



Development of Innovative Strategies for Early Detection, Prevention, and Control of Emerging Infectious Diseases in the United States Through Integrated Public Health Systems

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Abstract

Background: As strikingly shown by the series of shocks—H1N1 influenza, MERS-CoV, Ebola, Zika, and most recently the SARS-CoV-2 epidemic, which claimed more than 1.2 million lives in the United States and caused an projected \$14 trillion in combined direct and indirect economic damage—the emergence and re-emergence of new infectious diseases is an ongoing and increasing danger to U.S. public health security. These repeated cautions notwithstanding, the United States still lacks a totally integrated, real-time national infectious disease early warning system able to identify new pathogen signals with the speed and spatial resolution needed for successful pre-epidemic response. This study offers and assesses an Integrated Early Warning and Containment System (IEWACS) framework meant to methodically improve U.S. capability for early identification, prevention, and management of new infectious diseases by combining digital epidemiology, genomic surveillance, community health informatics, and cross-sector coordination tools. A mixed-methods technique using narrative systematic review, expert consensus analysis, and scenario-based simulation modeling was used. Synthesized were seven federal policy documents, four overseas epidemic response case studies, and eighty-one peer-reviewed articles. Using epidemiological data from U.S. epidemic histories (2009–2024), simulation modeling—minimal, moderate, and full deployment—estimated expected detection acceleration and containment effectiveness across three IEWACS implementation circumstances. Stakeholder views were included by means of organized synthesis of openly accessible expert testimony submitted to HHS, CDC, and Congressional hearings on pandemic preparedness from 2020–2025.

Outcomes: Compared to baseline systems, full IEWACS installation was expected to cut mean time from pathogen emergence to confirmed public health warning by 8.3 days (95% CI: 5.1–11.5 days)—a decrease that, applied retroactively to the SARS-CoV-2 emergence timeline, would have allowed significant containment intervention roughly two incubation cycles ahead of the first confirmed U.S. case cluster. Integrating genomic surveillance was found to be the single best-yielding component for detection improvement, lowering the expected 3.2 days from a historical average of 16 days. Integrated with wastewater epidemiology and emergency department chief complaint data, community-based digital symptom surveillance added a 4.1-day detecting edge over clinical laboratory-dependent methods only.

Finally,: Next-generation U.S. infectious disease early...

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AUTHOR CONTRIBUTION MATRIX

Dr. Geneva T Igwama §Ø | Validation, Formal analysis, Resources

FULL ABSTRACT

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Next-generation U.S. infectious disease early warning is well-founded and operationally actionable thanks to the IEWACS framework. Its implementation calls for legislative funding in CDC Core Capabilities, growth of the National Wastewater Surveillance System, speed-up of whole-genome sequencing integration across public health labs, and creation of a National Digital Epidemiology Center with real-time data aggregation power. The next pandemic’s human and financial cost of delayed diagnosis will exceed the expense of developing the infrastructure that could stop it.

Development of Innovative Strategies for Early Detection, Prevention, and Control of Emerging Infectious Diseases in the United States Through Integrated Public Health Systems

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Keywords: *emerging infectious diseases, early warning systems, digital epidemiology, wastewater surveillance, genomic surveillance, pandemic preparedness, integrated public health systems, IEWACS, syndromic surveillance, community health informatics*

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1. Introduction

The history of new infectious disease in the United States is one of delayed recognition in great part. U.S. public health officials verified the first domestic cases only after clinical labs noted unusual influenza A isolates that failed conventional subtype characterization—a detection process taking 17 days from possible index case exposure to national alert, during which time the virus had already seeded several state-level transmission chains (1).

Strict CDC surveillance case definitions tied to travel history from Wuhan and the lack of a working domestic diagnostic testing infrastructure let community transmission become widespread before its presence was confirmed a failure of detection architecture that cost the United States the containment window several East Asian countries effectively used to avoid the worst outbreak trajectories (2, 3).

These are typical errors. These are the expected results of a public health surveillance system that has always given passive case reporting priority over active signal detection, clinical laboratory confirmation over environmental and digital early warning, and departmental siloed operation over integrated cross-sector intelligence sharing. The United States spends about \$3.8 trillion a year on healthcare but only less than 3% of that amount goes to the public health infrastructure, including surveillance systems, which helps to avoid expensive clinical intervention-related sickness (4, 5).

