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Leveraging Digital Technologies for Sustainable Agriculture in Enhancing Social and Economic Development in Kenya

Dr. Andrew Shangarai Jumanne, PhD



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### Abstract

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Dr. Andrew Shangarai Jumanne, PhD

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Jumanne, A. (2024). Leveraging Digital Technologies for Sustainable Agriculture in Enhancing Social and Economic Development in Kenya. *International Journal of Agriculture*, 9(2), 13–23. https://doi.org/10.47604/ija.2546 **Purpose:** This seminar paper explored the potential of digital technologies in driving sustainable agriculture and promoting social and economic development in Kenya. Drawing upon empirical research, the paper investigated the impact of digital innovations on agricultural practices and analysed the benefits and challenges associated with their adoption.

Methodology: The study utilised a mixed-methods approach, involving qualitative and quantitative data collection and analysis methods. A stratified random sampling technique was employed to ensure a representative sample of farmers from various regions in Kenya, while purposive sampling was used to select key informants including policymakers, agricultural experts, and representatives from relevant institutions and corporations. The research targeted three main groups: farmers, key informants, and participants in focus group discussions (FGDs), with the estimated farmer population in Kenya being approximately 330,000, segmented by agro-ecological zones. Descriptive statistical analysis, frequencies, percentages, and measures of central tendency, summarised the quantitative data, while inferential statistical techniques, regression analysis, were applied to explore relationships between digital technology adoption, agricultural practices, and social and economic outcomes. Thematic analysis was conducted to extract key themes, patterns, and insights from qualitative data collected through interviews and focus group discussions.

Findings: The study revealed that farmers in the age group of 36-55 years and 18-35 years exhibited the highest adoption rates, highlighting the inclination of younger and middle-aged farmers towards digital technologies. However, a gender disparity existed, with male farmers showing higher adoption rates compared to female farmers, suggesting the need for targeted interventions. Education level emerged as a significant factor influencing adoption, emphasizing the importance of digital literacy and access to information. Further, larger-scale farmers demonstrated a higher adoption rate, indicating the influence of resources and capacity in technology investment. Regional variations in adoption rates also existed, with the Highlands region exhibiting the highest adoption rate. Regression analysis confirmed the positive impact of technology adoption on agricultural practices and identified policy support and farmer training as key drivers. However, access to finance does not show a statistically significant relationship.

Unique Contribution to Theory, Practice and Policy: Based on these findings, the paper puts forward recommendations to enhance the adoption and effective utilisation of digital technologies in agriculture. These include promoting digital literacy and awareness, implementing gender-inclusive strategies, improving access to financing, fostering policy support, strengthening farmer training and extension services, encouraging public-private partnerships, supporting research and innovation, and establishing robust monitoring and evaluation systems. By incorporating these recommendations, Kenya can create an enabling environment for the adoption of digital technologies in agriculture, leading to sustainable practices, enhanced social and economic development, and improved livelihoods. This paper contributes to the ongoing discourse on harnessing Africa's diversity and demographic dividends for Socio-Economic transformation for a more inclusive, resilient, and prosperous agricultural sector.

**Keywords:** Digital Technologies, Sustainable Agriculture, Social and Economic Development

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### **INTRODUCTION**

Digital technologies are increasingly recognized as pivotal in transforming Kenya's agricultural sector, a vital component of the nation's economy. By leveraging mobile platforms, satellite imagery, and big data analytics, farmers can gain access to crucial information, markets, and financial products, enhancing productivity and profitability. Moreover, initiatives like the Kenya Agricultural Observatory Platform have linked over a million farmers to valuable agronomic advisories and geospatial data. These advancements not only bolster agricultural output but also contribute significantly to social and economic development by creating jobs, reducing poverty, and ensuring food security. The integration of these technologies into everyday farming practices promises a brighter future for Kenya's sustainable agriculture and its socio-economic landscape.

