

Unlocking Productivity Potential: The Promising Role of Agricultural Robots in Enhancing Farming Efficiency

Abstract

Integration of advanced robotics in agriculture can exponentially boost productivity, alleviate labor shortages, reduce the environmental footprint, and increase the overall profitability of farming. Agricultural robots, often controlled by sophisticated algorithms and AI, offer precision farming capabilities that can enhance yield and quality, while minimizing waste and harmful impacts on the environment. They carry out numerous tasks such as sowing, watering, harvesting, and pest control, more efficiently than traditional methods. The usage of robotic systems enables 24/7 farming operations, overcoming the challenges posed by traditional human labor like working hours and physical exhaustion. The review also explores how this technology can help cater to the rising global food demand in a sustainable way, making it a promising solution to achieving food security in the face of increasing population pressures and climate change impacts. Despite the significant capital investment required for adopting these technologies, the potential long-term benefits, such as reduced operational costs and enhanced farm outputs, underscore their vital role in the future of farming and food production. Concluding that the integration of robotics in agriculture could bring about a revolution in farming practices, ushering in a future of enhanced productivity and sustainability in the sector.

Key word: Crop Monitoring, farming practices, farming technology, smart farming

Introduction

The field of agriculture has a profound history, tracing back approximately 10,000 years when early humans transitioned from nomadic hunter-gatherers to settled farmers. This dramatic shift, often referred to as the Agricultural Revolution, marked a critical turning point in human history, spurring the growth of civilizations and shaping societies as we know them today [1]. Throughout the centuries, the practice of agriculture has evolved significantly, influenced by technological advancements, societal changes, and the growing understanding of the natural world [2]. At the core of agricultural progression lies the concept of agricultural productivity. Simply defined, agricultural productivity is a measure of the efficiency of a farm in converting inputs (such as labor, land, and capital) into outputs (i.e., crops or livestock) [3]. It is an essential determinant of farm income, economic growth, and food security, particularly in developing countries where agriculture plays a critical role in their economies [4]. Numerous issues plague agricultural productivity. Climate change, for instance, poses a significant challenge due to unpredictable weather patterns, increasing occurrences of droughts and floods, and rising temperatures [5]. The escalating problem of labor shortage in agriculture, driven by urbanization and the decreasing appeal of farming as a profession, further aggravates productivity challenges [6]. The constant need to feed a growing global population further exacerbates the strain on agricultural productivity [7]. In response to these challenges, agricultural practices are rapidly evolving, incorporating technology to optimize productivity. A paradigm-shifting development within this technological evolution is the introduction of agricultural robots. Agricultural robots, also known as AgBots or smart farming robots, are designed to automate slow, repetitive, and dull tasks for farmers, allowing them to focus more on improving overall farm efficiency [8]. These automated solutions vary from autonomous tractors to robotic weeders, drone-based

monitoring systems, and harvest robots. The potential of agricultural robots in enhancing farming efficiency is immense. By taking over tedious tasks, these robots can address labor shortages while increasing precision and efficiency [9]. In addition, through advancements in artificial intelligence (AI) and sensor technology, agricultural robots can provide unprecedented levels of data for more informed decision-making, ultimately boosting productivity [10]. This topic's significance is underscored by the crucial role of agriculture in global food security and the economy. As the global population is projected to reach nearly 10 billion by 2050, the necessity for efficient, sustainable, and productive agricultural practices has never been more critical [11]. The implications of this research extend beyond productivity, affecting areas such as climate change, rural livelihoods, and societal well-being. The purpose of this review is to provide a comprehensive overview of the promising role of agricultural robots in enhancing farming efficiency. It aims to outline the current state of agricultural robotics, identify the potential benefits and challenges of adopting this technology, and explore future opportunities for growth and innovation. In doing so, it will contribute to the broader discourse on technological innovation in agriculture, providing valuable insights for researchers, policymakers, farmers, and all stakeholders within the agricultural value chain.

