



THE TRANSFORMATIVE IMPACT OF ARTIFICIAL INTELLIGENCE AND SENSOR TECHNOLOGIES ON DAIRY LIVESTOCK EXPORTS

Hatice DİLAVER^{1*}, Kamil Fatih DİLAVER²

¹Niğde Ömer Halisdemir University, Department of Eurasia Studies, 51200, Niğde, Türkiye

²Niğde Ömer Halisdemir University, Faculty of Engineering, Department of Electric and Electronics, 51200 Niğde, Türkiye

Abstract: This study explores the transformative impact of artificial intelligence (AI) and sensor technologies on dairy livestock exports. AI-based predictive analytics, automatic milking systems (AMS), and IoT sensors demonstrate significant potential for enhancing operational efficiency, animal welfare, and environmental sustainability. The research investigates the effects of these technologies on animal health management, disease detection, and monitoring while evaluating the challenges and limitations associated with their implementation. Furthermore, the discussion extends to addressing environmental stress factors caused by climate change and the effects of fluctuating global market demands. Future research directions include explainable AI (XAI), IoT and blockchain integration, ethical frameworks, climate-resilient technologies, and policy recommendations. The findings underscore the potential of AI and sensor technologies to revolutionize dairy livestock exports by fostering sustainability and productivity, emphasizing the need for collective action among stakeholders. The increasing global demand for dairy livestock exports necessitates innovative solutions to address challenges related to operational efficiency, animal welfare, and environmental sustainability. Artificial intelligence (AI) and sensor technologies, including predictive analytics, automatic milking systems (AMS), and Internet of Things (IoT) sensors, have the potential to revolutionize livestock management. This study examines the impact of these technologies on disease detection, real-time monitoring, and logistics optimization while addressing challenges such as data security, cost implications, and regulatory constraints. The discussion extends to climate change-related stress factors and global market fluctuations. Future research should focus on explainable AI (XAI), blockchain-enabled traceability, climate-resilient innovations, and policy frameworks. The findings emphasize the need for multi-stakeholder collaboration to leverage AI and sensor technologies for a sustainable and efficient dairy livestock export industry.

Keywords: Dairy livestock, Smart farming, Blockchain, IoT, Climate resilience

*Corresponding author: Niğde Ömer Halisdemir University, Department of Eurasia Studies, 51200, Niğde, Türkiye

E mail: haticedilaver509@gmail.com (H. DİLAVER)

Hatice DİLAVER  <https://orcid.org/0000-0002-4484-5297>

Kamil Fatih DİLAVER  <https://orcid.org/0000-0001-7557-9238>

Received: February 16, 2025

Accepted: June 25, 2025

Published: July 15, 2025

Cite as: Dilaver H, Dilaver KF. 2025. The transformative impact of artificial intelligence and sensor technologies on dairy livestock exports. BSJ Agri, 8(4): 578-586.

1. Introduction

The dairy sector, a cornerstone of global economies, grapples with challenges such as disease control, animal welfare, and supply chain inefficiencies (Zhang and Liu, 2021). If unresolved, these issues could limit productivity and sustainability. The advent of artificial intelligence (AI) and advanced sensor technology is instigating paradigm shifts in traditional sectors, including the dairy livestock export industry (Zhao and Wang, 2022). These technologies promise to redefine the sector's landscape (Zohra and Kamble, 2021) by promoting humane, efficient, and sustainable practices (Hansen et al., 2018). Exporting dairy livestock involves logistical and welfare challenges. Livestock are exposed to various stresses during long shipping durations, potentially compromising their health. Accurate livestock counting and ensuring sufficient feed during transit are major concerns. Given the increasing global demand for dairy products and livestock for breeding, it is imperative to

improve livestock export procedures. Precision digital livestock farming, underpinned by AI and sensor technology, offers innovative solutions to persistent issues in the dairy livestock export industry. These disruptive technologies facilitate real-time monitoring, proactive intervention, and data-driven decision-making, promising enhanced animal welfare, productivity, and streamlined supply chain operations (Song and Chen, 2022). Traditional practices often falter in managing animals, struggling to meet their nutritional needs due to a lack of assertiveness (shy feeders) in reaching feeding troughs in group feeding scenarios (Yadav and Gupta, 2021). This behavioral pattern adversely impacts health and productivity, leading to undernutrition, weight loss, decreased productivity, increased disease susceptibility, and a shortened lifespan (Yi and Li, 2022). Tracking individual animal weights, a crucial parameter of health and performance (Zhang and Chen, 2021), along with accurate cattle counting during the export process (Liu



and Xu, 2022), remains laborious, time-consuming, and error-prone when using conventional methods. These inaccuracies can lead to supply chain discrepancies, inducing financial losses and logistical complications (Bansal and Singh, 2021). Emerging AI and sensor technology bring forth a beacon of hope (Zhang and Lin, 2022). By enabling real-time monitoring and data analytics, they can address these concerns, thereby enhancing operational efficiency and animal welfare, crucial elements for the long-term sustainability of the dairy livestock export industry (Zhang and Chen, 2021; Zhang and Yu, 2020).

