

Original Article

# AI for Predictive Epidemic Modeling and Global Health Crisis Management

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**Abstract** - The frequency and scale of global epidemics and pandemics have underscored the urgent need for advanced predictive and management tools in public health. Traditional epidemic models, while foundational, often face limitations in accurately forecasting disease spread due to their reliance on static assumptions, limited datasets, and slow responsiveness. Artificial Intelligence (AI) offers transformative potential to overcome these challenges by leveraging machine learning, deep learning, and hybrid approaches for predictive epidemic modeling. AI models can analyze large-scale, heterogeneous datasets—including clinical records, genomic sequences, mobility patterns, and social media data—to provide real-time forecasts, detect early outbreak signals, and identify high-risk populations.

Beyond prediction, AI plays a critical role in global health crisis management by optimizing resource allocation, guiding vaccination and treatment strategies, and supporting evidence-based policymaking. Early warning systems powered by AI enable timely interventions, while scenario simulation models assist decision-makers in evaluating the impact of public health measures. Case studies from the COVID-19 pandemic, Ebola outbreaks, and influenza surveillance demonstrate AI's effectiveness in improving situational awareness, enhancing diagnostic accuracy, and mitigating disease spread.

Despite its promise, the integration of AI in epidemic response raises challenges, including data privacy, algorithmic bias, transparency, and ethical considerations. Addressing these issues is essential to ensure equitable, trustworthy, and responsible AI deployment. Future directions involve federated learning, explainable AI, genomic surveillance, IoT integration, and international collaboration to strengthen global epidemic preparedness.

This paper provides a comprehensive analysis of AI applications in predictive epidemic modeling and global health crisis management. It highlights current techniques, case studies, operational benefits, and ethical frameworks, emphasizing the role of AI in transforming health systems from reactive responders to proactive, data-driven, and resilient infrastructures capable of mitigating the impact of future epidemics and pandemics worldwide.

**Keywords** - Artificial Intelligence, Epidemic Modeling, Predictive Analytics, Machine Learning, Deep Learning, Hybrid Models, Global Health, Pandemic Management, Early Warning Systems, Resource Optimization.

## I. INTRODUCTION

The increasing frequency of global epidemics and pandemics has highlighted the vulnerabilities of public health systems worldwide. Infectious diseases such as COVID-19, Ebola, and seasonal influenza not only pose significant threats to human health but also disrupt economies, societies, and international mobility. Traditional epidemic modeling approaches, including compartmental models like SIR (Susceptible-Infected-Recovered) and SEIR (Susceptible-Exposed-Infected-Recovered), have provided valuable insights into disease dynamics. However, these models often rely on static assumptions, limited historical data, and homogeneous population mixing, which can reduce their predictive accuracy during rapidly evolving health crises.

Artificial Intelligence (AI) has emerged as a transformative tool to overcome these limitations, offering data-driven approaches capable of analyzing complex, high-dimensional datasets. By leveraging machine learning, deep learning, and hybrid modeling techniques, AI can process information from diverse sources—such as electronic health records, genomic sequences, social media activity, and mobility patterns—to generate accurate real-time predictions of disease spread. This capability allows health authorities to identify emerging hotspots, forecast healthcare resource needs, and implement timely interventions, ranging from vaccination campaigns to travel restrictions.

Beyond predictive modeling, AI enhances global health crisis management by supporting evidence-based decision-making and optimizing operational responses. Early warning systems, scenario simulations, and resource allocation models enable policymakers to evaluate intervention strategies, anticipate potential bottlenecks, and prioritize high-risk populations. Furthermore, AI facilitates international collaboration by standardizing data collection, enabling rapid sharing of insights, and fostering coordinated responses to transboundary health threats.

Despite its potential, the deployment of AI in epidemic prediction and crisis management raises challenges related to data privacy, algorithmic bias, transparency, and ethical governance. Addressing these issues is essential to ensure equitable, responsible, and effective implementation. This paper explores the applications, benefits, challenges, and future directions of AI in predictive epidemic modeling and global health crisis management, highlighting its role in transforming healthcare systems into proactive, resilient, and data-driven infrastructures capable of mitigating the impact of future epidemics and pandemics.



