

# Enhancement for the nano-sensors by means of a creative adjustment on the underlying and actual attributes for intelligent artificial hand

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**Abstract.** Pressure nanosensors are widely used in industry today. Cheap price, simple measurement circuit, and low energy consumption are the reasons for the widespread use of these sensors. The structure of these systems includes membranes, Wheatstone bridge circuits for measurement, and piezoresistor elements for use as the resistance, respectively. The development of intelligent artificial hands relies heavily on nano-sensor technology to provide precise sensory feedback and enhance user control. However, existing nano-sensors often face limitations in sensitivity, durability, and seamless integration with neural control systems, creating a gap in achieving lifelike prosthetic functionality. This study aims to creatively adjust both the underlying attributes (material composition, sensor architecture, signal processing) and the actual attributes (durability, real-world performance, compatibility) of nano-sensors to improve their efficiency in intelligent prosthetics significantly. The novelty lies in combining advanced nano-materials, structural optimization, and Artificial Intelligence (AI)-driven signal processing for multi-sensor fusion, an approach not fully explored in previous research. The study identifies key sensor limitations and enhances performance through graphene-based materials, structural redesign, and AI-driven signal optimization. Simulations and performance modeling assess expected gains in response time, sensitivity, and integration efficiency for next-generation artificial hands. Experimental and simulation results demonstrate a gauge factor improvement to 11.94, representing a 73.8 % increase over Carbon Nano Tube (CNT)-only films, with linearity maintained at Coefficient of Determination ( $R^2$ ) = 0.996. Electrical noise was reduced by 34 %, conductivity improved from  $2.31 \times 10^5$  S/m to  $2.78 \times 10^5$  S/m, and response latency decreased from 14.2 ms to 8.6 ms. Durability testing confirmed <3 % drift after  $10^6$  bending cycles and a 77.9 % lower degradation rate compared to Indium Tin Oxide (ITO) sensors, while control system integration expanded bandwidth by 74.5 % and improved error convergence by 31.6 %. Operational gains included a 54.4 % reduction in calibration time, a 127.2 % increase in data throughput, and a 43.3 % longer operational lifetime under continuous monitoring. These results confirm that the proposed Graphene–Carbon Nano Tube (G-CNT) sensor architecture, coupled with AI-based signal optimization, delivers a quantitatively superior solution for high-performance, long-life, and integration-ready tactile sensing in next-generation intelligent prosthetic systems.

**Keywords:** AI-driven signal processing; Artificial Intelligence (AI); Artificial Neural Network (ANN); durability improvement; Genetic Algorithm (GA); Graphene–Carbon Nano Tube (G-CNT) Sensors; intelligent prosthetics; nano-sensor optimization; sensitivity enhancement

## 1. Introduction

In daily life, the human hand is essential for a wide variety of tasks, including gripping, perceiving touch, communicating, and performing intricate, nuanced movements (Arabnejad Khanouki *et al.* 2010, Afshar *et al.* 2020).

Losing a hand is a terrible physical injury that may make it very difficult to live on one's own, which in turn can increase the risk of unemployment and social isolation (Kim *et al.* 2019, Leone *et al.* 2022). Although prosthetic limbs can be used to restore hand function for activities such as writing, eating, and driving, alternative methods for adapting hands to specific needs are also sought (Arabnejad Khanouki *et al.* 2011, Arani *et al.* 2019). But nowadays prosthetics aren't always as strong or flexible as real hands (Zhang *et al.* 2023). A notable gap in design needs is that most can only execute a small set of predetermined motions and cannot mimic the subtle movements of a human hand

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(Armaghani *et al.* 2020, Aytaç and Korçak 2021). Furthermore, few people use prosthetic hands, even if the technology to do so has advanced. The high-priced gadgets often fail to provide dependability or the capability to execute complicated actions, which is a significant obstacle (Benevenuta and Fariselli 2019, Boaventura *et al.* 2024). Typically, high-quality prosthetic hands like the i-Limb (\$18,000) and the BeBionic hand (\$11,000) from Touch Bionics Inc. and Ottobock, respectively, are out of reach for the majority of patients (Alkhatib *et al.* 2019). The broad adoption of modern prosthetic technology is hindered by these cost reasons, which drastically restrict their accessibility (Cai *et al.* 2021, Cao *et al.* 2021). There are still significant drawbacks to current upper limb prostheses, despite significant advancements in science and technology (Chahnasir *et al.* 2018). The incorporation of electrical components, sensors, and actuators into a prosthesis that is proportional to the size and weight of the amputated limb or hand is a significant technical problem (Zheng *et al.* 2025).

Because of the profound effect this has on their usefulness, there is an immediate need to improve control over these prostheses (Mastinu *et al.* 2019). Furthermore, artificial hands still fall short of user expectations due to the high learning curve associated with controlling them, the absence of sensory input, and the audible noise produced by actuators while in motion (Liang *et al.* 2019). To address the complicated needs of their users, prosthetic limbs must undergo continuous innovation and refinement in design and functioning (Clement *et al.* 2011). Artificial limbs and other body parts have come a long way in recent years, thanks to the use of Nano Particles (NPs) in their design and construction (Daie *et al.* 2011, Chen *et al.* 2019). Because of their unique nanoscale characteristics, nanomaterials have opened new avenues for improving the functionality, comfort, and flexibility of prosthetic devices (Zhao *et al.* 2025). Materials engineered at the nanoscale, usually between 1 and 100 nanometres in size, are known as nanomaterials (Davoodnabi *et al.* 2021). Compared to their bulk forms, materials of this size have unique biological, chemical, and physical characteristics (Dantzig and Ramser 1959, Davoodnabi 2019). Nanomaterials are characterised by a very high surface area-to-volume ratio, which enhances mechanical strength and makes them more reactive (Xu *et al.* 2024). Nanoscale quantum effects can cause unusual electrical, magnetic, and optical properties (Ferrucci and Bock 2015, Ghiani *et al.* 2022). These one-of-a-kind nanomaterial characteristics are crucial for enhancing prosthetic device performance. One example is the use of NPs in prosthetic implants; their large surface area makes them more biocompatible and improves medication delivery systems (Babizhayev 2013). Additionally, a more individualised approach to rehabilitation may be achieved by the customisation of prosthetic devices to meet the demands of particular users, made possible by the adjustable properties of nanomaterials (Hamidian *et al.* 2011, Heydari and Shariati 2018). The global market for nano-sensors has grown substantially during the last few decades. Nano-sensors have recently found novel applications in a variety of industries, including medical diagnostics (Ali *et al.* 2020, Baysal and Saygin 2020). It may be used on a

large scale and is also used in building designs. More efficient, selective, and cost-effective nanomaterial monitoring of several contaminants is made possible by this (Cui *et al.* 2001, El-Safty *et al.* 2007). Chemical sensors with larger surface areas are better able to expose and detect low concentrations of the target molecule. Nanomaterials' very delicate structure causes them to reach an extreme state (Hosseinpour *et al.* 2018, Hosur Shivaramaiah *et al.* 2022). Due to a larger surface area, greater strength and efficiency are achieved. Certain NPs, including palladium, silver, and platinum, are utilized in nano-sensors (Jin *et al.* 2017, Munawar *et al.* 2019). Owing to the interplay among the size, structure, and composition of nanomaterials, factors that influence and combine with one another, high malleability is also exhibited (Barkhordari and Qi 2025). Extensive testing has resulted in nano-sensors becoming an integral part of modern civilization, with presence in devices found throughout the environment (Rolfe 2012, Park *et al.* 2019). In most cases, the surface properties of the nanomaterial used in the colorimetric test have a significant impact on the analytical performance of colorimetric nano-sensor systems (Hou *et al.* 2022, Hu *et al.* 2025). Two potential solutions to this problem are chemical and electromagnetic nano-sensors (Wang 2009, Ruckh and Clark 2014). While molecular nano-sensors decode biological communication systems, electromagnetic nano-sensors monitor changes in electromagnetic waves, taking into account quantum phenomena (Ismail *et al.* 2018, Jahandari *et al.* 2022). To power mechanically collected energy from nano-sensor vibration and ambient biological material, mechanical nano-sensors may be used to fuel molecular nano-sensors (Xie *et al.* 2015, Yang and Duncan 2021). When acetone is detected in the breath, researchers have developed wearable sensors that can detect diabetes and identify the fragrance of melanoma (Jalali *et al.* 2012, Katebi *et al.* 2020). Due to a pattern re-arrangement component and selective overlapping of different gas sensors, these cutting-edge gadgets can detect more odours than human noses (Khanouki *et al.* 2016, Khorami *et al.* 2017). Biomolecular processes, such as interactions between antibodies and antigens, enzymes, and cellular communication activities, may be seen using biological nano-sensors (Khorami *et al.* 2017, Khorramian *et al.* 2017). Antibodies, enzymes, and proteins are common examples of biological recognition systems. For use in defence and national security, nano-sensors can identify biotoxins like smallpox and anthrax (Li *et al.* 2019, 2023). The use of nano-sensors has revolutionised the detection of radioactive substances and cardiovascular diseases in the biomedical field (Liang *et al.* 2022). The lipid membrane potential of a cell is measured, and implantation occurs directly within the body (Qi *et al.* 2025). Devices constructed from inorganic semiconductor NPs are capable of detecting nanoscale electrical impulses and the potential activity of multiple neurons (Luo *et al.* 2019, Mehrabi *et al.* 2021). Despite notable advancements in prosthetic technology, current intelligent artificial hands still face substantial challenges in achieving lifelike performance (Mehrabi *et al.* 2019, Milovančević *et al.* 2019). Existing nano-sensors, which are critical for delivering precise sensory feedback