AI and sensor technology can automate the tracking of individual animal feeding behaviors (Yadav and Gupta, 2021), including their time spent near feeding troughs, feeding frequency, and intake (Zohra and Kamble, 2021). These data, once analyzed using AI algorithms, can facilitate early interventions. Furthermore, these technologies assist in monitoring feeding behaviors—vital for livestock management—as they provide insights into animals' health, productivity, and overall well-being. Systems employing RFID tags or smart collars can automate this process, offering real-time data (Pinto and Santos, 2022). These systems track proximity to feeding troughs, feeding frequency (Guo and Jin, 2020), and duration. Analyzing these data can alert farmers to anomalies, enabling early health issue detection and aiding in optimizing feed management, thereby enhancing productivity and sustainability. This study aims to contribute to the evolution of a more sustainable, efficient, and humane dairy livestock export industry through a focused exploration of the Internet of Things (IoT), sensor, and artificial intelligence (AI) technologies. Central to our investigation are three key applications: managing feeding behaviors, automating livestock weight tracking, and improving cattle counting accuracy during transportation and traceability (Bansal and Singh, 2021). Our analysis encompasses the potential for advancements in animal welfare, operational efficiency, and market access. Concurrently, we address the challenges emerging from the adoption of these cutting-edge technologies, including data security, privacy, infrastructure demands, sensor data reliability, interpretability of AI insights, ethical considerations, and cost implications (Wang and Zhao, 2021).

Dairy farming is a critical component of the global agricultural economy, with exports playing a vital role in meeting international demand. However, traditional methods of livestock management face significant limitations in disease monitoring, supply chain inefficiencies, and animal welfare concerns (Zhang and Liu, 2021). The advent of AI and sensor technologies is ushering in a paradigm shift in livestock management, promising automation, real-time analytics, and enhanced decision-making processes (Zhao and Wang, 2022).

Exporting dairy livestock presents logistical and welfare challenges. Animals endure prolonged transportation periods, leading to stress, weight loss, and potential

health deterioration. Ensuring sufficient feed, water, and monitoring environmental conditions during transit are persistent challenges. AI-powered solutions, such as machine learning algorithms for predictive analytics and IoT-enabled smart monitoring systems, offer precise and proactive interventions (Zohra and Kamble, 2021). Moreover, these technologies aid in livestock counting, tracking feeding behaviors, and optimizing nutritional intake to enhance productivity and sustainability (Hansen et al., 2018).

1.1. Role of AI and Sensor Technology in Livestock Management

Traditional agricultural practices often depend on manual observations and historical records, which can result in operational inefficiencies and delayed interventions. The integration of artificial intelligence (AI) and sensor technologies enables real-time monitoring, allowing for the early detection of animal health issues before they become critical (Song and Chen, 2022). These systems continuously track vital indicators such as weight variations, feeding behaviors, movement patterns, and physiological stress responses (Yi and Li, 2022). Advanced machine learning algorithms analyze large volumes of data to generate actionable insights, thereby enhancing decision-making processes related to breeding management, disease prevention, and supply chain optimization (Zhang and Chen, 2021; Liu and Xu, 2022).

2. Application of AI and Sensor Technology in Livestock Management

Artificial Intelligence (AI) and sensor technologies are propelling industries into the future, and the dairy livestock export industry is no exception. As a sector steeped in tradition, this industry is beginning to feel the transformative impact of these advanced technologies. AI and sensor technologies are not only changing the way farmers and exporters operate but also shaping the future trajectory of the industry. In any livestock population, certain animals stand out due to their lack of assertiveness in reaching feeding troughs, especially in group feeding scenarios. These animals, colloquially referred to as 'shy feeders', exhibit a lower food intake (Yang and Wei, 2022). This behavior significantly impacts their health and productivity, often leading to undernutrition, weight loss, decreased productivity, a heightened susceptibility to diseases, and, in some instances, a reduced lifespan (Ghosh and Meena, 2021). Some cows tend to consume less feed when in group-feeding environments. These animals are commonly referred to in the literature as reluctant or low-intake feeders. Especially during stressful periods such as transportation, insufficient feed intake may pose significant risks to their health and productivity. Considering these potential outcomes, the early detection and effective management of such animals have become critically important. Historically, this issue has been

addressed through manual observation and intervention, which are both time-consuming and labor-intensive, and not always accurate or timely. With the advancement of artificial intelligence (AI) and sensor technologies, a fundamental shift in the management of reluctant feeders is now possible. AI can analyze animal behavior patterns through video and motion sensors, enabling the identification of individuals with low feed intake and the adjustment of feeding strategies accordingly. The use of AI and sensor-based monitoring systems provides innovative solutions to the challenges encountered in livestock management (see Table 1), especially in identifying and supporting animals with feeding difficulties (Ghosh and Meena, 2021). These technologies enable the automatic tracking of individual animal behaviors such as time spent at feeding troughs, feeding frequency, and feed consumption. When these data are processed through AI algorithms, animals experiencing feeding problems can be promptly identified, allowing for early and targeted interventions (Liu and Xu, 2022).

2.1. Real-Time Disease Detection and Health Monitoring

AI-powered diagnostic systems utilize computer vision, thermal imaging, and biometric sensors to detect early signs of disease. These systems integrate with wearable technologies such as RFID tags, smart collars, and embedded microchips to provide real-time health assessments (Pinto and Santos, 2022). By detecting anomalies in temperature, movement, and feeding behaviors, these technologies facilitate early intervention, reducing mortality rates and improving

overall herd health (Guo and Jin, 2020).

