evolution of traditional methods, enhancing crop improvement through AI, big data and smart technologies. As these tools become more accessible, breeding becomes faster, more precise and globally connected. Empowered by real-time data and predictive analytics, breeders can develop climate-resilient, high-yielding crops with unprecedented efficiency. To unlock its full potential, however, digital breeding must be supported by inclusive infrastructure, ethical practices and global collaboration. With strategic investment and shared commitment, it can play a pivotal role in securing a sustainable, food-secure future for all.

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