

# Optimization of Data Aggregation Techniques in IoT-Based Wireless Sensor Networks

Shaik Sadulla

Department of Electronics and Communication Engineering, KKR & KSR Institute of Technology and Sciences, Vinjanampadu, Guntur-522017, Andhra Pradesh, India.

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## ABSTRACT

This article delves into the optimization of data aggregation techniques in IoT-based wireless sensor networks (WSNs), aiming to improve network efficiency and extend sensor node lifespan. It explores various algorithms and methods used for aggregating sensor data, emphasizing their role in reducing communication overhead and conserving energy in IoT environments. Key challenges such as balancing aggregation accuracy, latency, and energy consumption are discussed, highlighting the need for adaptive and context-aware aggregation approaches tailored to specific IoT applications. The article presents a simulation and experimental framework for evaluating these techniques, outlining metrics and methodologies for performance assessment. A comparative analysis assesses the strengths, limitations, and suitability of different aggregation methods across diverse IoT scenarios, providing practical insights for implementing effective data aggregation strategies in WSNs. Finally, it concludes by outlining future research directions, advocating for advancements in optimization algorithms and collaborative efforts to meet evolving IoT demands and maximize the benefits of data aggregation.

Author's e-mail: sadulla09@gmail.com

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## INTRODUCTION

In the context of the Internet of Things (IoT), Wireless Sensor Networks (WSNs) are pivotal for gathering and transmitting large volumes of data from numerous sensors dispersed across various settings. However, managing the vast amounts of data generated by these sensors presents substantial challenges, particularly regarding energy consumption, bandwidth usage, and data processing [1]. Data aggregation techniques have been developed to tackle these issues effectively by consolidating data from multiple sensors into a more manageable and meaningful form. This approach not only reduces the amount of data that needs to be transmitted but also helps conserve energy and enhance overall network efficiency. WSN deployed for IoT based monitoring system is shown in Figure 1.

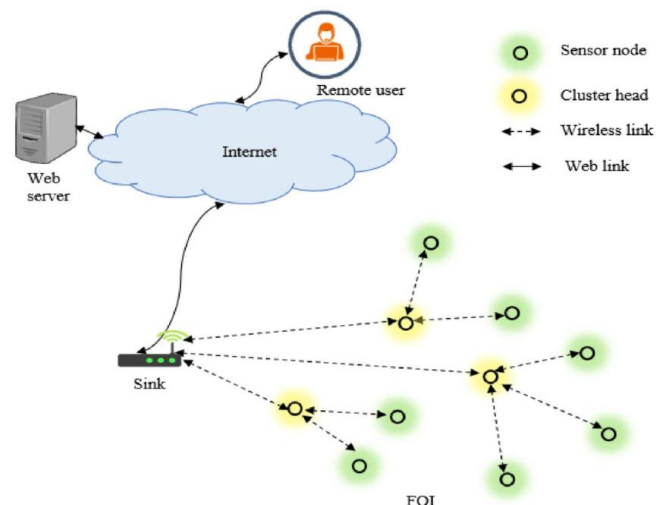
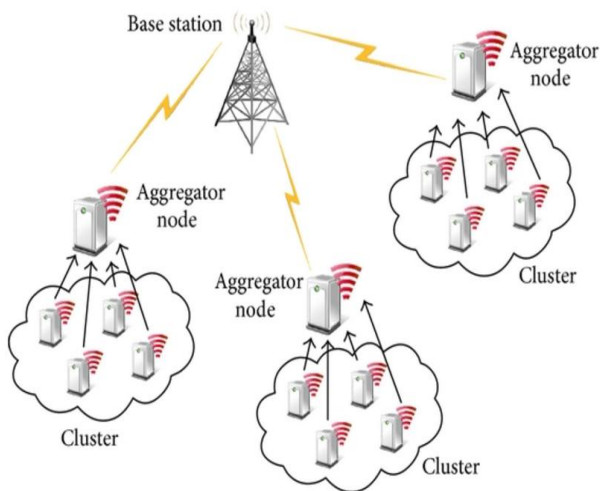


Figure 1. WSN for IoT based monitoring system

Data aggregation refers to the process of collecting and combining data from different sensor nodes to eliminate redundancy and produce accurate, concise information for end-users or central processing units [2]. This is especially crucial in IoT-based WSNs, where sensors typically operate on limited battery power and are often located in remote or difficult-to-access areas. By minimizing the volume of data that needs to be transmitted, data aggregation techniques can significantly extend the operational lifespan of these sensor networks, making them more efficient and cost-effective.