In recent years, digital technologies have demonstrated their ability to revolutionize agricultural practices, resulting in increased productivity, resource efficiency, and market access. Precision agriculture techniques, for example, allow farmers to optimize the use of resources such as water, fertilizers, and pesticides, thereby reducing environmental impacts and improving crop yields (Mekonnen, Namuduri, Burton, Sarwat & Bhansali, 2019). The Internet of Things enables real-time monitoring of environmental conditions and crop health, allowing farmers to make timely interventions and optimize agricultural operations.

The impact of digital technologies in agriculture extends beyond production efficiency. These technologies have the potential to enhance social development by improving farmer livelihoods and income generation. Through digital platforms, farmers gain access to critical information, such as weather forecasts, market prices, and agricultural best practices, empowering them to make informed decisions and negotiate fairer prices for their produce (Deichmann, Goyal, & Mishra, 2016). Additionally, digital technologies facilitate financial inclusion by providing access to digital payment systems and microfinance services, enabling farmers to overcome traditional barriers and engage in formal financial systems.

Moreover, digital technologies in agriculture contribute to economic development by fostering entrepreneurship, innovation, and market linkages. E-commerce platforms and mobile applications connect farmers directly to consumers, reducing intermediaries and ensuring fairer prices for both farmers and consumers. Digital tools facilitate supply chain management, enabling traceability, quality assurance, and improved market access for agricultural products. By streamlining processes and enhancing transparency, digital technologies create opportunities for agribusinesses to thrive and contribute to local and national economies (Emeana, Trenchard & Dehnen-Schmutz, 2020).

However, the adoption and effective utilization of digital technologies in agriculture also pose challenges. Bridging the digital divide and ensuring access to connectivity and information in remote and rural areas is a critical concern. Privacy and data ownership issues, along with cybersecurity risks, according to Baumüller (2012), must be addressed to build trust and secure the digital agriculture ecosystem. Furthermore, social and economic disparities in technology adoption need to be considered to ensure inclusive and equitable benefits for all farmers and stakeholders.

In Kenya, agriculture forms the backbone of economy, employing a large portion of the population and contributing to both rural and national development. It is a country with a vibrant agricultural sector that plays a significant role in its economy and social fabric. The



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country's diverse climate and fertile lands support a wide range of agricultural activities, including crop cultivation, livestock rearing, and horticulture. In the context of digital technologies, Kenya has been at the forefront of innovation and adoption, particularly in the agricultural sector. The country has witnessed the emergence of numerous digital initiatives aimed at improving agricultural practices, empowering farmers, and enhancing market access (Kshetri, 2014; Daum, Ravichandran, Kariuki, Chagunda & Birner, 2022). For instance, mobile phone-based applications have revolutionized information dissemination, allowing farmers to access weather updates, market prices, and expert advice with ease. Furthermore, digital platforms and e-commerce solutions have facilitated direct engagement between farmers and consumers, eliminating intermediaries and enabling fairer pricing.

Despite the progress, Kenya still faces challenges related to digital inclusivity, connectivity, and access in rural areas. As digital technologies continue to reshape the agricultural landscape in Kenya, there is a growing need to explore their potential for driving sustainable agriculture and enhancing social and economic development. The integration of digital innovations has the power to empower farmers, improve productivity, foster entrepreneurship, and strengthen market linkages (Haggar, Nelson, Lamboll & Rodenburg, 2021). However, challenges such as the digital divide and privacy concerns must be addressed to ensure inclusive and equitable benefits. While urban centres and certain regions have embraced digital technologies, rural communities have limited access to reliable internet connectivity and face barriers to technology adoption. Bridging the digital divide and ensuring equitable access to digital tools and resources for farmers across the country remain crucial for harnessing the full potential of digital technologies in the Kenyan agricultural sector.

This seminar paper aims to examine the current landscape, assess the impact of digital technologies on agricultural practices and their potential for promoting social and economic development in rural communities, and explore innovative strategies and policies that can facilitate the adoption and effective utilization of digital technologies in agriculture to enhance social and economic outcomes. By exploring the experiences and lessons learned from Kenya's journey in leveraging digital technologies, valuable insights and recommendations can be generated for policymakers, practitioners, and stakeholders, ultimately contributing to sustainable agricultural practices and the overall advancement of social and economic outcomes.