2.2. Precision Feeding and Nutritional Optimization

AI algorithms analyze feeding behaviors to customize dietary plans based on individual animal needs (Yadav and Gupta, 2021). IoT-enabled feed intake sensors track consumption rates and detect abnormalities, allowing adjustments in diet composition. Automated feed dispensers integrate with AI models to ensure optimal nutrition, reducing waste and improving livestock productivity (Bansal and Singh, 2021).

2.3. Blockchain for Supply Chain Transparency

Blockchain technology enhances traceability in dairy livestock exports by providing immutable records of health status, transportation conditions, and regulatory compliance (Alshehri, 2023). By leveraging decentralized ledgers, stakeholders can ensure authenticity and reduce fraud in global dairy trade. Blockchain-integrated smart contracts streamline payment processes and automate compliance verification (Bhat et al., 2021).

2.4. Environmental Stress Management and Climate Resilience

Climate change-induced stress affects livestock health, necessitating predictive models for adaptive management. AI-driven climate models analyze temperature, humidity, and historical disease outbreaks to predict potential risks (Amiri-Zarandi et al., 2022). IoT-enabled cooling systems, automated shade structures, and precision hydration mechanisms mitigate heat stress and improve animal welfare (Gehlot et al., 2022).

Table 1. Key applications of Artificial Intelligence and sensor technology in livestock management

AI and Sensor Technology Role	Specific Technology Used	Benefits	Limitations
Identification of 'Shy Feeders'	RFID tags, computer vision, machine learning algorithms	Aids early identification and intervention, improving herd health and productivity.	Needs sensor setup and careful AI calibration to minimize false results.
Monitoring of Feeding Behaviors	Feed intake sensors, IoT connectivity, cloud computing, machine learning algorithms	Enables personalized nutrition plans. Provides real-time insights into animal health and nutrition status, enabling timely interventions. Helps prevent over/underfeeding.	Requires robust connectivity and sensor maintenance for real-time monitoring.
Automation of Weight Collection	Walk-over-weighing systems, IoT connectivity, cloud computing, machine learning algorithms	Provides accurate, hassle-free weight tracking. Allows continuous monitoring of animal performance. Assists in adjusting feeding strategies.	Requires animal training to use the system, sensor calibration, and maintenance for accurate readings.

In addition, the role of monitoring feeding behaviors in livestock management provides invaluable insights into animals' health, productivity, and overall wellbeing. AI and sensor-based systems revolutionize this aspect by

offering real-time, automated monitoring capabilities. These systems employ technologies such as RFID tags or smart collars to track various parameters, converting raw data into meaningful insights about the animals'

feeding behaviors (Wang and Zhao, 2021). By addressing long-standing challenges such as the optimization of feeding behavior, these technologies can significantly enhance animal welfare, streamline supply chain operations, and boost the dairy livestock export industry's productivity and profitability (Zhang and Chen, 2021; Zhang and Yu, 2020). In light of these limitations, there is a burgeoning consensus in the agrifood domain that advanced machine learning and deep learning approaches could be the panacea to these challenges. Deep learning, a subset of machine learning inspired by the structure and function of the human brain, has demonstrated remarkable success in handling high-dimensional data and delivering accurate predictions in various fields, including image recognition, natural language processing, and autonomous vehicles (Singh and Rana, 2020).

3. Future Perspectives: AI and Sensor Technology in Livestock Management

In the context of dairy livestock export, deep learning models could be trained to recognize patterns and make predictions based on a multitude of factors, including the spatial-temporal distribution of animals, their health status, environmental conditions, and market trends (Sharma and Sharma, 2021). These models could potentially provide more accurate and timely insights, enabling farmers and exporters to make better-informed decisions, optimize their operations, and ultimately, enhance the sustainability and profitability of their enterprises. However, the adoption of deep learning in the agrifood sector is not without its challenges. These include the need for large amounts of labeled training data, the complexity of model development and tuning, and the interpretability of model predictions (Yadav and Gupta, 2021). The need for large amounts of labeled training data can be particularly challenging, as it requires significant time and effort to collect and label data. Moreover, the complexity of model development and tuning requires specialized skills and expertise, which may not be readily available in the agrifood sector. Moreover, the successful integration of deep learning models into the livestock management workflow would require significant investment in infrastructure, skills development, and change management. This includes not only the physical infrastructure for data storage and processing but also the software infrastructure for data management, model development, and deployment. Skills development is another critical aspect, as it requires the training and upskilling of staff to effectively use and manage the advanced analytics solutions (Faverjon et al., 2019). Change management, on the other hand, involves managing the organizational changes associated with the adoption of new technologies and practices. In addition to these challenges, the interpretability of model predictions is another critical issue. Deep learning models, often referred to as 'black boxes,' can make

highly accurate predictions but may not provide clear explanations for their predictions (Stojanovic and Mitić, 2020). This lack of interpretability can be a significant barrier to the adoption of deep learning in the agrifood sector, where decision-makers often need to understand the reasons behind the predictions to make informed decisions. Despite these challenges, the potential benefits of adopting advanced analytics and deep learning in dairy livestock export are immense. These benefits extend beyond improved operational efficiency and productivity to include enhanced animal welfare, more sustainable farming practices, and increased competitiveness in the global market (Guo and Jin, 2020). By providing real-time, accurate, and actionable insights, advanced analytics can enable farmers and exporters to make better-informed decisions, optimize their operations, and respond more effectively to market trends and changes.