There are various data aggregation techniques tailored to the specific needs of IoT-based WSNs. These techniques can generally be categorized into cluster-based aggregation, tree-based aggregation, and centralized aggregation [3]. Each category has its own set of advantages and is suited to different network structures and applications. For example, cluster-based aggregation organizes sensor nodes into clusters, with a designated cluster head responsible for aggregating data within each cluster. This method is particularly effective in reducing communication overhead and distributing energy consumption evenly among nodes. A cluster-based data aggregation process in the WSNs is shown in Figure 2.



**Figure 2. A cluster-based data aggregation process in the WSNs.**

Tree-based aggregation, in contrast, arranges the network in a hierarchical tree structure, where sensor nodes send their data to a parent node, which then aggregates and forwards the data up the tree. This technique is beneficial for reducing data latency and ensuring efficient data routing [4]. Centralized aggregation, which involves a central node or base station that collects and processes data from all sensor nodes, simplifies data management and processing but can create a bottleneck and increase energy consumption at the central node.

The selection of an appropriate data aggregation technique depends on several factors, including the specific requirements of the application, the network topology, and the nature of the data being collected. For instance, in environmental monitoring applications where sensors are widely dispersed, a cluster-based approach might be more appropriate. Conversely, for applications that require real-time data processing, tree-based aggregation might be favored due to its lower latency.

Despite the advantages, implementing data aggregation in IoT-based WSNs comes with several challenges. These challenges include ensuring data accuracy and reliability, managing dynamic network conditions, and balancing the trade-offs between energy efficiency and data latency[5]. Additionally, security issues such as data integrity and privacy must be addressed to prevent malicious attacks and unauthorized access to sensitive information.

In summary, data aggregation techniques are vital for optimizing the performance and efficiency of IoT-based Wireless Sensor Networks. By reducing data redundancy and conserving energy, these techniques enhance the longevity and reliability of sensor networks, making them more suitable for a wide range of applications. Ongoing research and development in this field aim to overcome existing challenges and improve the effectiveness of data aggregation methods, paving the way for more advanced and intelligent IoT systems.

### Data Aggregation Algorithms and Methods

Data aggregation is essential in IoT-based Wireless Sensor Networks (WSNs) to minimize energy consumption, enhance data precision, and reduce communication load. Various algorithms and methods have been devised to optimize data aggregation, tailored to meet different application needs and network structures [6]. These approaches can be categorized into cluster-based, tree-based, centralized, and hybrid methods.

Cluster-based data aggregation algorithms organize sensor nodes into clusters, with each cluster having a designated cluster head that collects data from its members and transmits it to the base station. An example of this approach is the Low-Energy Adaptive Clustering Hierarchy (LEACH) algorithm. LEACH periodically selects different cluster heads based on criteria like residual energy, ensuring balanced energy use across the network. This reduces the number of transmissions to the base station, thereby conserving energy and extending the network's operational life.

Tree-based data aggregation methods build a hierarchical tree structure where sensor nodes send their data to parent nodes. These parent nodes aggregate the data and forward it up the tree to the base station [7]. The Aggregation Tree (AT) algorithm exemplifies this method. In AT, nodes closer to the tree's root act as aggregation points, collecting data from child nodes. This approach effectively reduces data latency and ensures a structured data flow,

making it ideal for applications requiring real-time data processing.

Centralized data aggregation algorithms involve a central node or base station that gathers and processes data from all sensor nodes. This simplifies data management but can create a bottleneck at the central node, leading to increased energy consumption and potential delays [8]. The Centralized Energy-Efficient Clustering (CEEC) algorithm is one such example, where the base station handles all aggregation tasks. While centralized methods are straightforward and easy to implement, their scalability and energy efficiency are often limited.