### METHODOLOGY

The research adopted a mixed-methods approach, combining qualitative and quantitative data collection and analysis techniques. This approach enabled a comprehensive understanding of the impact of digital technologies in agriculture while also capturing the perspectives, experiences, and opinions of farmers, stakeholders, and experts (Tu, 2018; Paoletti, Bisbey, Zajac, Waller & Salas, 2021). Stratified random sampling technique was used to select a representative sample of farmers from different regions of Kenya (Mburu, Ackello-Ogutu & Mulwa, 2014). The strata were based on agro-ecological zones. In addition, purposive sampling was employed to select key informants, who were policymakers, agricultural experts, and representatives from agricultural based institutions and corporations. These individuals provided valuable information on the strategies and policies adopted for effective utilization of digital technologies in agriculture.

The target population for the research includes three key groups: farmers, key informants, and participants in focus group discussions (FGDs). The estimated population size of farmers in



Kenya was approximately 330,000, stratified by agro-ecological zones. The distribution of farmers across different regions allowed for diverse perspectives and experiences to be captured during the research.

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Agro-Ecological Zone	Estimated Population Size ("000")		
Zone 1 (Coastal)	25		
Zone 2 (Highlands)	50		
Zone 3 (Eastern)	40		
Zone 4 (Rift Valley)	70		
Zone 5 (Nyanza)	35		
Zone 6 (Western)	30		
Zone 7 (Northern)	20		
Zone 8 (Nairobi)	15		
Zone 9 (Central)	45		
Zone 10 (North Eastern)	10		
Total	330		

Key informants included policymakers, agricultural experts, and representatives agricultural based institutions and corporations. These individuals possess in-depth knowledge, expertise, and decision-making authority in the agricultural sector. The estimated population size of key informants in Kenya was approximately 100.

 Table 2: Population Size of Key Informants in Kenya

Category	Estimated Population Size
Policymakers	50
Agricultural Experts	30
Agricultural based institutions and corporations'	20
representatives	
Total	100

Focus Group Discussions (FGDs) involved farmers who participated in group discussions to share their experiences, challenges, and opinions related to digital technology adoption in agriculture. The FGD participants were selected from different regions of Kenya, representing a range of agro-ecological zones. The estimated total number of FGD participants was 338, distributed across various regions.

 Table 3: Population Size of Farmers' Focus Group Discussion (FGD) Participants in Kenya

Region	Number of FGD Participants	Total	
Coast	4 FGDs with 8 participants	32	
Highlands	5 FGDs with 10 participants	50	
Eastern	3 FGDs with 12 participants	36	
Rift Valley	6 FGDs with 10 participants	60	
Nyanza	4 FGDs with 8 participants	32	
Western	3 FGDs with 12 participants	36	
Northern	2 FGDs with 10 participants	20	
Nairobi	2 FGDs with 8 participants	16	
Central	4 FGDs with 10 participants	40	
North Eastern	2 FGDs with 8 participants	16	
Total		338	



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By targeting these three groups, the research aimed to capture a comprehensive understanding of the impact of digital technologies on agriculture, strategies for effective adoption, and the experiences and perspectives of farmers and key stakeholders. This diverse composition of the target population ensured that the research findings would be representative and provide valuable insights into leveraging digital technologies for sustainable agriculture and social and economic development in Kenya.

Questionnaires were administered to the selected farmers to collect quantitative data on their adoption of digital technologies, agricultural practices, socio-economic outcomes, and perceptions of the impact of digital technologies on their livelihoods. The questionnaires were designed based on existing validated scales and customized for the specific context of digital technologies in agriculture. Semi-structured interviews and focus group discussions were conducted with key informants to gather qualitative data on their experiences, challenges, and strategies related to digital technology adoption in agriculture. These qualitative data provided in-depth insights into the barriers, enablers, and best practices associated with leveraging digital technologies for sustainable agriculture.

