

Integrating Sustainable Technology, Artificial Intelligence, and Advanced Communication Networks for Climate Change and Disaster Management

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ABSTRACT

Since the start of the 21st century, the issues of climate change and disaster management have become major areas of concern in the world. The main factors responsible for the environmental imbalance that is taking place are rapid industrialisation, urbanisation without limits, and over-exploitation of natural resources. As a result, the weather is becoming more and more erratic and severe. In addition to becoming more frequent, the power of floods, droughts, cyclones and wildfires is also increasing. There are new problems in the world, and these problems can only be solved by new solutions.

The main factors which could have played the leading role in controlling the situation during the good times are sustainable technologies, smart monitoring systems and AI-driven prediction models. These resources are designed to 'slow down' the threats, attract the readiness and make the response easier. The report is about how the use of eco-friendly technologies, predictive AI and communication systems can provide the disaster management with the improvements they require

Keywords: Sustainable technology, Artificial intelligence (AI), Advanced communication networks, Climate change adaptation, Climate change mitigation, Disaster management

Introduction

Climatic changes, the effects of which are being visibly detected at an ever speedier rate, are causing numerous, never-seen-before problems for mankind all over the world (which is also true for the local societies). These disasters are accordingly at the core of such troubles as loss of life, the weakening of economic welfare, and sustainable development and financial stability being put at risk. All levels of government being forced, necessarily, to completely redo their methods of disaster management is the only solution – so, their collective imperative has to be a radical one: they need to stop using the traditional, eco-xity-

centrist response and instead opt for the one that is based on intelligent and even more technologically advanced adaptation made possible by the latest developments in clean tech, smart monitoring systems, AI, and modern communication networks. By abandoning the conventional, ecocentric approaches of disaster management for futurist, technology-driven methods that can be enabled by the latest developments in green technology, smart monitoring systems, AI, and modern communication networks, governments of all levels can not only meet but exceed the urgent need to cut climate-related risks. This work, "Integrating Sustainable Technology, Artificial

Intelligence and Advanced Communication Networks in Climate Change and Disaster Management", indeed, explores the opportunities that these existing artifices could bring together with those already conceived for handling catastrophe situations. It offers the role of a comprehensive analysis of sustainable methods, forecasting through AI, monitoring through IoT, and stable communication networks in tying together disaster risk reduction and early warning with rapid, resilient recovery under the climate change scenario. In this part, the authors have reviewed various articles related to WSAN/IoT (Adeel et al., 2019)¹, applying machine learning to climate change adaptation (Rolnick et al., 2019)², change detection using the remote-sensing-based prototype (Oh et al., 2023)³, an analysis of foundation models in cases of extreme environmental events (Zhao et al., 2025), and⁴ deep learning techniques in three-dimensional flood mapping (Jia et al., 2025).⁵ This investigation through both frontier technologies and real-world case studies allows a full picture of the promises and the issues that remain when trying to build disaster-resilient societies. The paper is structured according to academic conventions, featuring a full-fledged introduction, a description of climate change effects, a review of sustainable technologies pertinent to the issue, the elaboration of the intelligent monitoring system and AI, an evaluation of robust communication networks, a depiction of the situation on the ground, and a concluding part indicating research topics and inviting international cooperation.

Climate change is rapidly worsening with its far-reaching effects and numerous connections to the increase in the number and the intensity of natural disasters that spread all over the world. These are wildfires, droughts, hurricanes, and floods, just to name a few. Disasters of such kind threaten human lives, destroy the material well-being of people's lives, and jeopardise the possibility of lasting development and financial stability.

The Growing Impact of Climate Change on Catastrophes

Changing Global Weather Patterns

The climate change caused by humans has reshaped the weather all over the world, which has resulted in extreme events getting more powerful, and they can be less predictable and occur with higher frequency. The IPCC AR6 Synthesis Report (2023)⁶ strongly states that, among other things, the cases of drought, storms, wildfires, and flooding have become more powerful and frequent, and the report gives a warning of disastrous consequences if the emission of greenhouse gases is not reduced dramatically (Rolnick et al., 2019; Oh et al., 2023).^{2,3} The proof is enough: the disasters caused by climate change are becoming the new normal instead of being rare exceptions. Only in the period

from 2018 to 2022, floods were the cause of more than 20,000 deaths or disappearances, and the total financial losses exceeded \$189.8 billion USD. At this moment, about 1.81 billion people—almost 23% of the global population—are living in areas that are directly exposed to the risk of flooding (Jia et al., 2025).⁵ The impacts of climate change and urbanisation have led to the situation where most of the people and the important facilities are located in areas that are vulnerable, which, in turn, increases the vulnerability and makes the needed emergency relief more complex.

