

Review Paper on Advanced Floating Robotic System for Water Quality Monitoring

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ABSTRACT

Among all the emerging global environmental issues, the deterioration of water quality is of prime importance because it directly affects human health, aquatic ecosystems, and industrial processes. Traditional methods of manual sampling and laboratory-based analysis have intrinsic limitations due to high labor costs, low sampling frequency, and no real-time insights. As a result of recent developments within autonomous systems, IoT, and embedded sensing technologies, the development of advanced floating robotic systems with the capability of continuous automated monitoring of water bodies has been possible. These can integrate multi-parameter sensors, wireless data communication, GPS navigation, and intelligent processing units for collection and transmission of key water quality indicators like pH, turbidity, dissolved oxygen, temperature, and conductivity.

This review paper presents a comprehensive analysis of the technological evolution, design methodologies, and current state-of-the-art floating robotic platforms for water quality monitoring. The study will analyze the strengths and limitations of several sensor configurations, communication protocols, power management techniques, and robotic designs adopted in recent research. It further discusses the major challenges: environmental interference, sensor calibration issues, biofouling, power limitations, and long-term deployment constraints. By comparing the existing systems and identifying technological gaps, this paper looks into future opportunities comprising AI-based predictive analytics, low-cost sensor innovations, energy harvesting, and fully autonomous navigation. Conclusively, the study finds that floating robotic systems hold great promise for transforming real-time water quality assessment and thereby offering substantial support toward sustainable water resource management.

Keywords: Floating robotic system, Water quality monitoring, IoT sensors, Autonomous navigation, Environmental monitoring

INTRODUCTION

Water quality degradation has become one of the most critical environmental challenges due to rapid industrial growth, urban expansion, agricultural runoff, and inefficient wastewater management. Pollutants increasingly enter natural water bodies, leading to severe ecological damage, reduced biodiversity, and significant risks to human health. Traditional water-quality monitoring methods depend heavily on manual sampling and laboratory analysis, which, although accurate, suffer from several drawbacks such as delayed results, limited spatial coverage, high operational costs, and the inability to detect sudden real-time fluctuations.

Recent advancements in IoT devices, embedded controllers, autonomous robotics, and long-range communication networks have enabled the development of floating robotic platforms for automated water-quality assessment. These systems integrate sensors such as pH, dissolved oxygen, turbidity, temperature, and conductivity along with wireless technologies including GSM, LoRa, Wi-Fi, and ZigBee. Unlike fixed station-based monitoring, floating robots can navigate across water surfaces, collect distributed samples, detect spatial variations, and transmit data continuously to cloud platforms.