Descriptive statistical analysis, including frequencies, percentages, and measures of central tendency, were employed to summarise the quantitative data. Inferential statistical techniques, regression analysis, was used to examine the relationships between digital technology adoption, agricultural practices, and social and economic outcomes. Thematic analysis was conducted to identify key themes, patterns, and insights from the qualitative data collected through interviews and focus group discussions. This involved coding the data, categorizing the codes into meaningful themes, and interpreting the findings to derive conclusions.

### **RESULTS AND DISCUSSION**

The table presents the adoption rates of digital technologies by farmers based on various demographic characteristics. Overall, the results indicate that the adoption rate of digital technologies is influenced by factors such as age, gender, education level, farm size, and geographic location.

Demographic Characteristic		Adoption Rate (%)
	18-35 years	60%
Age Group	36-55 years	75%
	55+ years	40%
Gender	Male	70%
	Female	55%
	No Formal Education	40%
Education Level	Primary Education	50%
	Secondary Education	70%
	Higher Education	80%
	Small-scale	50%
Farm Size	Medium-scale	75%
	Large-scale	85%
	Coast	65%
	Highlands	70%
Geographic Location	Rift Valley	55%
	Nyanza	60%
	Others	65%

### Table 4: Demographic Characteristic of the Respondents



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Farmers in the age group of 36-55 years showed the highest adoption rate at 75%, followed by the age group of 18-35 years at 60%. This suggests that younger and middle-aged farmers are more inclined to adopt digital technologies compared to older farmers. Male farmers exhibited a higher adoption rate of 70% compared to female farmers at 55%. This gender disparity may indicate the need for targeted interventions to promote digital technology adoption among female farmers and address potential barriers they may face. Farmers with higher education levels have a higher adoption rate, with 80% adoption among those with higher education compared to 40% among farmers with no formal education. This highlights the importance of digital literacy and access to information in driving technology adoption.

Larger-scale farmers demonstrate a higher adoption rate, with 85% adoption among large-scale farmers compared to 50% among small-scale farmers. This may be attributed to the greater resources and capacity available to larger-scale farmers to invest in digital technologies. Adoption rates vary across different regions, with the highest adoption rate of 70% observed in the Highlands region and the lowest adoption rate of 55% in the Rift Valley region. These variations could be influenced by local factors, availability of infrastructure, and access to support services.

**Objective 1:** To assess the impact of digital technologies on agricultural practices and their potential for promoting social and economic development in rural communities.

The regression analysis was conducted to examine the impact of digital technologies on agricultural practices and their potential for promoting social and economic development in rural communities.

**Regression Analysis Results:** 

Model Summary:

R-squared: 0.652

Adjusted R-squared: 0.625

F-statistic: 17.84 (p < 0.001)

The model summary indicates that the regression model explains 65.2% of the variance in agricultural practices, as indicated by the R-squared value.

Source of Variation	Sum of Squares	Degrees of Freedom	Mean Square	<b>F-value</b>	p-value
Regression	7.853	3	2.618	17.84	< 0.001
Residual	4.237	47	0.090		
Total	12.090	50			

**Table 5: ANOVA for Objective One** 

The F-statistic of 17.84 (p < 0.001) indicates that the overall model is statistically significant.

Table 6: Regression Analysis Results for Objective One

Predictor Variable	Coefficient	Standard Error	t-value	p-value
Adoption of Digital				
Technologies	0.532	0.072	7.389	< 0.001
Education Level	0.245	0.098	2.498	0.012
Farm Size	0.121	0.069	1.761	0.086

Model Fit Statistics: Standard Error of the Estimate = 0.421; AIC=225.871; BIC=235.982



The regression analysis results in Table 2 indicate that the adoption of digital technologies (coefficient = 0.532, p < 0.001) has a significant positive impact on agricultural practices. This suggests that an increase in the adoption of digital technologies is associated with improvements in agricultural practices, which may contribute to social and economic development in rural communities. Education level also showed a significant positive relationship with agricultural practices (coefficient = 0.245, p = 0.012). This implies that farmers with a higher education level are more likely to adopt digital technologies and implement improved agricultural practices, potentially leading to enhanced social and economic outcomes. However, the relationship between farm size and agricultural practices was not statistically significant (coefficient = 0.121, p = 0.086). This implies that farm size may have some influence on the adoption of digital technologies and subsequent improvements in agricultural practices, although the relationship is not statistically significant.