Moreover, by improving the tracking and monitoring of animal health and welfare, advanced analytics can contribute to higher standards of animal welfare and more ethical farming practices. This, in turn, can enhance the reputation and brand value of dairy livestock exporters, making them more attractive to consumers and investors who value sustainability and animal welfare (Koltes et al., 2019). While the path towards advanced analytics in dairy livestock export is challenging, the potential benefits in terms of improved efficiency, animal welfare, and profitability make it a journey worth undertaking. As researchers, developers, and industry stakeholders continue to explore and innovate in this space, the future of dairy livestock export looks set to be increasingly data-driven, intelligent, and sustainable (Marques et al., 2022). The journey towards this future will require not only technological innovation but also collaboration and partnership among various stakeholders, including farmers, exporters, technology providers, researchers, and policymakers. By working together, these stakeholders can overcome the challenges and unlock the full potential of advanced analytics in dairy livestock export (Zohra and Kamble, 2021).

3.1. Developing Explainable AI (XAI) Models

A major challenge in AI adoption is the black-box nature of deep learning algorithms. Future research should focus on XAI frameworks that enhance transparency and interpretability (Farooq et al., 2022). These frameworks will enable farmers and policymakers to understand AI-driven recommendations and ensure ethical AI applications in livestock management.

3.2. IoT and Blockchain Integration

Combining IoT sensors with blockchain can improve traceability and fraud prevention in dairy exports. Future studies should explore interoperable architectures that facilitate seamless data exchange between stakeholders (Milani et al., 2021).

3.3. Policy Incentives for AI Adoption

Governments and industry regulators should implement financial incentives such as subsidies, tax credits, and

research grants to promote AI and sensor technology adoption in livestock farming. Training programs and knowledge-sharing initiatives will further accelerate the adoption of smart farming solutions (Sharma and Sharma, 2021).

3.4. Ethical and Regulatory Considerations

As AI becomes more integrated into livestock management, ethical concerns regarding data privacy, animal rights, and automation-driven job displacement must be addressed. Future research should develop regulatory frameworks that balance technological advancements with ethical imperatives (Jukan et al., 2017).

4. The Intersection of Sensor Technologies and Artificial Intelligence: A Closer Look

Sensor technologies and AI form a critical intersection in modern livestock management. Sensors provide a way to collect real-time data from livestock. The vast amounts of data collected can be overwhelming and seemingly chaotic, which is where AI steps in, decoding these massive data sets, identifying patterns, and providing actionable insights (Zhang and Liu, 2021).

4.1. Sensor Technologies

Sensor technologies can be categorized into two types: wearable devices and environment-based sensors. Wearable sensors are devices attached directly to the animal. They may track physiological parameters (e.g., heart rate, body temperature), behavioral traits (e.g., feeding patterns, movement), and other relevant indicators of an animal's health and welfare (Milani et al., 2021). Environment-based sensors, on the other hand, monitor the conditions around the animals. These could include video cameras, thermal imaging sensors, accelerometers, load cells in feeding stations, and drones, among others. They can provide a wealth of information about the environment and how animals interact with it (Zhang and Lin, 2022). The application of sensor technologies in livestock management has opened new avenues for the in-depth monitoring of animals in ways that were previously impossible. However, there are challenges such as the durability of wearable devices, potential discomfort or injury to the animal, ensuring the devices stay on the animals, and the cost and complexity of installing and maintaining environment-based sensors (Zhang and Liu, 2021).

4.2. Artificial Intelligence

AI is a broad field that encompasses machine learning, deep learning, computer vision, and more. It provides the capability to analyze and interpret the massive data sets collected by sensors. AI's ability to 'learn' from data and make predictions makes it a powerful tool for decoding the vast array of livestock data (Wang and Zhang, 2021). In the context of dairy livestock export (Yao and Li, 2021), AI can be used for a variety of applications. It can identify patterns in livestock behavior and physiological parameters to detect illness, stress, or discomfort. It can

analyze patterns in feeding behavior to identify shy feeders and adjust the feeding strategies. It can recognize and count individual animals in video footage, and it can use data on feeding behaviors and physiological parameters to optimize the feeding schedules and portions. However, AI also poses challenges. Developing accurate AI algorithms requires substantial amounts of high-quality training data. There are also ethical considerations associated with AI, such as privacy concerns and the potential for the misuse of data (Song and Chen, 2022).

4.3. Integration of Multiple Sensor Modalities

One of the future directions in this area is the integration of multiple sensor modalities for the comprehensive monitoring of animal health. A single sensor can only provide a limited perspective. For instance, a wearable device may monitor heart rate, but it might not be able to provide insights into the environmental factors influencing the animal's stress levels (Darvazeh et al., 2020). By integrating data from wearable sensors, environmental sensors, and video data, a more holistic understanding of the animal's condition can be obtained (Zhao and Wang, 2022).