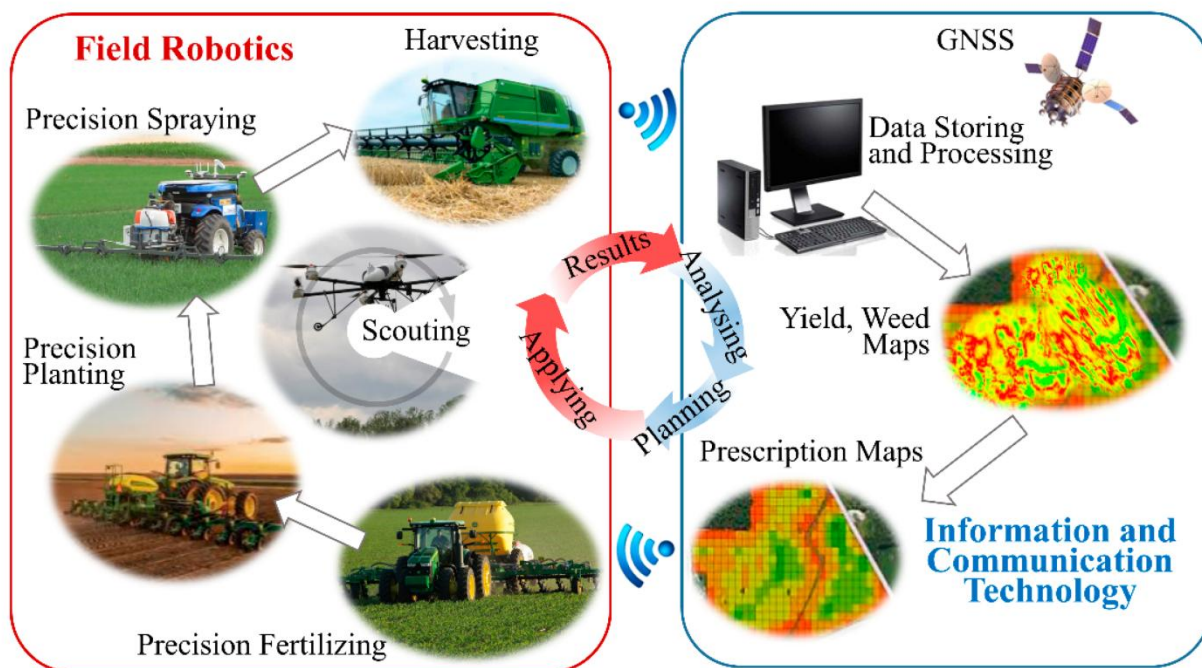


Image 1: Different Types of Robot for Agricultural Purposes (Source- <https://www.mdpi.com/>)

Challenges in Agricultural Productivity

The process of farming and food production is as old as human civilization itself. Traditional farming methods have evolved over millennia, from basic hand cultivation to the use of draught animals and simple machinery [12]. These conventional methods, while enduring, carry inherent drawbacks. They are labor-intensive, often imprecise, and can lead to land degradation due to

overuse and poor management [13]. Such techniques, while sufficient in the past, are now increasingly insufficient to meet the growing demand for food. Traditional farming is characterized by its heavy reliance on human labor and is subject to its inefficiencies. Human error, inconsistent performance, and the physical limitations of manual labor are significant challenges [14]. Furthermore, these traditional techniques often lead to soil degradation and overuse of water and chemicals, contributing to environmental harm [15]. Climate change poses an additional and formidable challenge to agricultural productivity. Changing weather patterns, increased temperature, and more frequent extreme weather events directly impact crop yields [16]. For example, higher temperatures can accelerate the plant's metabolic rates, shorten the growing season, and reduce yields. Likewise, droughts and floods, expected to increase in frequency due to climate change, have devastating effects on crops and livestock [17]. On top of the environmental challenges, agriculture faces a labor shortage crisis. Globally, the trend towards urbanization, coupled with the decreasing appeal of farming as a profession, has led to an increasing scarcity of agricultural labor [18]. In many countries, the agricultural labor force is aging, and younger generations are often reluctant to take up farming, seeking opportunities in cities instead. This labor shortage can lead to unharvested fields, limiting agricultural productivity. Another significant challenge is the continuous need to feed a growing global population. According to the United Nations, the world's population is expected to reach nearly 10 billion by 2050 [19]. This growth, coupled with changing dietary patterns in many parts of the world, translates into increasing demand for food and pressure on agricultural systems. It is estimated that food production will need to increase by 70% by 2050 to meet this demand, a target that current productivity levels will struggle to meet [20]. Low agricultural productivity also carries significant economic implications. Agriculture is the main source of income for many rural communities around the world, and low productivity can lead to poverty and food insecurity [21]. At the national level, countries heavily dependent on agriculture can experience economic stagnation due to low productivity [22]. Additionally, low productivity can lead to higher food prices, affecting the affordability of food and potentially leading to social unrest [23]. To address these challenges, the global agricultural sector needs to undergo a profound transformation. Adopting more efficient and sustainable farming methods, leveraging technology, and investing in agricultural research and development are critical to increasing productivity, improving farmers' livelihoods, and ensuring food security.

