

Agriculture 4.0: Technological Adoption, Drivers, Benefits and Challenges in Italy. A Descriptive Survey

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Abstract: This study aims to examine the current state of awareness regarding Agriculture 4.0 (A4.0) among Italian agricultural enterprises and to analyse variations in adoption levels, expressed needs, perceived benefits, challenges and barriers to digitalisation. Drawing on data from a descriptive survey conducted among Italian farms in 2024, this study presents findings from 1,248 respondents. The results indicate varying levels of adoption of A4.0 solutions, with monitoring systems and connected vehicles being the most widely implemented. The primary drivers for A4.0 adoption include farm management, operational control, and the enhancement of production efficiency, all of which are associated with significant perceived benefits. However, challenges such as limited interoperability and skill shortages hinder A4.0 implementation, while financial and structural constraints remain major barriers for farms seeking to transition to A4.0. This study offers valuable insights to inform policymakers, industry stakeholders, and researchers in fostering a more effective and inclusive digital transformation in the Italian agricultural sector.

Keywords: Agriculture 4.0, Smart Farming, Digital Agriculture, Survey

JEL codes: M15, O32

1. Introduction

Agriculture 4.0, also known as “digital agriculture”, “smart farming” or “smart agriculture”, is defined as the integration of advanced digital technologies - such as the Internet of Things (IoT), robotics, Artificial Intelligence (AI), and Big Data analytics - into the agricultural sector (Fragomeli et al., 2024). This concept is grounded in the broader framework of Industry 4.0, which is responsible for the transformation of manufacturing processes (Da Silveira et al., 2021). Agriculture 4.0 (hereby A4.0)

represents a significant departure from both traditional and precision agriculture by leveraging automated, interconnected, and data-driven systems (Sharma et al., 2022).

The transition to digitalised agricultural systems is increasingly considered as pivotal for addressing the global challenges facing society today. Rapid population growth, urbanization, industrialization, loss of arable land, freshwater scarcity, and environmental degradation have escalated concerns regarding food security (Abbasi et al., 2022). To meet the rising global demand for food, agricultural practitioners must enhance productivity while minimising pressure on natural resources such as water, land, and energy (Sharma et al., 2022).

This highlights the urgent need for efficient, data-driven agricultural practices that optimise resource usage and improve productivity (Fragomeli et al., 2024). Moreover, agriculture is both a major contributor to greenhouse gas emissions and a sector vulnerable to the impacts of climate change (Sott et al., 2020). Integrating digital technologies offers the potential to mitigate the environmental footprint of agricultural practices while bolstering farmers' resilience to climate change (Balasundram et al., 2023).

Technologies such as robotics, smart irrigation and IoT sensors can promote more sustainable agricultural practices by reducing emissions, optimising resource use, and enabling real-time monitoring of crop conditions (Assimakopoulos et al., 2024). The environmental benefits of A4.0 are closely tied to economic advantages, as digital solutions improve operational productivity, reduce resource waste, and generate cost savings (Zul Azlan et al., 2024). Additionally, from a social perspective, the digitalisation of agriculture empowers farmers by providing better decision-making tools and improving working conditions (Zhai et al., 2020).

According to Papadopoulos et al. (2024), for instance, recording and mapping technologies, combined with guidance and controlled traffic farming technologies, could lead to reductions of up to 80% in fertiliser use. Furthermore, VRT (Variable Rate Technologies) could achieve a 60% decrease in fertiliser consumption and an 80% reduction in pesticide use, while also potentially boosting yields by 62%. Additionally, robotic systems and smart machines could reduce labour by 97% and diesel consumption by 50%.

Thus, A4.0 represents a transformative approach that addresses environmental, economic, and social challenges, contributing to the development of more sustainable and resilient agricultural systems (Maffezzoli et al., 2022b).

Despite the promising role that A4.0 solutions could play in mitigating sustainability challenges while improving productivity, their uptake remains limited and fragmented (Osrof et al., 2023). Literature relates the uneven adoption rate to different factors.

Recent empirical contributions confirm that farmers' intentions to adopt new solutions go beyond purely economic considerations and are shaped by a combination of personal attitudes and perceived obstacles. For instance, Giampietri et al. (2020) emphasize the role of farmers' trust, experience and knowledge in the adoption of risk management practices, highlighting the importance of transparency about costs and benefits in adoption incentivization. Menozzi et al. (2015) highlight the relevance of farmers' attitudes and perceived control in adopting sustainable farming practices, stressing the need for better communication and collaboration within the agricultural supply chain to increase A4.0 adoption.

Meanwhile, data from farm-level surveys show how age, gender, education and farm size remain significant influencing factors for choices regarding, for example, climate change adaptation strategies (Onyenekwe et al., 2023).

Despite the ongoing discussions in the literature regarding the factors that favour or hinder the spread of A4.0, the influence of specific contexts, as countries and types of farms and farmers, remains a

compelling area of investigation (Fragomeli et al., 2024; Da Silva et al., 2023). Therefore, the authors emphasise the need for a country-specific investigation on: i) farmers' awareness of A4.0; ii) the main challenges and barriers in the adoption as well as iii) the sustainability benefits perceived. We believe that building a comprehensive knowledge around the gap between A4.0 technologies, their promised technical advantages and the actual implementation along with the feasibility of realising the related sustainability benefits, is fundamental to inform key decision makers (e.g., policy makers). This knowledge can help in shaping proper strategies which place farmers and their context-specific needs at the centre.

To this end, a survey was conducted targeting Italian farms to assess the current level of digitalisation in the agricultural sector, with a specific focus on the key dimensions influencing the adoption and implementation of A4.0 solutions. The following research questions were formulated to explore the state-of-the-art of A4.0 in Italy:

- RQ1: What is the level of adoption and awareness of A4.0 solutions in Italy?
- RQ2: What are the primary factors driving agricultural enterprises to adopt A4.0 solutions?
- RQ3: To what extent have the achieved benefits aligned with the expressed needs?
- RQ4: What are the most significant hindering factors to farmers' adoption of A4.0 solutions?

This study reveals that, while A4.0 awareness is high, adoption is uneven, with greater uptake of A4.0 solutions such as monitoring systems and connected vehicles. Adoption is mainly driven by improvements in farm management rather than operational efficiency.

Benefits generally meet or exceed expectations, particularly in optimizing technical inputs and water use, which yield both economic and environmental gains. Social sustainability effects remain debated, with some evidence of labor market benefits, though concerns persist over potential job displacement.

Despite the benefits, adoption is hindered by challenges such as interoperability, lack of skills, uncertain return on investments and limited technical support. Financial and structural barriers - especially for small farms - and poor connectivity in remote areas further constrain A4.0 uptake. This study recommends targeted policy support, training, and agrifood supply chains stakeholder collaboration to overcome these barriers and accelerate digital transformation in Italian agriculture.

The remainder of the paper outlines as follows: the first section develops a review of the existing literature on main A4.0 solutions and applications along with the factors connected to their spread, section 2 presents a literature review covering the evolution of technologies in agriculture, the main driving technologies and their applications, challenges and benefits. Section 3 explains the methodology adopted, while results are described in section 4. Finally, sections 5 and 6 discuss the main results and draw conclusions from the authors' work.

2. Literature Review

2.1 The evolution of agricultural technologies

Over the years, agriculture has evolved through distinct technological phases, progressing from Agriculture 1.0 to Agriculture 4.0 (Zhai et al., 2020). Traditional agriculture, Agriculture 1.0, relied heavily on manual labour and animal power. The transition to Agriculture 2.0 began in the 19th century with the introduction of mechanised farming and steam engines, which significantly increased the efficiency of agricultural activities. This second phase was also characterised by an extensive use of

chemical fertilizers and pesticides, leading to environmental degradation and resource overexploitation. In the 20th century, Agriculture 3.0 emerged, leveraging advancements in computing and electronics to automate processes and enhance precision, also reducing dependency on chemicals. Today, A4.0 marks the era of smart farming, integrating digital technologies to create highly interconnected and data-driven agricultural systems (Fragomeli et al., 2024).

These innovations enable farmers to make real-time, data-informed decisions, improving productivity, sustainability, and resource efficiency while minimising environmental impact. Several terms are used to denote A4.0, such as “digital agriculture”, “smart farming” and “smart agriculture” (Albiero et al., 2020).

As outlined by Sponchioni et al. (2019) and Maffezzoli et al. (2022b), Agriculture 4.0 can be defined as the evolution of precision farming, realised through the automated collection, integration, and analysis of data from the field, equipment sensors, and other third-party sources. While precision farming serves as a management system that aims at optimising crop production inputs at the field level (Bongiovanni and Lowenberg-Deboer, 2004; Pierce and Nowak, 1999; Gebbers and Adamchuk, 2010), A4.0, facilitated by the smart and digital technologies inherent in Industry 4.0, transforms previously isolated data silos into actionable knowledge, supporting farmers in decision-making both within their enterprises and across the broader agrifood supply chain. This shift from traditional to digital systems ultimately aims to enhance cost efficiency and profitability, fostering the transition to more sustainable agricultural systems from an economic, environmental and social perspective.

