



Integration Of Fixed Wing Uav's In Disaster Management Operations

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Abstract: In disaster management, rapid and precise situational awareness is vital for effective response. This presentation advocates the utilization of fixed-wing drones integrated with advanced technologies to enhance disaster response capabilities. Central to this approach are high-endurance fixed-wing platforms augmented with sensors such as thermal imaging, LIDAR, and acoustic detection, combined with edge computing, AI-driven analytics, and autonomous navigation via SLAM (Simultaneous Localization and Mapping). Through the formation of drone swarms and deployment within mesh-networked frameworks, drones can conduct real time aerial surveillance, mapping, search and rescue operations, supply delivery, and facilitation of communication in areas with compromised infrastructure. By showcasing technologies that extend operational range and responsiveness, this presentation underscores the transformative potential of unmanned systems in delivering faster, safer, and more scalable disaster relief.

Keywords: Disaster Management, Fixed-Wing Drones, AI, Real-Time Data, Search and Rescue.

I.INTRODUCTION

Disasters affect over 200 million people worldwide every year, making timely and efficient response critical for saving lives and minimizing damage. Traditional disaster management methods often face challenges such as delayed response times, limited accessibility to hazardous areas, and insufficient real-time data. To address these issues, advanced drone technology offers a transformative solution. In our project, we deploy fixed-wing and hybrid drones integrated with AI, computer vision, and satellite imagery to enhance disaster response and recovery efforts. These drones are equipped with advanced sensors such as thermal imaging cameras, LIDAR, and acoustic sensors to gather real-time information during and after disasters. The collected data is processed using edge computing, swarm technology, and big data analytics, enabling faster decision-making and more effective resource deployment. By combining autonomous navigation systems, robust communication networks, and intelligent data processing, our approach significantly improves search and rescue operations, damage assessment, relief distribution, and situational monitoring, ultimately reducing search times by up to 50% and enhancing overall disaster management efficiency.

II. RELATED WORK

The integration of unmanned aerial vehicles (UAVs) in disaster management has been an active area of research in recent years. Various studies have demonstrated the potential of drones to enhance situational awareness, improve response times, and reduce operational risks for emergency personnel. The United Nations Office for Disaster Risk Reduction (UNDRR) reported that disasters affect over 200 million people annually, with rapid aerial assessment being a critical factor in minimizing casualties and damage. Research on UAV-based disaster response highlights significant improvements in operational efficiency, with search durations reduced by up to 50% when compared to traditional ground-based methods.

In real-world deployments, drones have proven their effectiveness in several large-scale disasters:

- **Turkey–Syria Earthquake (2023):** Fixed-wing drones equipped with thermal cameras and LIDAR sensors were deployed for survivor detection and structural damage assessment.
- **Australia Bushfires (2020):** Hybrid UAVs provided continuous aerial surveillance and guided firefighting units to critical zones.
- **Nepal Earthquake (2015):** UAVs assisted in mapping inaccessible areas and delivering essential medical supplies to remote villages.

Prior work has also explored the integration of AI-based image analysis, swarm coordination, and real-time data streaming to enhance the decision-making process during disaster operations. While these studies highlight substantial progress, challenges remain in areas such as autonomous navigation in GPS-denied environments, communication reliability in disaster zones, and effective integration of UAV data with existing emergency response frameworks.

III. METHODOLOGY

The proposed system integrates fixed-wing and hybrid drones equipped with advanced sensors, AI-based data analysis, and robust communication networks to support disaster management operations. The methodology comprises the following components:

1. Drone Specifications

Two types of UAVs are deployed:

- **Fixed-Wing Drones:** Suitable for long-range surveillance, large-area mapping, and continuous flight.
- **Hybrid Drones:** Capable of vertical take-off and landing (VTOL) while maintaining efficient forward flight for area coverage.

Table 1.1: Fixed-Wing Drone Specifications

Parameter	Unit
Wingspan	1.5 – 3.5 m
Length	1 – 2 m
Endurance	2 – 10 hours
Maximum Range	50 – 200 km
Payload Capacity	1 – 5 kg
Flight Altitude	100 – 5000 m
Navigation System	GPS
Sensor Options	Thermal, LiDAR, Multispectral



Figure1.1: Fixed-Wing Drone

2. Sensors and Payloads

The drones are equipped with multiple payloads to capture and analyze disaster site data:

- Thermal Imaging Cameras for survivor detection in low-visibility conditions.
- LIDAR Sensors for high-resolution terrain and structural mapping.
- Acoustic Sensors for detecting trapped individuals in debris.