Agricultural Robotics

Agricultural robots, or agbots, represent an exciting development in the field of farming technology (Table 1). These machines can perform a wide variety of farming tasks, from seeding and harvesting to crop monitoring and weed control [24]. By employing advanced technologies such as computer vision, artificial intelligence (AI), and machine learning, these robots are capable of precise, efficient, and consistent agricultural operations [25]. The journey of agricultural robotics began in the 20th century with the mechanization of agriculture. This mechanization involved replacing manual and animal labor with machinery to perform tasks like planting, harvesting, and tilling [26]. The evolution of agricultural robots can be traced back to the advent of the first autonomous tractor in the late 20th century [27]. Over the years, technological advancements have led to more sophisticated machines that are capable of a wide range of tasks, improving productivity and reducing the demand for manual labor [28]. Agricultural robots today can be broadly divided into several types, based on their functions.

Autonomous Tractors: These machines can carry out ploughing, seeding, and harvesting tasks without the need for a human operator. They utilize GPS and sensor data to navigate fields and perform tasks with high precision, leading to optimized usage of resources like seeds, water, and fertilizers [29].

Drones for Crop Monitoring: These aerial robots are used for monitoring large fields, enabling early detection of plant diseases, pests, and nutrient deficiencies. Equipped with advanced sensors and cameras, these drones can capture detailed images, which are then processed using AI algorithms to provide insights into crop health [30].

Robotic Harvesters: Harvesting is a labor-intensive and delicate process, especially for fruits and vegetables that are susceptible to damage. Robotic harvesters can identify ripe produce using computer vision and carefully pick them without causing damage, making the harvesting process more efficient and less wasteful [31].

Robotic Weeders: Weeds compete with crops for resources, leading to reduced yields. Robotic weeders can identify and eliminate weeds using computer vision, without damaging the crops. This not only improves crop yield but also reduces the need for chemical herbicides, leading to more sustainable farming practices [32].

Table 1: Types and Applications of Agricultural Robots

Type of Robot	Application	Manufacturer	Current Status
Autonomous Tractors	Ploughing, planting, and harvesting crops	John Deere, Mahindra	In use and under development
Drone Technology	Surveying, crop monitoring, and precision spraying	DJI, Yamaha	In use and under development
Robotic Harvesters	Picking and harvesting crops	Abundant Robotics, FFRobotics	In use and under development
Weed Control Robots	Weed identification and removal	EcoRobotix, Naïo Technologies	In use and under development
Robotic Milkers	Automated milking of livestock	Lely, DeLaval	In use

Potential of Agricultural Robots to Enhance Productivity

Agricultural robots, in their various forms, represent a powerful tool for addressing many of the limitations and challenges inherent in traditional farming methods. The integration of these machines into farming operations can lead to significant enhancements in agricultural productivity. One of the most pressing challenges in traditional farming is its labor-intensive nature. Here, robots serve as a potent solution. With the ability to work tirelessly, agricultural robots can perform tasks such as seeding, harvesting, weeding, and crop monitoring, freeing up human labor for other essential tasks and reducing the physical burden on farmers [33]. Robots' capability to operate 24/7 also enhances productivity by enabling farming operations even during off-hours. The incorporation of robotic technologies also introduces an unprecedented level of

