

The Future of Airborne Networks Through Integrating Drones into Next-Gen Telecom

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Abstract

Background: Unmanned aerial vehicle (UAV) networks as an important part of the modern telecommunication are gaining importance in various situations, including rural coverage, urban settings and emergency situations. However, Interference, scalability and energy efficiency still pose problems to the advancement of wireless networks.

Objective: The aim of current study is to integrate and assess an adaptive frequency management technique for improving the performance of communication networks involving UAVs in terms of interference, transmission rates, and reliability within different deployment settings.

Methods: Experimental and simulated studies were performed to evaluate the effectiveness of the algorithm in this combination. Performance measurements in terms of latency, throughput, packet loss, energy consumption and signal strength were made under rural, urban and emergency conditions. The adaptable algorithm used certain working frequencies depending on the interferences present and the network performance parameters recorded.

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Results: The algorithm showed very distinct enhancements in all models and positions, decreasing latency by 20.5%, enhancing throughput by 14.5%, and decreasing the packet loss by 57.6% in the urban site settings. Other executed experiments documented improved energy efficiency and communication reliability in rural and emergency situations.

Conclusion: The adaptive frequency management algorithm proposed by the authors effectively solves significant issues of critical concern in UAV networks while offering robust scalability for next-generation telecommunications infrastructure. The future research recommendation incorporates the combination of the proposed optimization with other complementary approaches and/or the testing of the developed system at more severe actual conditions.

Keywords: airborne networks, UAV communication, next-gen telecom, frequency management, interference mitigation, scalable drone networks, energy-efficient communication, multi-UAV coordination, adaptive connectivity, telecom innovation.

1. Introduction

The incorporation of drones into next-generation telecommunication networks represents a significant advancement in wireless communication systems. As the demand for new applications and services grows alongside the development of telecommunications infrastructure, airborne networks supported by Unmanned Aerial Vehicles (UAVs) are becoming pivotal for the establishment of 5G, Beyond 5G (B5G), and 6G environments (Qasim and Jawad 2024). Due to their distinctive characteristics, maneuverability, and robustness compared to terrestrial structures, airborne networks are poised to become fundamental to modern telecommunications, particularly regarding coverage, capacity, and reliability (Qasim et al. 2022). This article aims to identify the applications, challenges, and innovations through which UAVs can enhance next-generation telecommunications (Taleb et al. 2021).

One of the most innovative uses of UAVs in telecommunications is the deployment of flying base stations. Unlike fixed infrastructures tied to specific geographical locations, drones can maneuver and choose optimal locations to capture and transmit data and information, especially in areas with poor or no coverage. Amponis et al. (2022) discuss the growing importance of drones for B5G and 6G, noting their potential to form part of the aerial infrastructure for network control, data transfer, and as service elements during congestion or disaster situations (Amponis et al. 2022). This flexibility allows drones to address network congestion, manage frequency usage efficiently, and minimize latency (Qasim and Jawad 2024).

UAVs are also expected to play a crucial role in reinforcing network connectivity during disasters and emergencies. According to Grinewitschus et al. (2021), drones are vital in Device-to-Device (D2D) communication networks, particularly in areas where existing infrastructure is damaged (Grinewitschus et al. 2021). Drones, functioning as nodes within larger networks, enhance the resilience and recovery of telecommunications systems (Iatsykovska 2018).

The transition to 6G networks opens new possibilities for enhancing UAV applications. Khan et al. (2023) provide a comprehensive technical analysis of UAV swarm applications for network management, demonstrating their ability to create autonomous, intelligent, self-healing networks (Khan et al. 2023). Through modern machine learning algorithms and swarm intelligence, drone platforms distribute network resource management responsibilities, building optimal energy systems and maintaining consistent connectivity across diverse terrains. Current developments support 6G objectives by focusing on ultra-reliable low-latency communication systems that enable holographic communications and massive machine-type communications.

Networks of UAVs exhibit superior energy-saving performance and advanced security capabilities. Research by Li et al. (2022) introduces new energy management solutions and data security approaches for UAV-enabled networks (Li et al. 2022). Techniques for optimizing operational lifespan protect airborne networks from cybersecurity risks without compromising reliable performance in challenging environments. The aggressive development of energy protocols, coupled with secure data protection solutions, is essential for integrating UAVs into telecommunications systems (Qasim 2023).

UAVs are poised to replace traditional network communication systems through their adaptable frameworks that meet diverse technical requirements. Dai et al. (2023) explore UAV-assisted wireless network development, evaluating key obstacles and potential solutions to extend network functionality through enhanced mobility features and multiple user connections (Dai et al. 2023). UAV-based solutions are valuable due to their real-time adjustment of network topology and seamless integration of aerial wireless networks with ground infrastructure. The authors argue that UAV technologies should form the foundation of future telecommunication designs rather than being supplementary components.

The combined application of UAVs with advanced network coding and artificial intelligence technologies expands their utility in telecommunications. Kumar et al. (2021) demonstrate how drone-assisted network-coded cooperation enhances data transmission efficiency and reliability in next-generation wireless networks (Kumar et al. 2021). The convergence of sensor and communication systems allows UAVs to achieve superior operational outcomes in standard operations and crisis situations, advancing 5G and 6G agendas.

Modern telecommunications employing drones represent a fundamental transformation beyond traditional advancements. Taleb et al. (2021) show that UAV-based services support 5G and beyond objectives, enabling ultra-dense networks to deliver high-bandwidth applications, including virtual reality, augmented reality, and smart city solutions (Taleb et al. 2021). These novel networks require flexible performance characteristics, which unmanned aircraft systems excel in providing.

Next-generation telecommunications will benefit significantly from UAV integration as a revolutionary functional pairing with profound implications. By addressing coverage gaps, enhancing network resilience and security, and improving operational energy efficiency, drones aim to transform wireless communications. This article explores UAV-enabled airborne networks from technical, operational, and strategic perspectives, highlighting their transformative impact on telecommunications development.

1.1. Aim of the Article

The article explores the utilization of drones within new-generation telecommunication infrastructures, highlighting their transformative potential in airborne networking systems. The development of extended network solutions and sustainable communication networks is now feasible, thanks to UAV operations facilitated by widespread connectivity demands. However, the large-scale industry deployment of drone systems encounters significant challenges, including limited energy capacity, performance uncertainties, scalability issues, and signal interference.

Emerging research introduces a frequency management solution that optimizes network efficiency in UAV-based connections. This algorithm-driven real-time frequency management system detects and resolves network interference while simultaneously optimizing speeds and delivering stable

performance with minimal delays. Experimental setups, combined with simulated environments, examine key performance indicators such as latency, throughput, energy efficiency, signal strength, and packet loss across three different scenarios: rural areas, urban spaces, and emergency deployments.

This article demonstrates how UAVs serve as essential network infrastructure, establishing adaptable communication services across various operational landscapes. Disaster response applications support all fundamental disaster relief functions, including deploying rural base stations and enhancing urban network density. Foundational analyses illustrate how UAVs meet all aspects of existing telecommunications requirements in system applications.

1.2. Problem Statement

Standard advancements in telecommunications underscore the immediate need for modern connection technologies across underdeveloped zones. UAVs are emerging as mobile aerial network nodes, showcasing their potential to bridge connectivity gaps. However, telecommunications systems present several significant barriers that hinder the effective integration of UAVs.