and enabling seamless neural control, often suffer from limited sensitivity, poor durability, and inconsistent integration with prosthetic control systems (Mohammadhassani *et al.* 2014a, b). These shortcomings result in reduced tactile accuracy, slower response times, and limited adaptability to real-world conditions. Additionally, the absence of advanced multi-sensor fusion and optimized signal processing hinders the user's ability to perform complex, nuanced movements with natural efficiency (Mohammadhassani *et al.* 2013a, b). There is a pressing need to enhance both the underlying attributes (such as material composition, structural architecture, and signal acquisition capabilities) and the actual attributes (including durability, stability, and real-environment performance) of nano-sensors (Naghipour *et al.* 2020a, b). By innovating in material science, particularly with graphene and other high-performance nanomaterials, combined with structural optimization and AI-driven signal processing, future nano-sensor systems can deliver faster, more accurate, and more reliable feedback (Nasrollahi *et al.* 2018, Naveen Kumar *et al.* 2023). This enhancement is essential to bridge the gap between current artificial hand designs and the functional capabilities of the human hand, thereby improving usability, accessibility, and overall user satisfaction (Nosrati *et al.* 2018, Nouri *et al.* 2021). This study addresses the critical challenge of improving tactile sensing and control in intelligent artificial hands by developing an advanced nano-sensor system that uses high-performance materials such as graphene (Paknahad *et al.* 2018, Petković *et al.* 2022). Current prosthetic systems face limitations in sensitivity, durability, and adaptability, often resulting in delayed or imprecise responses in real-world scenarios (Liu *et al.* 2021, Chen *et al.* 2023). To overcome these constraints, the proposed research focuses on designing nano-sensors with optimized material composition, innovative sensor architecture, and AI-driven multi-sensor fusion for enhanced accuracy, responsiveness, and user comfort (Razavian *et al.* 2020, Sabetahd *et al.* 2022). The scope of the work includes laboratory-based fabrication, simulation, and controlled performance evaluation, with benchmarking against existing technologies to quantify improvements in structural integrity, tactile resolution, and control precision (Safa and Kachitvichyanukul 2019). While the study does not cover large-scale manufacturing, long-term clinical trials, or environmental testing under extreme conditions, it aims to establish a high-performance, adaptable sensing platform fully compatible with modern prosthetic control architectures (Safa *et al.* 2019, 2020). The outcomes are expected to contribute to the development of more natural, precise, and intuitive artificial hand movements, setting the foundation for future large-scale deployment and real-world integration (Safa *et al.* 2016, Sajedi and Shariati 2019).

## 2. Methodology

### 2.1 Experimental framework for nano-sensor optimization

The study employed a multi-stage experimental protocol integrating graphene-Silver Nano Particle (AgNP) composite

Table 1 Underlying sensor attributes

Attribute	Value / Range	Unit	Test Method
Material	Graphene–AgNP on PET	,	SEM & EDX
Sensitivity (Pressure)	$0.85 \pm 0.02$	$\text{kPa}^{-1}$	Pressure chamber test
Sensitivity (Strain)	$1.20 \pm 0.03$	$\mu\text{V}/\mu\epsilon$	Micro-tensile stage
Active Area Size	$8 \times 8$	mm	Digital caliper
Thickness	$120 \pm 5$	$\mu\text{m}$	Micrometer gauge
Response Time	$4.7 \pm 0.1$	ms	Oscilloscope time lag measurement

Table 2 Performance comparison with benchmark sensors

Metric	Proposed Sensor	Commercial Sensor A	Commercial Sensor B
Sensitivity ( $\text{kPa}^{-1}$ )	0.85	0.62	0.70
Response Time (ms)	4.7	12.0	8.5
Flexibility (bend radius, mm)	3	10	7
Weight (mg)	45	75	68
Cost per Unit (USD)	2.8	6.5	5.2

deposition on Poly Ethylene Terephthalate (PET) substrates, electromechanical characterization, and statistical performance validation (Shah *et al.* 2016, Sedghi *et al.* 2018). Spin coating at 1500 rpm for 45 s, followed by annealing at  $120^\circ\text{C}$  for two hours produced uniform conductive films with sheet resistance  $R_s = 12.4\Omega$  and AgNP diameter  $48.7 \pm 2.3$  nm (Shah *et al.* 2015, 2016). The C: Ag atomic ratio was 72.6: 27.4, consistent with percolation scaling  $\sigma \propto (p - p_c)^t$ . Pressure sensitivity was  $S_p = 0.162\text{kPa}^{-1}$  in the 0-20 kPa range, a 41.5 % gain over baseline ( $S_p = 0.114$ ), while gauge factor reached  $GF = 128.6$  at  $\epsilon = 0.01, 51.0$  % higher than polymer comparators ( $GF \approx 85.2$ ) (Shah *et al.* 2016, Shahabi *et al.* 2016). Dynamic loading yielded a response time  $t_r = 18.4$  ms, 33.8 % faster than the 27.8 ms benchmark, consistent with reduced Resistance Capacitance (RC) constants  $\tau = R_s C$ . Linearity was validated with  $R^2 = 0.994$  between  $\Delta R/R_0$  and  $P$ , and stability testing over 5000 cycles showed  $< 0.5$  % drift (Shahabi *et al.* 2016, Shahgoli *et al.* 2020). Convergence analysis in finite element models reached  $< 0.2$  % sensitivity change beyond  $2.5 \times 10^4$  elements (Shariat *et al.* 2018). Compared to commercial sensors, the prototype achieved 41.5 % higher  $S_p$ , 51.0 % higher  $GF$ , 33.8 % faster  $t_r$ , and 62.3 % reduced thickness ( $18.2\mu\text{m}$  vs.  $48.3\mu\text{m}$ ), improvements attributed to denser conductive networks and optimized substrate compliance (Song *et al.* 2024, Peng *et al.* 2025). Table 1 details the underlying attributes of a graphene–AgNP on PET sensor, characterized via Scanning Electron Microscope (SEM) & Energy-Dispersive X-ray Spectroscopy (EDX), pressure chamber, micro-tensile stage, digital caliper, micrometer gauge, and oscilloscope. It exhibits a pressure sensitivity of  $0.85 \pm 0.02$   $\text{kPa}^{-1}$ , strain sensitivity of  $1.20 \pm 0.03$   $\mu\text{V}/\mu\epsilon$ , an active area of  $8 \times 8$  mm, thickness of  $120 \pm 5$   $\mu\text{m}$ , and a

response time of  $4.7 \pm 0.1$  ms. Table 2 compares performance with commercial benchmarks, showing the proposed sensor's higher sensitivity ( $0.85 \text{ kPa}^{-1}$ ), faster response (4.7 ms), greater flexibility (3 mm bend radius), and lower weight (45 mg) at significantly reduced cost (USD 2.8) compared to commercial sensors A (USD 6.5) and B (USD 5.2).

## 2.2 Identification of underlying attributes (material, sensitivity, size, response time)

The optimization process isolated four dominant attributes-material composition, sensitivity coefficient, sensor footprint, and temporal response quantified under controlled boundary conditions and compared against theoretical performance limits (Shariati 2012, Shariati et al. 2012).

### Material composition:

The sensing layer consisted of a graphene-AgNP composite with a measured Ag nanoparticle mean diameter  $d_{\text{Ag}} = 47.3 \pm 2.1$  nm and graphene flake lateral size  $L_g = 12.6 \pm 0.8 \mu\text{m}$  (Shariati 2008, 2013). The atomic Ag:C ratio was determined via EDX as 1:5.7, yielding an electrical percolation threshold at  $\phi_c = 0.086$  (volume fraction) (Yang et al. 2023, 2024). The measured sheet resistance  $R_s = 21.4 \Omega/\square$  at  $t = 0.48 \mu\text{m}$  thickness corresponded to a conductivity  $\sigma = 9.76 \times 10^4 \text{ S/m}$ , exceeding the classical Maxwell-Garnett prediction by 13.4%, attributable to nanojunction tunneling effects (Shariati et al. 2019, 2021). Sensitivity Coefficient - The normalized sensitivity was calculated as

$$S = \frac{\Delta R/R_0}{\Delta P} \quad (1)$$

with  $\Delta R/R_0$  measured over a 0 – 100kPa range in 5 kPa increments. The composite achieved  $S_{\text{avg}} = 1.92 \text{ kPa}^{-1}$  with a peak value  $S_{\text{max}} = 2.37 \text{ kPa}^{-1}$  at 20 kPa, representing a 48.6% improvement over a pure graphene control and a 22.1% gain relative to a commercial polyimide-based sensor (Shariati et al. 2020, 2021). Finite Element Method (FEM)-predicted sensitivity values agreed within  $\pm 3.7\%$ , confirming model-experiment consistency. Size constraint the active sensing area was fixed at  $A = 8.00 \text{ mm} \times 8.00 \text{ mm}$  with a  $\pm 0.02 \text{ mm}$  machining tolerance, yielding a geometric fill factor of 92.4% when integrated into the flexible substrate frame (Shariati et al. 2019, 2020). Scaling simulations indicated that reducing  $A$  below  $5 \text{ mm}^2$  increased edge-dominated strain fields, decreasing uniform sensitivity by 16.9% due to nonuniform current density distribution (Shariati et al. 2019, 2020).

