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Novel Disease Event Monitoring Tools

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Disclosures

- None of the presenters at this session have received financial support or in-kind support from a commercial sponsor.
- The presenters have potential conflicts of interest to declare.

Presenter Disclosures

Nature of relationship(s)	Name of for-profit or not-for-profit organization(s)	Description of relationship(s)
Funded grants or clinical trials	International Society for Infectious Diseases	Funding for research program support
Funded grants or clinical trials	PandemicTech	Funding for research program support



This presentation was peer-reviewed to ensure that principles of scientific integrity, objectivity, and balance have been respected.

Polling Question 1

Have you used digital disease event monitoring tools over the past 18 months?

Learning Objectives

- 1. Identify the role of informal disease event monitoring and reporting
- 2. Define specific applications of informal disease event monitoring systems for outbreak detection
- 3. Discuss data limitations in real-time epidemic forecasting



- 1. Identify the role of informal disease event monitoring and reporting
- 2. Define specific applications of informal disease event monitoring systems for outbreak detection
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PHOTOGRAPH: GETTY IMAGES

Published Date: 2019-12-30 23:59:00

Subject: PRO/AH/EDR> Undiagnosed pneumonia - China (HU): RFI

Archive Number: 20191230.6864153

UNDIAGNOSED PNEUMONIA - CHINA (HUBEI): REQUEST FOR INFORMATION

A ProMED-mail post

http://www.promedmail.org

ProMED-mail is a program of the International Society for Infectious Diseases http://www.isid.org

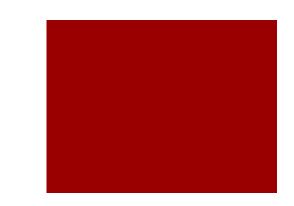
[1]

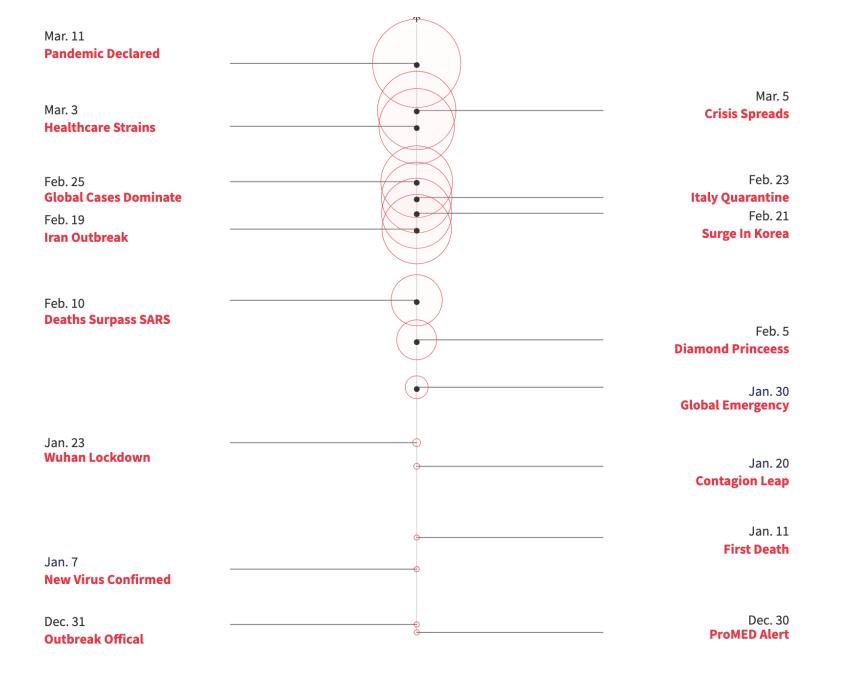
Date: 30 Dec 2019

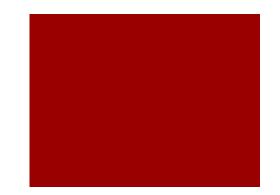
Source: Finance Sina [machine translation]

https://finance.sina.cn/2019-12-31/detail-iihnzahk1074832.d.html?from=wap

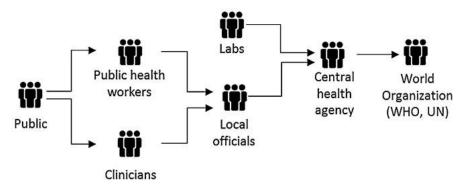
Wuhan unexplained pneumonia has been isolated test results will be announced [as soon as available]







A. Traditional indicator-based surveillance systems



B. Event-based Internet biosurveillance systems

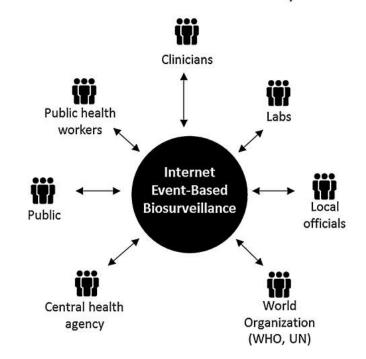






Table 2
List of Event-Based Internet Biosurveillence Systems Identified.