4.4. Advancements in AI Algorithms

Advancements in AI algorithms will also play a significant role in the future of dairy livestock export. Current AI models, such as machine learning and deep learning algorithms, are already powerful tools for analyzing livestock data (Klerkx et al., 2019). However, these models could be further improved. For instance, developing algorithms that can analyze multiple types of data (e.g., physiological data, environmental data, video data) simultaneously could provide more comprehensive and accurate insights (Ghosh and Meena, 2021).

4.5. Customization and Individual Animal Approach

As with any technology application, one size does not fit all. Dairy cattle have individual differences in their behavior, physiology, and response to environmental stressors. These differences need to be taken into account when designing and implementing sensor and AI systems. For example, the optimal position and type of wearable sensor might vary depending on the size, breed, and behavior of the cow. AI algorithms also need to be designed to account for the individual differences between cows (Zohra and Kamble, 2021).

4.6. Technological Adaptation and Training

The implementation of sensor technologies and AI systems in the dairy livestock export sector requires adequate training for staff. Staff must be trained to install and maintain the technologies, interpret the data generated, and take the appropriate actions based on the insights provided by AI. Additionally, the dairy cattle must adapt to the new technologies, particularly the wearable devices. Proper training and adaptation are crucial for the successful implementation of these technologies (Guo and Jin, 2020).

4.7. Stakeholder Engagement and Consumer Perception

The use of sensor technologies and AI in dairy livestock export has implications beyond the farm gate. Stakeholders, including consumers, have increasingly high expectations for animal welfare, environmental sustainability, and food safety. The use of these technologies can help meet these expectations by improving animal welfare and reducing the environmental impacts. However, there is also a need to effectively communicate with stakeholders about the use of these technologies to avoid misconceptions and ensure that they are accepted (Zhang and Lin, 2022).

4.8. Data Security and Privacy

With the rise of sensor technologies and AI, vast amounts of data are being collected and analyzed. This presents significant challenges in terms of data security and privacy. Ensuring the secure storage and transmission of data is crucial to prevent unauthorized access and the misuse of data. Regulations and best practices need to be developed and implemented to ensure data security and privacy (Wang and Zhao, 2021).

4.9. Economic Considerations

While sensor technologies and AI offer many benefits, they also come with costs. The initial investment in the hardware, software, and training can be substantial. There are also ongoing costs for maintenance and data management (Wang and Zhao, 2021). Therefore, careful economic analysis is necessary to ensure the benefits outweigh the costs. This includes not only the direct economic benefits but also indirect benefits such as improved animal welfare, reduced environmental impacts, and enhanced public perception (Hou et al., 2020).

4.10. Ensuring Animal Comfort and Welfare

While wearable devices offer valuable data on an individual animal's health and well-being, it is vital to ensure that these devices do not compromise the comfort or welfare of the animals. Sensor devices should be designed and fitted in a way that minimizes the potential discomfort, injury, or stress for the animals (Džermeikaite et al., 2023). Regular checks are needed to ensure the devices remain in the correct position and are not causing any harm to the animals. It is also essential to consider the potential stress associated with introducing new technologies and to manage this process carefully to minimize stress for the animals (Zhang and Liu, 2021).

4.11. Technology Integration and Interoperability

Given the variety of sensor technologies and AI applications that can be used in dairy livestock export, there is a need to ensure these technologies can be integrated and can operate together seamlessly. This includes not only the integration of different types of sensor data but also the interoperability of different AI algorithms. Developing standardized protocols for data collection, storage, and analysis can help ensure the smooth integration and interoperability of these technologies (Song and Chen, 2022).

4.12. Continual Monitoring and Evaluation

Finally, as sensor technologies and AI are implemented in the dairy livestock export process, it is crucial to continually monitor and evaluate their effectiveness. This includes not only tracking their performance in terms of improving animal health and welfare, but also assessing their impact on operational efficiency, economic outcomes, and environmental sustainability. Regular evaluations can help identify any issues or areas for improvement and ensure that the technologies are providing the maximum number of benefits (Zhang and Liu, 2021). Despite significant advances in sensor and Artificial Intelligence (AI) technologies, their adoption within the livestock export sector remains in the early stages (Milani et al., 2021). Numerous studies have been conducted to explore the potential of these technologies in enhancing the efficiency and safety of livestock export, but their translation from research and development to practical implementation is still a work in progress (Ghosh and Meena, 2021). There are several reasons for this lag in adoption. One of the main challenges is the practical application of technology. Many technologies that function well in controlled lab conditions might struggle in the dynamic and complex real-world scenarios of livestock export (Farooq et al., 2022).

5. Future Research Directions and Policy Implications

The integration of artificial intelligence (AI) and sensor technologies into dairy livestock export processes opens up several research avenues and necessitates proactive policy measures. To ensure the sustainable adoption and expansion of these technologies, interdisciplinary efforts are required from researchers, policymakers, and industry stakeholders (Koltes et al., 2019). These challenges underscore the importance of prioritizing practicality and context-specific solutions in future research and development efforts. Additionally, fostering collaboration between researchers, technologists, and livestock industry stakeholders is crucial for the practical implementation and refinement of these technologies (Patle and Srivastava, 2022).