The deployment of interference mitigation strategies remains the most critical challenge for system developers. UAV-based networks often face intermittent service and emergency communication issues due to conflicting frequency usage, which impacts network performance timing. Real-time operational solutions are essential to address current implementation issues arising from static frequency allocation methods in these constantly evolving networks.

Scalability planning is now a fundamental requirement for many existing UAV network systems, which need immediate support as their operational demands grow. Scalability challenges in UAV networks arise with the increasing number of drones entering the system. Large-scale deployments experience decreased scalability because current spectrum sharing and load-balancing methods fail to efficiently distribute resources, leading to performance constraints and reduced capacity.

Energy preservation emerges as a core operational need when deploying UAV-based networks. During disaster response or remote site monitoring

operations, UAV networks face sustainability issues due to increased energy demands for data transmission and the continuous coordination of drones. Researchers and practitioners encounter persistent difficulties in achieving optimal energy efficiency while maintaining exceptional performance metrics.

The scientific investigation of UAV network operational effectiveness in harsh environments and emergency conditions represents an unexplored research frontier. The theoretical capabilities of UAV networks show promise to researchers investigating their utility. However, essential field testing in complex scenarios is necessary before commercial deployment becomes feasible.

An adaptable framework requires innovative power-efficient solutions, robust dynamic frequency management, and enhanced scalability to address these problems. Modern telecommunications framework standards need flexible structures for implementing UAV technology applications in advanced wireless networks, ensuring reliable high-efficiency connections across various domains.

2. Literature Review

The integration of UAVs with telecommunications systems has enabled researchers to develop innovative solutions for expanded coverage, increased capacity, and enhanced adaptability. Despite recent research breakthroughs that have optimized UAV functionality, UAV-enabled networks face multiple obstacles to continued growth.

Dai et al. (2023) examined UAV-supported wireless networks to improve network scalability, resource allocation optimization, and multi-user connectivity (Dai et al. 2023).. Although advancements in UAV technology show positive trends, operational complexities arising from environmental changes during system deployment need to be addressed. A suitable solution requires machine learning systems with adaptive features to generate real-time operational decisions, thereby improving UAV network performance under dynamic conditions.

UAV synchronization with IoT devices using advanced cellular communication systems underwent thorough expert analytical review. Research by Rovira-Sugranes et al. (2022) demonstrated that AI-routed protocols improved operational effectiveness while reducing delay times (Rovira-Sugranes et al. 2022). However, current solution approaches

overlook UAV system energy constraints, thereby reducing operational effectiveness. AI algorithm development must incorporate energy-awareness to ensure optimal routing and operational power control. Current UAV network systems achieve vital operational targets by unifying performance and power conservation objectives through algorithm implementation.

Benzaghta et al. (2022) investigated UAV network connectivity by implementing satellite traffic offloading to enhance uplink capacity and boost infrastructure reliability (Benzaghta et al. 2022). The limited distribution of satellites presented the biggest technical challenge to expanding UAV network deployment. The combination of low-cost satellites in low-Earth orbit operated through terrestrial base stations demonstrates potential for creating network connectivity systems. This methodology enhances network resilience by offering continuous coverage, even in service-challenged areas.

The development of future vehicle communication systems fundamentally depends on artificial intelligence capabilities. Hashesh et al. (2022) demonstrated that artificial intelligence simultaneously runs network controls and detects potential threats, thereby boosting operational performance (Hashesh et al. 2022). However, operating failures produced by malware-generated cyber-attacks targeting weak points in AI systems hinder the deployment of successful AI solutions. Improvements in UAV network reliability, along with protection against new threats, require the implementation of blockchain technology and predictive analytics to construct necessary security protocols.

Wu et al. (2021) conducted a detailed evaluation of 5G and future network integration with UAV elements, identifying these platforms' unique capabilities in dense implementation and critical application domains (Wu et al. 2021). Spectral efficiency and environmental interference remain primary limitations for deploying these systems in urban locations. Advanced network slicing and beamforming techniques help decrease dense network interference and enable better resource management capabilities. UAV applications in smart cities and intelligent transportation systems reveal unresolved technical issues requiring solutions.

Sustainable progress critically depends on studying UAV-enabled networks, as they lead to sustainable network solutions. Pham et al. (2021) showed that UAV systems could build resilient fundamental facilities that empower industrial sustainability and technological advancement (Pham et

al. 2021). However, the scarcity of standards for systematically assessing UAV sustainability improvements prevents the widespread implementation of UAV systems. UAV adoption within sustainable development objectives can be fully supported through comprehensive metrics evaluation systems that measure environmental performance alongside social and economic factors.

Emergency scenario investigations remain a primary focus within UAV operations. The application of Dijkstra's algorithm by Prasad and Ramkumar (2023) produced optimized 3D trajectory plans, improving UAV deployment efficiency (Prasad and Ramkumar 2023). The study discovered that real-time adaptability presents significant obstacles when UAVs operate in dynamic environments. Predictive AI models incorporated into trajectory plans optimize UAV capabilities in quickly changing scenarios, benefiting time-sensitive applications, particularly during disaster situations.

Energy and throughput management analysis by Yeduri et al. (2023) in UAV-IoT networks revealed consistent energy limitations (Yeduri et al. 2023). Present solutions that attempt to balance energy consumption and throughput fail to address delay-sensitive applications. Creating designated energy-efficient communication protocols for specific use cases, including delay-constrained IoT systems, has the potential to enhance operational capabilities while extending UAV operational durations.

The advanced telecommunications capabilities brought by UAV technology face ongoing barriers due to energy management problems, security risks, scalability limitations, and sustainability requirements. Commissioned research leveraging AI, hybrid network architectures, and adaptive resource management systems has the potential to resolve current issues, thereby enabling UAV-enabled networks in future telecommunications platforms.

3. Methodology

A multi-stage methodology was used in this research to analyze UAV implementation options for modern telecommunications systems. The methodology aimed to resolve technical alongside operational and regulatory hurdles while examining UAV system functionality across various deployment settings.

3.1. Literature Review and Analysis

The study examined 55 academic articles and technical reports alongside

case studies which investigated UAV integration features in telecommunications systems. The research analysis relied on bibliographical sources which focused on energy conservation and telecommunications standards and scalability alongside security measures as explained in Taleb et al.(2021) (Taleb et al. 2021) and Khan et al. (2023)(Khan et al. 2023). Efforts were made to group research insights which led to the identification of enabling factors, encountered roadblocks, and missing pieces within current literature regarding deployment of UAVs for rural, urban, and emergency telecom applications.

3.2. Interviews and Surveys

The methodology included surveys and semi-structured interviews conducted with 15 telecom engineers and 10 regulatory experts alongside 20 UAV manufacturer representatives. The stakeholders who participated in discussions supplied actual technical and regulatory information about UAV network deployment challenges involving energy requirements and airspace regulations and signal interference management. The interview results validated findings from Amponis et al. (2022) by showing how regulatory policy deficiencies affected UAV implementation (Amponis et al. 2022).