System Name	Category	Founding Country	Year Started	Currently online	Prototy
Argus	Moderated	USA	2004	No	No
BioCaster	Automatic	Japan	2006	No	No
GeniDB	Automatic	Japan	2012	No	No
GODSN	Automatic	USA	2006	No	No
MiTAP	Automatic	USA	2001	No	Yes
EWRS	Moderated	EU	1998	Yes	No
Proteus-BIO	Automatic	USA	2000	No	Yes
MedCOLLECTOR	Automatic	Portugal	2009	Yes	No
nEmesis	Moderated	USA	2013	No	Yes
EpiSPIDER	Automatic	USA	2006	No	No
Digital Disease Detection Dashboard	Moderated	USA	2014	No	Yes
MedISys and PULS	Automatic	EU	2004	Yes	No
GPHIN	Moderated	Canada	1997	Yes	No
HealthMap	Automatic	USA	2006	Yes	No
InSTEDD	Moderated	USA	2006	Yes	No
ProMED-mail	Moderated	USA	1994	Yes	No
Foodborne Chicago	Moderated	USA	2013	Yes	No
GET WELL	Automatic	Sweden	2010	Yes	No
Google Trends	Automatic	USA	2009	Yes	No
De Grote GriepMeting ^a	Moderated	Netherlands Belgium	2003	Yes	No
Gripenet ^a	Moderated	Portugal	2005	Yes	No
Influweba	Moderated	Italy	2008	Yes	No
FluSurvey ^a	Moderated	UK	2009	Yes	No
InfluenzaNet ^a	Moderated	EU	2009	Yes	No
Influmeter ^a	Moderated	Denmark	2012	Yes	No
GripeNet ^a	Moderated	Spain	2012	Yes	No
Grippenet ^a	Moderated	France	2012	Yes	No
Halsorapport ^a	Moderated	Sweden	2011	Yes	No
FluSurvey.le ^a	Moderated	Ireland	2013	Yes	No
FluTracking	Moderated	Australia	2006	Yes	No
Reporta	Moderated	Mexico	2009	Yes	No
Flu Near You	Moderated	USA	2011	Yes	No
	Moderated	Puerto Rico		Yes	No No
Salud Boricua	Moderated	Brazil	2012 2011	Yes	No No
Dengue na Web					
Crowdbreaks	Moderated	USA	2012 2011	Yes	Yes
Doctor Me App	Moderated	Thailand		Yes	No
GermTrax	Automatic	USA	2012	Yes	No
SickCity	Automatic	USA	2009	No	No
FluDetector	Automatic	UK	2009	No	No
Infovigil	Moderated	Canada	2009	No	Yes
M-Eco	Automatic	EU	2010	No	Yes
NowTrending (previously MappyHealth)	Automatic	USA	2012	Yes	No
FluTrackers.com	Moderated	USA	2006	Yes	No
GoViral	Moderated	USA	2013	Yes	Yes
BlueDot (previously BioDiaspora)	Moderated	Canada	2008	Yes	No
Mo-Buzz	Moderated	Singapore	2013	Yes	No
Sickweather	Automatic	USA	2011	Yes	No
FluCaster	Automatic	USA	2013	Yes	No
GOARN	Moderated	USA	2000	Yes	No
HealthTweets.org	Automatic	USA	2013	Yes	No

a Collectively can be called InfluenzaNet.

Pros	Cons
 Fast detection and reporting 	 Information captured might not be accurate or significant
 Not constrained to certain events 	 Sources may present biased information
 Multiple sources of information (e.g., clinicians, labs, media reports, internet blogs) 	Uses broad case definitions
 Leverages publicly available information, increasing transparency 	 No standard data format, takes additional time to synthesize

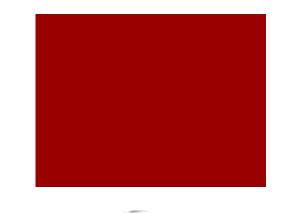
- ProMED (Program for Monitoring Emerging Diseases)
- Global Public Health Intelligence Network (GPHIN)







- MediSys
- EpiSPIDER
- BioCaster
- Wildlife Disease Information Node
- EIN
- GOARN
- Epi-X
- GHSA
- EpiCore
- Global.health