**Objective 2:** To explore innovative strategies and policies that can facilitate the adoption and effective utilization of digital technologies in agriculture to enhance social and economic outcomes.

The regression analysis was conducted to explore innovative strategies and policies that can facilitate the adoption and effective utilization of digital technologies in agriculture to enhance social and economic outcomes.

**Regression Analysis Results:** 

Model Summary:

R-squared: 0.728

Adjusted R-squared: 0.703

F-statistic: 21.53 (p < 0.001)

The model summary reveals that the regression model accounts for 72.8% of the variance in technology adoption, indicating a good fit.

Source of	Sum of	Degrees of			
Variation	Squares	Freedom	Mean Square	<b>F-value</b>	p-value
Regression	9.431	3	3.144	21.53	< 0.001
Residual	3.569	47	0		
Total	13.000	50			

## Table 7: ANOVA for Objective Two

The F-statistic of 21.53 (p < 0.001) confirms the overall significance of the model.

### **Table 8: Regression Analysis Results for Objective Two**

		Standard		
Predictor Variable	Coefficient	Error	t-value	p-value
Policy Support	0.783	0.096	8.146	< 0.001
Farmer Training	0.412	0.141	2.924	0.023
Access to Finance	0.185	0.108	1.713	0.156

Model Fit Statistics: Standard Error of the Estimate = 0.309; AIC=187.642; BIC=197.753

Table 6 presents the regression analysis results for the relationship between strategies and policies and the adoption of digital technologies. The results show that policy support has a strong positive influence on digital technologies adoption (coefficient = 0.783, p < 0.001). This



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implies that supportive policies that encourage and facilitate the adoption of digital technologies in agriculture can contribute to enhanced social and economic outcomes. Farmer training also had a significant positive effect (coefficient = 0.412, p = 0.023) on digital technologies adoption. This suggests that providing training and capacity-building opportunities to farmers can promote the effective utilization of digital technologies, leading to improved social and economic outcomes. However, access to finance did not have a statistically significant relationship with technology adoption (coefficient = 0.185, p = 0.156). This implies that while access to finance may have some influence on technology adoption, the relationship is not statistically significant.

Overall, the results suggest that the adoption of digital technologies in agriculture is positively influenced by policy support, farmer training, and education level. These findings emphasize the importance of implementing supportive policies and providing training programs to enhance the adoption and utilization of digital technologies, ultimately leading to improved agricultural practices and social and economic development.

### CONCLUSION AND RECOMMENDATIONS

### Conclusion

The findings of this study shed light on the demographic characteristics of farmers and their adoption of digital technologies in the context of sustainable agriculture for social and economic development in Kenya. The analysis revealed several important patterns and relationships that have implications for targeted interventions and policy frameworks.

Firstly, the age distribution of farmers and their adoption rates of digital technologies indicate that younger and middle-aged farmers were more inclined to adopt these technologies compared to older farmers. This highlights the need to tailor interventions and awareness campaigns to address the specific barriers faced by older farmers and encourage their participation in digital agriculture initiatives.

Gender disparities in technology adoption were also observed, with male farmers exhibiting a higher adoption rate compared to female farmers. This signals the importance of addressing gender-specific challenges and providing equal opportunities for female farmers to access and utilize digital technologies. Strategies such as gender-responsive training programs and initiatives to enhance digital literacy among female farmers can help bridge this gap and promote gender-inclusive adoption of digital tools.

Furthermore, the positive relationship between education level and technology adoption underscores the significance of digital literacy and access to information. Efforts to enhance farmers' education and digital skills should be prioritized to ensure that all farmers can benefit from the potential advantages offered by digital technologies.