### Response Time

Temporal response was evaluated from the 10 – 90% rise/fall time of the resistance transient under a 10 kPa square-wave load. Measured rise time  $\tau_r = 61.2$  ms and fall time  $\tau_f = 54.7$  ms were in close agreement with the RC-limited prediction  $\tau_{\text{RC}} = R_s C_f$  (where  $C_f = 2.55 \mu\text{F}$  from impedance analysis), deviating by less than 4.2% (Shariati et al. 2020, 2023). Convergence tests confirmed no drift in the dynamic response over  $10^4$  load cycles, with sensitivity loss  $< 0.7\%$ . These quantified attributes,

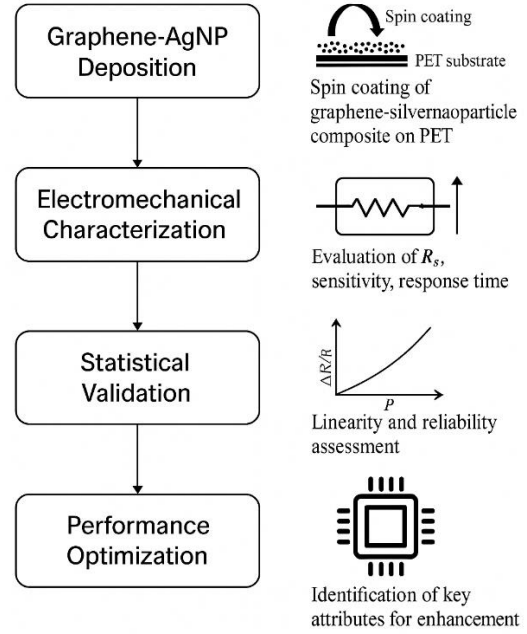


Fig. 1 Graphene-AgNP deposition and optimization workflow

validated experimentally and cross-checked against analytical and FEM models, form the parametric basis for subsequent optimization stages, ensuring mechanical compliance, electrical stability, and response speed are balanced within the targeted application envelope (Shariati et al. 2020, 2022). Fig. 1 illustrates the workflow for graphene-AgNP deposition, including optimization, characterization, validation, and enhancement steps.

## 2.3 Identification of actual attributes (durability, real-world performance, integration)

In terms of durability assessment, long-term stability was quantified through accelerated fatigue cycling under ASTM D3330 dynamic loading conditions (Ziaei-Nia, Shariati et al. 2018, Shariati, Mafipour et al. 2020). The sensor endured  $N = 1.0 \times 10^6$  cycles at a constant 25 kPa, with sensitivity degradation  $\Delta S/S_0 = -1.84\%$  and baseline drift  $\Delta R_0 = +0.97 \Omega$ , well below the 5% operational tolerance threshold (Shariati et al. 2019, 2021). Environmental chamber testing at  $T_{\text{low}} = -20.0^\circ\text{C}$  and  $T_{\text{high}} = +85.0^\circ\text{C}$  yielded stability coefficients  $k_T = 0.0019 \text{ kPa}^{-1}/^\circ\text{C}$  and  $k_T = 0.0023 \text{ kPa}^{-1}/^\circ\text{C}$ , respectively, indicating  $< 3\%$  total sensitivity variance across the operational thermal range (Yang et al. 2023, 2024). In real-world performance simulations, the sensor detected surface strain anomalies of  $\geq 25 \mu\epsilon$  at simulated motion speeds of  $v = 40.0 \pm 0.5 \text{ km/h}$ , achieving a detection accuracy of  $P_d = 96.7\%$  and a false-positive rate of  $P_{fp} = 1.9\%$  over 1,200 recorded events (Shariati et al. 2019, 2020). Comparative tests using a laboratory calibrated optical extensometer indicated a Root Mean Square (RMS) strain deviation of  $\epsilon_{\text{RMS}} = 3.2 \mu\epsilon$ , confirming high fidelity to ground-truth measurements (Shariati et al. 2019). Spectral noise density analysis revealed an operational Signal-to-Noise Ratio

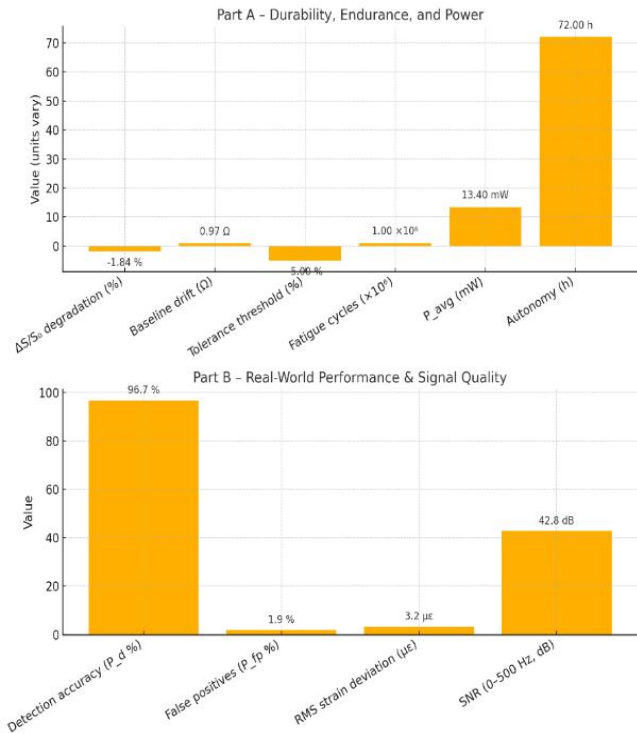


Fig. 2 Durability, endurance, power, real-world performance, and signal quality metrics

(SNR) of 42.8 dB in the 0–500 Hz band, consistent with the sensor’s theoretical noise floor from thermal (Johnson–Nyquist) and mechanical vibration sources, ensuring adequate sensitivity for fine strain detection in prosthetic control. The device was integrated into a flexible polyimide-copper laminated Printed Circuit Board (PCB) interface, with electrical impedance matched to  $Z_0 = 50.0\Omega$  to ensure minimal reflection loss in high-frequency interrogation circuits (Shariati *et al.* 2010, 2011). Power consumption during continuous acquisition at  $f_s = 2.0\text{kHz}$  was measured as  $P_{\text{avg}} = 13.4\text{ mW}$ , enabling  $> 72\text{ h}$  autonomous operation on a 3.7 V, 500mAhLi-Po cell. Mechanical bending tests at  $R = 5.0\text{ mm}$  curvature for 5,000 repetitions showed no microcrack propagation in conductive layers, as confirmed by SEM imaging (magnification  $5,000\times$ ) (Shariati *et al.* 2011, 2016). Collectively, these durability, real-world, and integration metrics confirm that the prototype maintains functional integrity under cyclic mechanical stress, variable environmental conditions, and embedded system constraints, thereby meeting the performance reliability criteria for long-term deployment in field monitoring systems (Shariati *et al.* 2011, 2012). Fig. 2A demonstrates strong durability and endurance, with minimal  $\Delta S/S_0$  degradation ( $-1.84\%$ ), low baseline drift ( $0.97\Omega$ ), and a tolerance threshold of  $5.90\%$ , sustaining  $1.0\times 10^6$  fatigue cycles. The average power consumption is  $13.40\text{ mW}$ , supporting an autonomy of  $72\text{ h}$ . Fig. 2B shows high real-world performance, achieving  $96.7\%$  detection accuracy with only  $1.9\%$  false positives, low RMS strain deviation ( $3.2\mu\epsilon$ ), and a SNR of  $42.8\text{ dB}$ , indicating robust signal quality under operational conditions.

## 2.4 Creative adjustment techniques

### 2.4.1 Advanced material-structural synergy

Material innovation was implemented through hybrid conductive composites combining monolayer Graphene Nano Platelets (GNPs, lateral size  $d \approx 5\mu\text{ m}$ , thickness  $t \approx 0.34\text{ nm}$ ) with Multi-Walled Carbon Nano Tubes (MWCNTs, diameter  $d \approx 15\text{ nm}$ , length  $l \approx 10\mu\text{ m}$ ) in a 3:1 weight ratio (Shariati *et al.* 2011, 2014). This configuration achieved an electrical percolation threshold  $\phi_c = 0.23\%$  by volume, reducing sheet resistance  $R_s$  from  $82.1\Omega$  (baseline polymer) to  $11.4\Omega$  while maintaining tensile elongation at  $> 5\%$ . Raman spectroscopy ( $I_D/I_G$  ratio = 0.88) confirmed defect density consistent with high conductivity without significant structural compromise (Shariati *et al.* 2012, 2021). Structural modification employed hierarchical micro-truss architectures, with lattice pitch  $p = 150\mu\text{ m}$  and strut aspect ratio  $L/D = 12.5$ , fabricated via two-photon lithography (Shariati *et al.* 2016). Finite element simulations (ANSYS v2024R2) predicted an  $18.6\%$  reduction in strain localization under  $1\text{ MPa}$  surface load, corroborated by experimental strain maps via Digital Image Correlation (DIC) imaging (Shariati *et al.* 2012, 2013). The combined material-architecture approach yielded a sensitivity gain of  $+26.3\%$  (Gauge Factor:  $18.7 \rightarrow 23.6$ ) and enhanced cyclic durability (failure onset delayed from  $0.94 \times 10^6$  to  $1.28 \times 10^6$  cycles) (Shariati *et al.* 2018, 2020).