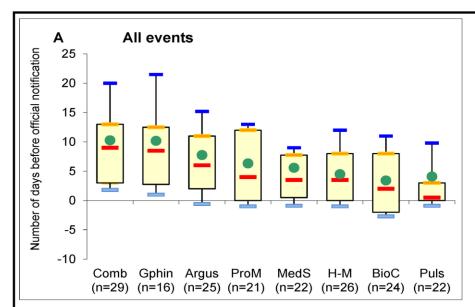


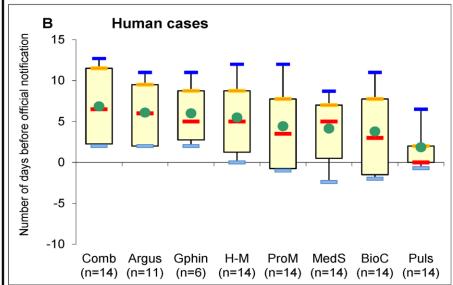




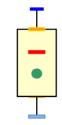








LEGEND



90th Percentile 3rd Quartile

Median

Mean

1st Quartile

10th Percentile

Comb = Combined virtual system

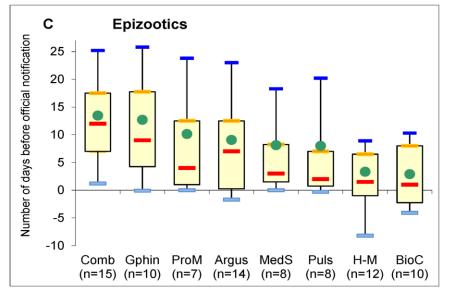
ProM = ProMED

MedS = MedISys

H-M = HealthMap

BioC = BioCaster

(n=..) = n event detected by the system





CDC Stands Up New Disease Forecasting Center

Press Release

For Immediate Release: Wednesday, August 18, 2021

Contact: Media Relations

(404) 639-3286

Today, the Centers for Disease Control and Prevention (CDC) is announcing a new center designed to advance the use of forecasting and outbreak analytics in public health decision making. Once established, the Center for Forecasting and Outbreak Analytics will bring together next-generation public health data, expert disease modelers, public health emergency responders, and high-quality communications, to meet the needs of decision makers. The new center will

US CDC. 2021.



Learning Objectives

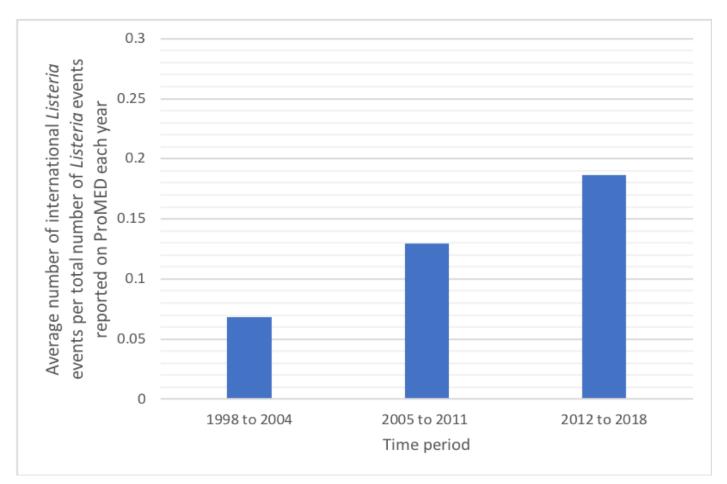
- 1. Identify the role of informal disease event monitoring and reporting
- 2. Define specific research applications of informal disease event monitoring systems for outbreak detection
- 3. Discuss data limitations in real-time epidemic forecasting

An Aside...

- Travel Medicine
- ProMED-AMR

Results

Figure 3: Average number of international *Listeria* events per total number of *Listeria* events reported on ProMED each year over three time periods

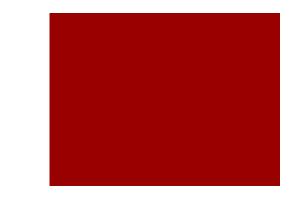




Results

Table I: Characteristics of *Listeria* events reported on ProMED

Events	N=Event Counts (%)
Total Events	123 (100%)
 Outbreaks (two or more human cases) 	81 (65%)
Sporadic cases	13 (11%)
 Precautionary recalls* 	29 (24%)
Hospital-acquired events	10/123 (8%)
Events involving multiple countries	21/123 (17%)
Case-fatality rate, overall	487/2,383 (20%)





Mapping the Risk of International Infectious Disease Spread (MRIIDS)

A User-friendly Tool for Outbreak Response and Epidemic Preparedness

nttp://mriids.org

https://github.com/ISIDOrg/MRIIDS/wiki/Mapping-the-Risk-of-International-Infectious-Disease-Spread

Disease outbreak information is assessed with automated intelligence capabilities that incorporates population density information to estimate the number of cases for specific outbreak events.

Additionally, the algorithm provides a risk projection to describe where infectious diseases cases are most likely to arrive from and depart to for specific countries. The information generated by the platform is accessible free of charge and incorporated into a platform with extensive end-user testing.