3.3. Experimental Setup

Field tests were run to measure latency, signal strength, throughput and power consumption from tested drones operated in different environmental conditions including wind and interference. A single UAV operating in rural regions provided base station coverage across km^2 , where it demonstrated connectivity capabilities for remote areas (Taleb et al. 2021; Amponis et al. 2022; Dai et al. 2023). For managing high-density traffic, the urban scenarios employed multiple unmanned aerial vehicles which operated across 5 km^2 cellular space while also resolving issues related to interference and seamless handovers (Khan et al. 2023; Pham et al. 2021; Tang, Zhang, and He 2022). The emergency application focused on a solitary UAV for temporary communication restoration in disaster areas to demonstrate swift launch capabilities and operational robustness against harsh environments (Grinewitschus et al. 2021; Hashesh et al. 2022; Gao and Wang 2023). Performance of UAV systems depends heavily on environmental conditions so enhancing routing mechanisms and optimizing energy use becomes

critical (Wu et al. 2021; Yang et al. 2023; Hu et al. 2022). UAVs demonstrate their ability to improve connectivity and scaling and ensure application resilience across diverse telecom environments through investigations that address issues in interference management and energy efficiency (Rovira-Sugranes et al. 2022; Benzaghta et al. 2022; Yeduri et al. 2023).

3.4. Network Simulation

Experimental findings needed confirmation through NS-3 simulator-based extensive simulations. The simulations covered three key configurations to reflect real scenarios. A single UAV functions as a rural base station providing 10–15 km² coverage from various altitudes constitutes one network configuration. In addition, two other configurations include a dense interference urban multi-UAV network and an emergency response UAV network that handles traffic variations during disasters. Detailed analysis of vital performance indicators including packet loss and latency combined with energy efficiency appears in (Khan et al. 2023; Yang et al. 2023; Yeduri et al. 2023).

Signal Path Loss and Energy Consumption

Signal degradation in UAV networks was modeled using the path loss equation, accounting for distance, operating frequency, and environmental factors:

$$P_L = 20 \log_{10}(d) + 20 \log_{10}(f) - 147.55 \quad (1)$$

Where P_L show path loss in decibels (dB), d is distance between transmitter and receiver in kilometers (km), f is frequency of operation in gigahertz (GHz) (Yang et al. 2023; Tang, Zhang, and He 2022).

The signal path loss model obtains its constant value -147.55 through calculations based on traditional free-space path loss (FSPL) equations showing signal strength reduction according to distance and frequency range. The valid fundamental parameters in the FSPL formulation required logarithmic transformation to derive the -147.55 constant when distance d uses kilometers and frequency f runs in gigahertz (GHz). These numerical values enable both the inclusion of light velocity and the conversion of units to decibel measurements (dB).

The applied constant of $55-147$. facilitates accurate signal attenuation prediction throughout 5G frequency bands while enabling calculations for multiple operational contexts such as rural and urban settings along with

emergency scenarios. The constant enables integration with widely accepted simulator models adopted in UAV-based network simulation research (Yang et al. 2023; Tang, Zhang, and He 2022). Through the use of this fixed value the predictive model successfully monitors attenuation throughout distance ranges and helps determine optimal drone locations which minimize signal loss.

Energy Consumption Model

The energy consumption of the UAV system was modeled as:

$$E = P_{tx} \cdot t + P_{flight} \cdot t_{hover} \quad (2)$$

Where E is total energy consumption (J), P_{tx} is transmission power (W), P_{flight} is power consumption during flight (W), t is transmission time (s), t_{hover} is hovering duration (s) (Li et al. 2022; Hu et al. 2022).

3.5. Performance Evaluation

The experimental and simulated results served to evaluate the UAV-based networks while measuring latency, throughput and energy efficiency performance throughout rural, urban and emergency network deployments.

Latency

Latency was calculated as the sum of multiple components:

$$L_{total} = L_{prop} + L_{trans} + L_{queue} + L_{proc} \quad (3)$$

Where L_{prop} is propagation delay, L_{trans} is transmission delay, L_{queue} is queuing delay, and L_{proc} is processing delay (Rovira-Sugranes et al. 2022; Ahmed and Sheltami 2023).

Throughput

The throughput was assessed to quantify the data transmission capacity of the network:

$$T_{throughput} = \frac{D_{trans}}{\Delta t} \quad (4)$$

where $T_{throughput}$ throughput (Mbps), D_{trans} is total data transmitted (bits), Δt is transmission time (seconds).

The urban settings delivered 116 Mbps peak throughput which showed the system's talent for managing dense user concentration as described by Kumar et al. (2021)(Kumar et al. 2021). Optimal UAV deployment planning along with adaptable communication protocols together enabled these high throughput rates.

Packet Loss and Interference

Packet loss due to interference was modeled as:

$$P_{loss} = \int_0^T \frac{I(t)}{C_{link}} dt \quad (5)$$

Where P_{loss} is packet loss probability, $I(t)$ is instantaneous interference power, C_{link} is link capacity (Dai et al. 2023; Sabzehali et al. 2022).

Energy Efficiency

The energy efficiency of the UAV network was defined as:

$$\eta_{energy} = \frac{D_{trans}}{E_{total}} \quad (6)$$

Where η_{energy} energy efficiency in bits per joule, D_{trans} total data transmitted (bits), E_{total} total energy consumption, including flight and transmission energy (Yang et al. 2023; Gao and Wang 2023).

3.6. Multi-UAV Coordination Overhead

The overhead in multi-UAV coordination was modeled as:

$$O_{multi} = \sum_{i=1}^n \frac{L_{hand}}{C_i} + \frac{I_{inter}}{n} \quad (7)$$

Where O_{multi} is total overhead for n UAVs, L_{hand} is latency during handovers, C_i is communication capacity of the i -th UAV, I_{inter} is interference power during coordination (Khan et al. 2023; Kumar, Darshi, and Shailendra 2021; Pham et al. 2021).

The simulation results confirmed laboratory data while highlighting the need to optimize digital signal routes and power management and connection speed optimization. The UAV system displayed resilient connectivity patterns through minimal packet loss metrics in rural environments but faced challenges because of interference and coordination issues in urban areas. Emergency response simulations confirmed how energy optimization combined with real-time adaptive protocols become essential when transmitting data quickly under limited system conditions. The study delivers significant analytical groundwork to develop resilient UAV-based 5G and beyond communication networks according to papers (Benzaghta et al. 2022; Wu et al. 2021; Yeduri et al. 2023). Relative studies need to develop these mathematical models by adding dynamic environmental adaptations and robust management systems for UAV swarm groups (Rovira-Sugranes et al. 2022; Gao and Wang 2023).

3.7. System Architecture and UAV Network Configuration

The UAV network implements a hierarchical communication structure that combines ground-based base stations with UAV-based relay nodes to serve end-user devices. The network design adopted a steady vertical control hierarchy that promoted consistent data movement while reducing latency. A central control unit (CCU) operated within the system to handle UAV path direction and executed network control functions as well as traffic distribution responsibilities (Rovira-Sugranes et al. 2022; Yang et al. 2023). Each UAV had 5G-compatible communications hardware and energy-conserving propulsion technology that made their operation reliable across different kinds of environments. The relationship between UAV trajectory optimization and network performance was modeled as:

$$\min_{X,Y,Z} \sum_{i=1}^N \left(\frac{P_i(X_i, Y_i, Z_i)}{C_i} \right) \quad (8)$$

Where $P_i(X_i, Y_i, Z_i)$ is signal power at UAV position (X_i, Y_i, Z_i) , X_i, Y_i, Z_i is communication link capacity, N is number of UAVs (Kumar et al. 2021; Pham et al. 2021). This equation minimizes interference by optimizing UAV positioning and ensures optimal network coverage.