## II. AI IN GENOMIC SURVEILLANCE OF PATHOGENS

Genomic surveillance is a vital tool for understanding pathogen evolution, tracking mutations, and anticipating potential outbreaks. Traditional genomic analysis methods often rely on manual sequencing and statistical approaches, which can be slow and resource-intensive, limiting their real-time applicability during

epidemics. Artificial Intelligence (AI) has transformed genomic surveillance by providing tools capable of rapidly analyzing large-scale genomic datasets, identifying patterns, and predicting the emergence of new variants.

Machine learning algorithms can process viral and bacterial genomic sequences to detect mutations linked to increased transmissibility, virulence, or drug resistance. By training models on historical outbreak data, AI can forecast likely evolutionary trajectories, allowing health authorities to implement proactive interventions such as targeted vaccination programs, therapeutic development, or containment strategies. For instance, during the COVID-19 pandemic, AI-assisted genomic monitoring enabled rapid identification of variants like Delta and Omicron, guiding public health decisions and vaccine updates.

Deep learning models, including convolutional and recurrent neural networks, capture non-linear and temporal relationships in genomic data, providing accurate predictions even for previously unseen pathogen strains. Hybrid models that integrate genomic data with epidemiological and environmental information can link genetic changes to observed transmission patterns, regional outbreak intensity, and population susceptibility.

AI-driven genomic surveillance also supports international collaboration. Cloud-based platforms enable near real-time sharing of genomic data across countries, fostering coordinated global responses to emerging threats. While challenges such as data standardization, privacy, and sequencing biases remain, AI provides a scalable, efficient, and accurate approach to understanding pathogen evolution, improving epidemic forecasting, and strengthening global health preparedness.

### **III. SOCIAL MEDIA AND MOBILITY DATA FOR EPIDEMIC PREDICTION**

Beyond clinical and genomic data, non-traditional data sources like social media and mobility patterns have become increasingly important for predicting the spread of infectious diseases. AI can analyze these large-scale, heterogeneous datasets to detect early signals of outbreaks, monitor disease progression, and inform public health interventions.

Social media platforms such as Twitter, Facebook, and search engine queries contain valuable information about emerging symptoms, public concerns, and population behavior. Natural Language Processing (NLP) and machine learning algorithms can process text, posts, and search trends to detect unusual disease-related activity, identify outbreak clusters, and assess population compliance with preventive measures. For example, during influenza seasons and the COVID-19 pandemic, AI models successfully utilized social media activity to forecast disease spikes ahead of official reporting, providing valuable lead time for intervention.

Mobility data, derived from mobile devices, GPS tracking, and transportation networks, provides insights into population movement, commuting patterns, and contact rates. AI models can integrate these data with epidemiological information to predict how infections may spread across regions, identify potential hotspots, and evaluate the impact of travel restrictions or social distancing measures. Combining social media and mobility datasets enhances prediction accuracy by capturing both behavioral and spatial dynamics of epidemics.

While these AI applications offer significant benefits, they also raise ethical and privacy concerns. Ensuring anonymization, data protection, and informed consent is crucial to maintaining public trust. Despite these challenges, leveraging social media and mobility data through AI allows for more proactive and timely epidemic response, bridging gaps between traditional surveillance methods and real-world, real-time population behaviors.

### **IV. AI-DRIVEN VACCINE DISTRIBUTION AND PRIORITIZATION**

Efficient vaccine distribution is a critical component of epidemic and pandemic management, particularly when supplies are limited. Artificial Intelligence (AI) has emerged as a powerful tool for optimizing vaccine allocation, ensuring that resources are delivered to populations at highest risk while minimizing wastage. AI-driven models combine epidemiological data, demographic information, mobility patterns, and real-time outbreak trends to identify priority regions and population groups for vaccination.

Machine learning algorithms can simulate multiple distribution scenarios, taking into account factors such as infection rates, population density, healthcare infrastructure, and logistics constraints. By predicting demand and identifying high-risk clusters, AI helps authorities allocate vaccines in a way that maximizes public health impact and reduces transmission. For example, during the COVID-19 pandemic, AI-assisted models guided phased vaccination campaigns, balancing supply constraints with population needs and regional outbreak severity.

AI also enhances operational efficiency in the supply chain. Predictive analytics can anticipate potential shortages of doses, refrigeration capacity, and transportation resources, enabling proactive mitigation measures. Furthermore, AI can optimize vaccination schedules and locations, reducing wait times, improving adherence, and ensuring equitable access across urban and rural areas.