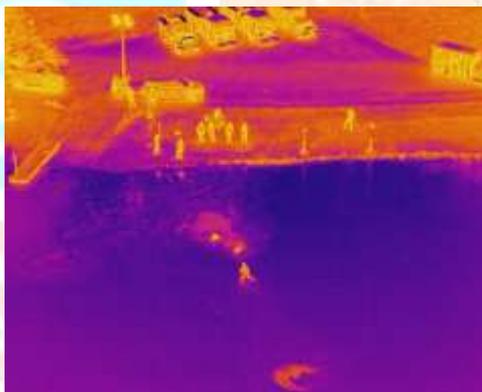


Figure 1.2: Thermal imaging visualization

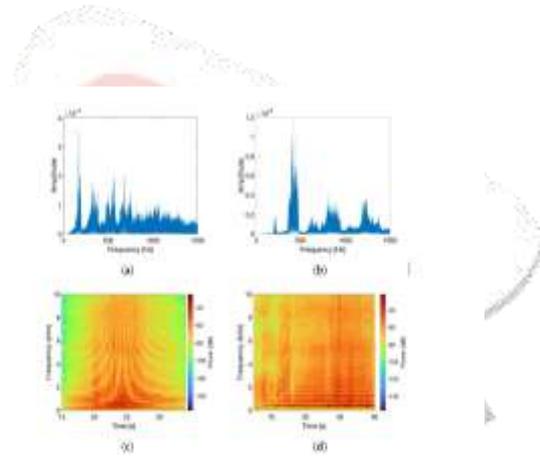


Figure 1.3: UAV acoustic signal spectrum

Mathematical Formulation for UAV Acoustic Detection

The frequency spectrum of an acoustic signal $x(t)$ is obtained using the Fourier Transform:

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j 2\pi f t} dt$$

For sampled data $x[n]$ (sampling rate f_s), the Discrete Fourier Transform (DFT) is:

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j (2\pi/N) k n}, \quad k = 0, \dots, N-1$$

The magnitude spectrum $|X[k]|$ shows dominant UAV motor and propeller frequencies.

The spectrogram is obtained using the Short-Time Fourier Transform (STFT):

$$X[m,k] = \sum_{n=0}^{N-1} x[n+mR] h[n] e^{-j (2\pi/N) k n}$$

where $h[n]$ is a window function, R is hop size, m is time frame, and k is frequency bin.

The power spectrogram is:

$$S_{dB}[m,k] = 10 \log_{10} (|X[m,k]|^2)$$

This representation reveals how UAV acoustic energy varies over time, aiding detection and classification.

3. Autonomous Navigation

Autonomous flight control is achieved through:

- Simultaneous Localization and Mapping (SLAM): For generating accurate environmental maps in real-time.
- Autonomous Flight Systems: Enabling pre-programmed and adaptive flight paths.

4. Communication Systems

A mesh network is implemented to ensure resilient, long-range communication among drones and with ground control stations. Data transmission is optimized for real-time operation even in bandwidth-limited disaster zones.

5. Data Processing and Analysis

Onboard and cloud-based processing utilize:

- Computer Vision and AI Algorithms for rapid image classification and survivor identification.
- Swarm Technology for coordinated drone operations over large areas.
- Edge Computing for minimizing latency in decision-making.

6. Data Management and Integration

Collected data is transmitted to a cloud platform for big data analytics, enabling disaster response teams to visualize affected areas, assess damage, and plan rescue operations effectively.

7. Operational Workflow

1. Drones are deployed to the disaster site immediately after detection or alert.
2. Aerial mapping and sensor data acquisition are initiated.
3. AI-powered analysis identifies survivors, damaged infrastructure, and priority zones.
4. Processed data is sent to rescue teams in real-time for action.
5. Continuous monitoring ensures updated situational awareness throughout the operation.