Cascading and Compound Events

One aspect that makes climate-related disasters more complex is the combination of compound and cascading events. For example, when a flood comes after a cyclone or drought is made worse by heatwaves, the after-effects can intertwine and amplify so that they can extend socially, ecologically, and economically to a much greater extent than those caused by the single events (Zhao et al., 2025).⁴ Among others, these compounding effects underpin the essential requirement for holistic disaster risk management systems that can not only foresee but also monitor and react to the coexisting hazards.

The Necessity for Proactive Disaster Management

One of the main weaknesses of traditional disaster management that relies on reactions is the huge size and complexity of climate-related disasters that have become modern times. Such a change needed is obvious: the first preventative, data-driven, and collaborative management that is able to make use of the latest technological developments to foresee, adjust to, and lessen the changing risks (Adeel et al., 2019; Rolnick et al., 2019)^{1,2} would be the correct way to go.

Sustainable Development and Community Resilience Against Disasters

Sustainable Infrastructure and Renewable Energy

Sustainable development practices greatly contribute to both lowering the risk of disasters and a community's capacity to bounce back. The use of renewable energy systems such as wind, solar and hydropower in the basic infrastructure not only leads to low emissions but also ensures energy stability during and a few hours after a disaster (Rolnick et al., 2019).² More distributed energy production via microgrids and decentralised renewable resources improves the number of redundancies, reduces the chance of service interruption over wide areas, and therefore allows more rapid recovery. Besides this, environmentally safe city design, which is made up of permeable surfaces, green areas and adaptable building codes, will encourage the reduction of the community's exposure to hazards such as the heatwave or flood. In the areas that are prone to flooding, practices like wetland

restoration, installing green roofs, and using sustainable drainage systems can mitigate runoff, decrease the peak of floods, and make natural water management much better (Jia et al., 2025).⁵

Circular Economy and Waste Management

Disaster resilience, particularly in cities, requires efficient waste handling, as the aftermaths of disasters usually involve both the debris and the contamination deepening the disaster effects. Such types of waste management as circular economy, which are based on resource recovery, recycling and waste reduction, are not only helpful in the recovery period from an environmental point of view but also make a positive contribution to the local sustainable economies and the public health (Rolnick et al., 2019).²

Community Involvement and Policy Frameworks

Disaster management systems with the power to bounce back from challenges need the support of a regulatory environment that is conducive, good policy instruments, and vibrant community participation. It is the responsibility of the governments at the national and local levels, among other things, to offer incentives for resilient construction, ensure the latest building standards are in place through inspections, and encourage community-led risk reduction programmes. Through the engagement of local people in planning, supervision, and reaction, interventions, in general, become more socially acceptable and more adapted to the peculiarities of the local area, and they are kept for the long run (Adeel et al., 2019).¹

Data-Informed Disaster Management Through Intelligent Monitoring Systems

IoT Technologies and Wireless Sensor Networks (WSNs)

The confluence of the Internet of Things (IoT) with wireless sensor networks (WSNs) has opened the gates wide to the next era in disaster monitoring, providing real-time data in high volume and in high resolution from various sources. Wireless Sensor Networks (WSNs) are made of independent sensor nodes, each equipped with a microcontroller, environmental sensors, and wireless radios, thus making it possible to deploy them in inaccessible areas where they can measure pressure, temperature, vibrations, and humidity (Adeel et al., 2019).¹ The data platforms enabled by the IoT allow different data streams—such as the UAV (Unmanned Aerial Vehicle) feed, satellite images and ground sensors—to flow effortlessly into synchronised monitoring platforms. Such instruments make possible the constant oversight of the risk indicators; they also constitute a quick damage assessment tool in the aftermath of the event and early warning capabilities (Adeel et al., 2019; Zhao et al., 2025).^{1,4} For example, the installation of WSNs in the landslide-susceptible regions has been the main factor leading to the

issuing of multi-level alerts based on immediate geological data, thus evacuation orders being accurate and timely. In the case of a flood in the city, IoT-based monitoring can locate the critical water levels and the failing infrastructure that can lead to the targeted allocation of resources and interventions.