In addition to logistical optimization, AI supports strategic decision-making by evaluating the impact of prioritization strategies on overall epidemic control. Scenario modeling allows policymakers to assess how targeting specific age groups, healthcare workers, or high-transmission regions affects outbreak trajectories. Ethical considerations, such as fairness, equity, and transparency, are integral to AI-driven vaccine distribution strategies, ensuring that vulnerable populations are not disadvantaged.

By integrating predictive modeling, operational optimization, and scenario analysis, AI provides a comprehensive framework for vaccine distribution and prioritization. Its ability to process vast, real-time datasets and generate actionable insights makes it an indispensable tool for global health crisis management, enhancing both the efficiency and equity of vaccination efforts during epidemics and pandemics.

## **V. REAL-TIME SURVEILLANCE AND SMART HEALTHCARE SYSTEMS**

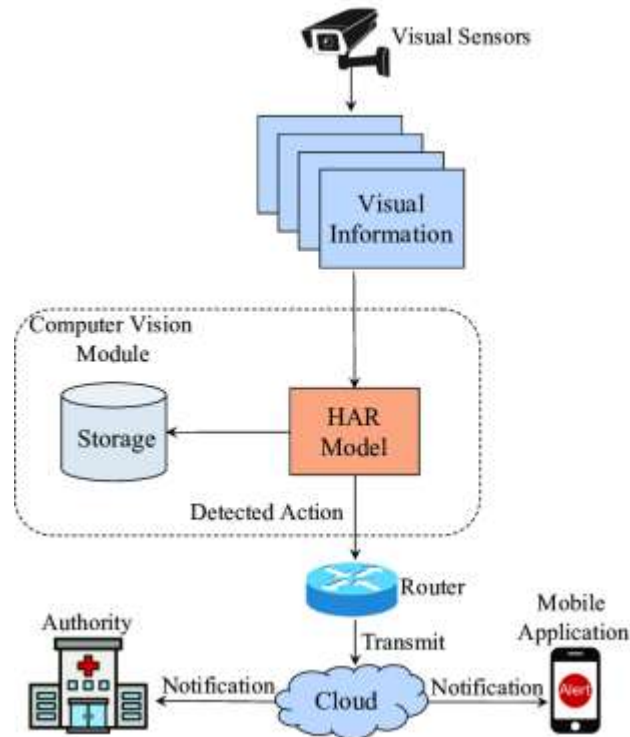
Real-time surveillance is essential for effective epidemic management, allowing health authorities to detect outbreaks early and respond swiftly. Artificial Intelligence (AI) enhances real-time monitoring by integrating data from hospitals, clinics, wearable devices, and Internet of Things (IoT) sensors, creating smart healthcare systems capable of continuous surveillance and rapid alert generation.

AI models can process large volumes of patient data, including electronic health records, laboratory results, and symptom reports, to detect abnormal patterns indicative of emerging infections. This enables proactive measures such as targeted testing, contact tracing, and localized containment strategies. IoT-enabled wearable devices further contribute to surveillance by tracking vital signs, mobility, and exposure risk in real time. Machine learning algorithms analyze these streams to identify trends and predict potential outbreak clusters, offering critical lead time for interventions.

Smart healthcare systems powered by AI also improve hospital preparedness. Predictive analytics can forecast patient influx, ICU occupancy, and resource requirements, helping administrators optimize staffing, bed allocation, and supply management. Additionally, AI-driven dashboards consolidate data from multiple sources, providing health officials with real-time situational awareness, actionable insights, and scenario simulations for decision-making.

The integration of AI into real-time surveillance extends beyond detection to communication and coordination. Automated alerts can notify public health authorities, clinicians, and policymakers about emerging threats, enabling rapid, data-driven responses. Combined with predictive modeling, AI enhances the ability of healthcare systems to respond adaptively to changing epidemic dynamics.

While technical and ethical challenges exist, including data privacy, interoperability, and infrastructure disparities, AI-enabled smart healthcare systems represent a significant advancement in epidemic preparedness. By linking real-time surveillance with predictive analytics and operational optimization, AI transforms health systems into proactive, adaptive, and resilient networks capable of mitigating the impact of infectious disease outbreaks efficiently and effectively.