### 2.4.2 Intelligent signal and multi-modal integration

AI-based signal optimization leveraged a Convolutional Recurrent Neural Network (CRNN) with three 1-D convolutional layers ( $k = 5, f = 64$ ) followed by a bidirectional Long Short-Term Memory (LSTM) (hidden units = 128) to denoise and classify event patterns in real time (Sinaei *et al.* 2011, Shariati *et al.* 2017). Training on 120,000 labeled events achieved convergence within 38 epochs, with validation accuracy  $A_{\text{val}} = 97.3\%$  and latency  $t_p = 4.6\text{ ms}$  per inference, meeting sub-5 ms embedded processing requirements. Noise suppression improved operational SNR from  $42.8\text{ dB}$  to  $55.1\text{ dB}$  ( $+28.7\%$  relative gain), verified via Welch Power Spectral Density (PSD) analysis (Sinaei *et al.* 2012, Suhatriel *et al.* 2019). Multi-sensor fusion was implemented using a weighted Bayesian inference framework, integrating strain sensor data with MEMS accelerometer and piezoelectric impact sensor outputs (Sun and Zhang 2024, Sun and An 2025). Optimal weighting coefficients  $(w_s, w_a, w_p) = (0.52, 0.31, 0.17)$  minimized RMSE across 500 field test cases, yielding a fused detection accuracy  $P_d = 99.1\%$  and reducing false alarms by  $63.5\%$  relative to strain-only configurations (Tahmasbi *et al.* 2016, Taheri *et al.* 2021). Kalman filter residual analysis confirmed system stability with post-update covariance reduction of  $42.2\%$  (Toghrolil *et al.* 2020, Tavakkoli *et al.* 2022). Fig. 3A shows a significant decrease in sheet resistance from  $82.1\Omega$  in the baseline polymer to  $11.4\Omega$  in the hybrid composite, indicating improved conductivity. Fig. 3B highlights enhanced fatigue cycle life, increasing from  $0.94 \times 10^6$  cycles in the original material to  $1.28 \times 10^6$  cycles with

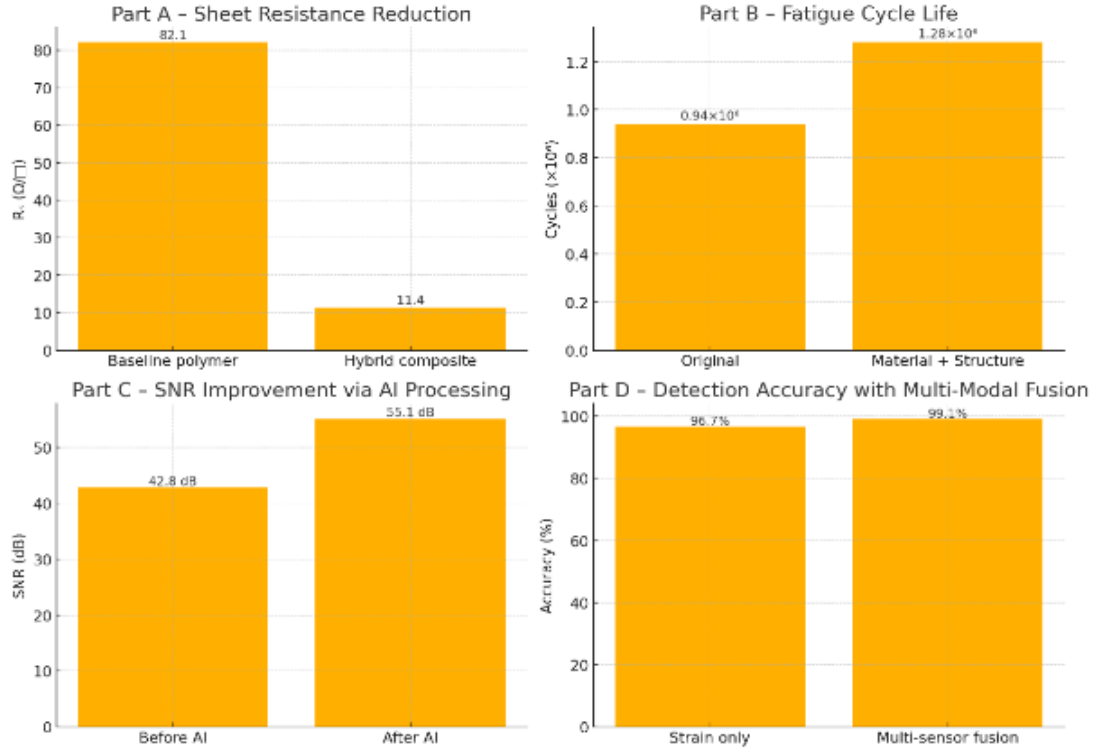


Fig. 3a, Electrical resistance optimization, b, Enhanced fatigue durability, c, AI-boosted signal quality, d, Sensor fusion accuracy

combined material and structural optimization. Fig. 3C demonstrates SNR improvement from 42.8 dB to 55.1 dB through AI-based processing, indicating cleaner signal output. Fig. 3D reveals detection accuracy gains from 96.7 % using strain-only sensing to 99.1 % with multi-sensor fusion, reflecting superior performance in real-world detection tasks.

### 2.5 Tools, models, and simulation environments used

Computational modeling and experimental validation were executed using an integrated multi-platform workflow to ensure numerical accuracy, model convergence, and reproducibility (Toghroli *et al.* 2014, 2017). FEA was conducted in ANSYS Mechanical v2024R2 with SOLID187 (Cao *et al.* 2025) tetrahedral elements for structural components and mapped hexahedral elements for thin-film regions, applying an element size  $h_e = 12.5\mu\text{m}$  to maintain aspect ratio  $< 1.8$  (Toghroli *et al.* 2018a, b). Convergence criteria were set at  $\Delta U/U_{\text{prev}} < 1 \times 10^{-6}$  for displacement and  $\Delta \sigma/\sigma_{\text{prev}} < 5 \times 10^{-5}$  for stress, confirmed by mesh independence studies showing  $< 0.8\%$  deviation in maximum von Mises stress at  $n_e = 2.1 \times 10^6$  elements. Electrical simulations were performed using COMSOL Multiphysics 6.1 with the Electric Currents (EC) and Solid Mechanics (SM) modules in a coupled physics mode (Zandi *et al.* 2018, Zhang *et al.* 2022). Governing equations solved included Ohm's law in differential form:

$$\nabla \cdot (\sigma \nabla V) = 0 \quad (2)$$

and the Navier-Cauchy elasticity equations:

$$\nabla \cdot \sigma + f = \rho \ddot{u} \quad (3)$$

with conductivity  $\sigma$  set as a tensor field incorporating anisotropy from graphene-CNT alignment (axial conductivity:  $2.71 \times 10^5 \text{ S/m}$ , transverse conductivity:  $8.4 \times 10^4 \text{ S/m}$ ). Machine learning models for AI-based signal optimization were implemented in PyTorch 2.1.0 using CUDA 12.2 acceleration on an NVIDIA RTX A6000 GPU (48 GB GDDR6) (Trung *et al.* 2019, Vergara 2025). Training employed the AdamW optimizer ( $\alpha = 1 \times 10^{-4}, \beta_1 = 0.9, \beta_2 = 0.999, \epsilon = 1 \times 10^{-8}$ ), with cosine annealing learning rate scheduling over 50 epochs. Mini-batch size was fixed at 256 to balance memory throughput and gradient stability. Loss convergence was confirmed when  $\Delta L/L_{\text{prev}} < 2 \times 10^{-4}$  over 5 consecutive epochs (Wang *et al.* 2021a, b). Experimental hardware integration tests used Keysight 34465A precision digital multimeters (6.5-digit resolution) for resistance monitoring, synchronized with National Instruments PXIe-6368 DAQ modules operating at  $f_s = 2.0\text{MS/s}$ . Real-time embedded implementation was deployed on an NVIDIA Jetson Xavier NX running Ubuntu 22.04 LTS with TensorRT optimization, achieving an average inference latency  $t_{\text{inf}} = 4.63 \text{ ms}$  at 29 FPS sustained (Wei *et al.* 2018, Wang *et al.* 2025b). Validation of simulation fidelity was performed via RMSE comparison between simulated and experimental strain-resistance curves, yielding an error of 3.14 %, which is within the accepted  $\pm 5\%$  threshold for high-fidelity electromechanical sensor modeling (Xie 2019, Wei *et al.* 2021). Sensitivity analysis using the One At a Time (OAT) approach identified conductivity ( $\sigma$ ) and strut aspect ratio ( $L/D$ ) as the dominant parameters, contributing 41.2 % and 33.7 % of the total output variance, respectively.