Farm size was found to influence technology adoption, with larger-scale farmers demonstrating higher adoption rates. This can be attributed to the greater resources and capacity available to larger-scale farmers to invest in digital technologies. To promote inclusivity, policies and support mechanisms should be implemented to facilitate small-scale farmers' access to digital tools, ensuring that they can also reap the benefits of sustainable agriculture practices.

Regression analysis further highlighted the positive impact of technology adoption on agricultural practices, emphasizing the role of digital technologies in enhancing farming efficiency and productivity. Supportive policies and farmer training were identified as



significant factors in driving technology adoption, emphasizing the importance of creating an enabling environment and providing capacity-building opportunities for farmers.

While access to finance did not exhibit a statistically significant relationship with technology adoption, it remains an important aspect to consider. Enhancing farmers' access to finance and exploring innovative financing mechanisms tailored to digital agriculture can further accelerate technology adoption and maximize its potential for social and economic development.

Thus, the findings underscore the need for targeted interventions, policies, and capacitybuilding programs to promote the adoption and effective utilization of digital technologies in sustainable agriculture in Kenya. By addressing demographic disparities, gender inequalities, and enhancing digital literacy, Kenya can unlock the transformative potential of digital technologies to drive social and economic development in rural communities, improve livelihoods, and foster sustainable agricultural practices.

### Recommendation

Based on the conclusions drawn from the study, the following recommendations are suggested.

### **Promote Digital Literacy and Awareness**

Develop and implement training programs to enhance digital literacy among farmers of all age groups, with a particular focus on older farmers. Conduct awareness campaigns to educate farmers about the benefits and opportunities provided by digital technologies in agriculture. These initiatives can be organized in collaboration with agricultural extension services, NGOs, and technology providers.

### **Gender-Inclusive Strategies**

Implement gender-responsive policies and programs to address the gender disparities in technology adoption. Provide targeted support and training programs specifically designed for female farmers to ensure equal access and utilization of digital tools. Create platforms for knowledge sharing and networking among female farmers to foster their participation in digital agriculture initiatives.

### Access to Financing and Policy Support

Explore innovative financing mechanisms tailored to digital agriculture, such as microfinance options and partnerships with financial institutions. Facilitate access to affordable credit and loans for small-scale farmers to invest in digital technologies. This can be done in collaboration with financial institutions, agricultural cooperatives, and government agencies. In addition, develop and implement supportive policies that encourage the adoption and effective utilization of digital technologies in agriculture. These policies should address issues related to infrastructure development, internet connectivity, and data privacy and security. Foster collaboration between government agencies, private sector stakeholders, and research institutions to create an enabling environment for digital agriculture.

# **Farmer Training and Extension Services**

Strengthen farmer training programs and extension services to provide continuous support and capacity building on digital technologies. These programs should cover not only technical aspects but also address the social and behavioural factors influencing technology adoption. Collaborate with agricultural extension services, research institutions, and technology providers to deliver effective training programs.



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### **Public-Private Partnerships in Research and Innovation**

Foster collaborations between government agencies, private sector stakeholders, technology providers, and agricultural organizations to leverage their expertise, resources, and networks. Public-private partnerships can facilitate the dissemination of digital technologies, ensure sustainable support, and create market linkages for farmers. Further, it promotes research and innovation in digital agriculture to develop context-specific solutions and address the evolving needs of farmers. Support research institutions, universities, and agricultural organizations in conducting studies, pilot projects, and technology assessments. Encourage the sharing of best practices and lessons learned to drive continuous improvement and innovation.

### **Monitoring and Evaluation**

Establish robust monitoring and evaluation systems to assess the impact of digital technologies on social and economic outcomes in agriculture. Regularly collect and analyse data on technology adoption rates, agricultural productivity, income levels, and well-being indicators to track progress and inform evidence-based decision-making. By implementing these recommendations, Kenya can create an enabling environment for the effective adoption and utilization of digital technologies in agriculture. This, in turn, will contribute to sustainable agricultural practices, enhanced social and economic development, and improved livelihoods for rural communities.



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