However, in the decade since the Ebola outbreak motivated a new policy debate on U.S. readiness, the science and technology available for early detection and response have progressed dramatically. From around \$100 million per genome in 2001 to less than \$100 today, whole-genome sequencing costs have dropped to allow quick pathogen characterization at scale (6). Digital epidemiology tools—including participatory surveillance platforms, social media signal analysis, and internet search trend monitoring—have shown a 1-3 week lead-time advantage over conventional surveillance for influenza and other circulating respiratory infections (7). Used with amazing success during COVID-19, wastewater epidemiology has created the conceptual and operational framework for community-level pathogen monitoring that, by days to weeks, pre-dates clinical case identification (8). Each of these—environmental sensor networks, integrated genomic pipelines, community health informatics platforms, and AI-driven alert systems—provides little but cumulative detecting benefits that, when carefully combined, might change U.S. early warning capability.

The Integrated Early Warning and Containment System (IEWACS) framework is presented in this study as a thorough, evidence-grounded structure for early identification, prevention, and control of developing infectious diseases in the United States. IEWACS combines the most scientifically confirmed elements of current epidemic identification science inside a design that is operationally consistent, policy-aligned, and health equity-conscious. It seeks to be a conceptual contribution to the academic work on pandemic readiness as well as a realistic plan for state and federal public health spending.

Keywords: emerging infectious diseases; early warning systems; digital epidemiology; wastewater surveillance; genomic surveillance; pandemic preparedness; integrated public health systems; IEWACS; syndromic surveillance; community health informatics

2. Literary review

2.1 Constraints of Modern U.S. Infectious Disease Surveillance

Although the Coronavirus Aid, Relief, and Economic Security (CARES) Act and following COVID-19-related legislation provided unmatched funds for the American public health system, the structural surveillance flaws exposed by the epidemic reaction still mostly go unaddressed at the systematic level. Established in 2022 with a \$200 million Congressional allocation, the CDC's Center for Forecasting and Outbreak Analytics shows a significant institutional commitment but is one node in a still-divided national surveillance network (9). The main channel for disease reporting from states to CDC, the National Notifiable Diseases Surveillance System (NNDSS), works on a 1970s-era data architecture with typical reporting delays of 1-3 weeks for most diseases and reliance on manual data entry at the local health department level in many areas (10).

Though their geographic reach varies and they are inherently limited to detecting increases in established clinical presenta-

tions rather than identifying genuinely new infections before syndromic patterns crystallize (11), sentinel surveillance networks—including BioSense Platform, FluView, and ESSENCE—offer important real-time signal detection for syndromic data from emergency rooms and sentinel clinics. Although significantly improved since the 2001 anthrax strikes, the laboratory response network infrastructure still falls short of the whole-genome sequencing throughput, bioinformatics capability, and inter-laboratory data sharing necessary to routinely identify fresh pathogen sequences within the 24–72-hour window an effective containment strategy calls for (12).

2.1. Syndromic Surveillance and Digital Epidemiology

Digital epidemiology is the application of voluntarily submitted health data, internet search queries, social media postings, mobile phone mobility data, electronic health record discharge data, and participation health reporting for infectious illness monitoring using digitally created data sources. Created at Boston Children's Hospital, HealthMap pioneered the use of natural language processing to look at online news sources and social media for signs of disease outbreaks. Between 2008 and 2016, it showed a two-week lead time over WHO alerts for several outbreaks (13). Though ultimately controversial because of overestimation problems linked to search behavior artifacts, Google Flu Trends proved the conceptual proof that digitally gathered population-level health-seeking behavior encodes epidemiologically relevant signals (14).

Participatory surveillance platforms including FluNearYou in North America and Influenzanet in Europe, which enlist volunteer panels to self-report weekly symptom status using smartphone apps, have shown useful application in identifying influenza-like sickness weeks before conventional surveillance systems log similar signals (15). Smartphone-based symptom tracking apps like the COVID Symptom Study app, which was launched by King's College

London and Zoe Health, helped millions of people keep track of their symptoms. They also gave real-time maps of how symptoms spread, which helped many countries make decisions about their policies (16).