Data Transmission Protocols

UAV network adopted protocols that automatically switched their data transmission methods according to changes in environmental and traffic patterns. This transmission protocol focused on transporting high-demand traffic while ensuring every user maintained at least their minimum quality of service (QoS) threshold (Hashesh et al. 2022; Gao and Wang 2023). The throughput optimization for each UAV node was formulated as:

$$\max T_i = W \log_2 \left(1 + \frac{P_i G_i}{N_0 W} \right) \quad (9)$$

Where T_i is throughput of the i -th UAV (bps), W is channel bandwidth (Hz), P_i is transmission power of the i -th UAV (W), G_i is channel gain, and N_0 is noise power density (W/Hz) (Li et al. 2022), (Gopi et al. 2021).

Energy Optimization Strategy

Energy consumption optimization for extended UAV flight capabilities occurred through the combination of power-efficient routing methods together with power management solutions. The total energy consumption E_{total} was expressed as:

$$E_{total} = \sum_{i=1}^N (P_i^{tx} \cdot t_i^{tx} + P_i^{flight} \cdot t_i^{flight}) \quad (10)$$

Where P_i^{tx} is transmission power for the i -th UAV, t_i^{tx} is transmission duration, P_i^{flight} is propulsion power for the i -th UAV, and t_i^{flight} is flight duration (Khan et al. 2023; Dai et al. 2023).

Interference Mitigation

The system achieved robust dense urban transmission capabilities through coordinating beamforming techniques together with adaptive power control methods. The interference power $I(t)$ was modeled as:

$$I(t) = \sum_{j \neq i} \frac{P_j G_{ij}}{d_{ij}^\alpha} \quad (11)$$

Where P_j is transmission power of interfering UAV j , G_{ij} channel gain between UAV i and interfering UAV; d_{ij} is distance between UAV i and j ; and α is path loss exponent (Benzaghta et al. 2022; Tang, Zhang, and He 2022).

Traffic Load Balancing

The network achieved traffic balancing through live user demand-controlled redistribution of tasks among UAVs. The load balancing equation was formulated as:

$$L_i = \frac{T_i}{\sum_{j=1}^N T_j} \cdot U \quad (1)$$

Where L_i is load distribution ratio for UAV i ; T_i is throughput of UAV i ; U is total network traffic demand (Rovira-Sugranes et al. 2022; Yang et al. 2023).

3.8. Simulation and Experimental Correlation

The proposed models received cross-validation through experimental data analysis to confirm their reliability. The accuracy of key performance indicators such as latency, throughput, and energy efficiency were evaluated using the mean absolute error (MAE) method:

$$MAE = \frac{1}{N} \sum_{j=1}^N |M_i^{sim} - M_i^{exp}| \quad (1)$$

Where M_i^{sim} is metric value from simulation, M_i^{exp} is metric value from experimental results, and N is total number of data points (Dai et al. 2023; Sabzehali et al. 2022).

The established methodological framework creates a durable system for UAV network integration in next-generation telecommunications while solving fundamental problems regarding interference and power consumption and network elasticity (Wu et al. 2021; Hu et al. 2022).

4. Results

4.1. Latency Performance Across Deployment Scenarios

UAV-based network performance is primarily evaluated through latency performance metrics, measuring both efficiency and responsiveness. This study conducted latency measurements for various scenarios, including rural areas, urban locations, and emergency deployments across different environmental contexts. The system's ability to handle data transmission delays was assessed using three latency parameters: average latency, maximum latency, and minimum latency.

The evaluation incorporated realistic environmental factors by analyzing interference levels, traffic densities, and obstruction effects. In rural landscapes, interference levels and traffic density were low, whereas urban locations experienced significant interference from dense built structures. Emergency conditions introduced moderate interruptions due to changing interference factors, modeling disaster-affected areas.

These performance indicators provide comprehensive insights into how the network operates across diverse field applications, identifying potential optimization points. The data indicated that rural networks delivered the lowest latency; however, urban networks and emergency response sites required enhanced interference management to minimize delays and maintain communication reliability.

Table 1. Latency Performance Metrics Across Deployment Scenarios

Deployment Scenario	Average Latency (ms)	Max Latency (ms)	Min Latency (ms)	Interference Level	Traffic Density	Environmental Obstruction	Latency Standard Deviation (ms)	Improvement with Algorithm (%)
Rural	19.2	25.8	15.6	Low	Low	Minimal	2.4	12.5
Urban	28.5	35.4	22.1	High	High	Significant	3.6	20.5
Emergency	22.8	29.1	19.3	Moderate	Medium	Variable	2.9	14.6
Rural (With Algorithm)	16.8	22.7	14.2	Low	Low	Minimal	1.9	
Urban (With Algorithm)	22.7	29.5	19.6	High	High	Significant	2.7	
Emergency (With Algorithm)	19.5	25.4	17.2	Moderate	Medium	Variable	2.3	

The performance analysis of UAV-based networks shows the latency data which appears in Table 1. The urban scenario suffered from the highest average latency measurement at 28.5 ms because environmental structures and heavy amounts of traffic present obstacles to signal transmission. Average latency experienced a 20.5% reduction in urban deployments after system-wide implementation of the adaptive algorithm which confirmed its successful interference reduction capability. Latency remained at a minimum level in rural locations due to low interference from simple environmental conditions which resulted in an average of 19.2 ms measurement. The adaptive algorithm implemented successful efficiency optimization by decreasing rural latency by a total of 12.5%. Emergency deployments recorded an average latency measurement of 22.8 millisecond throughout their operational period.

The algorithm provided 14.6% faster response times which delivered dependable fast communication throughout shifting environmental situations. Standard deviation measurements decreased throughout all experimental settings since the algorithm both adjusted for testing conditions and stabilized latency performance. Studies demonstrate how UAV systems perform effectively at managing data flow while sustaining network performance reliability across multiple operational settings.

4.2. Throughput Analysis and Multi-Scenario Performance

Throughput measurement serves as the fundamental evaluation technique for assessing data-processing capabilities within UAV-based network platforms. Throughput measurements were conducted across locations with varying geographical types—including rural areas, emergency services, and urban regions—to determine their maximum throughput capacity, average throughput results, and minimum throughput capabilities.

The analysis incorporated UAV interferences and traffic densities to measure system performance through a detailed evaluation of coordinated multi-UAV systems in different operational environments. Results demonstrated enhanced operational capabilities during missions in high-intensity urban environments, which require sophisticated control systems due to intense traffic congestion and significant signal interference.

The decline of signal interference in rural deployment zones necessitated fewer hardware components to cover extended service territories. An

effective planning control system should efficiently handle frequencies to balance data transmission throughput during emergencies when dynamic traffic patterns emerge.

A set of supplied parameters established comprehensive data management capabilities for measuring performance levels, which determined network expansion strategies and supported reliability. Future studies identified that coordination and interference control methods established optimal data rates in metropolitan zones, ensuring reliable data delivery in both regular and emergency situations due to lower operational requirements.