Hybrid data aggregation algorithms merge the benefits of cluster-based, tree-based, and centralized methods to mitigate their individual drawbacks. These algorithms adapt to changing network conditions and application needs, providing a balanced approach to data aggregation [9]. The Hybrid Energy-Efficient Distributed (HEED) clustering algorithm, for instance, combines clustering and multi-hop routing to optimize energy use and aggregation efficiency. HEED periodically selects cluster heads based on residual energy and communication cost, ensuring balanced energy consumption and extended network lifespan.

Beyond these primary methods, advanced techniques like compressive sensing and machine learning are being integrated into data aggregation processes. Compressive sensing exploits the sparsity of sensor data to reduce the number of transmissions, while machine learning algorithms predict data trends and anomalies, improving the accuracy and efficiency of aggregation.

### Challenges and Considerations in Optimizing Data Aggregation

Optimizing data aggregation in IoT-based Wireless Sensor Networks (WSNs) is critical for improving efficiency and extending network life. However, several challenges and considerations must be addressed to optimize data aggregation effectively. One major challenge is energy efficiency. Sensor nodes generally have limited battery life, so data aggregation processes must minimize energy consumption to extend the network's operational duration [10]. Effective data aggregation algorithms need to balance reducing communication overhead with maintaining data accuracy.

Ensuring data accuracy and integrity during the aggregation process is another significant challenge. When data from multiple sensors is combined, there is a risk of losing important information or introducing errors. It is crucial to ensure that aggregated data remains accurate and reliable, especially for applications requiring high precision, such as healthcare and environmental monitoring.

Scalability is also a key consideration. As the number of sensor nodes in a network grows, the data aggregation algorithm must manage the additional data without significantly increasing computational complexity or energy consumption. Scalable algorithms are necessary

to handle the expanding size and density of IoT deployments effectively.

Another challenge involves network dynamics. IoT-based WSNs often operate in environments where network topology can change due to node mobility, failures, or environmental factors. Data aggregation algorithms must be robust and adaptive to these changes, ensuring reliable data transmission despite network fluctuations.

Security and privacy are also crucial concerns in optimizing data aggregation. Aggregated data can be sensitive, so protecting it from unauthorized access and tampering is essential. Incorporating encryption techniques and secure communication protocols into data aggregation processes is necessary to safeguard data integrity and privacy.

Lastly, latency and delay are important considerations, especially for real-time applications. Data aggregation should not introduce significant delays that could affect the timeliness of information. Algorithms must be designed to ensure that data aggregation and transmission occur within acceptable timeframes.

In conclusion, optimizing data aggregation in IoT-based WSNs involves addressing challenges related to energy efficiency, data accuracy, scalability, network dynamics, security, and latency. By considering these factors, researchers and engineers can develop robust data aggregation algorithms that enhance the performance and reliability of IoT-based sensor networks, ensuring they meet the demands of various applications.

### Simulation and Experimental Setup for Evaluation

Establishing a simulation and experimental environment to evaluate data aggregation techniques in IoT-based Wireless Sensor Networks (WSNs) involves several essential steps to ensure accurate and reliable results. Firstly, choosing the appropriate simulation tools and platforms is crucial. Tools like NS-3, OMNeT++, and MATLAB are widely used for WSN simulations due to their comprehensive frameworks and extensive libraries. These tools enable the modeling of various network scenarios, including different node densities, topologies, and communication protocols. Defining simulation parameters, such as the number of nodes, communication range, and energy models, is vital to accurately mirror real-world conditions.

Beyond simulations, practical experimentation using actual hardware is important for validating the results obtained from simulations. This involves deploying sensor nodes in a controlled environment to observe the performance of data aggregation algorithms under real conditions. Common sensor platforms like Arduino, Raspberry Pi, and commercial IoT sensors are suitable for these experiments. Setting up a testbed that includes various sensors, gateways, and base stations helps in assessing the practical challenges and performance of the algorithms in a real-world setting.