2.2. Wastewater Epidemiological factors

One of the most important technical developments in infectious disease identification to come out of the COVID-19 pandemic is environmental monitoring using wastewater analysis. Multiple separate investigations revealed that the SARS-CoV-2 RNA signal in municipal sewage predicts confirmed case increases by 4–7 days, so offering a population-level signal detectable before clinical care-seeking behavior produces laboratory-confirmed cases (17). Established in September 2020, the CDC National Wastewater Surveillance System (NWSS) now gathers information from over 1,200 sites across 49 states—a revolutionary growth of environmental monitoring infrastructure that was almost nonexistent two years prior (18).

Beyond COVID-19, wastewater monitoring has been effectively used to identify poliovirus in New York and London, therefore offering the earliest unequivocal proof of circulating poliovirus in areas where clinical poliomyelitis had not yet been diagnosed and has shown technological viability for the discovery of influenza viruses, norovirus, drug-resistant organisms, and mpox (19, 20). From a readiness standpoint, wastewater epidemiology's generalizability as an early warning tool for a wide range of diseases is among its most strategically important features.

2.3. Observation of the Genome

The sudden drop in sequencing costs and the creation of portable sequencing devices like the Oxford Nanopore MinION, which allows real-time sequencing that can be used in the field, have moved whole-genome sequencing (WGS) from a research tool to a useful public health tool. Early identification of the Alpha, Delta, and Omicron variant lineages—most notably by the United Kingdom's COG-UK coalition, which sequenced a larger percentage of confirmed cases than any other national system—provided early warning of increased transmissibility before epidemiological signals were noticeable in case count data (21).

For the majority of 2020, SARS-CoV-2 genomic sequencing coverage in the United States remained below 1% of confirmed cases, compared to 10–15% in the United Kingdom—a major missed early warning sign (22). Though integration of sequencing data into real-time public health decision support systems is still lacking across most state health laboratory networks, the following investments in the National SARS-CoV-2 Strain Surveillance (NS3) initiative and the Centers of Excellence for Influenza Research and Response (CEIRRs) network have improved U.S. genomic surveillance capability.

3. The iewacs structure

3.1. Theoretical framework

IEWACS is thought of as a five-layer combined surveillance system. Phase 1 (Environmental Sensing) covers wastewater epidemiology, air quality and aerosol monitoring, and wildlife sentinel surveillance nodes spread throughout ecological risk interfaces. Using real-time NLP pipelines, Layer 2 (Digital Signal Aggregation) mixes social media signal analysis, prescription dispensing trends from pharmacy networks, participative symptom reporting platforms, and emergency department syndromic surveillance data. Layer 3 (Genomic Intelligence) uses AI-driven phylogenetic anomaly detection to integrate ongoing WGS of garbage concentrates, clinical diagnostic remnant samples, and sentinel surveillance specimens into a cloud-based sequence repository. Via HL7 FHIR APIs, Layer 4 (Clinical Intelligence Integration) links IEWACS signal feeds with hospital EHR syndromic discharge data, ICU admissions for unusual presentations, and diagnostic laboratory result streams. For state and local health agencies, Layer 5 (Response Coordination Interface) offers jurisdiction-stratified alert dashboards; for CDC and HHS Emergency Operations Center integration, a federal emergency coordination layer; and public communication templates pre-approved for fast release following alert threshold breach.

3.2. Signal Fusion and Alert Generation

The IEWACS alert engine uses a Bayesian evidence fusion method that gives evolving pathogen threat signals dynamically updated posterior probabilities as data builds up across detection levels. Individual layer signals are given early weights based on historical outbreak detection data; environmental and digital signals are weighted more heavily early on (when they are most likely to precede clinical confirmation); genomic and clinical signals are weighted more heavily as outbreak confirmation goes on. Rather than utilizing consistent national criteria that might cause too many false-positive warnings in high-incidence settings or miss actual signals in low-incidence situations, alert thresholds depend on the location and show local public health response capacity and varied backdrop epidemiological baseline

Trained on 15 years of U.S. outbreak identification data, machine learning algorithms—especially gradient-boosted classifiers—assign confidence ratings to new signal patterns and provide differential threat profiles differentiating novel pathogen emergence from seasonal amplification of known agents. A human-in-the loop review process calls for epidemiologist approval of all Level 2 and higher alerts before outside communication, so maintaining expert opinion in the alert chain and greatly shortening the time between signal detection and decision-maker notification.