Recent advancements in A4.0 are marked by emerging trends that are shaping the future of farming, with a particular focus on enhancing efficiency, sustainability, and resilience. A key forthcoming development is the transition toward Agriculture 5.0, which extends the foundations of A4.0 by incorporating human-centric, sustainable, and resilient principles derived from Industry 5.0 (Abbasi et al., 2022). This shift refines the collaboration between humans and machines, aiming to improve efficiency while reducing environmental impact through circular economy strategies (Fragomeli et al., 2024). Alongside this evolution, digital twin technology has gained prominence as a tool for optimising agricultural operations (Peladarinos et al., 2023; Escribà-Gelonch et al., 2024), creating real-time virtual replicas of farms that enable monitoring, predictive analytics, and improved system integration (Polymeni et al., 2023). By simulating real-world agricultural processes, digital twins can support decision-making in areas such as crop growth, soil composition, and climate adaptability (Peladarinos et al., 2023). At the same time, the increasing challenges posed by climate change have accelerated the adoption of climate-smart agricultural (CSA) practices, which focus on building resilience against environmental concerns, reducing greenhouse gas emissions, and ensuring long-term food security through adaptive resource management (Balasundram et al., 2023).

2.2 Key technologies and applications

There are various ways to categorize the key technologies driving A4.0, as different literature studies highlight several aspects of innovation in the field. Internet of Things (IoT) enables the connection of agricultural devices and machinery, allowing real-time monitoring and automation of farm operations (Assimakopoulos et al., 2024; Abbasi et al., 2022). Sensors and wireless sensor networks collect critical data on soil conditions, weather, and crop health, supporting precision farming (Ahmed et al., 2024). Artificial Intelligence (AI) and Machine Learning process large datasets to optimise resource use, detect plant diseases, and predict yields, making farming more data-driven and efficient

(Ahmed et al., 2024; Balyan et al., 2024). AI-driven systems are increasingly capable of autonomous decision-making, on-farm reinforcement learning, and real-time adaptation, significantly transforming how decisions are made at the farm level (Khanna et al., 2024). Robotics and automation include autonomous machines and drones that assist in tasks such as planting, harvesting, and spraying, reducing labour dependency and increasing precision (Ahmed et al., 2024; Balyan et al., 2024). Data analytics and Big Data play a crucial role in processing vast amounts of information collected from farms, offering insights for better decision-making (Abbasi et al., 2022). Cloud and edge computing ensure that agricultural data is processed efficiently and securely, reducing latency and enabling real-time responses in smart farming systems (Abbasi et al., 2022). Blockchain technology enhances transparency and traceability in the agricultural supply chain by securely recording transactions and ensuring data integrity (Ahmed et al., 2024).

While the technologies discussed above form the foundations of A4.0, they are not typically deployed in isolation. Instead, they are combined into integrated digital solutions, translating technological capabilities into practical tools for farming and therefore addressing different agricultural needs. Such integrated solutions include Decision Support Systems (DSS) (Araujo et al., 2021), monitoring systems (Dayioğlu and Turker, 2021), mapping solutions (Karunathilake et al., 2023), Variable Rate Technologies (VRT) (Dayioğlu and Turker, 2021), connected vehicles (Karunathilake et al., 2023), telemetry systems (Papadopoulos et al., 2024), robotics and drones (Araujo et al., 2021). These solutions are further described in the methodology section, where their identification, based on a review of scientific and grey literature, forms a key step of the survey design. Investigating adoption at the solution level, rather than at the level of individual technologies, better reflects how farmers actually implement digital tools in practice.

As with key technologies and solutions, the applications of A4.0 have been categorised in different ways, reflecting the broad range of domains in which digital technologies can support agricultural practices. Water and irrigation management involves smart irrigation systems, IoT-based sensors, and climate monitoring tools to optimise water use, ensuring efficient irrigation and drought adaptation (Ahmed et al., 2024; Javaid et al., 2022). Soil and crop health monitoring utilizes remote sensing, drones, and AI-driven diagnostics to assess soil fertility, detect diseases, and manage agrochemical and fertilizer use with precision (Yousaf et al., 2023). Predictive analytics for climate adaptation and yield forecasting apply Machine Learning and Big Data analytics to anticipate weather patterns, pest outbreaks, and crop productivity, helping farmers make data-driven decisions to mitigate risks (Kumar Kasera et al., 2024). Autonomous machinery and robotics enhance efficiency by using automated tractors, drones, and harvesting robots to perform tasks such as soil preparation, planting, and harvesting with minimal human intervention (Oliveira et al., 2021). Controlled-environment agriculture includes greenhouse cultivation, hydroponics, and aquaponics, which optimise growing conditions and reduce dependency on natural weather cycles, ensuring year-round food production (Maffezzoli et al., 2022b). Livestock monitoring and regulation employs wearable sensors, automated feeding systems, and AI-based health tracking to improve animal welfare, optimise breeding, and prevent diseases (Ahmed et al., 2024). Finally, supply chain optimisation focuses on product tracking, storage management, and food processing, incorporating blockchain and automation to enhance traceability, reduce waste, and streamline logistics from farm to consumer (Kumar Kasera et al., 2024).

To summarise, these technological solutions, applied in a diverse range of domains, can result in a set of improvements for farmers. Such improvements, later investigated through a survey, encompass different dimensions. A4.0 solutions can support farmers with improved forecasting capabilities and improved farm management and control; support planning and scheduling activities, while also facilitating and streamlining workforce processes; optimise the use of technical inputs (water, pesticides, fertilizers), enhance efficiency and reduce losses due to pests and diseases. Finally,

188 through monitoring and measurement, they enable increased awareness on farm operations and
189 improve the quality of agricultural products.

190 All these enhancements can lead to substantial economic, environmental and social benefits.

191 2.3 Sustainability benefits

192 A4.0 yields significant economic, social, and environmental benefits, thereby fostering a profound
193 transformation of the agricultural sector. Economically, it enhances resource use efficiency by
194 optimising the application of water, fertilizers, and pesticides, reducing waste, and increasing
195 agricultural yields. This leads to greater profitability for farmers and more cost-effective farming
196 practices (Zul Azlan et al., 2023; Abbasi et al., 2022). Additionally, the automation and digitalisation
197 of farm operations, such as harvesting, sowing, and irrigation, result in time and cost savings,
198 improving operational efficiency (Pradel et al., 2022). The introduction of predictive models and real-
199 time data analysis can help farmers forecast adverse conditions like disease outbreaks or extreme
200 weather, thereby improving the resilience of agricultural systems and ensuring production even in
201 challenging circumstances (Zul Azlan et al., 2023). From an environmental standpoint, smart farming
202 practices significantly reduce agriculture's ecological footprint. Precision agriculture technologies,
203 AI-driven crop monitoring, and automated machinery facilitate the efficient use of resources, leading
204 to reduced fuel consumption, lower greenhouse gas emissions, and improved water conservation
205 (Cambra Baseca et al., 2019). Moreover, the deployment of technologies such as drones and IoT-
206 based environmental monitoring systems supports soil health management, optimises nutrient use
207 efficiency, and strengthens climate resilience (Abbasi et al., 2022). By minimising waste and
208 promoting environmentally responsible practices, A4.0 emerges as a key driver of sustainable
209 agricultural development (Zul Azlan et al., 2023).

210 From a social perspective, A4.0 plays a crucial role in enhancing the well-being of rural communities,
211 agricultural workers, and consumers. By promoting more efficient and sustainable farming practices,
212 A4.0 strengthens food security, mitigates food shortages, and reduces waste (Jin et al., 2020).
213 Furthermore, the integration of advanced technologies equips farmers with improved decision-
214 making tools, contributing to higher living standards by lowering labour costs and enhancing working
215 conditions (Da Silveira et al., 2021). Additionally, A4.0 enhances product quality and traceability,
216 ensuring food safety and addressing consumer concerns (Zul Azlan et al., 2023). The integration of
217 advanced digital monitoring technologies can in fact support the verification of environmental and
218 social standards along the food supply chain (Meemken et al., 2024). These systems not only
219 strengthen sustainability management but also offer new avenues for accountability and trust in food
220 systems. However, they further raise important questions about equity and data access, which merit
221 further attention as digital monitoring expands (Meemken et al., 2024). Despite such promising social
222 benefits, scholars have also drawn attention to the danger of overly optimistic narratives that see these
223 innovations as universal solutions. Klerkx et al. (2020) emphasize the need to account for the social
224 and ethical implications of A4.0 transitions, particularly in terms of labor displacement, rural
225 depopulation, power concentration, and the marginalization of alternative, potentially more
226 accessible technologies.