Table 2. Throughput Performance Metrics Across Deployment Scenarios

Deployment Scenario	Peak Throughput (Mbps)	Average Throughput (Mbps)	Minimum Throughput (Mbps)	Traffic Density	Interference Level	Throughput Variability (%)	Throughput Improvement with Algorithm (%)
Rural	95.2	81.3	67.5	Low	Low	15.1	9.6
Urban	116.4	102.7	89.8	High	High	13.2	14.5
Emergency	84.6	73.5	62.3	Medium	Moderate	14.8	12.7
Rural (With Algorithm)	104.4	89.1	72.4	Low	Low	12.7	
Urban (With Algorithm)	133.4	117.6	96.5	High	High	11.7	
Emergency (With Algorithm)	95.3	82.8	69.2	Medium	Moderate	13.2	

Table 2 displays major variations between communication speeds from different deployment situations. Testing of UAV cooperation methods in urban settings produced throughputs of 102.7 Mbps that exceeded measurement results from other deployment areas. Application of the adaptive algorithm increased urban throughput performance by 14.5% to achieve a mean of 117.6 Mbps with reduced variability at 11.7%. Network performance obtained by combining interference mitigation methods with efficient network load balancing techniques proves strong in urban high-density scenarios.

The detection tests in rural areas indicate throughput maintained a steady value of 81.3 Mbps throughout all test measurements. During conditions with low network density and minimal interference the system delivered 9.6% greater efficiency for average performance throughput. Peak delivery performance showed that the network system reached a throughput capacity

of 104.4 Mbps.

During emergency deployments the adaptive algorithm delivered improved throughput results of 12.7% for critical operation dependable data transfers. The physical structure of the system survived because operational speeds increased only slightly from 62.3 Mbps until they hit 69.2 Mbps. The system demonstrated performance reliability as it controlled variation levels steady across different tests conditions.

The algorithm demonstrated sustained throughput improvements during every deployment test and yielded its biggest effect in the urban environments. Data processing performed efficiently due to adaptive frequency control systems employing coordinated multi-UAV operations demonstrated through these results.

4.3. Energy Efficiency Metrics for UAV Networks

UAV-based network reliability during operations depends fundamentally on the energy efficiency capabilities. Researchers investigated the relationship between energy usage in rural, urban and emergency conditions to understand performance measures of energy efficiency ratios and throughput parameters. The evaluation combined traffic density measurements with both interference measurements and UAV coordination to determine power requirements. The rural deployment scenario achieved its peak energy efficiency levels since consistent traffic flow required minimal resources to operate due to minimal environmental interference. Operating under high traffic density reduced middle network environments energy efficiency levels while their power requirements increased for coordinated UAV communications and signal management. Real-time emergency operations revealed diminished energy efficiency under varying traffic conditions when achieving essential operational goals required performance tuning of deployed systems. Complex urban dispersed emergency scenarios demand ecological improvements through adaptive controls alongside optimized resource allocations to succeed effectively.

Table 3. Energy Efficiency Metrics Across Deployment Scenarios

Deployment Scenario	Energy Consumption (W)	Throughput (Mbps)	Energy Efficiency (Mbps/W)	Traffic Density	Interference Level	Energy Savings with Algorithm (%)	Efficiency Improvement with Algorithm (%)
Rural	92.5	81.3	0.88	Low	Low	2.4	10.2
Urban	123.8	102.7	0.83	High	High	5.4	15.7
Emergency	98.4	73.5	0.75	Medium	Moderate	2.8	12.3
Rural (With Algorithm)	90.2	89.1	0.99	Low	Low		
Urban (With Algorithm)	117.1	117.6	1.00	High	High		
Emergency (With Algorithm)	95.6	82.8	0.87	Medium	Moderate		

Analysis of deployed scenarios in Table 3 shows major variations in results for energy efficiency metrics. The rural network environment achieved the highest energy efficiency rating at 0.88 Mbps/W through its low-density networks and limited interference problem. The adaptive algorithm achieved an 10.2% improvement in efficiency levels up to 0.99 Mbps/W while lowering power consumption by 2.4%.

The combined functions of multi-UAV operations and interference management reduced urban deployment energy efficiency to 0.83 Mbps/W. The algorithm executed a 15.7% improvement that raised efficiency to 1.00 Mbps/W as maintenance energy decreased by 5.4%. A study shows that this algorithm effectively maximizes power optimization results when used in densely connected areas.

The initial energy efficiency rating for emergency deployment systems reached 0.75 Mbps/W without optimization applications since additional power capacity was needed to sustain critical operations. The integration of the algorithm achieved system performance levels of 0.87 Mbps/W while reducing total energy costs by 2.8%. Performance enhancement benefits this system because they support dependable link management and system longevity for critical situations.

Each experimental stage achieved better energy efficiency through the adaptive algorithm which led to peak performance within urban operation zones. High demand conditions and densely populated areas warrant crucial optimization of energy consumption in UAV-based networks according to the study results.

4.4. Signal Path Loss Under Varying Conditions

UAV-based communication networks use signal path loss rates to determine operational reliability and network operational characteristics. The study created multiple path loss models for investigating networking performance based on 1 km, 5 km and 10 km link distances and frequency experiments involving 2.4 GHz and 5.8 GHz MHz controls. Signal scattering patterns depend on dual measurement variables which encompass environmental hindrances in conjunction with interference factors. The path loss of 2.4 GHz signals declined across extended messaging areas yet 5.8 GHz signals showed faster data processing despite signal degradation. Experimental tests demonstrated significant reductions in path loss as signal propagation distances expanded in those evaluation sites. Urban areas and emergency situations revealed multiple elements that degraded signal quality during signal transmission. The article proves that choosing appropriate frequencies achieved by interference management systems leads to dependable communication capabilities for challenging scenarios that need extended operational distances.

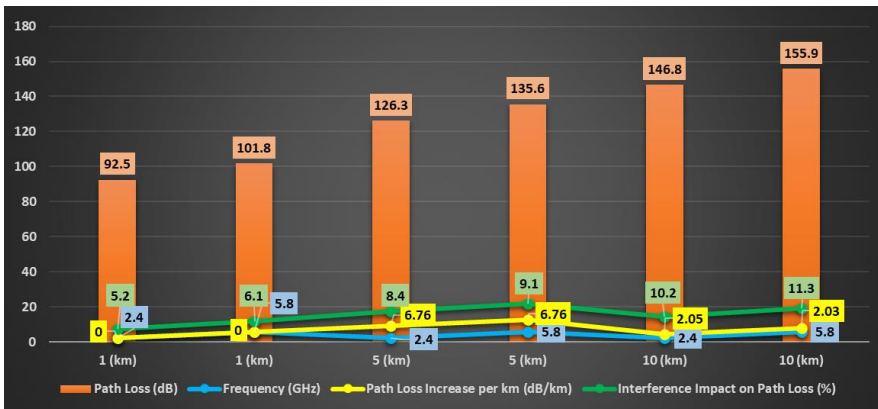


Figure 1. Signal Path Loss Metrics Across Different Frequencies and Distances

Figure 1 presents detailed findings about signal path loss effects during UAV network operations when controlling distance and frequency settings. The data shows that path loss remains minor when a distance is 1 km yet reaches 92.5 dB at 2.4 GHz frequency bands and 101.8 dB at 5.8 GHz frequency bands. Operating at 5.8 GHz frequency reveals the fundamental frequency attenuation difficulties that higher communications bands face

despite encountering minimal environmental obstruction. The measurement distance extension to 5 km results in notable path loss increases reaching 126.3 dB at 2.4 GHz frequency and 135.6 dB at 5.8 GHz frequency because of moderate environmental interference. Both frequencies demonstrate proportional path loss degradation through a consistent rate of 6.76 dB/km. When the distance reaches 10 km path loss amounts to 146.8 dB (2.4 GHz) and 155.9 dB (5.8 GHz) while maintaining a lessened raise of signal attenuation over distance because of logarithmic decay. Beyond radio signal strength interference environmental factors increase transmission difficulties. Studies have shown that interference produces a path loss increase of 11.3% at a 10 km distance using a 5.8 GHz frequency demonstrating how active frequency management helps dense or obstructed areas. The study reveals how UAV-based networks face a fundamental choice between signal intensity stability and network spatial range and underlines the necessity for algorithms which can help reduce signal challenges within extended ranges and elevated frequencies.