precision and efficiency into farming. For instance, autonomous tractors equipped with GPS and sensors can plant seeds and apply fertilizers with high accuracy, minimizing waste and maximizing crop yields [34]. Drones equipped with advanced imaging technologies can monitor crops on a detailed level, enabling early detection and treatment of diseases, pests, or nutrient deficiencies, thus ensuring optimal crop health [35]. Agricultural robots also play a significant role in promoting sustainable farming practices. For instance, robotic weeders can identify and eliminate weeds with high precision, reducing the need for chemical herbicides that can harm the environment [36]. Similarly, robots' precision in applying fertilizers and pesticides means less chemical runoff into nearby water bodies, contributing to the preservation of aquatic ecosystems [37]. Labor shortages in farming, a growing issue worldwide, can also be addressed by the implementation of agricultural robots. Robots can perform labor-intensive tasks without breaks, reducing the need for manual labor. They can also operate in harsh conditions where it might be challenging for human workers, further alleviating the labor shortage problem [38]. Various case studies demonstrate the impact of agricultural robots on farm productivity. A research project at Harper Adams University in the UK found that an entirely automated barley farm achieved a yield of 4.5 tonnes per hectare, demonstrating the potential of complete automation in farming. Similarly, the adoption of agricultural robots has profound economic implications. Despite the high initial investment, the increased productivity and reduced labor and input costs can lead to significant economic gains in the long run [39]. Robots can operate continuously, providing a higher return on investment compared to human labor or traditional machinery [40].

Technological Advances and Innovations in Agricultural Robots

Agricultural robots represent the fusion of multiple technological domains, each evolving rapidly to augment the capabilities of these machines. Advances in artificial intelligence (AI), machine learning, sensor technologies, and big data analytics, along with the development of the Internet of Things (IoT), have significantly expanded the potential of agricultural robots. In the field of AI and machine learning, significant strides have been made. These technologies empower agricultural robots with the ability to learn from data, adapt to new situations, and make decisions autonomously. For example, AI algorithms allow robotic weeders to distinguish between crops and weeds accurately, enabling precise and efficient weed control [41]. Machine learning techniques also enable drones to identify symptoms of plant diseases from aerial images, facilitating early intervention [42]. This autonomous decision-making capability is key to enhancing the effectiveness and efficiency of agricultural robots. The development of IoT in agriculture, termed as agri-IoT, has further enhanced the capabilities of agricultural robots. IoT connects farming equipment, sensors, and software over a network, enabling real-time data exchange and remote control of equipment [43]. This enables farmers to monitor and manage their operations more effectively. For instance, sensors on an autonomous tractor can send soil moisture data to a central system, which then commands the tractor to adjust irrigation levels accordingly [44]. Sensor technologies play a critical role in agricultural robots. Advanced sensors enable robots to perceive their environment, detect obstacles, and carry out tasks with high precision. For example, Lidar sensors help autonomous tractors navigate fields and avoid obstacles. Hyperspectral sensors on drones can measure plant health by detecting light reflected from crops, enabling early detection of plant stress [45]. This sensor-driven capability enables agricultural robots to execute tasks with high precision, significantly improving farming efficiency. Big data and analytics are another crucial aspect of agricultural robotics. With sensors

and IoT devices generating vast amounts of data, big data analytics techniques are necessary to extract actionable insights. For instance, data from drones monitoring a large field can be analyzed to assess crop health, predict yields, and guide farming decisions [46]. Big data analytics can also identify patterns and trends, enabling predictive maintenance of robots and reducing downtime.