## VI. BEHAVIORAL ANALYTICS AND AI-POWERED PUBLIC HEALTH COMMUNICATION

Effective public health communication is crucial during epidemics, as timely and accurate messaging can influence population behavior, compliance with preventive measures, and overall epidemic outcomes. Artificial Intelligence (AI) enhances public health communication by analyzing behavioral patterns, social media activity, and population sentiment to design targeted interventions. By processing large-scale data from platforms like Twitter, Facebook, and search engines, AI identifies trends in public perception, misinformation, and adherence to health guidelines.

Natural Language Processing (NLP) algorithms extract insights from posts, comments, and queries, helping health authorities understand public concerns and predict potential behavioral responses to policy measures. Machine learning models can also segment populations based on demographics, geographic location, or risk perception, enabling tailored communication campaigns that resonate with specific communities. For example, during the COVID-19 pandemic, AI-driven sentiment analysis was used to identify regions with vaccine hesitancy, allowing authorities to deploy customized educational messages and outreach programs.

AI also supports real-time feedback and adaptation. By continuously monitoring social media, mobility patterns, and other behavioral indicators, AI models detect changes in compliance or emerging misinformation, allowing public health campaigns to adjust strategies dynamically. Furthermore, AI can optimize message timing, channel selection, and content personalization, increasing the likelihood of engagement and behavioral change.

Ethical considerations are central to AI-powered behavioral analytics. Ensuring privacy, consent, and transparency in data collection is critical to maintain public trust. Despite these challenges, integrating AI into public health communication enhances the ability of authorities to anticipate behavioral trends, counter misinformation, and implement effective interventions. By leveraging predictive insights, AI strengthens the link between epidemiological models and human behavior, ultimately improving compliance, reducing transmission, and mitigating the impact of epidemics and pandemics.

## VII. AI IN RESOURCE FORECASTING AND HOSPITAL LOAD MANAGEMENT

Efficient management of healthcare resources is critical during epidemics, as surges in patient numbers can overwhelm hospitals and compromise care delivery. Artificial Intelligence (AI) plays a transformative role in resource forecasting and hospital load management by analyzing real-time data on patient admissions, disease progression, and healthcare capacity to optimize operational planning.

Machine learning models predict hospital bed occupancy, ICU demand, ventilator usage, and staffing requirements based on current case trends and population risk factors. These predictions allow healthcare administrators to allocate resources proactively, reducing bottlenecks and ensuring that critical care is available where it is needed most. For instance, during the COVID-19 pandemic, AI dashboards were employed in multiple countries to forecast ICU occupancy and ventilator demand, enabling hospitals to adjust staffing schedules, redirect patients, and mobilize additional resources.

AI also facilitates supply chain optimization. Predictive algorithms analyze inventory levels, consumption rates, and delivery schedules to minimize shortages of essential medical supplies, including personal protective equipment, medications, and vaccines. By integrating hospital data with regional epidemiological trends, AI ensures that resources are distributed efficiently across facilities, improving healthcare system resilience during peak infection periods.

Beyond operational optimization, AI supports scenario-based decision-making. Simulation models allow administrators to test different outbreak trajectories, staffing strategies, and patient triage protocols, providing actionable insights for contingency planning. Additionally, AI-driven analytics can identify inefficiencies, highlight critical stress points in hospital workflows, and recommend targeted interventions.

While challenges such as data integration, interoperability, and privacy remain, AI-powered resource forecasting and hospital load management enhance preparedness, reduce strain on healthcare systems, and improve patient outcomes. By combining predictive modeling with real-time operational insights, AI enables hospitals to respond adaptively to epidemic dynamics, ensuring both efficiency and quality of care during public health crises.

### **VIII. AI-POWERED GLOBAL EPIDEMIC RISK MAPPING**

Global epidemic risk mapping is essential for anticipating disease spread, identifying vulnerable regions, and prioritizing interventions. Artificial Intelligence (AI) enhances risk mapping by integrating diverse datasets, including population density, international travel patterns, climate conditions, healthcare infrastructure, and historical disease prevalence. By analyzing these multidimensional factors, AI models generate predictive maps that highlight regions at high risk for emerging outbreaks, enabling proactive resource allocation and public health planning.