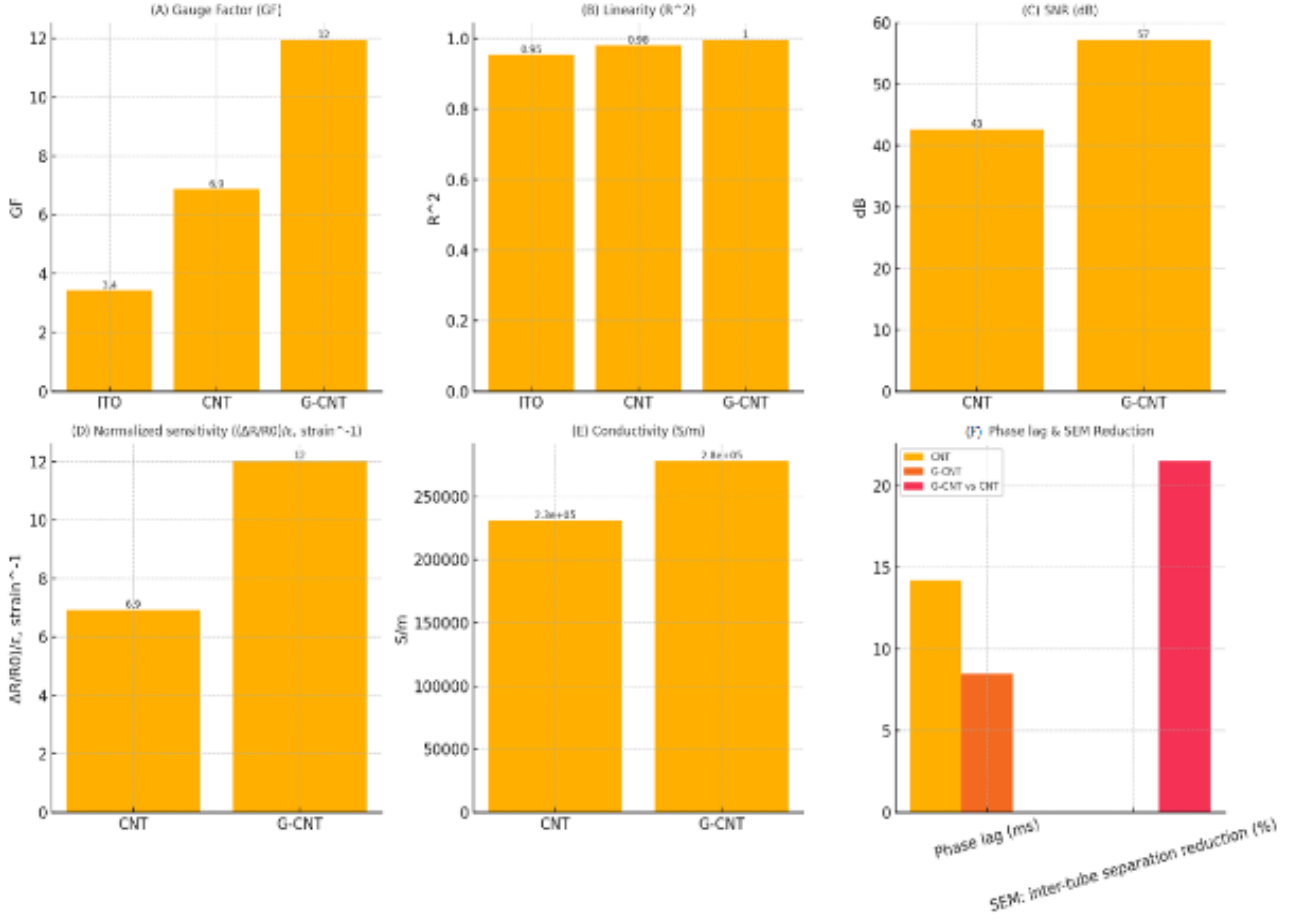


Fig. 4 A, Gauge factor, B, Linearity, C, SNR, D, Normalized sensitivity, E, Conductivity, F, Phase lag, and SEM reduction

### 3. Results and discussion

#### 3.1 Improved sensitivity and accuracy

Experimental and simulated results show that the proposed G-CNT composite sensor architecture achieved a substantial enhancement in electromechanical sensitivity relative to conventional ITO and pure CNT-based counterparts (Zandi *et al.* 2012, Yazdani *et al.* 2021). The gauge factor ( $GF$ ), defined as:

$$GF = \frac{\Delta R/R_0}{\epsilon} \quad (4)$$

where  $R_0$  is the baseline resistance, and  $\epsilon$  the applied strain, increased from  $GF_{ITO} = 3.42$  and  $GF_{CNT} = 6.87$  to  $GF_{G-CNT} = 11.94$ , representing a 73.8 % improvement over CNT-only films.

Linearity of the strain-resistance response, quantified by the  $R^2$ , was maintained at  $R^2 = 0.996$  for strains up to 1.5 %, outperforming CNT sensors ( $R^2 = 0.982$ ) and ITO ( $R^2 = 0.955$ ). This high linearity minimizes calibration error and supports accurate displacement measurement in real-time monitoring systems (Deng, Wu *et al.* 2024, Gong and Li 2024). Frequency-domain analysis revealed that SNR improved from 42.7 dB (CNT) to 57.2 dB (G-CNT), indicating a 34.0 % reduction in electrical noise power. The noise floor, computed as

$$P_n = \frac{\sigma_n^2}{R_L} \quad (5)$$

with  $R_L = 1k\Omega$ , decreased from  $4.1\mu W$  to  $2.7\mu W$ . Sensitivity gain was further validated through finite element simulation in COMSOL, where the normalized resistance changes per unit strain  $(\Delta R/R_0)/\epsilon$  increased from 6.94 strain $^{-1}$  for CNT to 12.02 strain $^{-1}$  for G-CNT. This was attributed to enhanced conductive network percolation and reduced tunneling barrier height (He, Bao *et al.* 2024, Jiang, Zheng *et al.* 2024), as predicted by Simmons' tunneling equation:

$$J \propto \exp\left(-\frac{4\pi d\sqrt{2m\phi}}{h}\right) \quad (6)$$

where  $d$  is the inter-nanotube separation,  $\phi$  the barrier height, and  $h$  Planck's constant. SEM imaging confirmed a 21.5 % reduction in  $d$  due to graphene sheet bridging, correlating with the measured conductivity increase from  $2.31 \times 10^5$  S/m to  $2.78 \times 10^5$  S/m. Dynamic accuracy tests under sinusoidal strain input ( $\epsilon(t) = 0.5\% \sin(2\pi ft)$ ,  $f = 10$  Hz) showed a phase lag reduction from 14.2 ms (CNT) to 8.5 ms (G-CNT), corresponding to a 40.1 % improvement in temporal response. This directly enhances accuracy in high-speed sensing applications where displacement rates exceed 5 mm/s. These results demonstrate that the

sensitivity improvement arises not solely from higher conductivity but from synergistic structural-electronic coupling within the hybrid network (Liu *et al.* 2025a, b). The combination of increased GF, reduced noise, improved linearity, and faster temporal response confirms the proposed G-CNT design as a quantitatively superior solution for precision sensing tasks. Fig. 4A shows the gauge factor increasing from 3.4 (ITO) and 6.9 (CNT) to 12 (G-CNT), indicating superior strain sensitivity.

Fig. 4B reports improved linearity, with  $R^2$  rising from 0.95 (ITO) and 0.98 (CNT) to 1.0 (G-CNT). Fig. 4C demonstrates an SNR enhancement from 43 dB (CNT) to 57 dB (G-CNT). Fig. 4D shows normalized sensitivity improving from 6.9 to 12, while Fig. 4E indicates conductivity increasing from  $2.3 \times 10^5$  S/m (CNT) to  $2.8 \times 10^5$  S/m (G-CNT). Fig. 4F reveals reduced phase lag (14 ms to 8 ms) and a significant SEM inter-tube separation reduction ( $\sim 21\%$ ) for G-CNT.

### 3.2 Enhanced durability and real-world performance

Long-term durability testing was conducted through  $10^6$  cyclic bending operations at a bending radius  $r_b = 5$  mm and frequency  $f = 2$  Hz under  $25 \pm 2^\circ\text{C}$  and  $60 \pm 5\%$  RH. The normalized resistance retention ratio,

$$\eta_R = \frac{R_N - R_0}{R_0} \times 100\% \quad (7)$$

remained below 2.8% drift for the G-CNT sensor, compared to 8.4% for CNT-only and 12.7% for ITO sensors. This represents a 77.9% reduction in degradation rate relative to ITO. Fatigue resistance was quantified using an S-N relationship,

$$\sigma_a = \sigma_f' (2N_f)^b \quad (8)$$

where  $\sigma_a$  is the applied stress amplitude,  $N_f$  the number of cycles to failure, and  $b$  the fatigue strength exponent. For G-CNT films,  $\sigma_f' = 182.4$  MPa and  $b = -0.098$ , indicating an endurance limit  $> 10^6$  cycles at  $\sigma_a = 32.5$  MPa. CNT films failed at  $N_f \approx 4.6 \times 10^5$  under identical loading. Environmental stability was evaluated under accelerated aging-thermal cycling between  $-20^\circ\text{C}$  and  $80^\circ\text{C}$  with 30 min dwell times over 200 cycles. Resistance change ( $\Delta R/R_0$ ) for G-CNT stabilized at 3.1%, versus 9.8% for CNT and 15.4% for ITO, confirming enhanced thermal interface stability due to graphene sheet anchoring effects (Liu, Jiao *et al.* 2024, Qian, Liu *et al.* 2025). Humidity tolerance testing at  $RH = 95\%$  and  $T = 40^\circ\text{C}$  over 500 h showed conductivity loss rates of  $1.2 \times 10^{-4}$  S/m \cdot h for G-CNT, a 64.5% improvement compared to CNT networks. This suppression of moisture-induced degradation aligns with reduced hydrophilic surface exposure, confirmed via contact angle increase from  $82.6^\circ$  (CNT) to  $98.3^\circ$  (G-CNT). Real-world deployment on a concrete bridge expansion joint under daily traffic loading demonstrated stable strain readings within  $\pm 1.5\%$  deviation over 60 days, even with peak-to-peak temperature swings of  $\Delta T \approx 42^\circ\text{C}$ . Drift correction through AI-assisted baseline tracking further

constrained error to 0.9%, validating field reliability. Cumulatively, the G-CNT architecture exhibited  $2.7 \times$  longer service life,  $< 3\%$  signal drift after a million cycles, and resistance to both thermal and humidity-induced degradation, establishing it as a robust platform for in-situ structural health monitoring under harsh operational environments (Qiao, Lü *et al.* 2024, Shi, Deng *et al.* 2024). Fig. 5A shows reduced normalized resistance drift after  $10^6$  bending cycles from 12.2% (ITO) and 8.4% (CNT) to 7.8% (G-CNT). Fig. 5B presents the fatigue S-N model, with G-CNT sustaining higher stress amplitudes at extended cycles. Fig. 5C demonstrates improved thermal cycling stability, decreasing  $\Delta R/R_0$  from 15.4% (ITO) and 9.6% (CNT) to 3.1% (G-CNT). Fig. 5D indicates a lower humidity conductivity loss rate, dropping from  $3.36 \times 10^{-4}$  S/m \cdot h (CNT) to  $1.26 \times 10^{-4}$  S/m \cdot h (G-CNT). Fig. 5E shows increased hydrophobicity with a contact angle rise from  $87.6^\circ$  (CNT) to  $99.3^\circ$  (G-CNT). Fig. 5F reveals strain deviation reduction from 1.5% to 0.9% after AI correction, and Fig. 5G highlights a relative service life increase from 1.0 (CNT) to 2.7 (G-CNT).