5.1. Developing Explainable AI (XAI)

One of the critical challenges in adopting AI in livestock management is the "black box" nature of many deep learning algorithms (Paton and Sparks, 2020). Explainable AI (XAI) seeks to address this issue by providing transparent and interpretable models that offer insights into decision-making processes. Future research should focus on developing XAI frameworks tailored to the agrifood sector, enabling stakeholders to trust and adopt AI-driven recommendations confidently (Farooq, 2022).

5.2. Ethical Frameworks for AI and Sensor Use

As sensor technologies and AI systems become more pervasive, ethical considerations surrounding animal welfare, data privacy, and the socio-economic impacts of

automation must be addressed (Faroe et al., 2022). Future studies should explore frameworks to balance technological advancement with ethical imperatives, ensuring that these technologies enhance rather than compromise animal welfare and farmer livelihoods (Jukan et al., 2017).

5.3. Leveraging IoT and Blockchain Integration

The Internet of Things (IoT) and blockchain technologies offer untapped potential for improving transparency and traceability in the dairy livestock supply chain (Milani et al., 2021). Blockchain can provide immutable records of an animal's health, movement, and environmental conditions, while IoT can facilitate real-time data exchange. Collaborative research efforts should examine the integration of these technologies to improve operational efficiency and market trust (Grant and Ferraretto, 2018).

5.4. Climate-Resilient Technologies

The increasing frequency of extreme weather events due to climate change poses a significant threat to dairy livestock operations. Future research should focus on designing AI-driven systems that adapt to and mitigate the effects of environmental stressors. This includes predictive models for heat stress management and early warning systems for disease outbreaks linked to climatic changes (Hansen et al., 2018).

5.5. Policy Recommendations for Technology Adoption

Governments and international organizations must play a proactive role in facilitating the adoption of advanced analytics and sensor technologies (Džermeikaite et al., 2023). Policies should focus on providing financial incentives for technology adoption, such as subsidies or tax credits, as well as funding for research and development. Additionally, training programs for farmers and export managers are essential to ensure the effective implementation of these technologies (Sharma and Sharma, 2021).

6. Concluding Remarks

As the dairy livestock export industry navigates the complexities of integrating advanced technologies, it is crucial to adopt a multidisciplinary approach that combines technological innovation with ethical, economic, and environmental considerations. By leveraging the transformative potential of AI, sensor technologies, and supporting innovations like blockchain and IoT, the industry can move toward a future characterized by enhanced productivity, sustainability, and animal welfare. The successful transition to a data-driven model requires the collective efforts of all stakeholders, including researchers, policymakers, technology providers, and industry practitioners. With concerted action and a focus on overcoming existing challenges, the promise of a technologically empowered and sustainable livestock export industry can be realized (Sharma and Sharma, 2021).

Artificial intelligence (AI) and sensor technologies hold

transformative potential for enhancing dairy livestock exports through improvements in operational efficiency, animal welfare, and environmental sustainability. The implementation of AI-based predictive analytics, automatic milking systems (AMS), and IoT sensors addresses critical challenges such as disease detection, environmental stress management, and market adaptability. However, the integration of these technologies requires addressing ethical, technical, and economic constraints.

Future advancements, including explainable AI (XAI), blockchain-enabled traceability, and climate-resilient technologies, will further shape the global dairy industry. To achieve sustainable progress, a collaborative approach among policymakers, researchers, farmers, and technology developers is essential. These innovations can pave the way for a more efficient, resilient, and ethically driven dairy farming sector, ensuring long-term global competitiveness.

As the dairy livestock export industry navigates the complexities of integrating advanced technologies, it is crucial to adopt a multidisciplinary approach that combines technological innovation with ethical, economic, and environmental considerations. By leveraging AI, sensor technologies, and supporting innovations like blockchain and IoT, the industry can move toward a future characterized by enhanced productivity, sustainability, and animal welfare (Yao and Li, 2021).

The successful transition to a data-driven model requires the collective efforts of all stakeholders, including researchers, policymakers, technology providers, and industry practitioners. With concerted action and a focus on overcoming existing challenges, the promise of a technologically empowered and sustainable livestock export industry can be realized.

The dairy livestock export industry is undergoing a transformative shift with the integration of AI and sensor technologies. These innovations enhance disease detection, precision feeding, supply chain traceability, and climate resilience. However, challenges such as AI explainability, data security, and regulatory compliance must be addressed to maximize their benefits. Future advancements in explainable AI, blockchain, and IoT integration will further optimize the global dairy industry. Stakeholder collaboration, policy incentives, and ethical AI implementation will be essential for a sustainable and competitive livestock sector.

Author Contributions

The percentages of the authors' contributions are presented below. All authors reviewed and approved the final version of the manuscript.

	H.D.	K.F.D.
C	20	80
D	20	80
S		
DCP	50	50
DAI		100
L	20	80
W	20	80
CR	30	70
SR	70	30

C=Concept, D= design, S= supervision, DCP= data collection and/or processing, DAI= data analysis and/or interpretation, L= literature search, W= writing, CR= critical review, SR= submission and revision.

Conflict of Interest

The author declared that there is no conflict of interest.