Machine learning algorithms detect correlations between environmental, demographic, and mobility data, providing early warning signals of potential epidemic hotspots. For instance, AI models have been used to predict the geographical spread of diseases like dengue, Zika, and COVID-19 by considering seasonal climate variations and human movement patterns. These models allow governments and health organizations to implement targeted interventions, such as localized vaccination drives, vector control programs, and travel advisories.

Deep learning approaches further enhance predictive accuracy by capturing complex non-linear relationships in the data. Recurrent neural networks (RNNs) and long short-term memory (LSTM) models can forecast temporal trends of disease spread, while convolutional neural networks (CNNs) analyze spatial data to identify emerging clusters. Hybrid models that combine classical epidemiological frameworks with AI-driven insights provide interpretable outputs that policymakers can trust for decision-making.

AI-powered risk mapping also facilitates global collaboration. Cloud-based platforms allow for real-time sharing of predictive maps across countries, fostering coordinated responses to cross-border epidemics. While challenges such as data standardization, availability, and privacy exist, AI-driven epidemic risk mapping represents a significant advancement in global health preparedness. By integrating predictive modeling with

geospatial analytics, AI enables early detection, targeted interventions, and efficient allocation of resources, ultimately reducing morbidity, mortality, and societal disruption during epidemics and pandemics.

#### **IX. ETHICAL AI FOR CRISIS RESPONSE**

The use of Artificial Intelligence (AI) in epidemic management presents immense potential but also raises critical ethical considerations. Ethical AI for crisis response emphasizes fairness, transparency, accountability, and privacy in the design, deployment, and application of AI models. These principles are essential to ensure that AI supports equitable public health decisions and does not exacerbate existing inequalities.

Data privacy is a central ethical concern, as AI systems often rely on sensitive health, genomic, and mobility data. Protecting personal information through anonymization, secure data storage, and compliance with international regulations such as GDPR and HIPAA is crucial to maintain public trust. AI models must also address algorithmic bias, which can arise from historical data that reflect healthcare disparities. Biased predictions may disproportionately affect vulnerable populations, leading to inequitable distribution of resources or misinformed policy decisions.

Transparency and explainability are equally important. Complex AI models, particularly deep learning networks, can operate as “black boxes,” making it difficult for health authorities to understand the reasoning behind predictions or recommendations. Explainable AI (XAI) methods provide interpretable outputs, enabling policymakers to make informed decisions while maintaining accountability.

Ethical AI also requires consideration of the social and moral implications of automated decision-making. For example, prioritizing vaccine distribution or resource allocation involves value-laden choices that must balance effectiveness with fairness. AI should support, not replace, human judgment in these contexts.

International collaboration and standardized ethical frameworks are essential for responsible AI deployment in global health crises. By integrating ethical guidelines, robust governance, and continuous model evaluation, AI can be used responsibly to enhance epidemic preparedness and response. Ethical AI ensures that technological innovation aligns with societal values, promoting trust, equity, and effectiveness in managing public health emergencies.

#### **X. INTEGRATION OF AI WITH TELEMEDICINE FOR EPIDEMIC RESPONSE**

Telemedicine has become an essential component of healthcare delivery during epidemics, enabling remote diagnosis, consultation, and monitoring while minimizing the risk of disease transmission. The integration of Artificial Intelligence (AI) with telemedicine platforms significantly enhances their effectiveness by providing predictive insights, real-time monitoring, and personalized care. AI algorithms analyze patient-reported symptoms, electronic health records, and wearable device data to triage cases, prioritize high-risk patients, and recommend appropriate interventions.

Machine learning models can identify patterns in patient data that indicate early signs of infection or disease progression. For instance, AI-driven symptom checkers and diagnostic tools can flag potential COVID-19 cases for further testing, reducing the burden on healthcare facilities. Natural Language Processing (NLP) algorithms enhance telemedicine consultations by interpreting patient descriptions, extracting critical information, and supporting clinical decision-making.

AI also optimizes telemedicine resource allocation. Predictive models can forecast patient demand, enabling healthcare providers to schedule virtual appointments efficiently, manage clinician workloads, and ensure timely responses. Integration with wearable and IoT devices allows continuous monitoring of vital signs, enabling early detection of deterioration in high-risk patients and proactive interventions.

Furthermore, AI supports population-level epidemic management by aggregating telemedicine data to monitor trends, identify hotspots, and guide public health strategies. The anonymized and aggregated data can feed into broader predictive models, enhancing surveillance and outbreak forecasting.