Barriers to the Adoption of Agricultural Robots

While agricultural robots present substantial benefits, several barriers limit their widespread adoption (Table 2). These hurdles span from technological challenges and high costs to regulatory issues, social acceptance, and a significant skills gap. The integration of robots into farming presents several technological challenges. For instance, agricultural robots often need to operate in complex and dynamic outdoor environments that can vary significantly, posing a significant challenge for the technology [47]. Also, despite advancements in AI and machine learning, the algorithms are still not perfect. They often struggle to deal with uncertainties or rare events, which could lead to errors in decision-making [48]. The need for reliable, high-speed internet for IoT applications remains unmet in many rural areas, limiting the deployment of network-dependent agricultural robots. Financial barriers also exist. The high upfront cost of agricultural robots makes them unaffordable for many small and medium-sized farms [49]. While the potential return on investment is high, it often requires several years of operation, posing significant financial risk. There can be additional costs for maintenance, repairs, and updates, further discouraging potential adopters [50]. Legal and regulatory issues also pose significant barriers. For instance, the operation of autonomous tractors and drones might need approval from transportation or aviation authorities. There are also concerns about data privacy and security related to the vast amounts of data generated by these machines [51]. Addressing these regulatory complexities requires time and can be a significant deterrent. Social and cultural acceptance of robots in farming is another significant barrier. Many farmers have deeply ingrained traditions and ways of doing things and might be hesitant to adopt technologies that fundamentally change their practices [52]. The adoption of agricultural robots requires a skilled workforce that can operate and maintain these machines. This skills gap poses a significant challenge, especially in regions where education and training opportunities are limited [53].

Table: 2 Barriers to the Adoption of Agricultural Robots

Barrier	Description	Potential Solutions
High Initial Cost	The cost of purchasing and installing robotics can be prohibitively high for small and medium-sized farms.	Government subsidies, leasing or sharing options, and cost reductions as technology matures
Technological Complexity	Understanding and managing robotic technologies may require skills and knowledge that many farmers do not currently possess.	Training programs, better user interfaces, and improved autonomous capabilities
Uncertain ROI	The financial return on investment (ROI) may not be clear, especially given potential maintenance costs and the risk of technology becoming obsolete.	Economic studies demonstrating long-term benefits, warranties, and upgrade options
Limited Technology Compatibility	Robots may not integrate well with existing farm infrastructure and equipment.	Standardization of robotic technologies, and development of adaptable, flexible systems

Regulatory Hurdles	Current regulations may not fully account for the use of autonomous robots in agriculture, particularly with respect to safety and liability.	Policy reform, industry standards, and insurance options
--------------------	---	--

Future Perspectives and Opportunities

Agricultural robots are poised to revolutionize the agricultural industry. The convergence of advanced technologies, evolving market demands, and the need for sustainable and efficient farming practices present a promising future for these intelligent machines.

Forecasts suggest a robust growth trajectory for the agricultural robots market. The global market is expected to reach \$20.3 billion by 2025, expanding at a compound annual growth rate of 24.2% during 2020-2025 [54]. This growth is driven by increasing labor costs, the need for precision farming, and the growing demand for high-quality food products. Emerging technological trends are expected to enhance the capabilities of agricultural robots significantly. Innovations in sensor technology, AI, machine learning, IoT, and big data analytics will continue to improve the precision, efficiency, and automation level of these machines [55]. For example, advancements in computer vision techniques could enable robots to identify and pick ripe fruits without causing any damage [56]. Similarly, improvements in IoT technologies could lead to the development of smart farms where multiple robots and sensors work together seamlessly. Opportunities for research and development abound in the realm of agricultural robotics. While much progress has been made, many technical challenges remain. For instance, developing robots that can work effectively under diverse and changing environmental conditions is a complex task that needs more research [57]. Similarly, there is a need for research on the socioeconomic implications of agricultural robotics, such as its impact on rural employment and income distribution [58]. The integration of robots into farming practices could reshape the agriculture industry in profound ways. They could transform farming from a labor-intensive activity to a knowledge-intensive one, requiring farmers to acquire new skills and knowledge [59]. Farms could become more like manufacturing plants, with high levels of automation and precision. Robots could enable a shift towards more sustainable farming practices by reducing the use of pesticides and optimizing resource use [60].

Conclusion

The integration of robotics into agriculture holds tremendous potential for enhancing productivity, efficiency, and sustainability. By overcoming traditional farming challenges, they can address labor shortages, climate change impacts, and the growing global food demand. Despite existing barriers such as high costs, regulatory complexities, and skills gaps, the future of agricultural robotics is promising. Emerging technological trends and ongoing research and development are paving the way for smarter, more capable agricultural robots. As these machines become more prevalent, they will inevitably reshape farming practices, transforming agriculture into a more knowledge-intensive, precise, and sustainable sector.