This approach also provides insights into the hardware's energy consumption, communication delays, and data accuracy.

Integrating simulation and experimental results is key to a thorough evaluation. Simulations offer a controlled environment to quickly and cost-effectively test various scenarios, while experimental setups validate these findings in real-world conditions. Evaluation metrics used in both simulations and experiments typically include energy consumption, data accuracy, latency, and network lifetime. Comparing these metrics across different data aggregation techniques allows researchers to identify the most efficient and robust algorithms for specific IoT applications. This comprehensive approach ensures that proposed solutions are not only theoretically robust but also practically viable, leading to more reliable and efficient IoT-based WSN deployments.

### Comparative Analysis of Data Aggregation Techniques

When conducting a comparative analysis of data aggregation techniques in IoT-based Wireless Sensor Networks (WSNs), the focus is on assessing their effectiveness, reliability, and scalability across different scenarios. Cluster-based methods, such as the Low-Energy Adaptive Clustering Hierarchy (LEACH), prioritize energy efficiency by organizing nodes into clusters with designated cluster heads responsible for aggregating and transmitting data. These approaches reduce the number of transmissions to the base station, thereby conserving energy and prolonging the network's lifespan. However, challenges may arise from uneven energy distribution among nodes and the overhead involved in frequent cluster head re-election. Tree-based aggregation techniques, exemplified by the Data Aggregation Tree (DAT) method, establish hierarchical structures where data flows from leaf nodes to the root through intermediary nodes that perform data aggregation. Such methods effectively minimize data redundancy and ensure a structured flow of information, making them advantageous in scenarios requiring timely data delivery with minimal delay. However, they can be susceptible to node failures, necessitating robust mechanisms for dynamic tree reconfiguration and fault tolerance.

Centralized and hybrid data aggregation techniques offer distinct benefits and trade-offs. Centralized approaches involve a single base station aggregating data from all nodes, simplifying data management and ensuring high accuracy. Yet, they may create bottlenecks and are less scalable due to increased energy consumption at the central node. Hybrid methods blend aspects of cluster-based, tree-based, and centralized techniques, adapting dynamically to network conditions to optimize performance. While these methods strike a balance between energy efficiency, scalability, and resilience, their implementation can be complex, requiring sophisticated algorithms for adaptive management.

Comparing these techniques involves evaluating performance metrics such as energy consumption, data accuracy, latency, and network longevity. This assessment helps in determining the most suitable data aggregation methods for specific IoT applications and deployment scenarios, ensuring robust and efficient operation of WSNs in diverse environments.

### Conclusion and Future Directions

In summary, data aggregation techniques are essential for optimizing the efficiency and performance of IoT-based Wireless Sensor Networks (WSNs). Our examination of cluster-based, tree-based, centralized, and hybrid methods reveals distinct advantages and challenges for each approach. Cluster-based techniques effectively conserve energy by grouping nodes but must address issues like uneven energy usage and overhead from frequent cluster head re-election.

Tree-based strategies efficiently manage data flow and reduce redundancy but require robust mechanisms to handle node failures. Centralized aggregation simplifies data management but may face scalability issues due to increased energy demands at the central node. Hybrid approaches offer a promising balance by adapting to varying network conditions while optimizing energy use and performance.

Looking forward, future research should focus on enhancing scalability, reliability, and security in data aggregation techniques. Leveraging advancements in machine learning and AI can help develop adaptive algorithms that dynamically optimize energy consumption and data transmission. Additionally, exploring blockchain technology for secure data aggregation could enhance data integrity and trust in IoT environments.

Furthermore, addressing real-time data processing challenges and improving edge computing capabilities will be crucial to reduce latency and enhance responsiveness in IoT applications. As IoT expands into smart cities, healthcare, and industrial automation, optimizing data aggregation will play a vital role in enabling innovative applications and ensuring sustainable network operations. By tackling these challenges and embracing emerging technologies, researchers and practitioners can advance data aggregation in IoT-based WSNs, paving the way for more resilient, efficient, and secure IoT ecosystems in the years ahead.

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