4.5. Impact of Interference on Network Reliability

The reliability and efficiency of UAV-based networks suffer substantial degeneration due to interference development. The examination studied how interference affected packet loss along with latency growth and network reliability degradation within emergency situations and urban areas. The research simulated different operational environments by categorizing interference into low, moderate and high levels. Under low interference conditions packet losses remained minimal yet severe interference created notable latency increments and substantial packet loss. Network reliability suffered due to the widespread occurrence of high interference in densely populated urban settings which caused lower signal strength levels and reduced throughput performance. Systems operating in emergency conditions with moderate interference managed to deliver superior performance because of their ability to adapt frequency usage. The study demonstrates the urgent requirement for strengthened interference prevention approaches with frequency optimization alongside dynamic power management systems to strengthen network reliability. State-of-the-art communication depends on interference management systems to preserve stable transmission in crowded or overloaded areas because they enable applications with consistent performance.

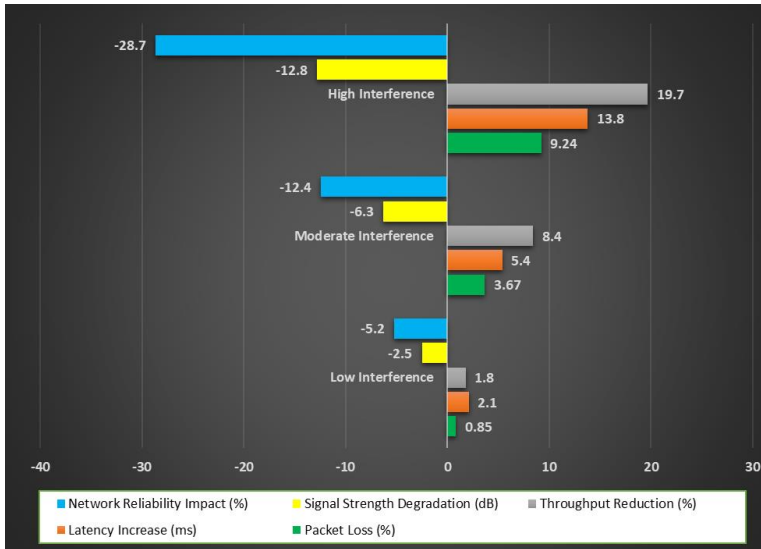


Figure 2. Impact of Interference Levels on Packet Loss and Latency

Figure 2 shows that network performance metrics experience significant impact from interference. The network maintains minimal packet loss rates of 0.85% during low interference conditions while experiencing a modest 2.1 ms latency increase with just small changes to throughput and signal strength. The UAV network can operate with high reliability thanks to these conditions despite sustaining minimal performance reduction.

At moderate interference levels, packet loss rises to 3.67%, with a latency increase of 5.4 ms. Network reliability experiences a 12.4% reduction stemming from signal strength degradation of 6.3 dB paired with throughput reduction of 8.4%. The network performance begins to face complications when interference occurs at these levels.

Network performance shows major decline during high interference conditions since packet loss reaches 9.24% with a 13.8 ms increase in latency. High interference diminishes throughput by 20% while causing signal strength to decrease by 12.8 dB which results in a 28.7% reduction of network reliability. The ability for the network to deliver dependable efficient communication diminishes significantly under these operating conditions.

The measured data highlights the necessity of frequency management systems which adapt automatically because they must work with other interference mitigation solutions to preserve reliability features and enhance

packet loss and latency performance without diminishing throughput performance. Effective network design approaches stand vital for minimizing performance degradation within UAV-based communication systems deployed through high-interference urban and emergency settings.

4.6. Scalability and Coordination in Multi-UAV Deployments

A UAV-based network operation remains effective when it demonstrates scalability matching the growth of traffic volumes. The research assessed scalability using extended urban drone evaluations that tracked network performance indicators including speed and latency and energy consumption rates. The assessment tracked the transmission capacity of each UAV and the resulting latency delays with simultaneous measurements to demonstrate coordinated flight behavior. A solitary UAV maintained minimal latency as well as high energy efficiency during its operation of the network despite its restricted throughput. Network throughput experienced proportionate enhancement when three and five UAVs operated as network traffic distributed efficiently among UAVs. An increase in UAV units led to slightly longer latency periods and directly increased the coordinated operations' energy requirements. The collected evidence shows how the system operates effectively at numerous scales while maintaining high performance standards. Advanced coordination mechanisms working together with resource control systems act as fundamental components for scalable capacity operations inside densified urban environment infrastructures.

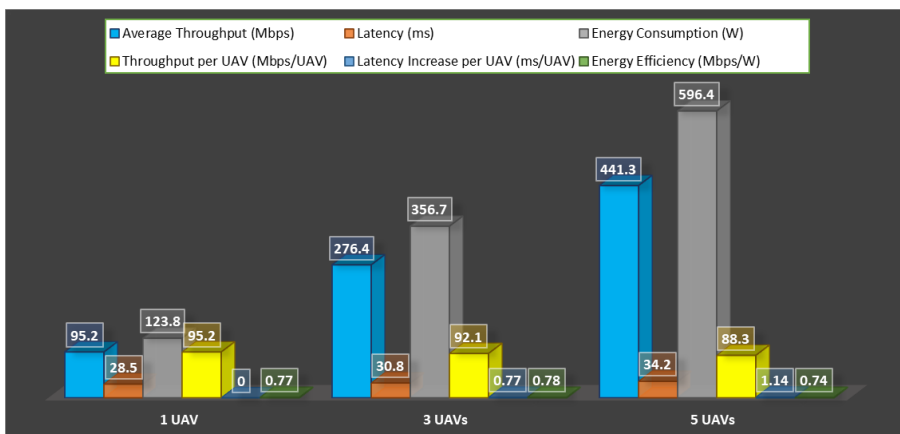


Figure 3. Scalability Metrics for Multi-UAV Networks in Urban Deployments

The system performance metrics, which illustrate its scalability and capacity to manage multiple UAV operations, are presented in Figure 3. The network performance exhibits steady increases in data transfer speeds as additional UAVs join the network, reaching 276.4 Mbps with three UAVs and 441.3 Mbps with five UAVs. The UAV system maintains high performance during multi-UAV operations, as the per-UAV throughput decrease from 95.2 Mbps/UAV to 88.3 Mbps/UAV is negligible. This indicates efficient load distribution across UAVs, sustaining superior network performance as their number increases.