227 In fact, while A4.0 promises numerous benefits, its impacts are not unilaterally positive. Muhl et al.
228 (2022) stress how digital agriculture may reinforce existing inequalities and that social issues like
229 food insecurity, often driven by broader social injustices, will not be solved by technological
230 development alone. The sustainability debate thus calls for an inclusive and responsible approach to
231 the use and development of these technologies, ensuring accessibility across different contexts (Muhl
232 et al., 2022).

The adoption of A4.0 technologies is hindered by a range of significant challenges and barriers, as highlighted by (Assimakopoulos et al., 2024, Da Silva et al., 2021; Da Silva et al., 2023; Fragomeli et al., 2024). An interesting classification of challenges is provided by Lezoche et al. (2020), where a distinction is made between organizational, social and technological challenges. Among organization challenges, one of the most frequently associated with A4.0 adoption is the high cost connected to the technology adoption, including the initial investment required for the implementation of the components, the ongoing maintenance costs, and the cost of skilled labour (Da Silva et al., 2023). These financial challenges are particularly burdensome for small-scale farms, which often lack the necessary capital or access to financing options to invest in such innovations (Assimakopoulos et al., 2024). Additionally, from a more social perspective, the complexity of modern agricultural technologies and the advanced skills required for their operation present further obstacles (Fragomeli et al., 2024). These barriers are not unique to the Italian context; similar challenges have been widely observed in other regions, particularly among smallholder farmers. For instance, Mhlanga et al. (2023) highlight the digital transformation obstacles in African agriculture, where factors such as limited infrastructure, insufficient digital literacy, lack of funding mechanisms, and farmer resistance significantly constrain adoption. In general, farmers with limited technological proficiency - especially older individuals or those with lower levels of formal education - may struggle to integrate digital tools into their daily operations (Assimakopoulos et al., 2024). It can be stated that, beyond costs, adoption is shaped by a complex interaction of operator characteristics (such as age, education, and digital skills), farm-level attributes (including size, income, and specialization), and the perceived attributes of the technologies themselves - such as their trialability, ease of integration, and perceived utility (Khanna et al., 2024).

From an organizational perspective, uncertain regulatory aspects and complex legal frameworks often hinder adoption (Lezoche et al., 2020), highlighting the role of manufacturers and governmental bodies as critical in mitigating these challenges.

Looking at technological challenges, a further barrier is often recognized in inadequate infrastructures, particularly in rural areas, where poor internet connectivity and restricted access to technical support networks hinder the full utilization of digital technologies (Da Silva et al., 2023; Fragomeli et al., 2024). Moreover, farmers already managing extensive daily responsibilities may perceive these new technologies as overly time-consuming or complex, together with concerns about lack of interoperability and issues about data security and privacy (Lezoche et al., 2020). Moreover, many farmers report a lack of accessible training programs, technical guidance, and support services, which prevents them from fully understanding and implementing digital tools (Da Silva et al., 2023).

These financial, educational, infrastructural, and institutional barriers underscore the multifaceted challenges associated with adopting A4.0 technologies. Addressing these issues through targeted policies, improved infrastructure, and comprehensive training initiatives is essential for promoting widespread and equitable adoption of digital farming solutions.

3. Research Methodology

The primary objective of this research is to assess the current state of digitalisation within the Italian agricultural sector, with a specific focus on different key dimensions that shape the adoption and

277 implementation of A4.0 technologies. To evaluate the state-of-the-art of A4.0 in Italy, the following
278 research questions have been formulated:

- 279 - RQ1: What is the level of adoption and awareness of A4.0 solutions in Italy?
- 280 - RQ2: What are the primary factors driving agricultural enterprises to adopt A4.0 solutions?
- 281 - RQ3: To what extent have the achieved benefits aligned with the expressed needs?
- 282 - RQ4: What are the most significant hindering factors to farmers' adoption of A4.0 solutions?

283 To address these research questions, the study examines the following dimensions:

284 *Adoption and awareness of A4.0 solutions:* assessing the extent to which identified A4.0 solutions have
285 been implemented across the sector and the level of awareness that Italian farms have regarding these
286 technologies.

287 *Drivers of digitalisation:* identifying the factors motivating farms to explore and implement the proposed
288 A4.0 solutions, highlighting key needs and expectations.

289 *Benefits achieved:* evaluating the advantages achieved through the adoption of A4.0 solutions with
290 regards to the specific needs expressed.

291 *Challenges encountered by farmers adopting A4.0 technologies:* examining obstacles that farms
292 encountered during the adoption and implementation process of A4.0 solutions.

293 *Inhibiting factors for non-adopting farmers:* investigating the underlying reasons for the hesitation or
294 inability of non-user farmers to adopt A4.0 solutions.

295 The last two categories are drawn from the literature on "challenges and barriers", which typically does
296 not distinguish between adopters and non-adopters. However, based on the authors' experience and
297 discussions with farmers and technology providers, this distinction was deemed necessary to better
298 reflect the obstacles faced by Italian agricultural enterprises in uptaking and using A4.0 solutions.

299 To address these objectives systematically, the research followed a structured methodology comprising
300 the following steps:

301 *Sample development.* The research referenced data from the 7th General Census of Agriculture of the
302 Italian National Institute of Statistics (ISTAT)¹ to identify a representative sample of Italian agricultural
303 enterprises. The sampling framework accounted for critical variables, including farm size, production
304 type, and geographic distribution, ensuring a diverse and comprehensive representation of the Italian
305 agricultural sector. The sample was drawn from three perspectives: (1) geographic distribution: Italian
306 farms were grouped in four main regions to capture macro-regional variations in farms geographical
307 distribution (Table 1). (2) Primary crop production: Italian farms have been classified based on their
308 primary agricultural products, determined by the proportion of Utilised Agricultural Area (UAA)
309 allocated to specific cultivations (Table 2). (3) Farm size: Italian farms have been categorised according
310 to their UAA size, allowing for an analysis of adoption patterns by operational scale (Table 3). A
311 proportionate stratified random sampling approach was employed, whereby the total population, as
312 defined by ISTAT, was divided into mutually exclusive strata. Each stratum was sampled in proportion
313 to its representation within the overall population. Within each stratum, participants were selected using
314 a random sampling method.

315

316 *Identification of a set of A4.0 solutions.* A tailored set of A4.0 solutions was developed in alignment with
317 the operational characteristics of the agricultural sector based on an analysis of scientific and grey
318 literature on this topic (Araújo et al., 2021; Dayioğlu and Turker, 2021; Karunathilake et al., 2023;
319 Papadopoulos et al., 2024). This set comprises the following A4.0 solutions:

¹ <https://www.istat.it/statistiche-per-temi/censimenti/agricoltura/7-censimento-generale/>

320 *DSS – Decision Support Systems*, that assist farmers in decision-making by optimising management and
321 agronomic choices based on field data, environmental, weather and soil data, and information provided
322 by the farmer. These systems can directly provide both management and agronomic advice to the users.

323 *Monitoring systems*, enabling the monitoring, often remotely and automatically, of environmental
324 conditions or other parameters related to crops.

325 *Mapping solutions*, allowing the mapping of soil and crops, providing spatial variability in soil, crop,
326 and hydrological characteristics, among others. These spatialised datasets can be used for various
327 purposes such as variable rate input applications, agronomic decision-making support, and operational
328 management.

329 *Variable Rate Technology (VRT) solutions* that enable field operations and the distribution of inputs
330 based on the spatial variability detected in the field and the needs of the soil and crop systems.

331 *Connected vehicles*, i.e. digitally connected machinery that is equipped with integrated digital
332 technologies, such as assisted driving, precision navigation systems, and auto-steering systems.

333 *Telemetry systems and solutions for vehicle and equipment monitoring*, that can locate, monitor and
334 provide assisted control of agricultural machinery, including auto-steering systems and telematic
335 solutions for fleet monitoring, predictive maintenance, and machinery efficiency improvement.

336 *Robotics*, i.e. solutions involving autonomous machinery capable of movement, decision-making, and
337 performing specific tasks and crop operations with little or no operator intervention.

338 *Drones*, i.e. solutions and services involving the use of drones for mapping crops and land through
339 cameras and sensors, monitoring crop health, and applying products or biological control agents.

340 For the purpose of this research, *Farm Management Information Systems (FMIS)* have been excluded
341 from the analysis, as they are classified as enabling technologies rather than core components of the
342 A4.0 paradigm. As Industry 4.0 evolves and digital technologies continue to expand and mature into
343 practical solutions for farmers, it becomes crucial to distinguish between core components of the
344 paradigm and enabling technologies. While enabling technologies play a vital role in supporting A4.0,
345 they are considered complementary rather than fundamental elements of the paradigm itself.