Developed based on data from the 2014-16 West Africa Ebola outbreak, the MRIIDS prototype was designed to be rapidly scalable by extending it to pathogens of significance to humans and animals on a global scale. The tool aims to inform key health decision-makers at national and regional levels of the risks of an outbreak spreading in real-time, and aids government and non-governmental decision-makers as they prepare for the possible arrival of an infectious disease threat to their region.

MRIIDS

Mapping the Risk of International Infectious Disease Spread

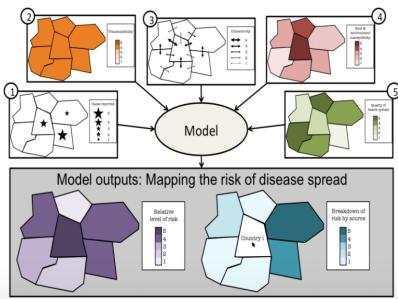
- User-friendly tool to estimate and visualize the risks of outbreak events spreading
- Aimed at helping decision makers with health resource allocation and infectious disease threat preparedness
- Uses multiple data streams including ProMED, international flight data, and health center data

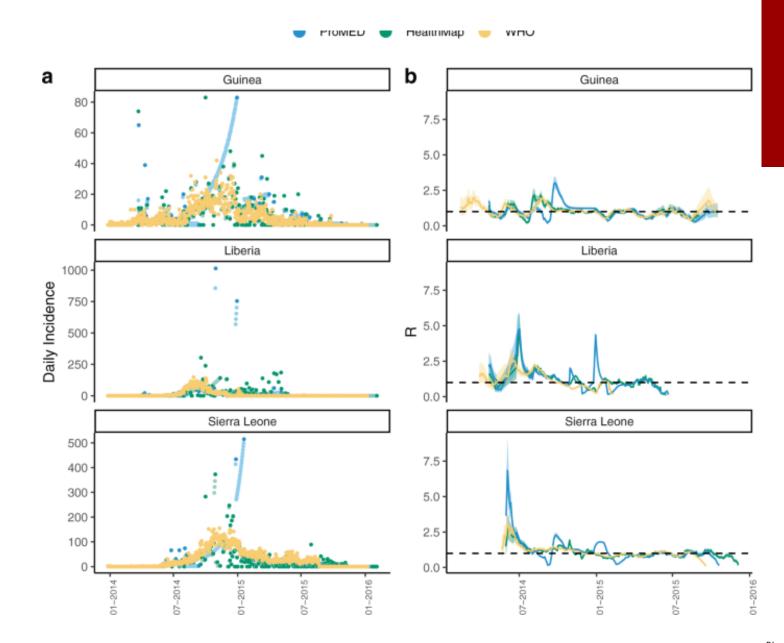
Potential end-users include government, public health experts, health care workers, NGOs, and

others

Project partners:

- The International Society for Infectious Diseases (ISID) and its Program for Monitoring Emerging Diseases (ProMED)
- Imperial College London
- heathsites.io
- HealthMap





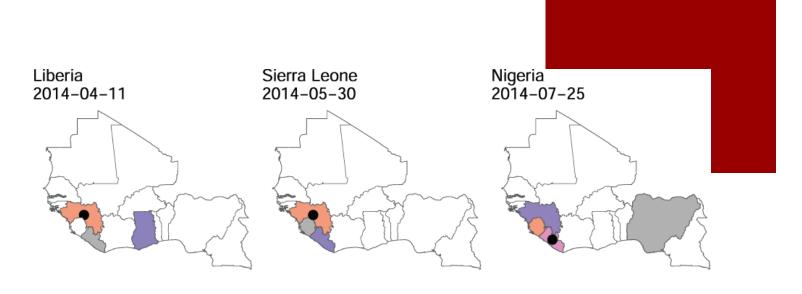
MRIIDS

Risk of Importation

Fig. 3: Relative risk of importation of the epidemic.

From: <u>Using digital surveillance tools for near real-time</u> mapping of the risk of infectious disease spread

For each country with non-zero incidence, the figure shows the relative importation risk (see Methods). Since we forecasted every 7th day, the risk of importation was estimated using forecasts closest to and before the date of the first case in that country reported in the data used. The date on which risk was estimated for each country is shown in the figure. Blue indicates low relative risk while deeper shades of red represent higher relative risk of acting as a source of importation. White is used to denote no risk. The estimates presented here use ProMED data with a 2 week calibration window. The country for which risk is estimated is shown in grey. The black circle represents the actual source of importation as retrospectively identified through epidemiological and genomic investigations. For each country, the figure shows only the risk of importation from other countries and does not show the risk of transmission within the country.