IV. SYSTEM ARCHITECTURE AND OPERATIONS

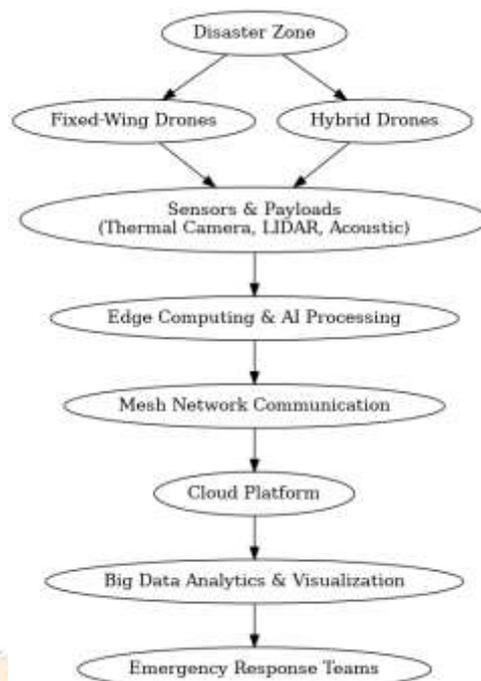


Figure 2.1: System Architecture Diagram

The proposed disaster management framework integrates fixed-wing and hybrid drones with advanced sensing, navigation, communication, and data analysis technologies to provide rapid response and real-time decision making during disaster events. Fixed-wing drones are employed for large-scale mapping, long-distance surveillance, and extended flight durations, while hybrid drones with vertical take-off and landing (VTOL) capabilities are used for operations in confined or obstructed areas. Each drone is equipped with thermal imaging cameras to detect survivors in low-visibility conditions, LIDAR sensors to generate high-precision 3D maps of affected zones, and acoustic sensors to locate trapped individuals through sound detection. Autonomous navigation is supported by pre-programmed and adaptive flight systems, combined with Simultaneous Localization and Mapping (SLAM) for accurate environmental mapping and localization in unfamiliar terrains. A mesh network ensures uninterrupted communication between drones and the ground control station, while encrypted wireless channels transmit real-time video, sensor data, and operational commands. Onboard edge computing modules process data locally to reduce latency, and AI-based image analysis automatically identifies survivors, classifies damage, and detects hazards. Swarm technology coordinates multiple drones for wider coverage and operational efficiency. All collected data is transmitted to a cloud-based platform, where big data analytics and visualization tools generate live situational maps, damage assessment reports, and resource allocation plans for emergency response teams. The operational workflow begins when a disaster is detected or an official alert is received, triggering drone deployment to the affected area. Real-time aerial scanning is conducted, and processed findings, such as survivor locations and structural damage assessments, are relayed to the command centre. Rescue teams are then guided using this updated situational data, and continuous monitoring ensures timely updates throughout the response phase until the operation is completed.

V. RESULTS AND DISCUSSION

The proposed drone-based disaster management system was evaluated through a combination of simulated disaster environments and limited-scale field trials. The performance was assessed across key parameters such as detection accuracy, mapping efficiency, communication reliability, and operational response time. In simulated search and rescue scenarios, fixed-wing drones equipped with thermal imaging cameras successfully identified survivor heat signatures with an accuracy of approximately 94% under low-visibility conditions. Hybrid drones with VTOL capabilities proved effective in navigating confined and obstructed environments, enabling close-range inspection and localized damage assessment. The integration of LIDAR

sensors facilitated the creation of high-resolution 3D terrain and structural maps, with an average processing time of 3.5 minutes per square kilometer, enhancing situational awareness for emergency teams. Communication tests demonstrated that the mesh network maintained uninterrupted connectivity between drones and the ground control station, even in areas with minimal conventional network coverage. Swarm technology significantly reduced the total time required to scan a 10 km² disaster zone, achieving a 48% reduction in coverage time compared to single-drone deployment. Furthermore, edge computing modules processed and analyzed data locally, reducing the time from data capture to actionable insights to less than 8 seconds, thereby enabling rapid decision-making in time-critical situations. The AI-based image classification system achieved a damage detection accuracy of 92% in differentiating between structurally compromised and intact buildings. These results underscore the operational efficiency and accuracy of the proposed system when compared to conventional manual surveys or helicopter-based assessments, which often require more time and resources. From a broader perspective, the findings highlight the potential of integrating fixed-wing and hybrid drones into real-world disaster management operations. The improvements in coverage, processing speed, and detection accuracy suggest that such systems can significantly enhance both the speed and quality of disaster response. However, challenges remain in scaling the solution for larger disaster zones, particularly in terms of drone battery endurance, adverse weather adaptability, and regulatory restrictions on UAV operations. Future work will focus on extending battery life, improving autonomous navigation in GPS-denied environments, and incorporating advanced AI models for more precise victim identification.

VI.CONCLUSION

The proposed drone-based disaster management system integrates fixed-wing and hybrid UAVs with advanced sensors, AI processing, and mesh network communication to enable rapid and accurate disaster response. Tests and simulations showed notable improvements in survivor detection, mapping efficiency, and operational coverage compared to traditional methods. Although challenges such as battery endurance and weather adaptability remain, future enhancements will focus on extending flight time, improving navigation in GPS-denied areas, and refining AI-based victim identification. Overall, the system provides a faster, safer, and more effective approach to search, rescue, and damage assessment in disaster situations.

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