References

1. Larsen, C. S. (2006). The agricultural revolution as environmental catastrophe: Implications for health and lifestyle in the Holocene. *Quaternary International*, 150(1), 12-20.
2. Nyong, A., Adesina, F., & Osman Elasha, B. (2007). The value of indigenous knowledge in climate change mitigation and adaptation strategies in the African Sahel. *Mitigation and Adaptation strategies for global Change*, 12, 787-797.
3. Leavy, J., & Poulton, C. (2007). Commercialisations in agriculture. *Ethiopian Journal of Economics*, 16(1), 1-37.
4. Pawlak, K., & Kołodziejczak, M. (2020). The role of agriculture in ensuring food security in developing countries: Considerations in the context of the problem of sustainable food production. *Sustainability*, 12(13), 5488.
5. Thornton, P. K., Ericksen, P. J., Herrero, M., & Challinor, A. J. (2014). Climate variability and vulnerability to climate change: a review. *Global change biology*, 20(11), 3313-3328.
6. Balogun, A. L., Adebisi, N., Abubakar, I. R., Dano, U. L., & Tella, A. (2022). Digitalization for transformative urbanization, climate change adaptation, and sustainable farming in Africa: trend, opportunities, and challenges. *Journal of Integrative Environmental Sciences*, 19(1), 17-37.
7. Gan, Y., Siddique, K. H., Turner, N. C., Li, X. G., Niu, J. Y., Yang, C., ... & Chai, Q. (2013). Ridge-furrow mulching systems—an innovative technique for boosting crop productivity in semiarid rain-fed environments. *Advances in agronomy*, 118, 429-476.
8. Gur, M. V., & Rylo, T. V. (2018). Robotic technology in agriculture.
9. Lakshmi, V., & Corbett, J. (2020). How artificial intelligence improves agricultural productivity and sustainability: A global thematic analysis.
10. Javaid, M., Haleem, A., Khan, I. H., & Suman, R. (2023). Understanding the potential applications of Artificial Intelligence in Agriculture Sector. *Advanced Agrochem*, 2(1), 15-30.
11. Joyce, A., Goddek, S., Kotzen, B., & Wuertz, S. (2019). Aquaponics: closing the cycle on limited water, land and nutrient resources. *Aquaponics Food Production Systems*, 19.
12. Lal, R., Reicosky, D. C., & Hanson, J. D. (2007). Evolution of the plow over 10,000 years and the rationale for no-till farming. *Soil and tillage research*, 93(1), 1-12.
13. Shao, S., Li, B., Fan, M., & Yang, L. (2021). How does labor transfer affect environmental pollution in rural China? Evidence from a survey. *Energy Economics*, 102, 105515.
14. Patankar, M. S., & Taylor, J. C. (2017). *Applied human factors in aviation maintenance*. Taylor & Francis.
15. Reynolds, T. W., Waddington, S. R., Anderson, C. L., Chew, A., True, Z., & Cullen, A. (2015). Environmental impacts and constraints associated with the production of major food crops in Sub-Saharan Africa and South Asia. *Food Security*, 7, 795-822.