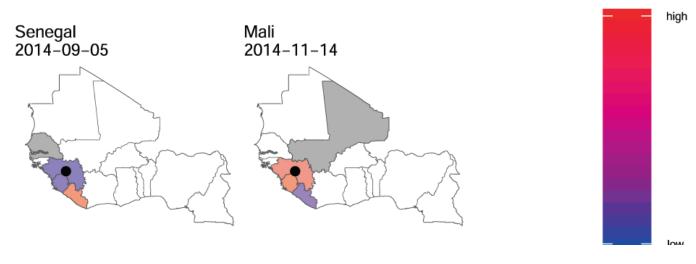


Table. Disease drivers identified in the literature and examples of data availability*			
Global data examples†	Regional data examples†		
Vaccine rumor surveillance, product distribution data from	US influenza vaccination rates,		
manufacturers, self-reported immunization status	measles vaccination rates from the		
	Mozambique Health Information		
	System		
· · · · · · · · · · · · · · · · · · ·	Climate data, social media reports of		
•	climate and air pollution effects on		
and the Global Environment, vulnerability to climate change	Twitter and Sina Weibo		
Night time lights, Gridded population of the world, mobile phone	National census data products,		
operator data	Twitter, world population		
International Monetary Fund, World Bank	National departments of economics		
Global agricultural lands, Center for International Earth Science	National departments of agriculture,		
Information Network, Global Forest Change 2000–2012, Global	croplands in western Africa, Africa		
Forest Watch, global livestock distribution densities	mining digital news reports, IMAZON		
	Deforestation Alert System		
Digital news, United Nations Global Pulse	NA		
Species distribution grids, digital news reports	State-level hunting data		
Natural disaster hotspots	News of impending natural disasters		
	(i.e., predicted hurricane landfall)		
Center for International Earth Science Information Network,	National census data		
· · · · · · · · · · · · · · · · · · ·			
	Syria Tracker		
Historical records, Transparency International, Cline Center for	NA		
Democracy			
Flight and shipping data	Regional distribution data of food		
	products		
*The table is purposely not exhaustive but provides a survey of types of available digital data that are associated with different drivers. NA, not applicable. †See online Technical Appendix Table 1 (http://wwwnc.cdc.gov/EID/article/21/8/14-1156-Techapp1.pdf) for available references.			
	Global data examples† Vaccine rumor surveillance, product distribution data from manufacturers, self-reported immunization status Numerous satellite products, National Oceanic and Atmospheric; Administration, Climatic Research Unit, Center for Sustainability and the Global Environment, vulnerability to climate change Night time lights, Gridded population of the world, mobile phone operator data International Monetary Fund, World Bank Global agricultural lands, Center for International Earth Science Information Network, Global Forest Change 2000–2012, Global Forest Watch, global livestock distribution densities Digital news, United Nations Global Pulse Species distribution grids, digital news reports Natural disaster hotspots Center for International Earth Science Information Network, Global Observatory Famine early warning system, digital news and social media Historical records, Transparency International, Cline Center for Democracy Flight and shipping data		

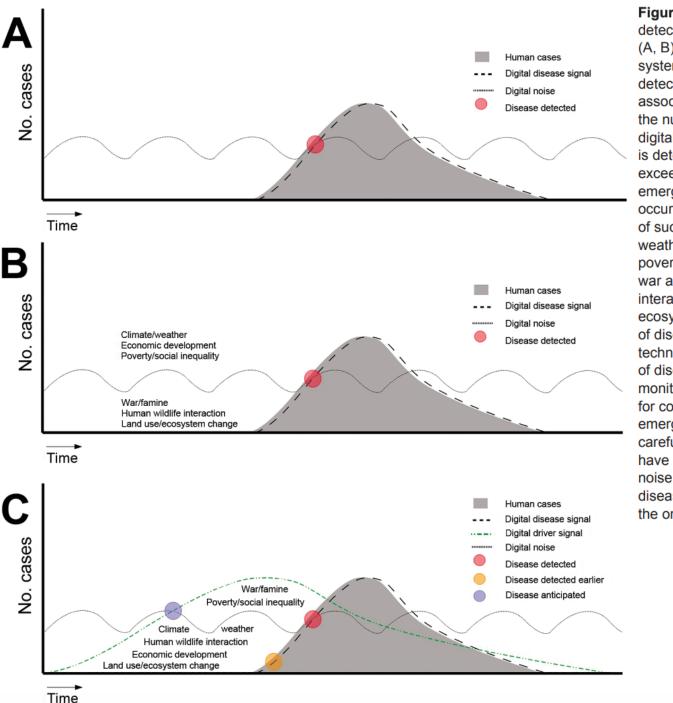


Figure 1. Surveillance and detection of disease by traditional (A, B) and digital (C) detection systems. A) Traditional disease detection, in which a close association exists between the number of cases and the digital disease signal. Disease is detected when the signal exceeds the noise. B) Disease emergence or outbreaks often occur following a driver. Examples of such drivers include climate and weather, economic development, poverty and social inequality, war and famine, human-wildlife interactions, land use and ecosystem changes. C) Detection of disease by using digital techniques. In this system, drivers of disease (not disease) are monitored, essentially to monitor for conditions suitable for disease emergence. Hypothetically, the careful surveillance of drivers that have been separated from digital noise could shorten the time to disease detection (as indicated by the orange dot).