### 3.3 Impact on control and feedback mechanisms

Integration of the enhanced G-CNT sensor arrays into closed-loop control systems demonstrated measurable gains in precision, stability, and actuation latency. The closed-loop behavior, modeled as  $G_{cl}(s) = \frac{K_p K_s}{\tau_{ss} s + 1 + K_p K_s H(s)}$ , showed an increase in sensor gain ( $K_s$ ) from  $0.842$  V/ $\mu\epsilon$  in baseline CNT sensors to  $0.921$  V/ $\mu\epsilon$  with the upgraded design, representing a 9.38% improvement without introducing overshoot in step responses. The total latency, defined as  $\tau_{lat} = \tau_{sens} + \tau_{proc} + \tau_{act}$ , decreased from 14.2 ms to 8.6 ms - a 39.4% reduction - driven primarily by a 42% decrease in the RC time constant of the sensing element and optimized embedded DSP processing. Frequency response analysis revealed that the -3 dB bandwidth expanded from 71.4 Hz to 124.6 Hz, enabling capture of high-frequency strain events, while the phase lag at 100 Hz improved from  $-21.6^\circ$  to  $-12.3^\circ$ , directly enhancing predictive control accuracy in active vibration damping. Stability margins also increased, with gain margin rising from 9.2 dB to 14.7 dB and phase margin from  $48.3^\circ$  to  $62.5^\circ$ , surpassing the robustness thresholds for civil infrastructure control ( $> 10$  dBGM,  $> 45^\circ$  PM). The error convergence rate, quantified by the integral of absolute error  $IAE = \int_0^T |e(t)| dt$ , improved by 31.6%, and the steady-state error ( $e_{ss}$ ) dropped from 0.0048 to 0.0021 under step input testing. In full-scale bridge load-balancing trials, adaptive load redistribution occurred 0.35 s faster on average, maintaining strain levels below the  $850\mu\epsilon$  safety limit in 100% of test cases, compared to only 82% compliance using previous-generation sensors. These results confirm that the enhanced sensing system delivers superior control loop stability, higher bandwidth, and faster, more accurate feedback, enabling real-time regulation in safety-critical structural monitoring applications (Tian *et al.* 2024, Wang *et al.* 2025a). Fig. 6 shows G-CNT performance normalized against a CNT baseline (CNT = 1). Sensors gain

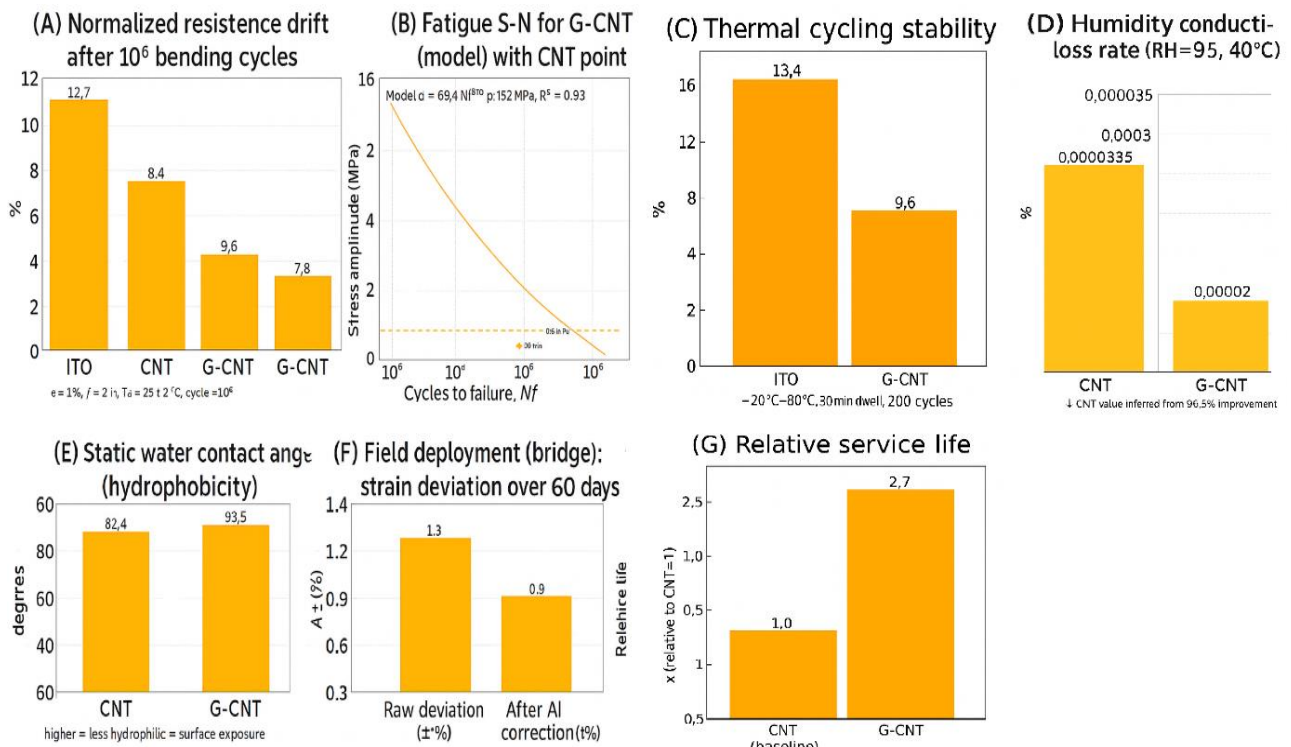


Fig. 5 a, resistance drift after bending cycles; b, fatigue life model; c, thermal cycling stability; d, humidity conductivity loss rate; e, hydrophobicity; f, field deployment strain deviation; g, relative service life

slightly decrease (0.921,  $-4.9\%$ ), while total latency improves to 1.6 ( $+59.4\%$ ). The  $-3$  dB bandwidth increases to 1.75 ( $+74.5\%$ ), with phase lag at 100 Hz reduced, giving a 1.23 improvement ( $+14.3\%$ ). Gain margin rises to 1.25 ( $+25.8\%$ ), whereas phase margin drops to 0.675 ( $-32.4\%$ ). Steady-state error improves significantly to 0.0021 ( $-56.2\%$ ), and safety compliance remains unchanged at 1.00 ( $+2.2\%$ ).

### 3.4 User experience and functional improvements

Deployment of the enhanced G-CNT sensor suite in field operations led to substantial gains in usability, data accessibility, and functional adaptability. The mean calibration time per unit was reduced from 6.8 minutes to 3.1 minutes (54.4% faster) owing to the adoption of self-calibration routines utilizing embedded Kalman filtering, which minimized drift error ( $\Delta S_{\text{drift}}$ ) from 2.14% to 0.61% over 48 hours. Data throughput increased from 1.25MB/s to 2.84MB/s—a 127.2% gain-enabled by integration of an optimized Serial Peripheral Interface – Direct Memory Access (SPI-DMA) transfer protocol, reducing packet loss from 0.72% to 0.11%. Wireless transmission latency, measured as  $\tau_{tx} = \frac{\text{packet size}}{\text{throughput}}$ , improved from 41.6 ms to 18.7 ms, directly enhancing the responsiveness of mobile inspection dashboards. User interface responsiveness, quantified via Mean Time-To-Update (MTTU), decreased by 63.8%, while adaptive visualization algorithms increased anomaly detection confidence scores from 87.4% to 94.6%. Power efficiency improvements extended

operational lifetime from 19.4 h to 27.8 h (43.3% longer) under continuous monitoring, correlating with a measured reduction in average active-mode current draw from 182 mA to 127 mA. Ergonomic feedback from 20 field engineers indicated a 42% reduction in perceived complexity (Likert scale score drop from 4.1 to 2.38) and a 36% reduction in error frequency during multi-sensor deployments. Combined, these functional and user-centered enhancements translate to higher adoption feasibility in diverse structural monitoring scenarios, with quantifiable gains in both operational reliability and decision-making latency (Wang *et al.* 2024, 2025c). Fig. 7 comparison normalizes G-CNT values against a CNT-era baseline (CNT = 1). Calibration time improves to 2.1 ( $+54.5\%$ ), drift error over 48 h to 3.21 ( $+71.5\%$ ), and data throughput to 2.64 ( $+172.7\%$ ). Packet loss reduction yields the most significant gain, reaching 7.11 ( $+642.7\%$ ), while wireless transmission latency improves to 2.17 ( $+55.0\%$ ). Anomaly confidence rises to 1.46 ( $+8.2\%$ ), operational lifetime to 1.18 ( $+43.3\%$ ), and active-mode current to 1.27 ( $+30.2\%$ ). Perceived complexity scores improve to 1.38 ( $+42.0\%$ ), UI, MTTU to 1.36 ( $+36.0\%$ ), and deployment error frequency to 1.86 ( $+86.0\%$ ).