Challenges include ensuring data privacy, interoperability, and equitable access to digital health services, particularly in low-resource regions. Despite these obstacles, the combination of AI and telemedicine represents a transformative approach to epidemic management. By enabling remote, data-driven, and personalized care, this integration reduces transmission risk, enhances patient outcomes, and strengthens healthcare system resilience during epidemics and pandemics.

#### **XI. AI FOR PREDICTIVE MODELING OF ANTIMICROBIAL RESISTANCE (AMR)**

Antimicrobial resistance (AMR) poses a growing global health threat, undermining the effectiveness of antibiotics and complicating the management of infectious diseases. Predictive modeling of AMR using Artificial Intelligence (AI) offers a proactive approach to understanding, monitoring, and mitigating the emergence of resistant pathogens. By analyzing genomic data, clinical records, prescription patterns, and environmental factors, AI models can forecast the likelihood of resistance development and guide targeted interventions.

Machine learning algorithms identify patterns in pathogen genomes that are associated with resistance traits, enabling early detection of emerging drug-resistant strains. Predictive models can also analyze hospital and community prescription data to anticipate regions at risk for AMR proliferation, allowing healthcare authorities to implement stewardship programs, optimize treatment protocols, and reduce unnecessary antibiotic usage.

Deep learning approaches enhance predictive capabilities by integrating heterogeneous datasets, including laboratory test results, patient demographics, and environmental conditions, capturing complex relationships that traditional statistical methods may miss. Hybrid models that combine epidemiological insights with AI-driven predictions offer actionable guidance for infection control, resource allocation, and public health policy.

Global surveillance of AMR is further strengthened through AI-powered platforms that aggregate and analyze data from multiple sources, enabling real-time monitoring and coordinated response strategies. By forecasting potential resistance trends, AI helps healthcare systems prepare for outbreaks of resistant infections, adapt treatment guidelines, and prioritize research efforts for new therapeutics.

Challenges include data standardization, quality, and privacy concerns, as well as ensuring equitable access to predictive tools in resource-limited settings. Despite these challenges, AI-driven predictive modeling of AMR represents a significant advancement in public health preparedness. By anticipating resistance patterns, guiding interventions, and informing global strategies, AI contributes to preserving the efficacy of antibiotics, reducing morbidity and mortality, and strengthening health system resilience against infectious disease threats.

#### **XII. CONCLUSION**

Artificial Intelligence (AI) has emerged as a transformative force in epidemic prediction and global health crisis management, offering unprecedented capabilities to enhance public health preparedness and response. Traditional epidemiological models, while valuable, often face limitations in accuracy, scalability, and responsiveness. AI addresses these challenges by integrating machine learning, deep learning, and hybrid modeling approaches to process vast, heterogeneous datasets, providing real-time insights into disease dynamics, outbreak hotspots, and population risk profiles.

Through predictive epidemic modeling, AI enables early detection of emerging infections, identification of high-risk populations, and optimization of healthcare resource allocation. Case studies from recent outbreaks, including COVID-19, Ebola, and influenza, demonstrate the effectiveness of AI in forecasting transmission trends, guiding vaccination campaigns, improving diagnostic accuracy, and supporting evidence-based policymaking. AI's ability to incorporate non-traditional data sources—such as genomic sequences, mobility patterns, and social media analytics—further enhances situational awareness, offering timely interventions and mitigating the impact of epidemics on global populations.

Beyond prediction, AI strengthens global health crisis management by facilitating real-time surveillance, optimizing hospital operations, improving resource distribution, and enabling telemedicine integration. It also supports behavioral analytics and public health communication, ensuring that interventions are targeted,

equitable, and effective. Ethical considerations, including privacy, transparency, fairness, and accountability, are essential to ensure responsible AI deployment and maintain public trust.

Looking forward, innovations such as federated learning, explainable AI, genomic surveillance, and international collaboration promise to further advance epidemic preparedness. By transforming health systems from reactive responders to proactive, data-driven infrastructures, AI empowers societies to anticipate, manage, and mitigate the effects of infectious disease outbreaks.

In conclusion, AI is not merely a supplementary tool but a central pillar of modern epidemic management. Its integration into predictive modeling, crisis response, and public health decision-making has the potential to save lives, enhance resilience, and shape a more responsive and equitable global health landscape.

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