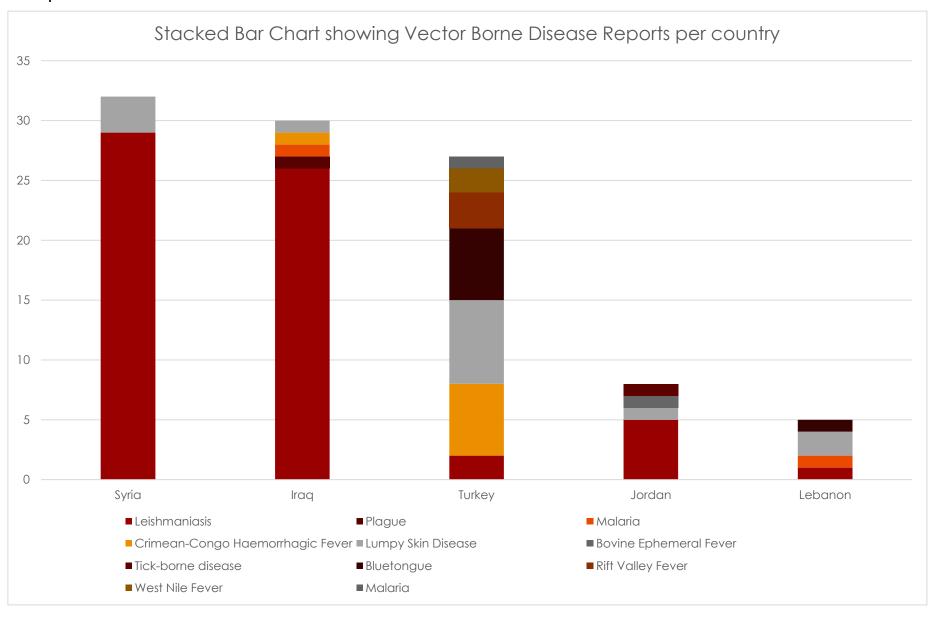
Table 2. Number of VBD events per pathogen and country for pre-conflict (2003-2010) and conflict (2011-2018) periods

- a. VBD reported in an animal that only affects animalsb. indicates a VBD reported in an animal that can also affect humans

Pre-conflict period		
Pathogen by country	Totals	
Iraq	8	
Leishmaniasis	8	
Jordan	1	
Epizootic hemorrhagic disease a	1	
Syria	2	
Leishmaniasis	2	
Turkey	12	
Crimean-Congo hemorrhagic	7	
fever	2	
West Nile virus	1	
Tularemia	1	
Bluetongue disease a	1	
Epizootic hemorrhagic fever ^a		
Total	23	

Conflict period	Conflict period		
Pathogen by country	Totals		
Iraq	30		
Leishmaniasis	26		
Plague	1		
Malaria	1		
Crimean-Congo hemorrhagic fever	1		
Lumpy skin disease a	1		
Jordan	8		
Leishmaniasis	5		
Bovine ephemeral fever ^a	1		
Tick paralysis b	1		
Lumpy skin disease a	1		
Syria	32		
Leishmaniasis	29		
Lumpy skin disease a	3		
Turkey	27		
Lumpy skin disease a	7		
Crimean-Congo hemorrhagic fever	6		
Bluetongue disease a	6		
Rift Valley fever b	3		
West Nile virus b	2		
Leishmaniasis	2 2		
Malaria	1		
Lebanon	5		
Lumpy skin disease a	2		
Malaria	1		
Bluetongue disease a	1		
Leishmaniasis	1		
Total	102		

Unpublished Data



Learning Objectives

- 1. Identify the role of informal disease event monitoring and reporting
- 2. Define specific applications of informal disease event monitoring systems for outbreak detection
- 3. Discuss data limitations in real-time epidemic forecasting

Table 1. Summary of Data Needs for Real-time Global Epidemic Forecasting

Aim	Data Needs	Examples of Open-Access Data Sources
Case counts	Case counts including confirmed, probable, and suspected cases Open sharing of case data	ProMED ⁹ HealthMap ¹⁰ World Health Organization ¹¹
Mobility	Movement of individuals and populations Flight and travel networks	Flowminder ¹² Flirt ¹³
Host susceptibility	Immunization coverage data: pediatric and adult	GHSA ¹⁴ World Health Organization ¹¹ UNICEF ¹⁵
Environmental susceptibility	Climate data such as temperature and precipitation Environmental characteristics, eg, flooding Vector mapping Ecological niche mapping	NOAA ¹⁶ NASA Earthdata ¹⁷ Natural Earth ¹⁸
Healthcare capacity	GPS latitude and longitude coordinates of hospitals, clinics, and health posts Number of beds Number of physicians Number of nurses Number of critical care beds	Healthsites.io ¹⁹ World Bank ²⁰
Population density and spatial demographic data	Census data Shapefiles for all countries Corresponding estimates of population sizes	LandScan ²¹ WorldPop ²² Facebook Population Maps ²³