346 *Survey design and implementation*. An online survey was developed and distributed targeting farms
347 identified through the sampling framework. The online format was chosen for its cost-efficiency, ease
348 of administration, and ability to minimise errors associated with manual data collection, as also reported
349 by van Selm and Jankowski (2006) and Maffezzoli et al. (2022a).

350 This survey consisted of seven sections:

- 351 1. *General information*, collecting foundational and demographic details about the respondent and
352 their agricultural enterprise.
- 353 2. *A4.0 awareness and implementation*, assessing the level of familiarity and the extent of adoption
354 of the proposed set of A4.0 solutions.
- 355 3. *Needs, benefits, and challenges*, exploring the specific needs driving the adoption of A4.0
356 solutions, the benefits achieved, and the challenges encountered during their implementation.
- 357 4. *Data management capabilities*, evaluating the farms' ability to collect, store, analyse, and utilize
358 data effectively to inform decision-making processes.
- 359 5. *Digital skills*, assessing the competences and level of expertise of farm operators in relation to
360 A4.0 solutions.
- 361 6. *Investments*, reviewing past investments, current expenditures, and anticipated future
362 investments in A4.0 solutions.
- 363 7. *Inhibiting factors*, identifying the barriers and constraints preventing or limiting the adoption of
364 A4.0 solutions.

The full set of questions included in each section of the survey is provided in Appendix A, located at the end of this manuscript.

Data collection. Data collection was conducted from September 2024 to December 2024. The process yielded a total of 1,248 valid responses, providing a robust dataset for detailed analysis. Tables 1, 2 and 3 report the sample of responses collected according to the critical sampling variables and table 4 provides a summary of the main descriptive statistics on collected data.

Table 1. Total population and sample size and their distribution across Italian regions (number of farms)

	Pop. size distr.	Pop. size distr. (%)	Sample size distr.	Sample size distr. (%)
North-west	113,972	10%	304	24%
North-east	187,429	16%	319	26%
Centre	179,230	16%	328	26%
South and Islands	652,392	58%	297	24%
Total	1,133,023	100%	1,248	100%

Table 2. Total population and sample size and their distribution across primary crop productions (UAA - Utilised Agricultural Area)

	Pop. size distr.	Pop. size distr. (%)	Sample size distr.	Sample size distr. (%)
Cereal crops	3,054,288	34%	31,923	19%
Vineyards	742,926	8%	92,693	56%
Fruit crops	444,805	5%	7,310	4%
Fodder crops	2,564,217	28%	3,469	2%
Olive cultivation	1,114,593	12%	4,723	3%
Vegetable crops	445,966	5%	4,490	3%
Legumes	85,007	1%	132	0.1%
Citrus fruits	149,863	2%	21,353	13%
Industrial plants	477,091	5%	562	0.3%
Total	9,078,756	100%	166,655	100%

Table 3. Total population and sample size and their distribution across farms' size (number of farms)

	Pop. size distr.	Pop. size distr. (%)	Sample size distr.	Sample size distr. (%)
0 hectares	12,499	1%	23	1%
Up to 0.99 hectares	228,481	20%	19	2%
From 1 to 1.99 hectares	209,662	18%	61	5%
From 2 to 2.99 hectares	128,381	11%	55	4%
From 3 to 4.99 hectares	147,320	13%	123	10%

From 5 to 9.99 hectares	160,133	14%	209	17%
From 10 to 19.99 hectares	109,545	10%	262	21%
From 20 to 29.99 hectares	45,118	4%	104	8%
From 30 to 49.99 hectares	41,167	4%	109	9%
From 50 to 99.99 hectares	32,487	3%	120	10%
From 100 onwards	18,230	2%	163	13%
Total	1,133,023	100%	1,248	100%

378 The tables presented above highlight a discrepancy between the sample distribution and that of the
379 overall population, resulting in an overrepresentation of farms located in Northern Italy and an
380 underrepresentation of those in the South and Islands. This imbalance may introduce a geographical
381 bias into the analysis. Moreover, the average Utilised Agricultural Area (UAA) of the sampled farms,
382 amounting to 22 hectares, is notably higher than the national average of 11.1 hectares reported by
383 ISTAT², indicating a sample skewed toward more structured and capital-intensive farming
384 operations. The sample also includes a disproportionately large share of vineyard farms, a sector
385 typically associated with higher profitability and investment capacity, which may further influence
386 the study's results.

387 However, these deviations do not necessarily compromise the validity of the findings. The research
388 primarily aims to investigate the adoption and perceived benefits of A4.0 solutions, an area where
389 more structured and technologically advanced farms are expected to play a pioneering role (Giua,
390 2022). Consequently, focusing on more innovative and capitalised enterprises allows for a more
391 detailed understanding of current trends, challenges, and potential impacts, which can serve as a
392 reference point for the broader agricultural sector as it transitions toward digitalisation.

393 Prior to presenting the results of the survey data analysis, the authors provide a table outlining the key
394 descriptive statistics of the collected dataset.

395

396 Table 4. Summary of main descriptive statistics of survey collected data

	Unit	Mean	Median	Std	Min	Max
Farm size	Ha	22.21	38.50	1,718.2 1	0	40,000
Farm annual turnover	EUR					
Less than €50,000	share	0.35				
Between €50,000 and €250,000	share	0.38				
Between €250,000 and	share	0.12				

² https://www.istat.it/it/files/2022/06/censimento_agricoltura_gismondi.pdf

€500,000						
Between €500,000 and €1,000,000	share	0.06				
Over €1,000,000	share	0.09				
Employees in farm	no.	3.69	11.75	6.60	0	950
A4.0 solutions adopted in farm	no.	2.68	4.00	1.65	0	8
Total amount spent on A4.0 solutions by farm	EUR					
Less than €5,000	share	0.23				
Between €5,000 and €15,000	share	0.17				
Between €15,000 and €30,000	share	0.13				
Between €30,000 and €50,000	share	0.09				
Between €50,000 and €75,000	share	0.08				
Between €75,000 and €100,000	share	0.07				
More than €100,000	share	0.23				

397

398 The descriptive statistics in the table above highlight that farm size distribution is skewed. This
399 asymmetry is commonly observed across many countries, as both very large and very small farms
400 coexist, often with significantly different spending capacities, as also noted by the OECD (Bokusheva
401 and Kimura, 2016).

402

403 4. Results

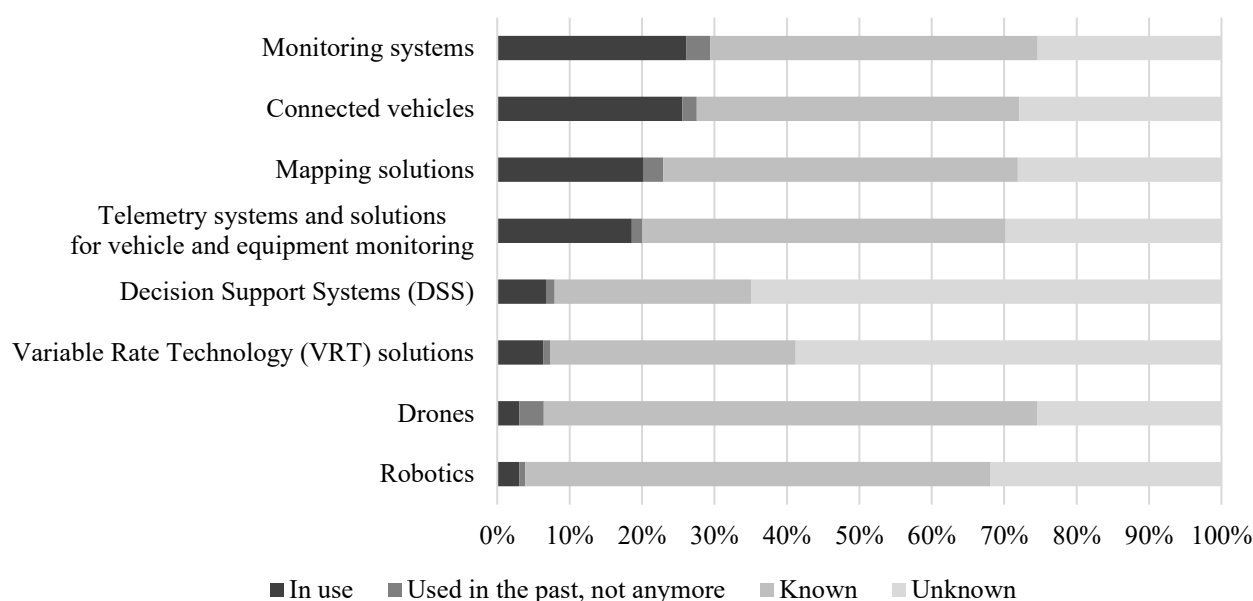
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405 4.1 A4.0 awareness and adoption level

406 The initial findings of this analysis focus on the current levels of adoption of A4.0 solutions among
407 survey respondents. A summary of these results is presented in Figure 1. To assess awareness of A4.0
408 solutions, a four-point scale was employed, ranging from low to high familiarity, following the approach
409 outlined by Maffezzoli et al (2022a). This scale effectively distinguishes varying levels of awareness
410 and facilitates cross-tabulation, allowing for the identification of patterns across different respondent
411 groups. The four levels of awareness are defined as follows: (a) *Unknown*, representing a complete lack
412 of familiarity, indicating no awareness of the existence of the proposed solution; (b) *Known*, denoting
413 limited familiarity and implying that the respondent has heard of the solution, but possesses only a
414 superficial understanding; (c) *Used in the past, not anymore*, indicating theoretical familiarity and
415 suggesting that the respondent has a solid understanding of the solution despite no longer using it; and

416 (d) *In use*, representing practical familiarity, meaning the respondent not only knows about the solution,
 417 but also employs it.