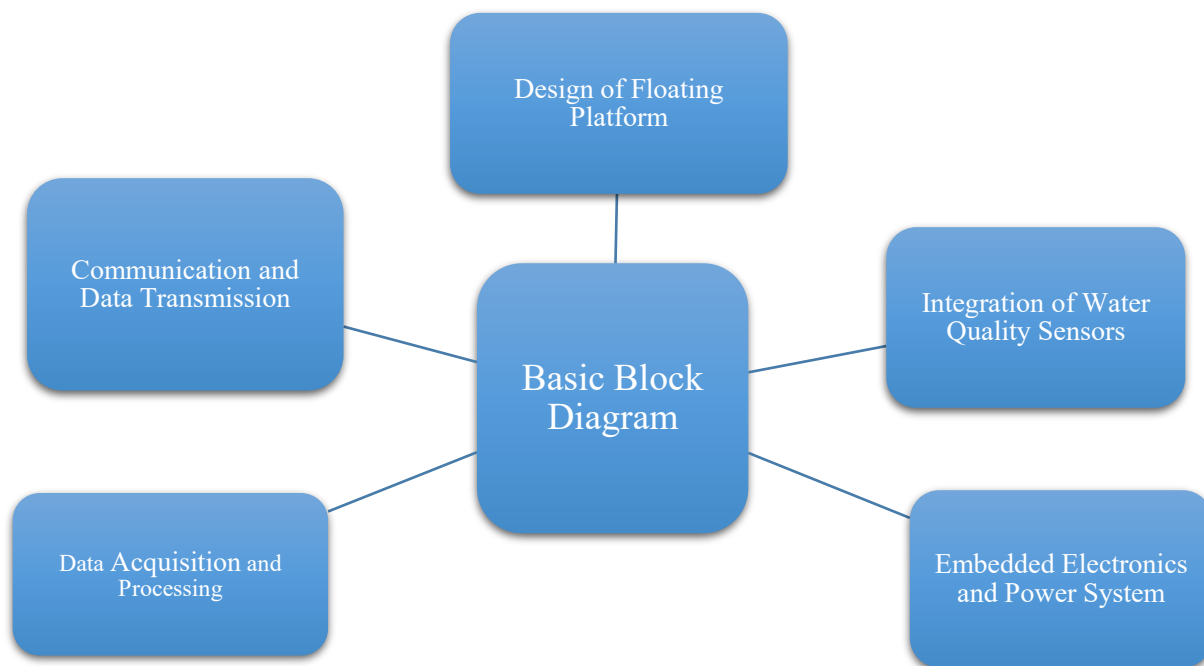
Despite these promising developments, floating robotic systems still face challenges such as sensor drift during long-term deployment, biofouling on sensor surfaces, unstable communication in remote locations, limited

battery capacity, and disruptions caused by water currents and floating debris. Moreover, the selection of appropriate sensors, power-management strategies, and reliable communication modules remains a crucial design consideration for ensuring long-term operational efficiency.

In recent years, researchers have increasingly focused on combining autonomous navigation with intelligent data processing to enhance system accuracy and reliability. Modern systems are exploring energy-harvesting techniques, multi-parameter sensor calibration, and AI-based analytics to improve predictive capabilities. Therefore, it is essential to critically examine the latest technological advancements, compare existing robotic solutions, and highlight the research gaps that need attention for developing more efficient and scalable water-monitoring platforms.

BASIC BLOCK DIAGRAM

Fig No.1 – Basic architecture of a floating robotic water quality monitoring system.



1. Design of Floating Platform

The first step is to design a buoyant, light floating structure that could operate on various water surfaces. Materials often used to make these structures buoyant and durable include PVC, acrylic sheets, HDPE, or 3Dprinted plastic. The design includes isolated chambers for electronics, batteries, sensors, and optionally solar panels. For autonomous movement, propulsion units can be included, such as DC motors or servo driven rudders.

2. Integration of Water Quality Sensors

The robotic platform is equipped with necessary sensors according to the requirements of different projects. Commonly used sensors include pH, turbidity, DO, temperature, and electrical conductivity. The sensors are housed in a dedicated chamber, wherein the water comes into direct contact with the sensors while protecting the internal electronics. Calibration using standard solutions is done to achieve accurate, real-time readings before actual deployment.

3. Embedded Electronics and Power System

It normally consists of one embedded processor or microcontroller, such as Arduino, ESP32, Raspberry Pi, or STM32, managing sensor data, controlling motors, and communicating. The power system usually consists of

rechargeable lithium batteries, sometimes complemented with solar panels for extended operation. Voltage regulators, waterproof connectors, and protective casings are used to maintain system safety and performance.

4. Data Acquisition and Processing

Sensor readings are acquired through analog or digital interfaces and then processed for filtering to remove noise or erroneous data. The conversion of raw sensor values occurs into standard measurement units, such as pH, °C, NTU, and mg/L. Some systems implement data averaging or the detection of outliers to improve the accuracy of readings. Data can be stored locally on an SD card or processed further for applications in real time.

5. Communication and Data Transmission

It involves the use of different communication technologies to transmit water quality data, depending on the deployment environment.