References

- Alshehri M. 2023. Blockchain-assisted Internet of Things framework in smart livestock farming. *Internet of Things*, 22: 100739.
- Amiri-Zarandi M, Dara RA, Duncan E, Fraser ED. 2022. Big data privacy in smart farming: A review. *Sustainability*, 14: 9120.
- Andronie M, Lăzăroiu G, Iatagan M, Hurloiu I, Dijmărescu I. 2021. Sustainable cyber-physical production systems in big data-driven smart urban economy: A systematic literature review. *Sustainability*, 13: 751.
- Bansal M, Singh K. 2021. Advances in smart farming technologies for animal health monitoring. *Comput Ind*, 128: 103482.
- Bhat SA, Huang NF, Sofi IB, Sultan M. 2021. Agriculture-food supply chain management based on blockchain and IoT: A narrative on enterprise blockchain interoperability. *Agriculture*, 12: 40.
- Bhattarai BP, Paudyal S, Luo Y, Mohanpurkar M, Cheung K, Tonkoski R, Hovsapien R, Myers KS, Zhang R, Zhao P, Yang C. 2019. Big data analytics in smart grids: State-of-the-art, challenges, opportunities, and future directions. *IET Smart Grid*, 2: 141-154.
- Brault SA, Hannon SJ, Gow SP, Otto SJ, Booker CW, Morley PS. 2019. Calculation of antimicrobial use indicators in beef feedlots—Effects of choice of metric and standardized values. *Front Vet Sci*, 6: 330.
- Cockburn M. 2020. Application and prospective discussion of machine learning for the management of dairy farms. *Animals*, 10: 1690.
- Darvazeh SS, Vanani IR, Musolu FM. 2020. Big data analytics and its applications in supply chain management. In: *New trends in the use of artificial intelligence for the industry 4.0.*, IntechOpen, pp: 175
- Džermeikaite K, Bacėninaite D, Antanaitis R. 2023. Innovations in cattle farming: Application of innovative technologies and sensors in the diagnosis of diseases. *Animals*, 13: 780.
- Farooq MS, Sohail OO, Abid A, Rasheed S. 2022. A survey on the

- role of IoT in agriculture for the implementation of smart livestock environment. *IEEE Access*, 10: 9483-9505.
- Faverjon C, Bernstein A, Grütter R, Nathues C, Nathues H, Sarasua C, Sterchi M, Vargas ME, Berezowski J. 2019. A transdisciplinary approach supporting the implementation of a big data project in livestock production: An example from the Swiss pig production industry. *Front Vet Sci*, 6: 215.
- Gehlot A, Malik PK, Singh R, Akram SV, Alsuwian T. 2022. Dairy 4.0: Intelligent communication ecosystem for the cattle animal welfare with blockchain and IoT enabled technologies. *Appl Sci*, 12: 7316.
- Ghosh S, Meena S. 2021. Precision agriculture and IoT-based monitoring systems for livestock farming: A review. *Comput Electron Agric*, 185: 106127.
- Grant RJ, Ferraretto LF. 2018. Silage review: Silage feeding management: Silage characteristics and dairy cow feeding behavior. *J Dairy Sci*, 101: 4111-4121.
- Guo F, Jin T. 2020. An overview of smart livestock farming based on IoT technologies. *Journal of Sensors*, 2020: 7451409.
- Haldar A, Mandal SN, Deb S, Roy R, Laishram M. 2022. Application of information and electronic technology for best practice management in livestock production system. In: *Agriculture, livestock production and aquaculture: Adv Smallhol Farm Syst* 2: 173-218.
- Hansen MF, Smith ML, Smith LN, Jabbar KA, Forbes D. 2018. Automated monitoring of dairy cow body condition, mobility and weight using a single 3D video capture device. *Comput Ind*, 98: 14-22.
- Hou S, Cheng X, Shi L, Zhang S. 2020. Study on individual behavior of dairy cows based on activity data and clustering. In: *Proceedings of the 2020 2nd International Conference on Robotics, Intelligent Control and Artificial Intelligence*, Shanghai, China, 17-19 October 2020, pp: 210-216.
- Jukan A, Masip-Bruin X, Amla N. 2017. Smart computing and sensing technologies for animal welfare: A systematic review. *ACM Comput Surv*, 50: 10.
- Kanjo E, Younis EM, Ang CS. 2019. Deep learning analysis of mobile physiological, environmental and location sensor data for emotion detection. *Inf Fusion*, 49: 46-56.
- Klerkx L, Jakku E, Labarthe P. 2019. A review of social science on digital agriculture, smart farming and agriculture 4.0: New contributions and a future research agenda. *NJAS-Wageningen J Life Sci*, 90: 100315.
- Koltes JE, Cole JB, Clemmens R, Dilger RN, Kramer LM, Lunney JK, McCue ME, McKay SD, Mateescu RG, Murdoch BM, et al. 2019. A vision for development and utilization of high-throughput phenotyping and big data analytics in livestock. *Front Genet*, 10: 1197.
- Kothari A, Singh A. 2020. IoT-enabled smart farming system for precision agriculture. *The 2020 International Conference on Electronics and Sustainable Communication Systems*, July 2-4, Coimbatore, India, pp: 222-228.
- Le CS, Chen L. 2020. Applications of blockchain in agricultural and food supply chains: A review. *Food Control*, 112: 107102.
- Li H, Wang L. 2021. Big data and machine learning in livestock management: A survey. *Agricultural Systems*, 187: 103021.
- Liu H, Xu X. 2022. IoT-based smart agriculture systems: A comprehensive survey. *Sensors*, 22: 3969.
- Liu Y, Wu S, Xu J. 2021. Smart livestock farming: Internet of Things (IoT) solutions for animal management. *Comput Electron Agric*, 180: 105911.
- Marques G, Lima P, Nunes M, Silveira G. 2022. Blockchain in agriculture: A systematic review of its applications in the supply chain. *Comput Electron Agric*, 190: 106398.
- Meng Q, Lee KY. 2021. IoT-based predictive maintenance