Change Detection and Remote Sensing

Remote sensing technologies that employ satellite and aerial imagery are the tools that cannot be done without when it comes to broad-scale disaster impact assessment and monitoring. One of the AI-powered change detection methods, like prototype-orientated unsupervised change detection (PUCD), makes it possible to detect changes brought about by a disaster automatically by comparing images taken before and after the event. Using foundational models such as DINOv2 along with sophisticated segmentation methods like the Segment Anything Model (SAM), PUCD systems are able to delineate the areas of the impact, the regions that are flooded and the structures that are damaged with great accuracy and a very low requirement for manual input. The mentioned methods are extremely helpful in the situations where a very quick reaction is needed but the disaster types and affected areas are different, or when a large amount of manual labelling is not feasible. In addition, works such as ExEBench provide the standardised protocols and datasets for testing the performance of foundational models in different scenarios of extreme events on earth, including heatwaves, storms, floods, and wildfires. These benchmarks empower the development and tuning of machine learning models that are specifically aimed at disaster prediction, observation, and identification.

Deep Learning for 3D Flood Mapping

In the past, flood delineation was heavily dependent on mapping that visually showed the areas that were affected by the flooding in a 2D fashion. Such maps, however, gave a very limited view of the flooding's severity and even failed to provide the depth of the water. Deep learning (DL) advances have made it possible to generate three-dimensional (3D) flood maps that not only allow for a more comprehensive visualisation of a flood but can also depict the floodwater's depth as well as the geographical area affected by the flood. Such 3D models employ the simulated output, the rainfall data, DEMs, and satellite images for the estimation of a flood's depth and area.

Task decomposition methods decouple flood extent identification from depth estimation, giving the possibility of modular optimisation and the easier integration with existing workflows. However, end-to-end DL models can directly draw 3D flood maps from multi-source inputs, which leads to the computational speed increase and

the error propagation reduction. All these new ways of thinking provide the support for the implementation of flood prediction in real time, urban risk assessment, and city planning, which, consequently, are the key to efficient and targeted disaster management. Still, with all the progress made, obstacles related to the model's interpretability, data scarcity, and the smooth integration of DL models with traditional hydrodynamic simulations remain. These difficulties have led to research efforts that are continually working on building stronger validation frameworks, increasing the adaptability of the model, and enriching datasets in order to secure the operational deployment and reliability of any 3D flood mapping.

The Role of Machine Learning and Artificial Intelligence in Disaster Management

Early Warning and Predictive Modelling

Machine learning (ML) and artificial intelligence (AI) are two technologies that have become the most influential instruments in not only adjusting to climate change but also in reducing it. In disaster management, AI-powered predictive models use real-time sensor inputs, historical datasets, and physical models to anticipate the progression, occurrence, and impacts of extreme events with gradually increased accuracy.

For example, energy systems are using ML algorithms to optimise the forecast of electricity production and consumption, which goes a long way in the integration of variable renewable sources and the reduction of the use of polluting standby generators. Similar methods are utilised in disaster management to forecast the severity and the beginning of such hazards as heatwaves, cyclones, and floods, allowing the taking of proactive measures such as the allocation of resources, evacuation, and the protection of infrastructure.

Some of the advanced ML models, such as hybrid physics-ML frameworks, deep neural networks, and transformers, are very capable of capturing the intricate non-linear relationships, i.e., across the various temporal and spatial scales. These models are highly effective when they combine the domain-specific knowledge, for example, the climatic and the weather drivers, for the optimisation of the system-level objectives like the reduction of cost, risk, and the control of emission.

Damage Assessment and Change Detection

Change detection that is automated is essential when rapid assessments of the disaster impacts are to be made, especially over large areas or those that are difficult to reach. Unsupervised change detection (UCD) methods, like those used in the PUCD framework, facilitate the comparison of pre-disaster and post-disaster remote sensing images to recognise the areas that are inundated, changes

in the landscape, and damaged structures without the requirement of a large number of pre-labelled datasets. UCD techniques can be made to work in any geography and in any type of disaster by using clustering methods and also by taking feature representations from base vision models. The adoption of semantic segmentation models by further stages helps to redefine the change maps obtained, thus enabling the accuracy of damage delineation and the facilitation of efficient rescue.

Evaluation and Benchmarking

Systematic assessments of the reliability and generalisation of AI models for disaster management are conducted through various benchmarking initiatives, such as ExEBench. ExEBench enables such a comprehensive evaluation of the performance of models across different times, spaces, and even spectral domains by combining datasets that represent a wide range of event types, e.g., heatwaves, floods, cold waves, cyclones, storms, wildfires, and extreme precipitation. These sorts of benchmarks play a critical role in guaranteeing stability, uncovering potential biases, and facilitating the creation of models that can not only efficiently cope with rare or extreme situations but also be able to learn from them since these events provide very little data. Moreover, the establishment of public datasets and standard metrics for evaluation serves as an important bridge for the transfer of knowledge between the field of operations and research, thus facilitating innovation at large.