- Wi Fi for shortrange monitoring
- Bluetooth for local testing and device configuration
- GSM/GPRS for long range mobile connectivity SIM800, SIM900
- LoRa for low power, long distance transmission
- GPS modules for location tracking and mapping

The processed data is then sent to the cloud platforms or custom dashboards for real time visualization and analysis.

LITERATURE REVIEW

Water quality monitoring using floating robotic systems has been a fast-growing field of research during the last decade, driven by enabling technologies in IoT, autonomous navigation, and miniaturized sensing. While most early works relied on static buoy-based systems, recent studies have emphasized mobility, multi sensor integration, and real time data transmission.

An early prototype of the floating buoy system with basic sensors for pH and temperature measurement was developed by Sharma et al. 2018. The system used simple RF communication, limiting the range of operation but demonstrating the feasibility of low-cost water monitoring platforms.

Gharat et al. (2019) proposed an IoT enabled buoy system using a GSM module for water quality data transmission from a distance. Their design significantly improved communication flexibility but unfortunately lacked the possibility of navigation itself and was also capable of providing only point based measurements, thus limiting its usage in larger water bodies.

Kumar and Singh (2020) designed a solar powered, autonomous floating robot that would be capable of monitoring several parameters: pH, temperature, and dissolved oxygen. Their effort leaned toward sustainability and renewable energy, while pointing out problems concerning sensor drift over long deployment.

Rahman et al. (2020) proposed a compact floating robot with Wi Fi communication integrated with a cloud-based dashboard. Though real time visualization was excellent in their system, its use was limited to network availability areas, hence restricting its deployment in remote water bodies.

Bhardwaj & Mehta, 2020, discussed low-cost sensor networks integrated with Arduino based floating modules. The design demonstrated that even low budget hardware is able to deliver reliable results; however, accuracy decreased when the surrounding environment was either highly polluted or highly turbid, causing saturation in the sensors.

Sivaraman et al. (2021) conducted an experiment with the LoRa long range communication systems. The floating unit successfully transmitted the data over more than 3 km with low power consumption. However, the system faced difficulties in transmitting large datasets and required stable gateways for continuous operation.

Ullah et al. (2021) proposed a solar powered floating platform with automated motorized propulsion. Their research emphasized dynamic sampling, whereby the robot patrols across different zones of a lake. This approach improved the spatial coverage, but battery management and motor endurance are the key limitations of the approach.

Zhou et al. (2022) introduced a multi node collaborative robotic system wherein small floating units communicate and coordinate their actions in a mesh network. This networked approach increased data density and improved the accuracy through spatial interpolation; however, synchronization among nodes was complex and computation heavy.

Patil et al. (2021) extended the earlier systems by incorporating GPS and ZigBee communication for spatial mapping of water quality in larger areas. Although data latency and dependency on line of sight communication remained major limiting factors, this system provided higher coverage.

Chen et al. (2022) integrated AI driven analytics into floating robotic systems, bringing predictive modeling and automated anomaly detection into operation. The platform adopted machine learning algorithms to improve data interpretation but required a computational load that is provided by more powerful processors onboard, thus increasing the cost and energy consumption.

Santos et al. (2022) developed a catamaran style floating robot equipped with ultrasonic sensors that could detect obstacles and plan autonomous routes. This increased stability and mobility made it fit for dynamic river environments; however, strong currents and debris present in the environment often interfered with the accuracy of navigation.

Li and Zhang (2023) focused on long term monitoring using antibiofouling coating and self cleaning mechanisms in underwater sensors. The results indicated the increased long term lifetime of sensors at greater manufacturing costs and system complexity. Recent research, in 2023-2024, has also started embracing hybrid robotic systems that incorporate surface floating platforms together with underwater drones for multi depth monitoring. These systems provide superior vertical profiling but also have important challenges related to coordination and energy consumption, and waterproofing electronics.