418 Figure 1. Agriculture 4.0 awareness level. Sample: 1,248 respondents



419

420

421 The data reveal varying levels of adoption and awareness of A4.0 solutions. Key findings show that
 422 approximately 26% of respondents currently implement *monitoring systems* and *connected vehicles*,
 423 making these among the most widely adopted A4.0 solutions. Meanwhile, 20% of respondents adopted
 424 *mapping solutions* and 19% employed *telemetry systems and solutions for vehicle and equipment*
 425 *monitoring*.

426 Adoption rates for *Decision Support Systems (DSS)* and *Variable Rate Technology (VRT) solutions* are
 427 notably lower, at 7% and 6% respectively. *Robotics* and *drones* show the lowest adoption rate,
 428 standing at only 3%, likely due to constraints related to cost, technical expertise, or perceived necessity.

429 Disaggregated data by farm size reveal that only 23% of farms with less than 10 hectares of UAA
 430 have adopted at least one A4.0 solution. Similarly, among farms with annual revenues below EUR
 431 50,000, the adoption rate stands at 21%. However, adoption increases substantially with scale: 66%
 432 of farms with a UAA between 100 and 199.9 hectares have adopted A4.0 technologies, and this figure
 433 rises to 82% for farms exceeding 200 hectares. A comparable trend is evident with respect to
 434 economic size, where adoption reaches 74% among farms with annual revenues between EUR
 435 500,000 and EUR 1,000,000, and rises further to 81% for those with revenues above EUR 1,000,000.

436 In contrast, our findings do not reveal substantial differences in A4.0 adoption based on the age or
 437 education level of farm managers. The only exception is among managers over the age of 65, who
 438 show a lower adoption rate (30%). Similarly, post-graduate degree holders are the only educational
 439 group exhibiting higher-than-average adoption rates (48%).

440 With regard to agricultural production types, enterprises primarily engaged in cereal cultivation report
 441 higher adoption rates of A4.0 solutions (49%), alongside vineyard and fodder farms (both at 43%).
 442 The relatively higher A4.0 adoption among cereal and fodder producers can be attributed to the
 443 extensive nature of these cropping systems, which can be particularly well-suited to the application
 444 of A4.0 solutions in optimizing operations over wide areas. Conversely, vineyard enterprises,

typically characterized by higher revenue margins, tend to possess greater financial capacity to invest in innovation, thereby facilitating the uptake of digital solutions and innovative technologies.

4.2 Needs expressed and benefits perceived from A4.0 implementation

To comprehensively analyse the key drivers that motivated respondents to adopt and implement A4.0 solutions, this study focuses on the specific needs expressed by farmers. These needs reflect both strategic and operational priorities, ranging from farm management and control to the optimisation of input consumption.

Figure 2 reveals a substantial level of awareness among respondents regarding the broad and multifaceted nature of the A4.0 paradigm. Rather than being perceived merely as an extension of precision agriculture - whose primary goal is to deploy technological solutions in the field to optimise input consumption and reduce costs - A4.0 appears to be increasingly recognised as a comprehensive framework for enhancing overall farm management and control, with positive effects along the overall agrifood supply-chain. This paradigm shift suggests that farmers view A4.0 not only to refine specific agricultural practices, but also as an integral component in fostering a more efficient and data-driven agricultural enterprise.

Among the ten most frequently expressed needs related to farm management and control, the most prominent include enhancing forecasting capabilities (41%), particularly in areas such as disease outbreaks, crop requirements, plant growth and yield projections, improving control and management processes within the farm enterprise (38%) with a focus on better decision-making and operational efficiency, optimising the planning and scheduling of agricultural activities (32%) and increasing awareness of ongoing farm activities and operations (31%). Similarly, in relation to the optimisation of input consumption, respondents identified key areas where A4.0 solutions could bring significant improvements, including optimising the use of technical inputs such as fertilisers and agrochemicals (28%) and enhancing the efficiency of machinery and equipment utilisation (26%), contributing to both cost reductions and operational sustainability. Furthermore, respondents expressed the need to streamline and optimise workforce processes (26%), ensuring that operators can perform tasks with efficiency and effectiveness, and to maximize water-use efficiency (24%), which is particularly critical in the context of recent meteorological events in Italy: in 2024, the country experienced heavy rainfall in the northern regions, while facing severe droughts in the south³.

Furthermore, respondents reported adopting A4.0 solutions for additional objectives, including reducing losses due to diseases, pests, and infestations (35%), a critical aspect of maintaining both yield stability and crop health, and improving the quality of the final agricultural product (20%) to meet regulatory requirements.

Figure 2. Needs expressed by respondents. Sample: 511 respondents who adopted at least one of the proposed Agriculture 4.0 solutions. Respondents could choose a maximum of 5 options.

³ Agro-meteorological Monitoring INDices (AgroMIND) map on Agricultural Drought (SPEI6) (<https://wonderful-bush-09061f403.5.azurestaticapps.net/AgroMIND.html>)

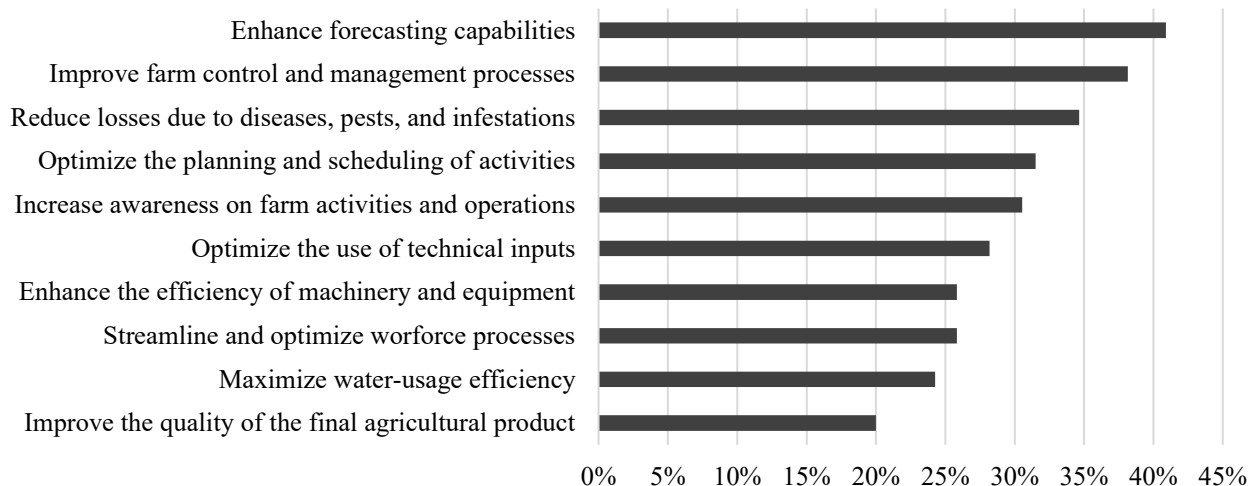
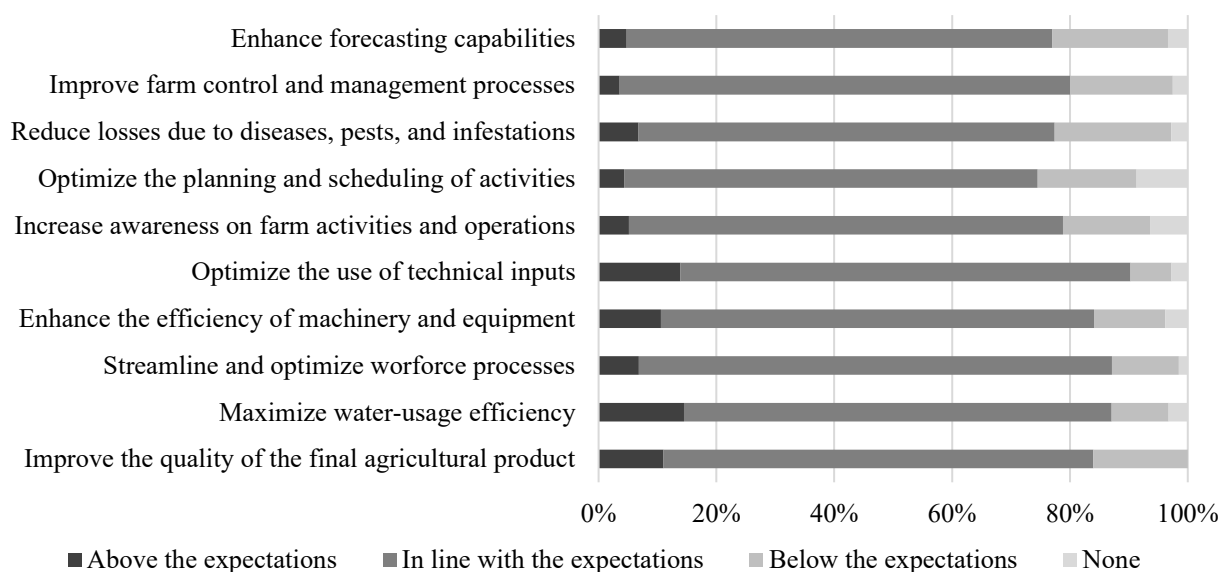