	Description	Data sources	URL
Financial Times: coronavirus tracker	The Financial Times analyses title scale of the COVID-19 outbreak including the collection and analysis of data on excess mortality (ie, numbers of deaths higher than the historical average) across the globe	WHO, the COVID Tracking Project, Johns Hopkins University, Our World in Data, US Centers for Disease Control and Prevention, and others	https://www.ft.com/content/ a2901ce8-5eb7-4633-b89c- cbdf5b386938
The Economist: tracking COVID-19 excess deaths across countries globally	Global excess death tracker	Human Mortality Database, World Mortality Dataset, and EuroMOMO	https://www.economist.com/graphic- detail/coronavirus-excess-deaths- tracker
The New York Times			
The pandemic's hidden toll	Excess deaths during the COVID-19 pandemic	Data are compiled from official national and municipal data for 24 countries	https://www.nytimes.com/ interactive/2020/04/21/world/ coronavirus-missing-deaths.html
Tracking the coronavirus at US colleges and universities	COVID-19 tracker at US colleges and universities; with no national tracking system, and statewide data available only sporadically, colleges have been making their own rules for how to tally infections	The New York Times surveyed more than 1900 US colleges and universities for COVID-19 information	https://www.nytimes.com/ interactive/2020/us/covid-college- cases-tracker.html
What we know about coronavirus cases in K-12 schools so far	Reporting focused on district-level and statewide COVID-19 case totals for public schools in the USA; the numbers presented are minimums because of differences in reporting	State and local health and education agencies or were identified by The Covid Monitor or the National Education Association and independently confirmed by The New York Times; The New York Times directly surveyed every school district in eight states: Colorado, Florida, Georgia, Illinois, Indiana, North Carolina, Texas, and Utah	https://www.nytimes.com/ interactive/2020/09/21/us/covid- schools.html
How full are hospital ICUs near you?	Occupancy levels in US ICUs	US Department of Health and Human Services (hospital capacity data); US Department of Homeland Security (hospital locations); and the Covid-19 Hospitalization Tracking Project, University of Minnesota Carlson School of Management	https://www.nytimes.com/ interactive/2020/us/covid-hospitals- near-you.html
The Atlantic: the COVID-19 tracking project	Collected, cross-checked, and published COVID-19 data from 56 US states and territories regarding testing, hospitalisation, and patient outcomes, providing ethnic demographic information and data on long-term-care facilities*	COVID-19 data from websites of US state or territory public health authorities; a public data API provides access to all their data at a national and state level	https://covidtracking.com
The Hindu: coronavirus India tracker	Collects, aggregates, analyses, and visualises state-level COVID-19 cases, deaths, and testing data from India and globally	Bing and Johns Hopkins University	https://www.thehindu.com/ coronavirus/
Zeit Online: coronavirus in Deutschland und bei Ihnen [coronavirus cases in Germany and globally]	Collects, aggregates, analyses, and visualises COVID-19 cases, deaths, patients in ICUs, and vaccinations	Robert Koch Institute, websites of German counties and states, and Johns Hopkins University	https://www.zeit.de/wissen/ gesundheit/corona-zahlen- deutschland-neuinfektionen-inzidenz- aktuelle-karte

Table: Examples of COVID-19 data collection, visualisation, and analysis

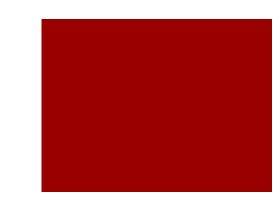
Desai et al. Lancet Digital Health. 2021.

Some additional thoughts

- Ethical Issues to Consider
- Zhao et al. JMIR. 2021 identified 6 domains
 - Awareness of implementing digital infectious disease surveillance
 - Digital integrity
 - Trust
 - Privacy and confidentiality
 - Civil rights
 - Governance

Extra Credit

- Wastewater Surveillance
- Building better buildings
- Genomic Surveillance



Extra Credit: Wastewater Surveillance

Notes from the Field: Early Evidence of the SARS-CoV-2 B.1.1.529 (Omicron) Variant in Community Wastewater — United States, November– December 2021