Recent research (2024) has involved the integration of edge computing into floating robots to allow onboard data processing to reduce delay and bandwidth usage. These systems can provide early warnings for spikes in pollution, but with higher computational demands, stronger processors and higher capacity batteries are required.

CRITICAL ANALYSIS OF LITERATURE

A deeper evaluation of existing studies reveals several important insights beyond the descriptive summaries presented in the literature review.

1. Communication Technology Trade-offs

Different communication modules show distinct advantages and limitations. GSM-based systems offer long-range communication but lack navigation capabilities. Wi-Fi-based robots provide real-time dashboards yet are confined to limited coverage areas. LoRa systems ensure long-range and low-power operation but cannot efficiently send large datasets.

2. Power and Operational Endurance

Solar-powered platforms increase deployment duration; however, their performance decreases during cloudy weather or monsoon seasons. Robots with propulsion systems experience high power consumption, forcing researchers to compromise between mobility and energy efficiency.

3. Sensor Accuracy and Stability

Low-cost sensors commonly used in budget-friendly systems tend to lose accuracy in highly polluted or turbid environments. Long-term deployments often experience sensor drift, biofouling on sensor surfaces, and reduced reliability of readings.

4. Autonomous Navigation Limitations

Although GPS and ultrasonic sensors improve navigation, environmental challenges such as floating debris, strong currents, and wave disturbances limit precision. Reliable autonomous route planning is still underdeveloped in most studies.

5. Advanced Systems Introduce New Challenges

AI-enabled and mesh-network systems allow predictive analysis and multi-robot coordination but increase system complexity, energy demand, and overall cost.

ANALYTICAL CONCLUSION

No existing system optimally balances mobility, long-range communication, sensor durability, energy efficiency, and cost. This gap highlights the need for hybrid power solutions, improved sensor calibration techniques, and robust intelligent navigation mechanisms.

LITERATURES ANALYSIS

The comparative study shows that existing floating robotic systems offer significant advancements in communication, energy efficiency, and autonomous monitoring. However, most systems still face challenges such as limited coverage, high energy demand, complex coordination, and reduced sensor accuracy over long term deployment. Overall, the literature highlights the need for a more robust, long range, and low maintenance water monitoring solution.

Table No. 1 Comparative Result Analysis of Existing Floating Robotic Systems

Author Name & Year	Advantages	Limitations
Gharat et al. (2019)	<ul style="list-style-type: none"> • IoT enabled buoy system • GSM based long distance data transfer • Improved communication flexibility 	<ul style="list-style-type: none"> • No navigation feature • Only point based measurements • Not suitable for large water bodies
Kumar & Singh (2020)	<ul style="list-style-type: none"> • Solar powered autonomous robot • Measures pH, temperature • Promotes renewable energy use 	<ul style="list-style-type: none"> • Sensor drift over long deployment • Reduced accuracy with time
Rahman et al. (2020)	<ul style="list-style-type: none"> • Compact floating robot • Wi Fi communication • Realtime cloud dashboard 	<ul style="list-style-type: none"> • Works only in Wi Fi coverage areas • Not suitable for remote locations