Figure 3 illustrates the perceived benefits derived from the adoption of A4.0 solutions, as evaluated in relation to the specific needs previously expressed by respondents. The findings indicate that, on average, the implementation of A4.0 technologies resulted in outcomes that aligned with initial expectations for most adopters (74% on average). Additionally, a subset of respondents (8% on average) reported that the benefits they experienced exceeded their initial expectations.

These results suggest that most farmers who invested in A4.0 solutions perceived their adoption as a successful means of addressing their agricultural needs, with reported benefits generally meeting anticipated outcomes. However, a smaller proportion of respondents indicated that the benefits they obtained were either below their expectations (14% on average) or entirely absent (4% on average), highlighting potential limitations in implementation effectiveness, technology adoption challenges, or contextual constraints that may have hindered the full realisation of expected advantages.

Furthermore, the analysis reveals that the perceived benefits were more pronounced in activities related to the optimisation of input consumption compared to those associated with farm management and control. Specifically, an average of 11% of respondents reported experiencing benefits that exceeded their expectations in the domain of input consumption optimisation. In contrast, only an average of 4% of respondents indicated that benefits surpassed expectations for farm management and control activities. This suggests that A4.0 solutions may be particularly effective in enhancing input efficiency, resource utilisation, and operational streamlining, whereas their impact on broader management and control functions may be more variable or dependent on additional contextual factors.

504 Figure 3. Benefits perceived by respondents. Sample: 511 respondents who adopted at least one of the proposed Agriculture 4.0 solutions



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506

507 Moreover, Italian farmers who have already adopted A4.0 solutions exhibit a significantly higher
 508 propensity to invest further in these technologies compared to non-adopters. Specifically, 20% of current
 509 users reported their intention to invest more than EUR 50,000 in A4.0 technologies within the next year,
 510 whereas only 3% of non-users indicated an equivalent investment plan. Furthermore, 27% of adopters
 511 expected to allocate between EUR 5,000 and EUR 30,000, compared to just 18% among non-adopters.
 512 Notably, 55% of non-users were unable to estimate their future investments, in contrast to only 26% of
 513 current users. These findings suggest that A4.0 adopters, having already perceived benefits (often
 514 exceeding expectations) are more inclined to pursue further technological advancement and exhibit a
 515 clearer strategic orientation toward digital transformation.

516

517 4.3 A4.0 implementation challenges and factors inhibiting A4.0 adoption

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519 This study also aims to assess the challenges encountered by respondents who have adopted at least one
 520 of the proposed A4.0 solutions, as well as the barriers faced by those who either could not or chose not
 521 to adopt any of these solutions.

522 Figure 4 presents the challenges encountered by farms that have implemented A4.0 solutions. The
 523 findings indicate that one of the most significant issues - reported by 36% of respondents - is limited or
 524 non-existent interoperability among the adopted solutions. Many farmers, indeed, struggle with
 525 integrating different digital tools within their existing farm management systems, leading to
 526 inefficiencies and operational difficulties (Khanna et al., 2024).

527 Following interoperability concerns, other notable challenges include the lack of appropriate skills to
 528 effectively utilise A4.0 solutions (30%) and the perceived inadequacy of return on investment (26%),
 529 suggesting that respondents may not see immediate or sufficient financial benefits from their A4.0
 530 investments, potentially discouraging further technological adoption. Furthermore, 26% of respondents
 531 indicate insufficient or unreliable technical assistance, which further limits A4.0 effectiveness together
 532 with operational challenges (20%) and inadequate connectivity (16%).

533 Interestingly, only 6% of respondents reported that they did not face any challenges during their A4.0
 534 implementation. This finding suggests that most adopters have encountered at least some difficulties in
 535 integrating and implementing A4.0 solutions, emphasising the need for targeted interventions to enhance

system compatibility, improve user experience and provide better support mechanisms for farmers transitioning to digital technologies.

Figure 4. Challenges encountered by respondents. Sample: 511 respondents who adopted at least one of the proposed Agriculture 4.0 solutions. Respondents could choose more than one option.

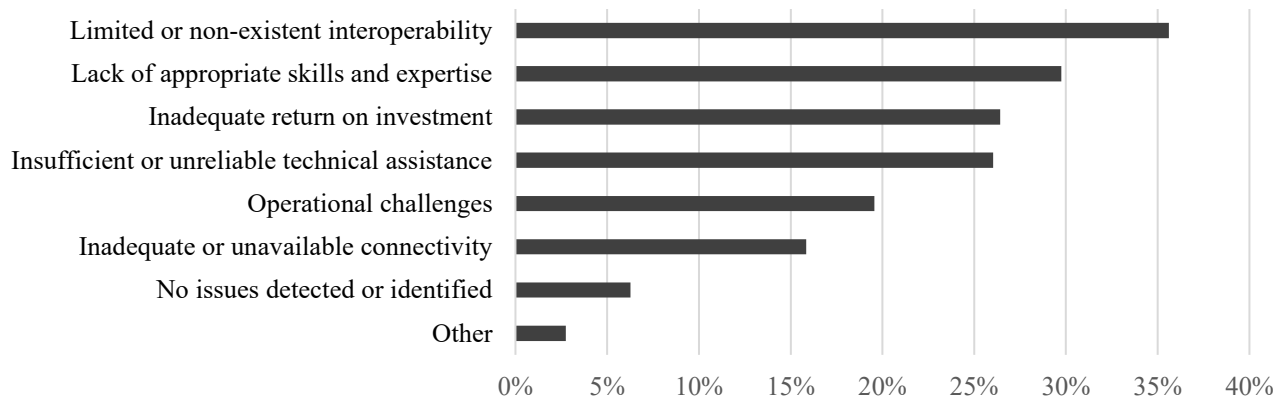


Figure 5 illustrates the key barriers that have prevented farms from adopting A4.0 technologies. One of the most frequently cited limitations is farm size, with 68% of respondents indicating that their farms are too small to justify investment in A4.0 solutions. This is not surprising, as Table 3 shows that in the Italian context, most farms (77%) are small or medium-sized.

Further constraints concern the possible exploitation of A4.0 solutions, with 59% of respondents believing that they would not fully exploit these solutions and 50% stating that their current agricultural technologies and management practices adequately meet their business needs, thereby reducing the perceived necessity of implementing A4.0 solutions.

Financial concerns also play a significant role, as 45% of respondents believe that the anticipated benefits do not justify the required investment, while 41% struggle to see the potential economic advantages of incorporating digital tools into their operations. Additionally, financial constraints further limit adoption, with 38% of respondents citing their inability to spread investment costs over time and 36% highlighting the difficulty of sharing these costs across multiple enterprises. Bureaucratic challenges also emerge as a deterrent, as 36% of respondents report difficulties in accessing financial incentives due to stringent requirements and administrative burdens, while 22% point to restricted access to credit lines as a further impediment.