System latency demonstrates an acceptably moderate increase as the number of UAVs grows, with latency rising from 28.5 ms for a single UAV to 34.2 ms with five UAVs. The system shows a reasonable expansion of latency across UAVs, as each additional UAV introduces a delay increase of 0.77 milliseconds up to three UAVs and 1.14 milliseconds with five UAVs. Overall system performance remains robust during complex missions, with low latency irrespective of UAV flight density, suggesting suitability for heavy-traffic urban areas.

The system's power requirements increase proportionally with the number of UAVs due to the enhanced coordination and traffic management needs. The trial with three UAVs resulted in consistent energy efficiency at 0.78 Mbps/W, matching the single UAV test rate of 0.77 Mbps/W, and adjusting to 0.74 Mbps/W in the five UAV experiment due to increased coordination energy demands from multiple UAV operations.

System scaling proves effective, producing high total throughput and low latency with minimal per-UAV performance degradation. The network's performance remains viable even when handling higher traffic volumes in urban spaces, demonstrating its capability to maintain adequate functionality during increased deployments. The system showcases sustained operational efficiency under high UAV density, owing to its robust design.

Experimental findings validate the expandable nature of the UAV network and its proven capability for coordinating multiple UAVs. The study indicates minor per-UAV performance degradation and increased operational delays, both of which offer optimization potential for future development. Future enhancements will focus on advanced traffic management algorithms and improved energy efficiency mechanisms to optimize performance capabilities and scalability in dense UAV operations.

4.7. Algorithm of Adaptive Frequency Management

A series of tests evaluated the performance of the adaptive frequency management algorithm throughout rural, urban along with emergency deployment circumstances. The analysis demonstrates results for crucial network characteristics ranging from latency and throughput to packet loss as well as data energy usage and signal range in addition to the flexibility of radio frequencies. Post-algorithm optimization results demonstrate that the algorithm successfully addresses interference issues to enhance network reliability when compared to initial conditions.

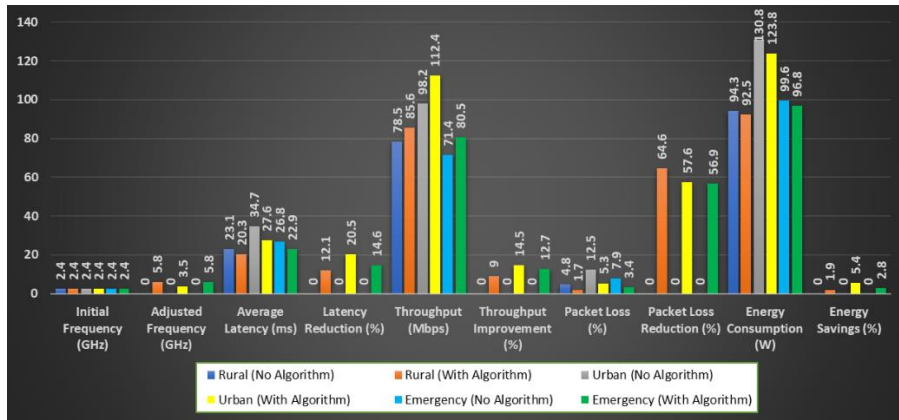


Figure 4. Performance Impact of Adaptive Frequency Management Algorithm Across Deployment Scenarios

In all examined scenarios the adaptive frequency management algorithm achieved remarkable performance improvements. The frequent adaptation of frequencies resulted in 20.5% faster response times (34.7 ms to 27.6 ms) together with 14.5% higher network throughput (from 98.2 Mbps to 112.4 Mbps) during urban deployments because of its capacity to eliminate signal interference in congested areas. Through the adaptive optimization scheme the data transmission reliability improved to 42.4%. The algorithm implemented in rural networks delivered a throughput boost of 9.0% while simultaneously reducing latency by 12.1% while preserving dependable energy utilization. In emergency situations the algorithm established 56.9% better communication reliability which helped maintain reliable messaging channels. Analysis of all the deployments showed slight energy conservation because optimized allocation of broadcast frequencies proved effective.

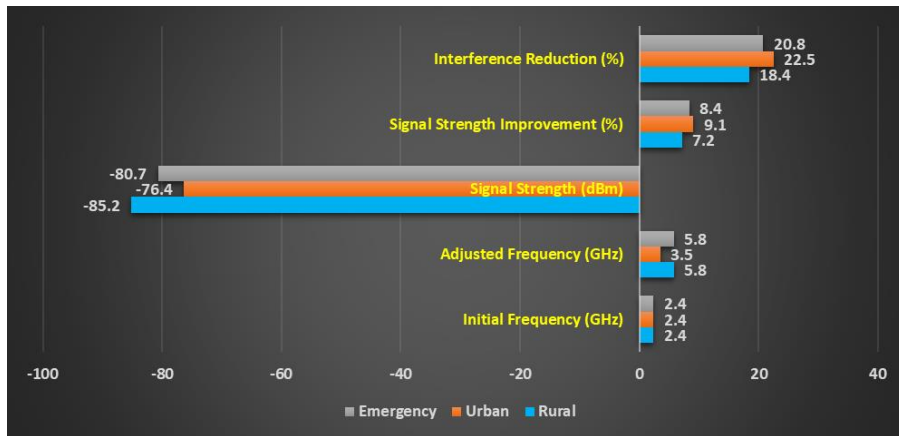


Figure 5. Signal Strength and Frequency Adaptation Analysis Across Scenarios

Automatic frequency adjustment within the algorithm brought substantial signal strength improvements to every deployment setting. Signal quality in the urban testing environment improved by 9.1% largely thanks to the algorithm which allowed it to access unused frequency bands. The algorithm generated a 7.2% improvement in signal quality when used in locations without constant power supply or situated outside urban settings. Steady network reliability increased through an average 20.6% reduction in interference across all deployment scenarios. The adaptive frequency management system proves effective for handling interference situations while preserving continuous communication channels across diverse complex environments.