16. Rosenzweig, C., Iglesias, A., Yang, X. B., Epstein, P. R., & Chivian, E. (2001). Climate change and extreme weather events-Implications for food production, plant diseases, and pests.
17. Altieri, M. A., Nicholls, C. I., Henao, A., & Lana, M. A. (2015). Agroecology and the design of climate change-resilient farming systems. *Agronomy for sustainable development*, 35(3), 869-890.
18. Maina, W. N., & Maina, F. M. P. (2012). Youth engagement in agriculture in Kenya: Challenges and prospects. *Update*, 2.
19. Gupta, G. S. (2019). Land degradation and challenges of food security. *Rev. Eur. Stud.*, 11, 63.
20. Schmidhuber, J., Bruinsma, J., & Boedeker, G. (2009, June). Capital requirements for agriculture in developing countries to 2050. In *FAO expert meeting on "How to Feed the World in (Vol. 2050, pp. 24-26)*.
21. Baiphethi, M. N., & Jacobs, P. T. (2009). The contribution of subsistence farming to food security in South Africa. *Agrekon*, 48(4), 459-482.
22. Salami, A., Kamara, A. B., & Brixiova, Z. (2010). *Smallholder agriculture in East Africa: Trends, constraints and opportunities* (p. 52). Tunis, Tunisia: African Development Bank.
23. Bellemare, M. F. (2015). Rising food prices, food price volatility, and social unrest. *American Journal of agricultural economics*, 97(1), 1-21.
24. Chougule, M. A., & Mashalkar, A. S. (2022). A comprehensive review of agriculture irrigation using artificial intelligence for crop production. *Computational Intelligence in Manufacturing*, 187-200.
25. Shaikh, T. A., Rasool, T., & Lone, F. R. (2022). Towards leveraging the role of machine learning and artificial intelligence in precision agriculture and smart farming. *Computers and Electronics in Agriculture*, 198, 107119.
26. Pingali, P. (2007). Agricultural mechanization: adoption patterns and economic impact. *Handbook of agricultural economics*, 3, 2779-2805.
27. Krishna, K. R. (2017). *Push button agriculture: Robotics, drones, satellite-guided soil and crop management*. CRC Press.
28. Rüßmann, M., Lorenz, M., Gerbert, P., Waldner, M., Justus, J., Engel, P., & Harnisch, M. (2015). Industry 4.0: The future of productivity and growth in manufacturing industries. *Boston consulting group*, 9(1), 54-89.
29. Pandey, P. C., Tripathi, A. K., & Sharma, J. K. (2021). An evaluation of GPS opportunity in market for precision agriculture. In *GPS and GNSS Technology in Geosciences* (pp. 337-349). Elsevier.
30. Rajagopal, M. K., & MS, B. M. (2023). Artificial Intelligence based drone for early disease detection and precision pesticide management in cashew farming. *arXiv preprint arXiv:2303.08556*.

31. Williams, H. A., Jones, M. H., Nejati, M., Seabright, M. J., Bell, J., Penhall, N. D., ... & MacDonald, B. A. (2019). Robotic kiwifruit harvesting using machine vision, convolutional neural networks, and robotic arms. *biosystems engineering*, *181*, 140-156.
32. Gianessi, L. P. (2013). The increasing importance of herbicides in worldwide crop production. *Pest management science*, *69*(10), 1099-1105.
33. Lenain, R., Peyrache, J., Savary, A., & Séverac, G. (2021). *Agricultural robotics: part of the new deal?: FIRA 2020 conclusions* (p. 80). éditions Quae.
34. Duckett, T., Pearson, S., Blackmore, S., Grieve, B., Chen, W. H., Cielniak, G., ... & Yang, G. Z. (2018). Agricultural robotics: the future of robotic agriculture. *arXiv preprint arXiv:1806.06762*.
35. Maddikunta, P. K. R., Hakak, S., Alazab, M., Bhattacharya, S., Gadekallu, T. R., Khan, W. Z., & Pham, Q. V. (2021). Unmanned aerial vehicles in smart agriculture: Applications, requirements, and challenges. *IEEE Sensors Journal*, *21*(16), 17608-17619.
36. Fennimore, S. A., & Cutulle, M. (2019). Robotic weeders can improve weed control options for specialty crops. *Pest management science*, *75*(7), 1767-1774.
37. Zia, H., Harris, N. R., Merrett, G. V., Rivers, M., & Coles, N. (2013). The impact of agricultural activities on water quality: A case for collaborative catchment-scale management using integrated wireless sensor networks. *Computers and electronics in agriculture*, *96*, 126-138.
38. Frank, F. D., Finnegan, R. P., & Taylor, C. R. (2004). The race for talent: Retaining and engaging workers in the 21st century. *Human resource planning*, *27*(3).
39. Murphy, K. R. (2013). Job performance and productivity. In *Psychology in organizations* (pp. 169-188). Psychology Press.
40. Javaid, M., Haleem, A., Singh, R. P., & Suman, R. (2021). Substantial capabilities of robotics in enhancing industry 4.0 implementation. *Cognitive Robotics*, *1*, 58-75.
41. Steward, B. L., Gai, J., & Tang, L. (2019). The use of agricultural robots in weed management and control. *Robotics and automation for improving agriculture*, *44*, 1-25.
42. Wu, H., Wiesner-Hanks, T., Stewart, E. L., DeChant, C., Kaczmar, N., Gore, M. A., ... & Lipson, H. (2019). Autonomous detection of plant disease symptoms directly from aerial imagery. *The plant phenome journal*, *2*(1), 1-9.
43. Munirathinam, S. (2020). Industry 4.0: Industrial internet of things (IIOT). In *Advances in computers* (Vol. 117, No. 1, pp. 129-164). Elsevier.
44. Zhu, X., Chikangaise, P., Shi, W., Chen, W. H., & Yuan, S. (2018). Review of intelligent sprinkler irrigation technologies for remote autonomous system. *International Journal of Agricultural & Biological Engineering*, *11*(1).
45. Thomas, S., Kuska, M. T., Bohnenkamp, D., Brugger, A., Alisaac, E., Wahabzada, M., ... & Mahlein, A. K. (2018). Benefits of hyperspectral imaging for plant disease detection and plant protection: a technical perspective. *Journal of Plant Diseases and Protection*, *125*, 5-20.