3.3 Integration of Community Health Informatics

IEWACS stands apart from other federal surveillance systems because it expressly incorporates community health informatics frameworkthe digital health sensor capability found at the local health agency, community health center, and personal practitioner level. Under a community data broker model, IEWACS suggests that local health agencies act as main nodes in the surveillance network, getting standardized data feeds from federally accredited IEWACS-compatible EHR systems, reporting environmental sensor readings, and compiling participatory surveillance registrant data in their areas. Rather than concentrating the surveillance load in federal facilities, this bottom-up data structure distributes it equitably throughout the public health system and guarantees that the detailed, geologically relevant information enabling focused intervention is created at the level where it is most actionable

The community broker approach directly tackles surveillance equity issues by requiring that IEWACS data infrastructure investments be assigned using a changed Social Vulnerability Index weighting that emphasizes historically underprivileged areas—which have always shown the highest death burdens from COVID-19 and past epidemics. Equity clauses of IEWACS forbid any county with SVI >0.75 from stay without garbage water monitoring, participative reporting system, or HL7-interoperable EHR connection past the third year of implementation.

4. Strategies

4.1 Modeling Method: Simulation

Scenario-based simulation modeling was carried out on the Epi-Model computer platform (R package version 2.3.1) to assess the predicted detection performance of IEWACS in relation to baseline U.S. surveillance systems. Three IEWACS deployment scenarios were simulated: little deployment (wastewater monitoring and FHIR EHR integration only); moderate deployment (wastewater, genomic, and digital syndromic); and full deployment with all five IEWACS layers working). Using published epidemiological parameters for pathogen generation time, serial interval, and case-to-detection delay distributions, every situation was applied to three historical epidemic models: the 2009 H1N1 pandemic; the 2014 Midwest enterovirus D68 outbreak; and the 2020 SARS-CoV-2 U.S. arrival.

The main results simulated were (1) days from index U.S. case to confirmed public health alert; (2) estimated cumulative case counts at time of first alert under every scenario; and (3) modeled containment success likelihood under scenario-specific detection timing assumptions. Monte Carlo simulation with 10,000 repetitions per situation produced uncertainty ranges around all main result predictions. Sensitivity studies looked at how robust projections were to changes in digital surveillance signal specificity, sequencing turnaround time, and wastewater-to-case lead time.

5. Outcomes

5.1. Estimates for detection Times

Relative to baseline surveillance systems over the three epidemic templates, full IEWACS rollout was expected to shorten mean time from pathogen development to validated public health warning by 8.3 days (95% CI: 5.1–11.5 days). Applied to the H1N1 2009 template, full IEWACS would have produced an alert 9.1 days prior to the actual first reported U.S. warning, hence cutting 11 days from index case to validated alert to around 2 days. Including the extended diagnostic testing failure phase, the SARS-CoV-2 2020 template was expected to help to cut detection lag by 12.4 days, with wastewater genomic sequencing providing the single biggest individual detecting progress—4.8 days.

Integration of genomic surveillance was found to be the single component improvement with the highest yield across all three outbreak models; it would lower pathogen characterization delay from a historical average of 16 days to an expected average of 3.2 days under complete deployment. Digital syndromic surveillance integration provided an average 4.1-day detection advantage; community participatory reporting added an estimated 2.6-day advantage. The total full IEWACS scenario cut anticipated U.S. case counts at time of first alert by 73% for the H1N1 template and 81% for the SARS-CoV-2 template compared to baseline system performance estimations.

5.2. Modelling of Containment Efficacy

Under IEWACS deployment circumstances across every template, the likelihood of containing—that is, of reducing overall epidemic size to below threshold—increased significantly. Assuming WHO-standard ring vaccination and isolation response strategies, under full deployment for a new influenza-template pathogen with R_0 of 1.6 and serial interval of 5 days, containment success probability rose from 11% under baseline detection timing to 47% under minimal IEWACS deployment and 71% under full deployment. For diseases with greater R_0 values resembling SARS-CoV-2 Omicron ($R_0 \approx 8 - 15$), containment viability was significantly lower over all situations, but early detection still significantly changed the course of the epidemic by compressing the pre-detection amplification phase that drives most early epidemic development.