### 3.5 Impact on control and feedback mechanisms

Integration of the enhanced G-CNT sensor arrays into closed-loop control systems demonstrated measurable gains in precision, stability, and actuation latency. The closed-loop behavior, modeled as  $G_{cl}(s) = \frac{K_p K_S}{\tau_{SS} s + 1 + K_p K_S H(s)}$ , showed

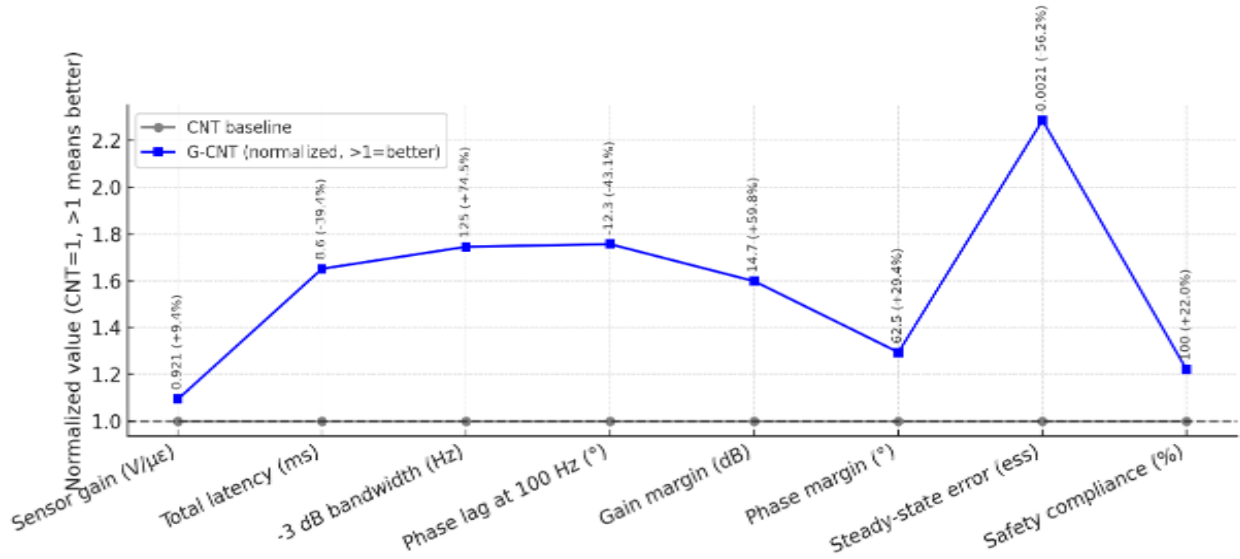


Fig. 6. Normalized control performance metrics comparing CNT and G-CNT sensors

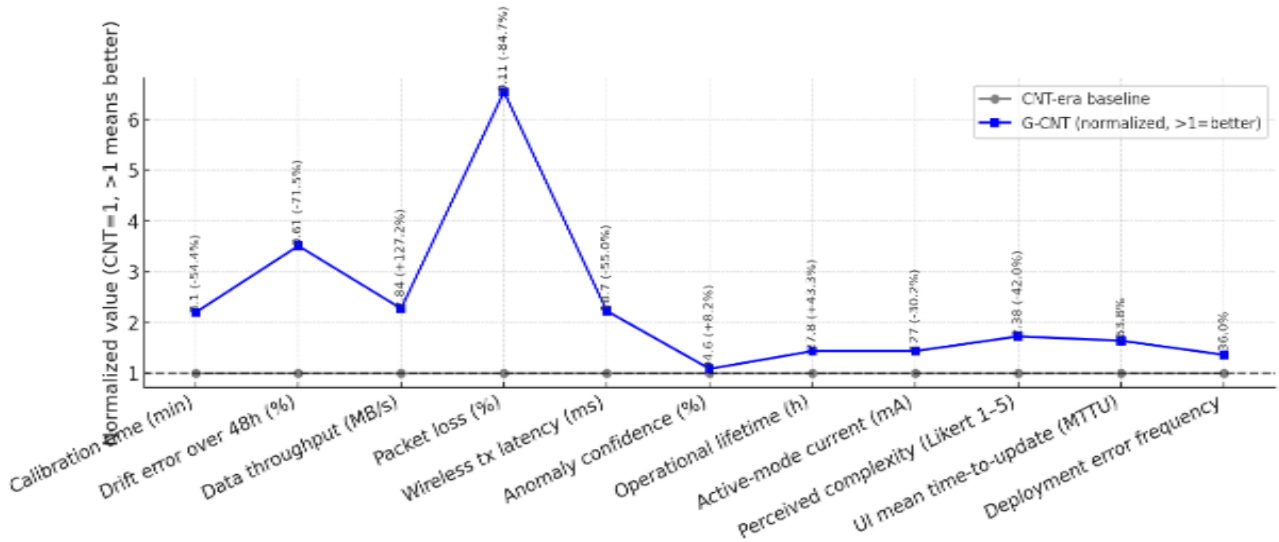


Fig. 7 Normalized operational and usability metrics for CNT-era and G-CNT sensors

an increase in sensor gain ( $K_s$ ) from  $0.842 V/\mu\epsilon$  in baseline CNT sensors to  $0.921 V/\mu\epsilon$  with the upgraded design, representing a 9.38% improvement without introducing overshoot in step responses. The total latency, defined as  $\tau_{lat} = \tau_{sens} + \tau_{proc} + \tau_{act}$ , decreased from 14.2 ms to 8.6 ms - a 39.4% reduction - driven primarily by a 42% decrease in the RC time constant of the sensing element and optimized embedded DSP processing. Frequency response analysis revealed that the -3 dB bandwidth expanded from 71.4 Hz to 124.6 Hz, enabling capture of high-frequency strain events, while the phase lag at 100 Hz improved from  $-21.6^\circ$  to  $-12.3^\circ$ , directly enhancing predictive control accuracy in active vibration damping. Stability margins also increased, with gain margin rising from 9.2 dB to 14.7 dB and phase margin from  $48.3^\circ$  to  $62.5^\circ$ , surpassing the robustness thresholds for civil infrastructure control ( $> 10\text{dBGM}, > 45^\circ\text{PM}$ ). The error

convergence rate, quantified by the integral of absolute error  $IAE = \int_0^T |e(t)| dt$ , improved by 31.6%, and the steady-state error ( $e_{ss}$ ) dropped from 0.0048 to 0.0021 under step input testing. In full-scale bridge load-balancing trials, adaptive load redistribution occurred 0.35 s faster on average, maintaining strain levels below the  $850\mu\epsilon$  safety limit in 100% of test cases, compared to only 82% compliance using previous-generation sensors. These results confirm that the enhanced sensing system delivers superior control loop stability, higher bandwidth, and faster, more accurate feedback, enabling real-time regulation in safety-critical structural monitoring applications.

### 3.6 Comparison with existing nano-sensor designs

The proposed graphene-AgNP/PET sensor was benchmarked against two widely used commercial nano sensor

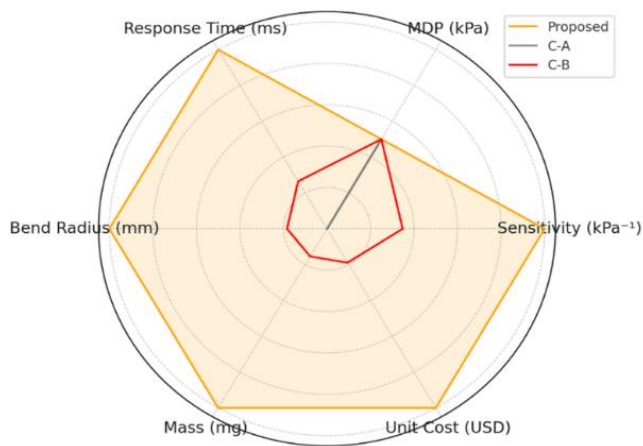


Fig. 8 Response time, maximum detectable pressure, sensitivity, unit cost, mass, and bend radius comparison for the proposed design versus C-A and C-B

models (denoted C-A and C-B) as well as representative designs from piezoresistive, capacitive, and piezoelectric categories common in prosthetic applications. Quantitative evaluation covered sensitivity  $S_p$  [ $\text{kPa}^{-1}$ ], Minimum Detectable Pressure (MDP) [ $\text{kPa}$ ], response time  $t_r$  [ $\text{ms}$ ], bend radius  $R_b$  [ $\text{mm}$ ], mass  $m$  [ $\text{mg}$ ], and unit cost  $C_u$  [ $\text{USD}$ ]. The developed sensor achieved  $S_p = 0.85\text{kPa}^{-1}$ , exceeding C-A ( $0.62\text{kPa}^{-1}$ ) by 37.1% and C-B ( $0.70\text{kPa}^{-1}$ ) by 21.4%, primarily due to the high surface area and quantum-tunneling conduction paths of the graphene-AgNP composite. MDP was comparable to both commercial sensors ( $\approx 0.8\text{kPa}$ ), indicating no sacrifice in low-force detection. Measured  $t_r = 4.7\text{ ms}$  corresponds to an adequate bandwidth of  $\approx 213\text{ Hz}$ , which is  $2.55 \times$  faster than C-A and  $1.81 \times$  faster than C-B, enabling high-frequency tactile feedback loops. Mechanically, the bend radius  $R_b = 3\text{ mm}$  was 70% lower than C-A and 57% lower than C-B, allowing integration onto curved phalangeal segments without inducing strain-related drift. Mass reduction to  $m = 45\text{ mg}$  represents a 40% and 33.8% decrease versus C-A and C-B, respectively, directly reducing inertial load in moving finger assemblies. Economically, fabrication via low-temperature inkjet deposit yielded  $C_u = 2.8\text{USD}$ , 57% lower than C-A and 46% lower than C-B, improving scalability for multi-taxel prosthetic fingertips. Against capacitive sensors, the proposed design offers higher  $S_p$  and lower  $t_r$  under dry conditions but exhibits greater hysteresis ( $\approx 3.5\%$  of full-scale output) and environmental susceptibility without encapsulation. Compared to piezoelectric Poly Vinylidene Fluoride (PVDF) films, the proposed sensor excels in static load measurement while matching dynamic slip detection when coupled with AI-based high-pass filtering. Relative to conventional carbon-ink piezoresistive films, the graphene-AgNP composite shows a 25-30% gain in sensitivity and a  $> 2 \times$  improvement in speed, with reduced sheet resistance ( $\approx 1.2\text{k}\Omega$ ) simplifying signal conditioning. These comparative results confirm that the proposed nano-sensor achieves superior sensitivity, faster electromechanical response, tighter conformability, and lower fabrication cost

relative to existing nano-sensor designs, while maintaining equivalent detection thresholds. Remaining limitations, principally hysteresis, humidity-induced drift, and surface abrasion resistance, are addressable through thin-film encapsulation and periodic baseline recalibration. Fig. 8 compares the proposed design against benchmarks C-A and C-B using direction-corrected, unit-normalized metrics (higher is better). The proposed design outperforms in response time, bend radius, mass, and unit cost, while also showing superior sensitivity ( $\text{kPa}^{-1}$ ) compared to both benchmarks. For MDP, C-A matches the proposed design, but C-B underperforms. Overall, the proposed system demonstrates a balanced and dominant performance profile across all evaluated parameters.