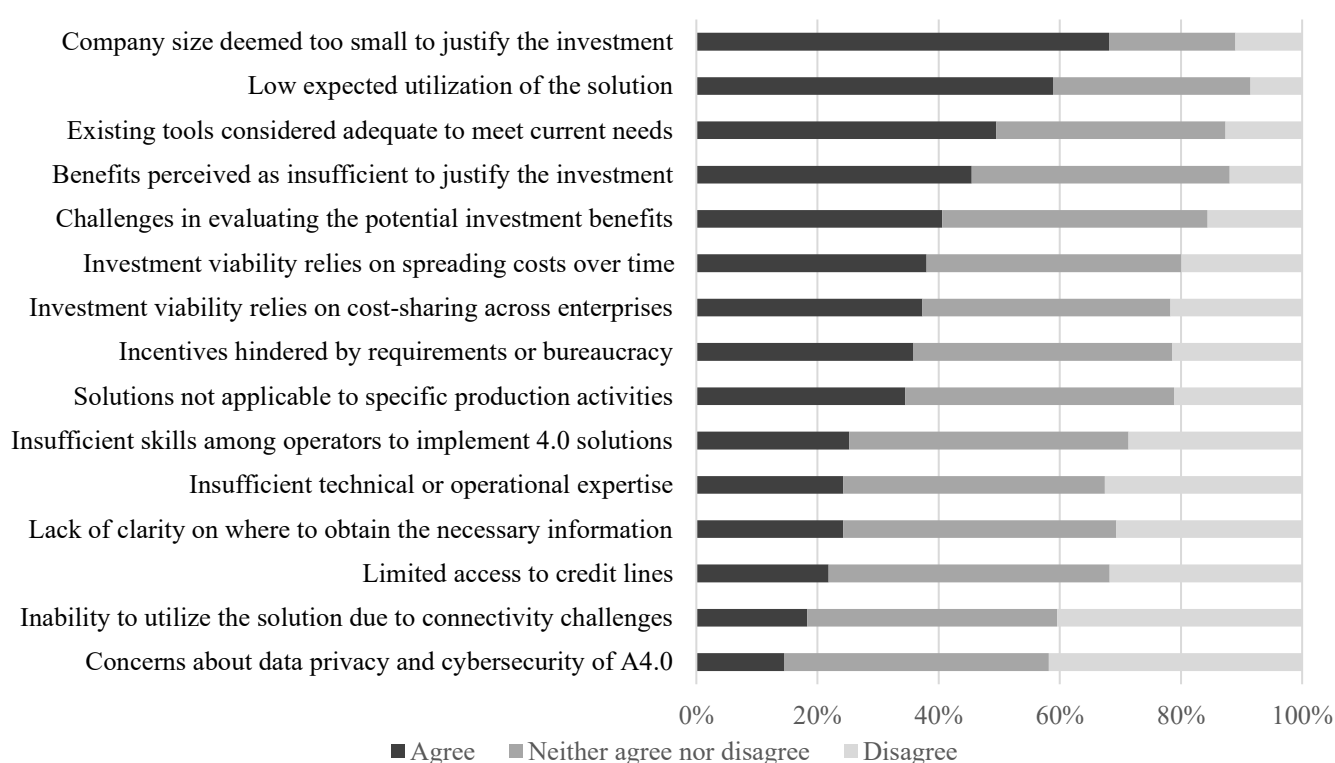
While digital skills were identified as a notable challenge among those who have already adopted A4.0 solutions, they appear to be a less pressing concern for non-adopters: only 25% of respondents cited a lack of necessary competencies as a barrier, while an equal proportion stated that their collaborators also lacked the required skills. Such discrepancy in how digital skills are perceived between adopters and non-adopters reflects an experience gap in A4.0 implementation: non-adopters seem to not yet acknowledge the digital skills challenge because they have not engaged with A4.0 deeply enough, whereas adopters have firsthand knowledge of the difficulties and their impact on agricultural activities. Furthermore, 24% of non-adopters indicated that they did not know where to access basic information about A4.0 solutions, underscoring the need for better dissemination of knowledge and educational resources.

Beyond financial and technical barriers, several other factors have contributed to the reluctance to adopt A4.0 technologies. A lack of applicability to specific agricultural production areas was cited by 34% of

571 respondents, suggesting that certain farming sectors or operational models do not align with the
 572 capabilities offered by the proposed A4.0 solutions. Connectivity issues also play a role, with 18% of
 573 respondents identifying poor internet access as a constraint, particularly where digital infrastructure may
 574 be insufficient. Additionally, concerns related to data security and privacy were reported by 15% of
 575 respondents, indicating a degree of hesitation regarding the management and protection of sensitive farm
 576 data in digital systems.

577 These findings highlight the multifaceted nature of the barriers impeding A4.0 adoption, encompassing
 578 economic, technical, infrastructural, and informational challenges. Addressing these concerns through
 579 targeted policies, financial support mechanisms, improved access to training, and enhanced digital
 580 infrastructure could facilitate broader adoption and ensure that a wider range of farms can benefit from
 581 the efficiencies and advancements offered by A4.0 solutions.

582 Figure 5. Inhibiting factors faced by respondents. Sample: 737 respondents who have not adopted any of the proposed Agriculture 4.0 solutions



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5. Discussion

587 This study examines the adoption and awareness levels of Agriculture 4.0 (A4.0) solutions, the drivers
 588 influencing technological adoption, the benefits obtained, as well as the challenges faced by A4.0 users
 589 and the inhibiting factors expressed by A4.0 non-adopters. A comprehensive understanding of these
 590 aspects is essential for policymakers, researchers, and industry stakeholders to identify obstacles and
 591 develop strategies aimed at facilitating the widespread integration of digital technologies in the
 592 agricultural sector. Such integration holds the potential to enhance productivity, efficiency, and
 593 sustainability within Italian agriculture.

594 The findings indicate that while there is widespread awareness of A4.0 solutions among Italian farmers,
 595 adoption levels vary significantly. These discrepancies are closely associated with the structural

596 characteristics of farming enterprises, particularly the size of the Utilised Agricultural Area (UAA) and
597 the level of annual turnover. Existing literature has consistently highlighted that the uptake of digital
598 agricultural technologies is contingent upon several structural and socio-economic factors, including
599 farm scale, crop specialization, farmer age, and educational background (Giua, 2022). At the national
600 level, our results corroborate this evidence, demonstrating that adoption rates tend to increase
601 proportionally with both the physical and economic size of farms. This trend is further reflected in
602 specific production types - such as cereals, fodder crops, and vineyards - where the extensive nature of
603 the former two may necessitate technological support, while the relatively higher revenue margins
604 typical of vineyard operations may facilitate investment in A4.0 solutions. Certain solutions, such as
605 monitoring systems and connected vehicles, have achieved higher acceptance, whereas others remain
606 unexploited. The primary motivation for adopting A4.0 solutions is predominantly associated with
607 macro-level farm management improvements, including enhanced forecasting capabilities and more
608 effective control and management processes, rather than in-field operational efficiencies, such as
609 optimising technical inputs and increasing machinery and equipment efficiency.

610 The analyses presented in this manuscript, which focus on the Italian agricultural sector, are broadly
611 aligned with the findings of international research. For instance, as reported by the United States
612 Department of Agriculture⁴, in 2023, 27% of U.S. farms or ranches employed A4.0 solutions for crop
613 management. Among the most widely adopted A4.0 solutions for crop management were automated
614 guidance systems (covering 58% of planted acres), yield mapping (44%), Variable Rate Technology
615 (37%), soil maps (22%) and drones and satellite imagery (7%) (United States Government
616 Accountability Office, 2024⁵). Similarly, in Germany, a survey conducted on Bavarian farmers reported
617 that the most widely adopted digital tools included weather and pests forecast models and apps (38%),
618 digital field records (21%), automated steering systems (21%), maps from satellite data (14%), with an
619 overall adoption rate estimated around 62% of the sampled agricultural enterprises (Gabriel and
620 Gandorfer, 2023).

621 This study also underscored the benefits of A4.0 solutions, which were generally perceived as aligning
622 with expectations, with some exceeding initial anticipations. This suggests a largely successful
623 implementation among adopters. Notably, the areas where respondents reported the greatest benefits
624 surpassing expectations were related to the optimisation of technical inputs and water management.
625 Consistent with the findings of Zul Azlan et al. (2023), Abbasi et al. (2022), and Pradel et al. (2022),
626 A4.0 solutions have demonstrated the potential to assist farmers in reducing input and water
627 consumption, thereby generating both economic advantages through cost reduction and environmental
628 benefits. Regarding the potential social sustainability benefits, Italian farmers have identified
629 “streamline and optimise workforce processes” among the ones more in line with expectations, with a
630 small share of farmers pointing out that A4.0 solutions disappointed their expectations. The broader
631 social sustainability implications of this perceived benefit remain debated in literature. Some studies
632 suggest a positive evolution in the agricultural labour market, potentially improving farmers' livelihoods
633 and creating new employment opportunities (e.g., Rotz et al., 2019). Other contributions, instead,
634 underline the need for specific studies on the yet unexplored consequences on the agricultural labour
635 market originated from the optimisation of farming activities, potentially reducing the demand for
636 unskilled workers (Rotz et al., 2019; Rose et al., 2021).