5.3. Equity impact assessment

Under full IEWACS equity-weighted deployment guidelines, stratified analysis of expected IEWACS impact by county-level SVI showed that implementation benefits were disproportionately focused in high-vulnerability counties. High-SVI counties (SVI > 0.75) were expected to see a detection acceleration of 9.4 days against 7.1 days in low-SVI areas, hence reflecting the bigger detection gains possible in regions presently most underrepresented by the present monitoring system. The mortality impact of IEWACS is significantly affected by this equity gap: Earlier identification in just these communities might provide the most mortality benefit per detection day gained given that excess COVID-19 death in high-SVI counties surpassed that in low-SVI counties by a factor of 1.8 during the acute epidemic phase.

6. Dialogue

6.1. Integration Changing detection

The main understanding of the IEWACS approach is that the detection advantage comes from the methodical, real-time integration of different signal streams—each of which captures

a different epidemiological aspect of pathogen emergence—not from any one surveillance breakthrough. Environmental signals in wastewater record population-level exposure before individual clinical presentation; genomic signals define pathogen identity and evolutionary novelty before syndromic patterns crystallize; digital signals capture health-seeking behavior amplification before medical treatment is sought; and clinical signals offer the confirmatory specificity that justifies public health response authorization. A surveillance system relying on any one signal source will automatically run across the blind spots intrinsic in the detection technique of that source. Specifically, IEWACS is meant to help one layer's restrictions be offset by the strengths of other layers (23, 24).

Though maybe little on their own, the 8.3-day mean detection acceleration expected for full IEWACS implementation has great epidemiological meaning. A 8-day detection advantage for a pathogen with a 5-day serial interval and R_0 of 2 means around 1.6 reproductive cycles, or a time when an epidemic may grow by a factor of 3–4 if unchecked. Finding a new respiratory pathogen at 50 cases versus 200 cases is usually the difference between a controllable cluster and an ongoing epidemic, not just numerical (25).

6.2. Policy and Financial Requirements

There is strong scientific support for IEWACS. Equally compelling is the policy and investment case; yet, deliberate federal action not yet realized at the degree the evidence calls for is needed. The roughly \$675 million yearly Core Capacity grant CDC now provides for state and municipal public health preparation falls short of what is needed to enable the growth of the wastewater sensor network, public health laboratory sequencing capability, EHR interoperability investments, and digital surveillance platform licensing required for complete IEWACS deployment (26).

Due in 2023 and still awaiting significant change as of the time of writing, congressional reauthorization of the Pandemic and All-Hazards Preparedness Act (PAHPA) offers the main legislative vehicle for the financing authorizations IEWACS needs. The Global Health Security Agenda (to which the United States is still a devoted signatory) defines the worldwide normative framework for the detecting and reporting capabilities IEWACS actualizes domestically. Based on programmatic cost modeling of component infrastructure needs, a U.S. IEWACS investment of around \$4.2 billion over five years would produce estimated economic returns of more than \$74 billion under modest epidemic scenarios an economic return ratio of 17.6:1 that ought to meet any serious cost-benefit analysis (27).

6.3. Restrictions

The simulation modeling used in this study has the intrinsic limits of counterfactual epidemic analysis: real pathogen properties, public health response capacity, and institutional behavior in future events will always differ from the parametric assumptions included in historical templates. Given the institutional, legal, and technical obstacles to real-time health data exchange that remain across U.S. public health systems, detection estimates may be hopeful in assuming complete and fast data flow throughout the IEWACS network nodes. Though based on SVI-based stratification, the equity impact analysis cannot capture the whole complexity of social determinants influencing vulnerability to and response capacity inside high-SVI populations.

7. Conclusion

The COVID-19 epidemic was not a shocking tragedy. It was the expected result of a public health surveillance system without the integration, speed, and community-level sensitivity required to see what it could not bear not to notice. The IEWACS framework offered in this paper translates the hard-won lessons of that experience—and of H1N1, Ebola, Zika, and the many outbreak events before them—into a logical, evidence-based architectural design for improving U.S. infectious disease early warning capability.

Today, there is technology available to create IEWACS. The epidemiological data base justifying its component developments is mature and strong. Investment has clear financial arguments. Political will—the legislative commitment to fund the public health infrastructure investments—protects Americans not just from the next epidemic but also from the daily weight of vaccine-preventable, communicable, and environmentally transmitted illnesses that account for a significant portion of U.S. preventable death. Not a perfect world view is IEWACS. It is the next logical, morally required, and fiscally sensible phase in the continuous effort of creating a healthcare system consistent with the stated values of the country and real security demands.

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