Author Name & Year	Advantages	Limitations
Bhardwaj & Mehta (2020)	<ul style="list-style-type: none"> • Low cost Arduino based system • Reliable basic monitoring • Budget friendly solution 	<ul style="list-style-type: none"> • Lower accuracy in polluted water • Sensor saturation in high turbidity
Sivaraman et al. (2021)	<ul style="list-style-type: none"> • Long range LoRa communication (>3 km) • Very low power consumption • Suitable for large areas 	<ul style="list-style-type: none"> • Cannot send large datasets • Needs stable LoRa gateways
Ullah et al. (2021)	<ul style="list-style-type: none"> • Solar powered robotic platform • Motorized propulsion for dynamic sampling • Better spatial coverage 	<ul style="list-style-type: none"> • Battery management issues • Limited motor endurance
Zhou et al. (2022)	<ul style="list-style-type: none"> • Multi node mesh networking • Higher data density • Better spatial interpolation 	<ul style="list-style-type: none"> • Complex node synchronization • High computational load
Patil et al. (2021)	<ul style="list-style-type: none"> • GPS + ZigBee communication • Large area spatial mapping • Better coverage 	<ul style="list-style-type: none"> • Data latency problems • Line of sight dependency
Chen et al. (2022)	<ul style="list-style-type: none"> • AI based analytics • Predictive modelling • Automated anomaly detection 	<ul style="list-style-type: none"> • High processor requirement • Increased cost and energy usage
Santos et al. (2022)	<ul style="list-style-type: none"> • Stable catamaran design • Ultrasonic obstacle detection • Autonomous route planning 	<ul style="list-style-type: none"> • Affected by strong currents • Debris reduces navigation accuracy
Li & Zhang (2023)	<ul style="list-style-type: none"> • Antibiofouling sensor coating • Self-cleaning mechanism • Longer sensor lifespan 	<ul style="list-style-type: none"> • High manufacturing cost • Increased design complexity
Hybrid Systems (2023–2024)	<ul style="list-style-type: none"> • Surface + underwater monitoring • Multidepth water profiling • Improved vertical analysis 	<ul style="list-style-type: none"> • Coordination complexity • High energy requirement • Waterproofing challenges

Author Name & Year	Advantages	Limitations
Edge Computing (2024)	<ul style="list-style-type: none">• Onboard data processing• Reduced transmission delay• Early pollution spike detection	<ul style="list-style-type: none">• High computational demand• Requires larger battery capacity

FUTURE RESEARCH ROADMAP

The analysis highlights several promising directions for enhancing next-generation floating robotic water-monitoring systems:

1. Adaptive Sampling Techniques

Developing algorithms that dynamically adjust sampling frequency based on detected pollution changes can improve data relevance and optimize energy usage.

2. Federated and Distributed Learning

Multiple robots can collaboratively train machine-learning models without transmitting raw data, reducing bandwidth usage and allowing large-scale predictive monitoring.

3. Hybrid USV AUV Monitoring

Integrating surface robots with underwater vehicles will enable multi-depth water profiling and more comprehensive environmental analysis.

4. Anti-Biofouling & Self-Cleaning Mechanisms

Incorporating coatings, ultrasonic cleaning, or mechanical wiping can significantly extend sensor lifespan during long-term deployments.

5. Energy Harvesting Innovations

Hybrid solar–hydro or wave-powered systems can support continuous monitoring and power-consuming components like motors and AI processors.

6. TRL (Technology Readiness Level) Evaluation

Assigning TRLs to different system types can help determine their maturity level and suitability for deployment in real-world conditions.

CONCLUSION

Advanced floating robotic systems represent an efficient and reliable technological means for continuous and real time water quality monitoring. Such robotic platforms are able to cover larger areas compared with traditional manual samplings, imply less labor, and allow quicker detection of the changes in water conditions. The integration of multiparameter sensors, wireless communication, and a microcontroller-based control unit enables such systems to operate autonomously and provide accurate environmental data.

The review of existing research shows significant progress in areas such as sensor accuracy, power management, and autonomous navigation. Solar powered designs, GPS based path planning, and cloud connected dashboards have further improved usability and long-term performance in these systems. However, some challenges remain; these are sensor calibration, biofouling, harsh weather conditions, and limited battery life.

In all, floating robotic systems hold immense promise for environmental monitoring and smart water management in the future. Further developments in sensor technology, AI based data analysis, and energy efficient designs will see these systems playing a key role in early pollution detection, ecosystem protection, and sustainable management of water resources.

Despite progress, the majority of existing systems struggle with long-term autonomous deployment due to energy constraints and biofouling-related sensor degradation. Addressing these challenges is crucial for scaling these robots into large water-management networks.

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