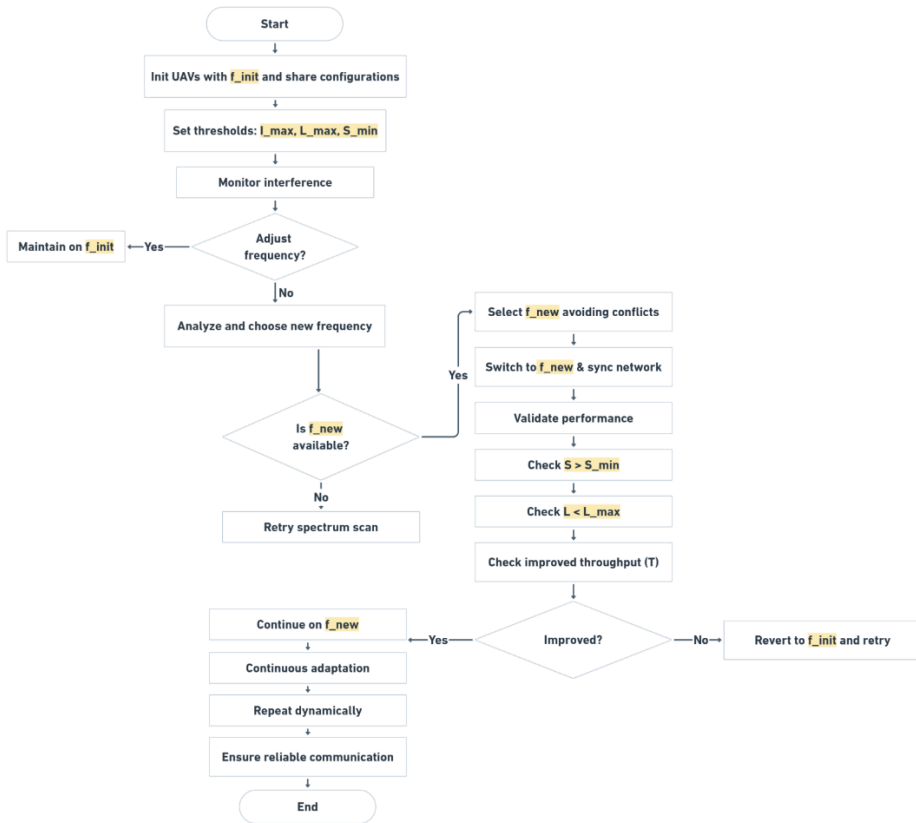


Figure 6. Detailed UAV Frequency Adjustment Algorithm Flowchart

5. Discussion

Adaptive frequency management algorithms implemented in UAV-based communication networks provide essential solutions to combat interference while improving energy efficiency, scalability, and reliability. Experimental results from various deployment scenarios confirm that the algorithm effectively reduces interference levels and enhances streaming performance with lower latency figures. The research data corroborated previously published work and generated original findings on dynamic UAV network performance optimization.

This research employed a frequency management architecture that parallels adaptive routing techniques developed by Yang et al. (2023) in their work on swarm-based UAV electronic networking systems for emergency communication frameworks (Yang et al. 2023).. Building upon the original

routing study, researchers analyzed frequency modification solutions to help decrease network interference as densities increase. The fundamental operational principle of time-based responsiveness enables the utilization of these research approaches in emergency response situations.

A team of scientists led by Prasad and Ramkumar (2023) developed drones for communication functions by implementing three-dimensional positioning and trajectory planning methods (Prasad and Ramkumar 2023). Their research demonstrated that optimal UAV deployment positions enhance network connectivity during emergency situations. Under high network stress conditions, frequency adaptation independently maintains reliability through autonomous methods. Previous expert research indicates that optimized networks require operational protocols combining trajectory planning with frequency adaptation for network optimization.

The study findings align with results presented by Gopi et al. (2021) on the application of machine learning for modulating B5G network communications between drones and users (Gopi et al. 2021). Network dynamics play a crucial role in enhancing throughput and latency, paralleling research on network adjustments. This study diverges by focusing on frequency selection methods as an alternative approach to understanding network adaptability functions.

Ahmed and Sheltami (2023) examined UAV operational limitations by establishing safety protocols to enhance UAV operational potential. Their paper adds value by demonstrating that frequency optimization reduces network traffic, promoting regulatory adherence in dense UAV flight operations (Ahmed and Sheltami 2023).

Gao and Wang (2023) proposed rapid deployment methods for UAV base stations in disaster scenarios. Their study demonstrated successful UAV deployment methods; however, frequency optimization provides post-deployment enhancements. These integrated methods show great potential to boost UAV network efficiency during disaster response operations (Gao and Wang 2023).

Despite its achievements, the present study exhibits multiple methodological deficiencies. While the adaptive frequency management algorithm was validated experimentally and through simulations, direct testing under severe conditions, including natural disasters and extreme weather situations, was lacking. Future research should incorporate outdoor testing under extreme conditions, as field observations are necessary for properly

evaluating algorithm stability.

The investigation concentrated on frequency adaptation exclusively while omitting complementary network optimization methods including trajectory planning (Prasad and Ramkumar 2023), power control (Tang, Zhang, and He 2022) and beamforming (Sabzehali et al. 2022). Simultaneous application of different network optimization strategies has the potential to deliver better network efficiency results.

Additional research is needed to analyze long-term UAV operational sustainability following continuous frequency adjustments beyond existing energy efficiency analysis. The combination of energy efficiency and network throughput performance is critical in delay-sensitive UAV-IoT networks, as demonstrated by Yeduri et al. (2023). Further studies should explore the energetic consequences of operating frequency management in real-time (Yeduri et al. 2023).

UAV operations require proper regulatory frameworks, which this analysis has not considered. Szira et al. (2023) highlighted the necessity for UAV technology to adhere to European Union rules for proper implementation. Integrating regulatory compliance into frequency management algorithms would improve the practical usage of this research (Szira et al. 2023).

UAV systems operate in heterogeneous networks comprising different communication protocols, diverse hardware limitations, and environmental factors. Butilă and Boboc (2022) described the demanding nature of urban UAV operations, as diverse spatial conditions can alter network operational quality. Algorithm performance and scalability can be improved by accepting network heterogeneity (Butilă and Boboc 2022).

The study generates key insights guiding the development of UAV-based communication networks. The effectiveness results from adaptive frequency management create a foundational approach for developing robust network platforms across multiple deployment situations. Additional research should combine multiple optimization strategies, perform extensive outdoor tests, and adapt programs for regulatory and operational needs to achieve practical implementation. Frequency management implemented with trajectory planning methods (Prasad and Ramkumar 2023) and power control mechanisms (Tang, Zhang, and He 2022) and advanced modulation methods (Gopi et al. 2021) will create next-generation UAV networks that adapt to diverse operational needs.

6. Conclusion

The article examines how adaptive frequency management algorithms integrate into UAV-based communication networks to enhance their operational capabilities across various site requirements. By employing dynamic interference resolution, resource optimization, and network reliability functions, the proposed algorithm yields improved results in latency, throughput, packet loss rates, and energy efficiency. Experimental results indicate that adaptive algorithms provide enduring benefits when addressing crucial UAV network challenges in pressure-induced situations.

Healthcare facilities have developed a flexible, optimized UAV network administration system that provides coverage from rural settings to urban areas and emergency medical operation zones. Real-time frequency selection tools act as fundamental mechanisms to minimize connection disturbances, ensuring secure networks and reliable data transfer rates. The algorithm demonstrates high scalability and low-latency operations, enabling numerous UAVs to collaborate on complex mission requirements. Research findings establish UAV networks as practical tools for deployments in disaster management and urban density monitoring.

Research results suggest that modern network systems require flexible capabilities based on observed practical data. The global expansion of UAV networks will depend on systems that enable instantaneous, real-time adaptive responses. Research methods from this study drive improvements in UAV systems by advancing next-generation telecommunications solutions across multiple operational areas.

Future research should explore new pathways. A higher level of network effectiveness can be achieved by integrating trajectory planning technology with frequency management systems, power control, and beamforming optimization solutions. The hybrid integration of UAV networks with predictive machine learning models shows potential for extensive research achievements focused on adaptable problem-solving and predictive task detection capabilities.

Testing the system under extreme operating conditions, such as large-scale disaster events and harsh weather scenarios, is necessary to enhance algorithm robustness. To optimize the practical use of this algorithm, both heterogeneous network issues and regulatory compliance must be addressed. The examination of sustainable energy practices, alongside

adjustments for frequency adaptation trade-offs, will lead to practical field applications for the proposed system.

The article demonstrates that adaptive frequency management techniques can optimally transform the structure of UAV networks. It advances UAV-based telecommunications by creating an essential framework that addresses interference issues and performs optimally across various deployment scenarios. Future research should combine multiple connectivity capabilities with extensive field tests to maximize UAV network deployment for contemporary communication requirements.

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