Future Research and Persistent Challenges

While ML and AI possess the potential to change the landscape of disaster management, their use in the field still encounters many hurdles. Some of these challenges are the incompatibility of AI results with current hydrodynamic models and decision-making systems, the need for more openness and understanding of models, and the lack of sufficiently good quality labelled data. Besides, the possibility of bias in models – which may arise from leaving out essential variables or having unbalanced data for training – calls for continuous confirmation of the validity of the models and communication with stakeholders. The next research that scientists would do should focus on AI that can be understood by the user, support networks that facilitate the involvement of the community and the different areas of knowledge, and models that can be transferred to different fields of study.

Advanced Communication Networks for Disaster Response

Device-to-Device Communication and Wireless Sensor Networks

In order to have a successful early warning system, a well-coordinated response and efficient disaster monitoring, the communication must be not only reliable but also of

low latency. The future technologies of device-to-device communication (D2D) protocols and wireless sensor networks (WSNs)⁷ promise to provide not only a high throughput but also a data interchange resilient to even heavily damaged or degraded environments. The main objective of the fifth-generation wireless (5G) networks is to offer better reconfigurability, flexibility, and resilience that, in turn, make it possible to support the traffic of both machine-type communication (MTC) and human-type communication (HTC).⁸ 5G technology is the enabler of media-independent handover, direct D2D communication, and seamless network transitions. That is why emergency personnel, mobile devices, and sensors can always be connected even when a crisis is going on.

Cognitive Radio and Software-Defined Radio

Software Defined Radio (SDR) technologies offer seamless communication between different standards of technologies, thus enabling the integration of civilian, military, and emergency networks on a single platform. SDRs can quickly modify their functions according to the changing operational requirements, can allocate dynamically the spectrum resources, and can even support more than one waveform simultaneously. Nevertheless, as for disasters, these problems keep persisting: high computational requirements, energy consumption, and cost. Cognitive Radio (CR) additionally makes communication more efficient by smartly handling radio frequency channels, by using resources in the best possible way, and by relieving congestion at the time when there is a high demand for a disaster situation. When used together, these technologies give the most comprehensive and reliable situational awareness and information exchange.

Mobile Disaster Modes and Indoor Positioning

Disasters often ruin the functioning of standard positioning systems which are specially designed for such places that are closed or have many obstacles. One of the main reasons for this downfall is the blockage that occurs in the area. As a result, an advanced indoor positioning system equipped with mobile device sensors and energy-efficient algorithms provides vital location services to the released officers as well as the victims of the disaster, thereby helping them in the search and rescue operation. Mobile phones are the main source of communication during a disaster in both developed and developing regions. Disaster-aware protocols and context-aware computing, when put together, enable the mobile to enter the “disaster mode”, which is a mode whereby the communication methods vary depending on the availability of the network, the battery life, and the mobility of the device. These properties, even when there is no traditional infrastructure, still give up-to-date information on the way to safety, the distribution of resources, and connectivity.

Future Directions and IoT Communication Standards

IoT (Internet of Things) systems use a variety of communication standards, such as 4G Long-Term Evolution (LTE)⁹ and Low Range Wide Area Network (LoRaWAN).

On one side, LoRaWAN is perfect for remote monitoring since it is a low-power, long-range transmission. However, its lower data rates limit the type of applications, for instance, video surveillance. On the other hand, 4G LTE has the advantage of more built-in security and higher data rates, but it still can experience considerable packet loss and delay at times of peak usage or when there is a failure in the infrastructure. The development of intelligent, low-power, high data rate architectures and the shift to 5G networks provide an excellent opportunity for better disaster communication. Besides that, integrating blockchain as a means for transparency and customising networks through the optimal parameter tuning are the new research fields that are set to revolutionise the disaster management system, according to

Practical Application: Case Studies in Technology

Chennai Floods (2015)

The 2015 Chennai floods caused by heavy rainfall beyond normal limits led to the waterlogging of extensive parts of the city and enormous material and human losses. Among other things, this calamity exposed inadequacies in the city's drainage system, urban planning, and early warning system.¹⁰ After the incident, a major step towards improving emergency response and forecasting capabilities was the installation of IoT-enabled sensors and real-time monitoring networks. Part of recovery planning and quick damage estimation was carried out with the help of remote sensing data and satellite images.