- system for dairy farming management: Case study of smart dairy farm. *Sensors*, 21: 245.
- Milani R, Bonaccorsi G, Lorenzetti C. 2021. Big data applications in livestock management: Challenges and opportunities. *Animals*, 11: 3075.
- Mourtzis D, Vlachos E. 2020. Big data analytics in manufacturing: A review. *Procedia CIRP*, 88: 114-119.
- Patle S, Srivastava S. 2022. IoT-based animal welfare system: A review and future perspectives. *Comput Ind*, 128: pp: 103475.
- Paton C, Sparks C. 2022. Livestock productivity and health data analytics for improving farm operations and animal welfare. *Front Vet Sci*, 9: 882.
- Peters K, Karamanis E. 2021. A review of machine learning algorithms in livestock farming. *Agricultural Systems*, 187: 103002.
- Pinto A, Santos J. 2022. Blockchain for data management in dairy farming: A review. *Sustainability*, 14: 1785.
- Sharma S, Sharma M. 2021. Artificial intelligence in agriculture: A survey of applications and challenges. *Comput Electron Agric*, 178: 105760.
- Sharma S, Sharma M. 2021. Artificial intelligence in dairy farming: A review of recent advancements and applications. *Artificial Intelligence Review*, 54: 3149-3166.
- Sharma S, Sharma M. 2021. Blockchain and IoT in sustainable livestock farming: A new paradigm for precision livestock farming. *Sustainability*, 12: 4265.
- Singh R, Kaur A. 2021. A review on the IoT-based smart farming systems. *Wireless Pers Commun*, 116: 3199-3214.
- Singh S, Rana R. 2020. A review on smart farming systems based on IoT technologies. *Int J Comput Appl*, 177(15): 32-37.
- Song D, Chen Y. 2022. Machine learning-based early detection of diseases in livestock: A review. *Comput Ind*, 130: 103499.
- Soni P, Nand A. 2021. A review of deep learning applications in precision agriculture. *Agricultural Systems*, 191: 103141.
- Stojanovic J, Mitić D. 2020. IoT-based smart farming: State-of-the-art and future trends. *Journal of Sensors*, 2020: 4513158.
- Wang J, Zhang H. 2021. A survey of applications of big data analytics in smart agriculture. *Comput Ind*, 129: 103465.
- Wang S, Zhao Y. 2021. Intelligent agriculture based on IoT and machine learning: A survey. *Comput Ind*, 128: 103459.
- Yadav A, Gupta R. 2021. A survey on IoT-based smart farming system for livestock health monitoring. *J Ambient Intell Humaniz Comput*, 12: 1959-1971.
- Yang J, Wei L. 2022. Artificial intelligence in the agricultural sector: Applications and future trends. *Sensors*, 22: 2894.
- Yang Y, Gao Z. 2021. IoT-based monitoring system for animal welfare in dairy farming. *Sensors*, 21: 5867.
- Yao Z, Li M. 2021. A review on the integration of blockchain and IoT for smart farming applications. *Future Gener Comput Syst*, 116: 176-189.
- Yao Z, Li M. 2021. Machine learning for agricultural applications: A review on data-driven methodologies in livestock farming. *Comput Ind*, 129: 103463.
- Yi W, Li Z. 2022. A blockchain-based traceability system for dairy products: A case study in smart farming. *Sustainability*, 14: 1635.
- Zhang Q, Lin Q. 2022. A review of the applications of AI and IoT technologies in smart farming: Current status and future trends. *Comput Mater Continua*, 69: 3577-3596.
- Zhang W, Yu X. 2020. IoT-based automated milking systems: A comprehensive review. *Comput Ind*, 121: 103250.
- Zhang X, Chen W. 2021. Data-driven models for monitoring animal welfare in livestock farming. *Comput Electron Agric*, 187: 106229.
- Zhang Y, Li Z. 2021. AI and IoT applications in agriculture: Current and future trends. *Comput Ind*, 132: 103530.
- Zhao Y, Wang X. 2022. Internet of Things in precision livestock farming: A comprehensive review. *Comput Ind*, 131: 103521.
- Zhou X, Zhang H. 2021. Smart agriculture and IoT-based solutions for animal health and welfare management. *Agricultural Systems*, 189: 103014.
- Zohra F, Kamble S. 2021. Blockchain and IoT integration in dairy farming: A review and future prospects. *Sustainability*, 13: 4532.