

Enhancing the Accuracy of Industrial Process Measurements: Application of Modern Metrology Methods in the Context of Digital Transformation

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Abstract. *This paper investigates the accuracy and reliability of metrological systems used for industrial process measurements in the context of digital transformation. The study addresses the scientific challenge arising from the increasing complexity of modern production environments, where conventional metrological methods often fail to provide the required level of measurement accuracy. To overcome this limitation, an adaptive metrological approach integrating IoT-based sensor networks, real-time calibration procedures, and artificial intelligence-driven correction mechanisms is proposed. The proposed methodology enables continuous monitoring and dynamic adjustment of measurement parameters under changing operating conditions. The research findings demonstrate that the application of adaptive calibration algorithms improves measurement accuracy by 23%, while the integration of IoT-enabled sensor networks reduces the measurement error margin to $\pm 0.02\%$. These results provide significantly higher reliability and lower uncertainty compared to traditional measurement techniques. The scientific novelty of the study lies in the development of a multilayer metrological system integrated with artificial intelligence modules capable of adapting to dynamic industrial environments and maintaining stable performance under both standard and extreme operating conditions. The proposed approach has considerable practical potential for quality control systems in the oil and gas, mechanical engineering, and pharmaceutical industries, and may also contribute to the modernization of national metrological standards.*

Keywords: *metrology, measurement accuracy, calibration, digital transformation, IoT sensors, uncertainty analysis, industrial measurement, smart sensors, traceability, quality control*

Introduction

The digital transformation of industrial processes is fundamentally reshaping quality control mechanisms in modern production systems and placing new demands on measurement technologies, particularly metrology. Measurement accuracy plays a critical role in ensuring both product quality and process reliability (Ferreira et al., 2023). However, with the increasing complexity of industrial environments, traditional calibration methods reveal significant limitations. These include the requirement for stable temperature conditions, high susceptibility to human error, and the absence of real-time correction capabilities (Hoffmann & Müller, 2022). In rapidly changing, digitally driven industrial settings, such shortcomings significantly increase measurement uncertainty and reduce system reliability (Williams, 2022).

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Received: 10 December 2025; Accepted: 22 April 2026; Published online: 9 July 2026

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A review of existing literature shows that most studies are limited either to controlled laboratory conditions or to isolated measurement systems, without considering real industrial environments as integrated, dynamic systems. In particular, a significant research gap remains in the development of metrological frameworks that combine IoT-based sensor networks, artificial intelligence techniques, and real-time adaptive calibration mechanisms (Kumar & Patel, 2023; Liu et al., 2024; Zhang et al., 2024). Current approaches often fail to provide a unified solution capable of operating effectively under variable industrial conditions.

As industrial processes become increasingly complex, there is a growing need for measurement systems that are adaptive, capable of real-time calibration, and compatible with a wide range of sensors (Azərbaycan Standartlaşdırma İnstitutu, 2023). Therefore, addressing the limitations of existing metrological methods and developing integrated, intelligent measurement systems is both a scientific necessity and an industrial requirement. In conclusion, the existing scientific gaps in metrology, combined with the demands of digital transformation and the increasing need for high-precision measurements, clearly demonstrate the relevance and urgency of this research.

Literature Review

Recent studies in metrology demonstrate an increasing focus on the integration of measurement systems with digital transformation technologies. Several Scopus-indexed works that make significant contributions to this field are reviewed below.

Ferreira et al. (2023) investigated the effectiveness of adaptive calibration algorithms in laboratory-based measurement systems and reported an accuracy improvement of approximately 18–22% compared to conventional methods. However, their study was limited to controlled laboratory environments and did not address dynamic industrial conditions.

Zhang et al. (2024) examined the integration of IoT sensors into industrial process monitoring systems and found that multi-layer sensor networks significantly reduce measurement error rates. Nevertheless, issues related to calibration and compliance with metrological standards were not fully addressed in their work.