Cyclone Fani (2019)

For instance, communication networks, along with predictive modelling, have been the lifeline during the passage of Fani, a tropical cyclone which ranks among the most destructive ones to have hit India recently. The usage of AI-powered early warning systems in conjunction with mass communication procedures made it possible to evacuate people on a large scale very quickly, which had a significant impact on the number of fatalities, hence less than what had been recorded before.¹¹ In this case, D2D communication and wireless sensor networks were instrumental, as they allowed the on-the-spot resource coordination and provided situational awareness.

Japan Earthquake and Tsunami (2011)

The 2011 Tōhoku earthquake and tsunami disaster in Japan has highlighted the need for the infrastructure of the crisis

era to be resilient and for smart monitoring systems to be in place. Seismic and tsunami detectors, along with automated warning systems, were essential elements of the whole network that allowed the emergency response and evacuation to be performed.¹² Moreover, the deployment of AI-based satellite detection of changes enabled rapid mapping of the affected areas, thereby facilitating the distribution of relief as well as the rebuilding process.

Global Impact and Benchmarking

Benchmarking projects like ExEBench have greatly amplified the capability to cross-check and measure the performance of AI models across various types of disasters.¹³ By way of illustration, implementing deep learning-based 3D flood mapping technology has demonstrated remarkable gains in the areas of computational efficiency, prediction accuracy, and practical utility, thus globally establishing the new standards for the agencies involved in managing disasters.

Future Trajectory: Developing a Technologically Advanced and Sustainable Framework

Real-Time Simulation and 5G-Enabled IoT

It is expected that 5G-enabled IoT networks will radically change the disaster management landscape, as they provide ultra-low latency, high bandwidth, and massive device connectivity. Such capabilities allow for on-the-fly aggregation and processing of multi-modal data streams, thereby enabling interactive disaster simulation, heightened early warning, and the agile deployment of the resources.¹⁴ The fabrication of AI-powered “digital twins”—virtual replicas of tangible surroundings—is permitting risk measurement, the portrayal of catastrophe situations, and the arrangement of response strategies. Besides, these efficient machines allow for public engagement, educators’ training, and the preparation of the community, which is the conversion of information into implementation.

Blockchain for Accountability and Transparency

Blockchain technology presents new avenues for improving accountability, traceability, and transparency in disaster relief efforts. By recording resource allocations, aid disbursements, and transactions on immutable ledgers, blockchain can minimise fraud, enhance trust, and streamline coordination among various stakeholders.¹⁵

Policy-Driven, Collaborative Strategies

Due to the complex and dynamic nature of the disaster management relationship with climate change, new policies, interdisciplinary research, and collaboration across national borders are a must-have. In addition, universities, civil societies, businesses, and governments not only need to coordinate their efforts to achieve common goals but also utilise integrated platforms, standardised benchmarks, and open data to maximise the impact of their combined

resources. Capacity building, education, and community involvement are indeed the main promoters of sustaining technological progress and also guaranteeing fairness in resource allocation for disaster management. Collective action founded on robust evidence and employing a flexible governance system is the principal instrument for installing societies that are resilient to climate change’s volatile nature.

Conclusion

Climate change has drastically altered the disaster risk landscape to the extent that the severity, interconnectedness, and recurrence of the extreme weather events have all increased. In order to resolve these complicated issues, a single technology-driven solution that employs not only one but four components, namely sustainable development, intelligent surveillance, advanced communication networks, and artificial intelligence, is essential.

This research quilts various contemporary and practical examples of climate change and disaster management that utilise and demonstrate the revolutionary nature of new technologies in this area. Consequently, the combination of remote sensing, deep learning, AI-powered predictive models, IoT systems, and wireless sensor networks has considerably elevated the capability for impact assessment, early warning provision, and resilient recovery realisation. Besides, the transition of communication networks, that is, everything from WSNs to 5G and even futuristic standards, has reached a stage where there is an absolute guarantee of both information dependability and the continuous flow of information during disasters.

However, there are still some difficult issues that remain alongside the very substantial accomplishments. Some of these issues include the problem of AI integration results with operational decisions, the requirement for model transparency in terms of interpretability, and data insufficiency. The way to success in the future will rely on choosing community-centred, inclusive strategies as a priority, investing in research and development, and cultivating global partnerships.

To summarise, the most efficient means of disaster management in the era of climate change is simply the combination of sustainable technologies, intelligent systems, and the implementation of stable networks. By promoting the values of collective responsibility, transparency, and innovation, the global society can garner a future which will not only be adaptive to forthcoming uncertainties but also be equitable, sustainable, and disaster-resilient.

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