46. Hafeez, A., Husain, M. A., Singh, S. P., Chauhan, A., Khan, M. T., Kumar, N., ... & Soni, S. K. (2022). Implementation of drone technology for farm monitoring & pesticide spraying: A review. *Information processing in Agriculture*.
47. Bechar, A., & Vigneault, C. (2016). Agricultural robots for field operations: Concepts and components. *Biosystems Engineering*, 149, 94-111.
48. Chermack, T. J. (2004). Improving decision-making with scenario planning. *Futures*, 36(3), 295-309.
49. Gil, G., Casagrande, D., Cortés, L. P., & Verschae, R. (2023). Why the low adoption of robotics in the farms? Challenges for the establishment of commercial agricultural robots. *Smart Agricultural Technology*, 3, 100069.
50. Lin, C. A. (2001). Audience attributes, media supplementation, and likely online service adoption. *Mass Communication & Society*, 4(1), 19-38.
51. Mohammad, A. S., & Pradhan, M. R. (2021). Machine learning with big data analytics for cloud security. *Computers & Electrical Engineering*, 96, 107527.
52. Hall, A., & Nahdy, S. (1999). *New methods and old institutions: The 'systems context' of farmer participatory research in national agricultural research systems: The case of Uganda*. Overseas Development Institute.
53. Angafor, G. N., Yevseyeva, I., & He, Y. (2020). Bridging the cyber security skills gap: Using tabletop exercises to solve the CSSG crisis. In *Serious Games: Joint International Conference, JCSG 2020, Stoke-on-Trent, UK, November 19–20, 2020, Proceedings 6* (pp. 117-131). Springer International Publishing.
54. Market Research Future. (2020). *Agricultural Robots Market Research Report - Global Forecast till 2025*.
55. Mishra, S., & Tyagi, A. K. (2022). The role of machine learning techniques in internet of things-based cloud applications. *Artificial intelligence-based internet of things systems*, 105-135.
56. Bogue, R. (2020). Fruit picking robots: has their time come?. *Industrial Robot: the international journal of robotics research and application*, 47(2), 141-145.
57. Shah, D., Yang, B., Kriegman, S., Levin, M., Bongard, J., & Kramer-Bottiglio, R. (2021). Shape changing robots: bioinspiration, simulation, and physical realization. *Advanced Materials*, 33(19), 2002882.
58. Herrero, M., Thornton, P. K., Mason-D'Croz, D., Palmer, J., Bodirsky, B. L., Pradhan, P., ... & Rockström, J. (2021). Articulating the effect of food systems innovation on the Sustainable Development Goals. *The Lancet Planetary Health*, 5(1), e50-e62.
59. Timmermann, C., & Félix, G. F. (2015). Agroecology as a vehicle for contributive justice. *Agriculture and human values*, 32, 523-538.
60. Linaza, M. T., Posada, J., Bund, J., Eisert, P., Quartulli, M., Döllner, J., ... & Lucat, L. (2021). Data-driven artificial intelligence applications for sustainable precision agriculture. *Agronomy*, 11(6), 1227.