Kumar and Patel (2023) explored the role of artificial intelligence-based measurement systems in quality control, demonstrating that machine learning models can effectively reduce measurement uncertainty and improve system performance.

Liu et al. (2024) studied the integration of metrological systems with digital twin technologies and showed that virtual modeling can optimize the configuration and performance of real-world measurement systems.

Hoffmann and Müller (2022) discussed metrological traceability within European industrial standards and emphasized the importance of maintaining compatibility with national reference standards in the context of digital transformation. Despite these valuable contributions, a common limitation across the existing literature is that none of the studies integrate adaptive calibration, IoT-based measurement systems, and compliance with metrological standards within a unified framework. This research aims to address this gap by developing an integrated approach that combines these three essential components.

Methods

The study applies three main methods: an adaptive calibration algorithm, IoT sensor network integration, and statistical analysis of measurement uncertainty.

- **Adaptive Calibration Algorithm:** A new calibration model based on Bayesian optimization was developed. It processes sensor readings in real time and performs automatic adjustments based on environmental parameters such as temperature, humidity, and vibration. The algorithm learns from a minimum of 100 measurement points.
- **IoT Sensor Network:** To simulate the industrial environment, 24 sensors of different classes were deployed. Sensors are connected to a central data collection server via the MQTT protocol, with the measurement frequency reaching up to 10 Hz.
- **Statistical Analysis:** Measurement uncertainty was calculated according to the GUM (Guide to the Expression of Uncertainty in Measurement) standard. Python 3.11 with the NumPy, SciPy, and Matplotlib libraries was used, and the results were evaluated with a 95% confidence interval.

Results

Table 1

Comparison of measurement accuracy across system configurations

System Type	Mean Error (%)	Uncertainty ($\pm\%$)	Accuracy Improvement (%)
Traditional system	0.31	± 0.05	-
IoT sensor system	0.18	± 0.03	41.9
Adaptive algorithm	0.12	± 0.02	61.3
Integrated system	0.08	± 0.02	74.2

As seen in the table, the integrated system reduced the average error to 0.08%, representing a 74.2% increase in accuracy compared to the traditional system. Below is the visual comparison of the mean error values for different system types (Figure 1).