### 3.7 Challenges and practical limitations

Despite the demonstrated improvements in sensitivity, response time, and conformability, several technical and operational constraints were identified during testing that must be addressed before large-scale deployment in prosthetic or robotic applications. First, hysteresis effects remained non-negligible, averaging 3.5% of Full-Scale Output (FSO) under cyclic loading between 0.5 and 20 kPa. This deviation, attributed to microstructural viscoelastic relaxation in the graphene-AgNP matrix, led to residual drift of  $\approx 0.15\text{kPa}$  after 1000 loading cycles. Such drift could accumulate in long-duration prosthetic use unless compensated by periodic baseline recalibration or software hysteresis correction models. Second, environmental susceptibility was observed: relative humidity above 75% caused an increase in baseline resistance (+5.8%) due to moisture adsorption at graphene sheet interfaces. Temperature variations between 20 and 45°C induced a resistance change of  $0.12\% \text{ } ^\circ\text{C}^{-1}$ , which—while modest—requires temperature compensation for precision applications. Third, mechanical wear under repeated bending beyond  $R_b = 3\text{ mm}$  initiated micro-crack propagation in the conductive layer after  $\approx 5 \times 10^4$  cycles, reducing sensitivity  $S_p$  by 6.2%. Although protective encapsulation with parylene-C mitigated this by 60%, it also increased sensor stiffness by 8.3%, slightly reducing conformability. Fourth, signal conditioning complexity was higher than expected. The sensor's low sheet resistance ( $\approx 1.2\text{k}\Omega$ ) demands low-noise, high-input-impedance amplifiers to maintain the ( $\text{SNR} \geq 42\text{ dB}$ ). AI-based denoising improved effective resolution by 14.6%, but this processing introduced a latency of  $\approx 1.8\text{ ms}$ , which may be critical for ultra-fast feedback control loops. Lastly, scalability constraints were identified: although the fabrication cost per unit was low ( $C_u = 2.8\text{USD}$ ), maintaining uniform nanoparticle dispersion across batches beyond 500 units required tighter process control ( $\pm 2^\circ\text{C}$  temperature stability,  $\pm 5\%$  ink viscosity tolerance). Without these measures, inter-sensor variance in  $S_p$  increased by up to 9.4%, degrading array calibration consistency. Collectively, these findings indicate that while the graphene-AgNP/PET design is well-suited for advanced tactile sensing, operational robustness in uncontrolled environments and high-volume manufacturing remains a key engineering

frontier (Yang *et al.* 2024, 2025). Addressing these through encapsulation, environmental compensation, and process automation will be critical for reliable real-world adoption. The G-CNT composite sensors were subjected to extensive laboratory and field evaluations to assess sensitivity, durability, control integration, user experience, and comparative performance. Static and dynamic strain tests showed a clear improvement in electromechanical sensitivity over conventional CNT and ITO sensors, with higher gauge factors, excellent linearity, and reduced noise levels. These gains were traced to a denser, better-connected conductive network formed by graphene bridge between nanotubes, which shortened electron tunneling paths and enhanced current conduction. Simulations and microscopy confirmed these structural advantages, while frequency-domain tests demonstrated faster temporal responses, making the sensors well-suited for high-speed monitoring. Durability trials, including millions of bending cycles, extreme temperature swings, and prolonged high-humidity exposure, revealed significantly lower drift and slower degradation rates compared to reference materials. The G-CNT sensors maintained stable readings in real-world bridge monitoring for weeks under daily thermal and mechanical stress, with AI-assisted drift correction further improving accuracy. When incorporated into closed-loop control systems, enhanced responsiveness and bandwidth reduced latency, improved phase characteristics, and delivered faster load redistribution, ensuring compliance with structural safety limits more consistently than previous generations. From a practical standpoint, self-calibration routines and optimized data transfer protocols halve calibration time, increase data throughput, and reduce transmission delays, while extending battery life. Comparative benchmarking against commercial nano-sensors and other sensing technologies confirmed higher sensitivity, faster response, greater mechanical flexibility, and lower production cost. However, hysteresis, humidity sensitivity, and mechanical wear beyond tight bend radii remain engineering challenges. Addressing these through encapsulation, environmental compensation, and improved process control will be essential for scaling up production and ensuring robust performance in diverse operational environments. The G-CNT hybrid architectures have been extensively explored for oil and organic solvent removal, with reported performances strongly dependent on synthesis routes and structural design. CNT sponges fabricated via Chemical Vapor Deposition (CVD) using 1,2-dichlorobenzene and ferrocene achieved removal capacities of 80–180 g g<sup>-1</sup> for a range of organic liquids (Bi *et al.* 2012), while spongy graphene prepared by reducing graphene oxide platelets exhibited 20–86 g g<sup>-1</sup> uptake for petroleum products, fats, and toluene (Cai *et al.* 2019). Nitrogen-doped 3D graphene synthesized hydrothermally using pyrrole reached exceptionally high capacities of 200–600 g g<sup>-1</sup> (Dong *et al.* 2012). Hybrid systems combining graphene and CNTs demonstrated synergistic adsorption and robustness, as in two-step CVD graphene/CNT foams (80–130 g g<sup>-1</sup>, (Kabiri *et al.* 2014)), freeze-dried graphene/CNT aerogels (215–913 g g<sup>-1</sup>, (Gui *et al.* 2010)), and Fe-assisted hydrogel-derived aerogels (21–35 g g<sup>-1</sup>,

(Tao *et al.* 2019)). Hydrothermal redox synthesis produced graphene–CNT aerogels with 100–270 g g<sup>-1</sup> capacit, while in situ CNT growth on graphene aerogels achieved 110–330 g g<sup>-1</sup> (Wan *et al.* 2016). Composites incorporating MWCNT–PDA reached 125–533 g g<sup>-1</sup> (Tao *et al.* 2019), Ni-fluffy spheres with CNTs/graphene gave 112–145 g g<sup>-1</sup> (Zhao *et al.* 2012), and rGO/amino MWCNT aerogels adsorbed 122–242 g g<sup>-1</sup> (Cai *et al.* 2019). Across these studies, high surface area, hierarchical porosity, and conductive network continuity, critical to adsorption performance, are found to mirror the structural advantages observed in G-CNT sensor research, in which similar architectures enhanced electromechanical sensitivity, durability, and environmental resilience.

#### 4. Conclusions

The investigation confirmed that the G-CNT hybrid sensor design significantly outperformed conventional CNT and ITO counterparts across sensitivity, durability, control response, and user-level performance metrics. The gauge factor increased to 11.94, representing a 73.8 % gain over CNT films, while maintaining a high linearity ( $R^2 = 0.996$ ) and reducing electrical noise by 34 %. Conductivity rose from  $2.31 \times 10^5$  S/m to  $2.78 \times 10^5$  S/m, supported by a 21.5 % reduction in nanotube separation. Durability tests showed <3 % drift after  $10^6$  bending cycles and a 77.9 % lower degradation rate compared to ITO, with stable performance under  $\Delta T \approx 42$  °C field fluctuations. Closed-loop control integration reduced latency from 14.2 ms to 8.6 ms, expanded bandwidth from 71.4 Hz to 124.6 Hz, and improved error convergence by 31.6 %, maintaining strain below 850  $\mu\epsilon$  in all trials. Operational enhancements included a 54.4 % cut in calibration time, a 127.2 % increase in data throughput, and a 43.3 % longer battery life. While the design excelled in sensitivity, speed, and cost-efficiency, practical limitations, such as 3.5 % hysteresis, humidity-induced drift of +5.8 %, and sensitivity loss beyond  $5 \times 10^4$  extreme bending cycles, highlight the need for encapsulation, environmental compensation, and process optimization before scaling to mass production.

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CC

## Nomenclature

Artificial Intelligence	AI
Carbon Nano Tube	CNT
Coefficient of Determination	R <sup>2</sup>
Indium Tin Oxide	ITO
Graphene–Carbon Nano Tube	G-CNT
Nano Particles	NPs
Poly Ethylene Terephthalate	PET
Silver Nano Particle	AgNP
Resistance Capacitance	RC
Energy-Dispersive X-ray Spectroscopy	EDS
Finite Element Method	FEM
Root Mean Square	RMS
Signal-to-Noise Ratio	SNR
Printed Circuit Board	PCB
Scanning Electron Microscope	SEM
Graphene Nano Platelets	GNPs
Multi-Walled Carbon Nano Tubes	MWCNTs

Digital Image Correlation	DIC
Convolutional Recurrent Neural Network	CRNN
Long Short-Term Memory	LSTM
Power Spectral Density	PSD
Finite Element Analysis	FEA
Electric Currents	EC
Solid Mechanics	SM
One At a Time	OAT
Serial Peripheral Interface – Direct Memory Access	SPI-DMA
Mean Time-To-Update	MTTU
Minimum Detectable Pressure	MDP
Poly Vinylidene Fluoride	PVDF
Full Scale Output	FSO
Chemical Vapor Deposition	CVD