637 Nevertheless, despite the perceived benefits of A4.0 solutions, their implementation remains constrained
638 by several challenges. These include interoperability issues, lack of adequate skills, return on investment
639 concerns and technical assistance limitations, which hinder correct A4.0 solutions implementation and
640 their benefits. In addition, several financial and structural constraints emerge as significant deterrents for

⁴ <https://downloads.usda.library.cornell.edu/usda-esmis/files/h128nd689/4j03fg187/fj237k64f/fmpc0823.pdf>

⁵ <https://www.gao.gov/assets/d24105962.pdf>

641 non-adopters. Among these, the lack of trust in A4.0 solutions appears to be the most critical barrier.
642 This skepticism is often linked to a perceived low utility of A4.0, a belief that existing tools are sufficient
643 to meet current needs, difficulties in assessing the potential benefits, and the generally small size of
644 agricultural enterprises - factors that collectively slow digital adoption in Italian agriculture. Economic
645 and financial obstacles seem to be less relevant: these include doubts about the feasibility of investments
646 that depend on cost-sharing over time or across multiple farms, as well as limited access to incentives -
647 often constrained by bureaucratic complexity (Cisilino and Licciardo, 2022). These financial constraints
648 pose a fundamental challenge particularly for small and medium-sized farms that may lack the capital
649 required for initial investments in A4.0 solutions. This issue is further exacerbated by the uncertainty
650 surrounding return on investment, making it difficult for farmers to justify the adoption of these solutions
651 without clear and measurable long-term economic benefits. In contrast, technical challenges appear to
652 be less influential: only a minority of non-adopters cite inapplicability to specific production processes,
653 lack of technical skills, or insufficient expertise as reasons for avoiding A4.0 solutions. Moreover,
654 connectivity issues emerge as a challenge for non-adopters, especially in marginal areas and on hills
655 across Italian regions, thus limiting the implementation of A4.0 solutions, as highlighted by Sozzi et al.
656 (2021). As also emphasised by Fragomeli et al. (2024) and Da Silveira et al. (2023), such obstacles
657 significantly impede the broader adoption of A4.0 solutions by limiting both the willingness and ability
658 of farmers to integrate these tools into their production systems. Furthermore, as highlighted by
659 Gonzales-Gemio and Sanz-Martín (2025), the inequality in access to A4.0 solutions could hinder the
660 adoption of sustainable agricultural practices. Digital platforms and monitoring solutions, for instance,
661 have the potential to substantially enhance the efficiency of carbon farming initiatives and contribute
662 more broadly to agricultural sustainability.

663 These findings are consistent with an analysis published by the General Secretariat of the Council of the
664 European Union⁶, which emphasizes that - compared to other sectors - the pace of digital adoption in
665 agriculture has been slower. This lag is attributed to several interrelated factors, including inadequate
666 infrastructure, substantial upfront investment requirements, a widespread lack of digital skills, and the
667 inherent complexity of the agricultural sector. The latter includes considerable variability in climate
668 conditions, soil types, crop systems, and farming practices, all of which pose additional challenges to
669 the effective implementation of A4.0 solutions.

670 The findings of this study are also aligned with emerging academic literature on the barriers to A4.0
671 adoption within the Italian agricultural sector. For example, Addorisio et al. (2025), based on interviews
672 with Italian farmers, underscore the critical role of stakeholder cooperation and targeted training
673 initiatives in addressing key impediments to adoption. These include limited interoperability among
674 A4.0 solutions, insufficient digital competencies, and a lack of adequate technical support. Similarly,
675 Giorgio et al. (2024) explore perceived advantages and challenges associated with digitalisation in
676 Northern Italy. Reported benefits include enhanced environmental sustainability, improved input
677 efficiency, reduced labour requirements, and lower operational costs. However, the study also identifies
678 persistent barriers such as limited digital skills, inadequate data management practices and issues with
679 interoperability. These findings suggest that policies should not only support equipment acquisition, but
680 also promote the development of farmers' human capital.

681 Addressing these challenges through targeted policy interventions, comprehensive training initiatives,
682 and improved system interoperability could substantially enhance A4.0 adoption rates, thereby ensuring
683 that a broader range of agricultural enterprises benefits from the efficiencies and advancements offered
684 by digital innovations. Moreover, collaboration among policymakers, technology providers, and industry

⁶ https://www.consilium.europa.eu/media/shxiaxmo/2024_971-art-agriculture-11-02-25.pdf

685 stakeholders is crucial in fostering an ecosystem that supports seamless integration, mitigates adoption
686 barriers, and maximizes the impact of digital agricultural innovations.

687 **Conclusions**

688 This study offers valuable empirical insights into the current state of Agriculture 4.0 (A4.0) adoption in
689 Italy, shedding light on drivers influencing the uptake of A4.0 solutions, the perceived benefits, the
690 challenges met by farmers who adopted A4.0 solutions and the barriers that prevented other agricultural
691 enterprises from adopting A4.0 solutions. By disaggregating results according to critical variables related
692 to farms (size, primary crop production and geographical localisation), this research contributes to a
693 more nuanced understanding of how the A4.0 paradigm is taking root within the Italian agricultural
694 sector. These findings provide a strong empirical foundation for informing public policy, guiding
695 investment strategies and designing initiatives that are tailored to the needs of diverse farming profiles.

696 Specifically, the results highlight the importance of structural variables such as farm size, crop
697 production and turnover in shaping adoption patterns, suggesting that public support mechanisms should
698 be differentiated accordingly. Small farms, which tend to face greater barriers in terms of investment
699 capacity and technical know-how, may benefit from targeted subsidies, tax incentives, and digital
700 infrastructure improvements, particularly in under-served rural regions. Moreover, the limited adoption
701 of certain A4.0 solutions underscores the need for broader outreach, technical assistance and knowledge
702 transfer mechanisms to ensure that innovation diffuses beyond a small subset of more-structured farms.
703 Training programs should also be adapted to the varying levels of digital literacy across the sector, with
704 modular content suited to both entry-level and experienced users.

705 Moreover, by identifying which types of farms might be most likely to adopt A4.0 solutions and which
706 barriers inhibit the uptake of digital tools, technology providers can refine their product design,
707 marketing strategies, and sales services. Companies may, for instance, enhance interoperability and user-
708 friendliness to address common usability challenges.

709 A promising avenue for future research would involve a comparative analysis of the levels of A4.0
710 adoption, associated needs, benefits, challenges, and barriers identified in this study with those observed
711 in other European countries and beyond.

712 Another potential research direction could focus on examining the impact of A4.0 solutions on
713 economic, environmental, and social sustainability to comprehensively assess the costs and benefits of
714 A4.0 implementation. This analysis could, in turn, contribute to bridging the gap between adopting and
715 non-adopting agricultural enterprises.

716 Nonetheless, these contributions should be considered in light of the following methodological
717 limitations arising from the survey administration method and the sample distribution compared to the
718 reference population. As with all Computer-Assisted Web Interviewing (CAWI) methods, this online
719 survey may exclude individuals without internet access or those less comfortable with technology.
720 Additionally, self-selection bias could skew the results, as participants are likely to be those with an
721 interest in the topic or familiarity with online surveys. Consequently, adoption rates of A4.0 solutions
722 reported in this study may be overestimated, while the perceived benefits and willingness to invest
723 further in digital technologies could reflect the attitudes of a smaller group of more innovation-oriented
724 farmers. Addressing these limitations in the future research would require efforts to reach less digitally-
725 involved segments of the Italian agricultural sector to enhance the external validity of the findings.

726 Moreover, the discrepancy between the sample size distribution and the population size distribution leads
727 to an overrepresentation of farms in the North and an underrepresentation of those in the South and
728 Islands, potentially introducing a geographical bias. Furthermore, the average UAA (Utilised
729 Agricultural Area) of the sampled farms (22 hectares) is significantly higher than the figure reported by

730 ISTAT⁷ (11.1 hectares), suggesting a selection of more structured agricultural enterprises. Additionally,
731 the greater representation of the vineyard sector, which is characterised by higher-than-average
732 profitability and greater spending capacity, could influence this study's findings.

733 This study was carried out within the AgriTech National Research Center and within the Smart AgriFood
734 Observatory - Politecnico di Milano & University of Brescia and received funding from the European
735 Union Next-GenerationEU (PIANO NAZIONALE DI RIPRESA E RESILIENZA (PNRR) –
736 MISSIONE 4 COMPONENTE 2, INVESTIMENTO 1.4 – D.D. 1032 17/06/2022, CN000000022). This
737 paper reflects only the authors' views and opinions, neither the European Union nor the European
738 Commission can be considered responsible for them.

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