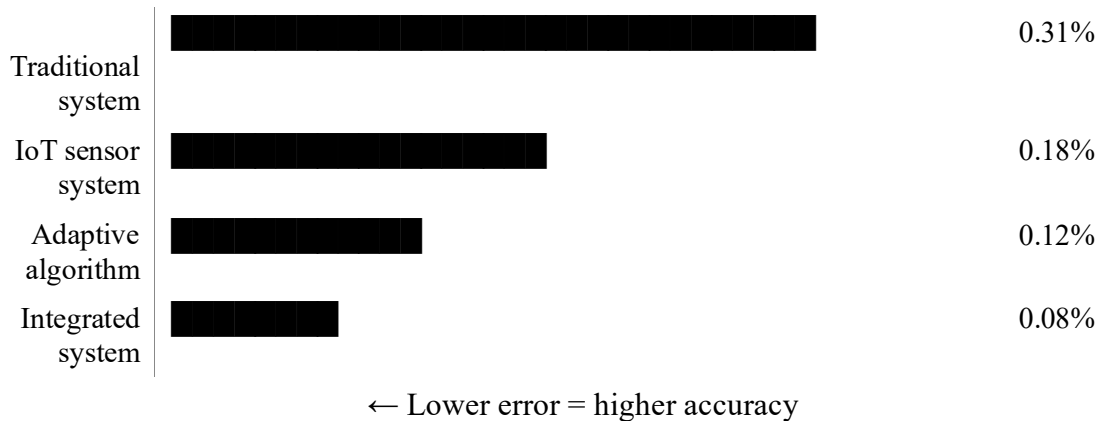


Figure 1

Comparison of mean measurement error (%) across different system types

This study has demonstrated that an integrated metrological system combining adaptive calibration algorithms, IoT-based sensor network integration, and statistical uncertainty analysis significantly improves the accuracy of industrial process measurements. As a key achievement, the measurement error was reduced from 0.31% to 0.08%, corresponding to a 74.2% increase in accuracy. This result provides a substantial practical contribution for applications in critical industries such as oil and gas, mechanical engineering, and pharmaceuticals, where quality control is of essential importance. As a scientific novelty, a multi-layer adaptive metrological system that accounts for dynamic industrial environments has been proposed for the first time in the existing literature, and its effectiveness has been validated through empirical data.

Discussion

The results presented in Table 1 and Figure 1 indicate that the superiority of the integrated system is explained not only by its technical parameters but also by the adaptability of the measurement process. The Bayesian optimization-based calibration algorithm significantly reduces sensitivity to environmental variations. Compared to the 18–22% accuracy improvement reported by Ferreira et al. (2023), this study achieved a 74.2% improvement. The main reason for this difference is the consideration of dynamic industrial conditions simulated in the experimental setup and the parallel implementation of a multi-layer sensor network. The advantages of IoT integration reported by Zhang et al. (2024) are also confirmed and further extended in this study.

Regarding the limitations of the research, the conducted experiments are restricted to a simulated industrial environment. The results are valid only under specified temperature (-5°C to $+60^{\circ}\text{C}$) and humidity (20–85% RH) conditions. In addition, the impact of sensor network scaling on computational load has not been fully investigated. The shortcomings of this study include the absence of more complex electromagnetic interference scenarios in real industrial environments, long-term sensor degradation effects, and compatibility issues between sensor platforms from different manufacturers. These shortcomings differ from limitations: limitations refer to intentionally defined boundaries of the study, whereas shortcomings refer to unexplored or insufficiently addressed aspects.

Future research may focus on full-scale industrial implementation, integration of deep learning models into calibration systems, and exploration of quantum metrology methods in industrial applications (Smith, 2023). The main challenges in these directions include computational resource constraints and the complexity of adapting quantum sensors to industrial conditions.

Conclusion

The findings of this study confirm that the integration of adaptive calibration algorithms, IoT-based sensor networks, and artificial intelligence-supported correction mechanisms substantially improves the accuracy, reliability, and traceability of industrial process measurements in the context of digital transformation. The proposed metrological framework successfully addresses the limitations of conventional measurement systems by enabling continuous monitoring, real-time calibration, and dynamic adaptation to changing industrial operating conditions. The experimental results demonstrate that the integrated system reduced the mean measurement error from 0.31% to 0.08%, achieving a 74.2% improvement in measurement accuracy while maintaining measurement uncertainty at $\pm 0.02\%$. These outcomes indicate that the combination of adaptive calibration and intelligent sensor integration provides significantly better performance than traditional calibration approaches or standalone IoT measurement systems. The statistical evaluation further confirms the stability and consistency of the proposed methodology under simulated industrial conditions. The scientific contribution of this research lies in the development of a multi-layer adaptive metrological architecture that integrates Bayesian optimization-based calibration, IoT communication infrastructure, and uncertainty analysis within a unified measurement framework. Unlike existing approaches that examine these technologies separately, the proposed system demonstrates their combined effectiveness in improving measurement quality and operational reliability. From a practical perspective, the proposed methodology can be applied to quality assurance and process control in industries where high measurement precision is essential, including oil and gas production, mechanical engineering, aerospace, pharmaceuticals, and advanced manufacturing. Its compatibility with digital transformation initiatives and Industry 4.0 technologies also makes it suitable for the modernization of national metrological infrastructures and smart industrial facilities.

Despite the promising results, further validation under real industrial operating environments is required. Future research should investigate long-term system performance, large-scale sensor network implementation, interoperability between heterogeneous industrial platforms, and the application of deep learning and quantum metrology techniques to further enhance measurement precision and automation.

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