Weekly / January 21, 2022 / 71(3);103-105

Amy E. Kirby, PhD¹; Rory M. Welsh, PhD¹; Zachary A. Marsh, MPH¹; Alexander T. Yu, PhD²; Duc J. Vugia, MD²; Alexandria B. Boehm, PhD³; Marlene K. Wolfe, PhD⁴; Bradley J. White⁵; Shannon R. Matzinger, PhD⁶; Allison Wheeler, MSPH⁶; Laura Bankers, PhD⁶; Kevin Andresen, MPH⁶; Cristal Salatas, MSGH⁶; New York City Department of Environmental Protection; Devon A. Gregory, PhD^{7,8,9}; Marc C. Johnson, PhD⁷; Monica Trujillo, PhD¹⁰; Sherin Kannoly, PhD⁸; Davida S. Smyth, PhD¹¹; John J. Dennehy, PhD^{8,9}; Nicolae Sapoval¹²; Katherine Ensor, PhD¹²; Todd Treangen, PhD¹²; Lauren B. Stadler, PhD¹²; Loren Hopkins, PhD^{12,13} (View author affiliations)

Extra Credit: Wastewater Surveillance

TABLE. Detection of mutations associated with the SARS-CoV-2 B.1.1.529 (Omicron) variant in wastewater — California, Colorado, New York City, and Houston, Texas, November 21-December 16, 2021

Location	Sample date	Test method	Results
California			
Sewershed A	Nov 25, 2021	Mutation-specific RT-PCRs targeting delHV69–70 and del143–145*	Both mutations detected at <1,000 genomic copies/gram wastewater solids
Sewershed B	Nov 30, 2021	Mutation-specific RT-PCRs targeting delHV69–70 and del143–145*	Both mutations detected at <1,000 genomic copies/gram wastewater solids
	Dec 2, 2021	Mutation-specific RT-PCRs targeting delHV69–70 and del143–145*	Both mutations detected at <1,000 genomic copies/gram wastewater solids
		Partial sequencing of S-gene using ARTIC v4 73R, 74L primers	Detected 9 bp insertion mutation in s214EPE and 3 bp N211I deletion
Sewersheds (10 sites)	Dec 17, 2021 10 of 10 sites	Mutation-specific RT-PCR targeting del143–145*	Mutations detected at >4,500 genomic copies/gram wastewater solids
Colorado			
Sewersheds (21 sites)	Dec 2, 2021 One of 21 sites	SARS-CoV-2-enriched tiled amplicon sequencing	Detected 13 of 17 Omicron-associated mutations
	Dec 6, 2021 Zero of 21 sites	SARS-CoV-2-enriched tiled amplicon sequencing	No Omicron-associated mutations detected
	Dec 9, 2021 Five of 21 sites	SARS-CoV-2-enriched tiled amplicon sequencing	Detected between four and 13 of 17 Omicron-associated mutations depending on the site
	Dec 13, 2021 12 of 21 sites	SARS-CoV-2-enriched tiled amplicon sequencing	Detected between six and 14 of 17 Omicron-associated mutations, depending on the site
	Dec 16, 2021 19 of 21 sites	SARS-CoV-2-enriched tiled amplicon sequencing	Detected between 12 and 14 of 17 Omicron-associated mutations, depending on the site
New York City			
Sewershed A	Nov 21, 2021	Short-read sequencing of S-gene amplicon ^{1,5}	Detected 12 Omicron-associated mutations including eight mutations unique to Omicron
	Nov 28, 2021	Short-read sequencing of S-gene amplicon ^{†,5}	Detected 12 Omicron-associated mutations including eight mutations unique to Omicron
Sewershed B	Nov 28, 2021	Short-read sequencing of S-gene amplicon ^{†,5}	Detected 12 Omicron-associated mutations including eight mutations unique to Omicron
Houston, Texas			
Sewersheds (39 sites)	Nov 29, 2021 Seven of 39 sites	SARS-CoV-2-enriched tiled amplicon sequencing using ARTIC v3 primers [¶]	Detected six Omicron-associated mutations
	Dec 6, 2021 25 of 39 sites	SARS-CoV-2-enriched tiled amplicon sequencing using ARTIC v3 primers [§]	Detected 14 Omicron-associated mutations
	Dec 13, 2021 35 of 39 sites	SARS-CoV-2-enriched tiled amplicon sequencing using ARTIC v3 primers [§]	Detected 18 Omicron-associated mutations

Abbreviation: RT-PCR = reverse transcription-polymerase chain reaction.

* https://www.protocols.io/view/quantification-of-sars-cov-2-variant-mutations-hv6-b2qmqdu6

† https://www.medrxiv.org/content/10.1101/2021.03.21.21253978v1

§ https://www.medrxiv.org/content/10.1101/2021.07.26.21261142v1

https://www.medrxiv.org/content/10.1101/2021.09.08.21263279v1

Polling Question 2

Would you consider consulting a digital disease event monitoring tool now after today's Rounds session?

Thank You

- Britta Lassmann
- Lawrence Madoff
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