

1 **Public Health Data Applications Using the CDC Tracking Network: Augmenting**
2 **Environmental Hazard Information with Lower-latency NASA Data**

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24 **Abstract**

25 Exposure to environmental hazards is an important determinant of health, and the
26 frequency and severity of exposures is expected to be impacted by climate change. Through a
27 partnership with the U.S. National Aeronautics and Space Administration, the U.S. Centers for
28 Disease Control and Prevention's National Environmental Public Health Tracking Network is
29 integrating timely observations and model data of priority environmental hazards into its publicly
30 accessible Data Explorer (<https://ephtracking.cdc.gov/DataExplorer/>). Newly integrated datasets
31 over the contiguous U.S. (CONUS) include: daily 5-day forecasts of air quality based on the
32 Goddard Earth Observing System Composition Forecast (GEOS-CF), daily historical (1980-
33 present) concentrations of speciated PM_{2.5} based on the Modern Era Retrospective analysis for
34 Research and Applications, version 2 (MERRA-2), and Moderate Resolution Imaging
35 Spectroradiometer (MODIS) daily near real-time maps of flooding (MCDWD). Data integrated
36 into the CDC Tracking Network are broadly intended to improve community health through
37 action by informing both research and early warning activities, including (1) describing temporal
38 and spatial trends in disease and potential environmental exposures, (2) identifying populations
39 most affected, (3) generating hypotheses about associations between health and environmental
40 exposures, and (4) developing, guiding, and assessing environmental public health policies and
41 interventions aimed at reducing or eliminating health outcomes associated with environmental
42 factors.

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47 **Plain Language Summary**

48 The well-being of people and communities can be impacted by hazards like polluted air
49 and flooding. This paper describes a new partnership to bring more timely information about
50 U.S. air quality and flooding into a centralized online environmental public health platform, the
51 CDC Tracking Network’s Data Explorer. Timely monitoring and forecasting of environmental
52 hazards have the potential to reduce harmful exposures by enabling early warnings and other
53 timely public health interventions. Currently, near real-time data may not be readily accessible to
54 public health practitioners, particularly those that need to react to public health emergencies,
55 because the data are fragmented across multiple systems in formats inconsistent with typical
56 public health uses. Developing data pipelines and processes to host this data in a usable and
57 standardized format and disseminating the data via an existing online platform assists in making
58 important information about environmental health hazards more readily accessible to public
59 health practitioners, decision makers, and partners.

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70 **Key Points**

- 71 • Timely information can be critical for public health decision making, but environmental
72 hazard data can be fragmented or hard to use.
- 73 • This work brings timely, routinely updated NASA data into CDC Tracking Network's
74 Data Explorer in familiar and accessible formats.
- 75 • For CONUS, daily NASA air quality forecasts, historical air quality data, and daily flood
76 maps are integrated into the Data Explorer.

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93 **1. Introduction**

94 Harmful environmental exposures are a significant contributor to morbidity and mortality
95 in the U.S. and numerous health outcomes are expected to worsen under climate change
96 [USGCRP, 2016; Watts et al., 2021]. The U.S. National Aeronautics and Space Administration
97 (NASA) uses satellites, aircraft, ground measurements, and Earth system models to study the
98 long-term state of the environment as well as produce near real-time (NRT) data about
99 environmental hazards such as natural disasters [NASA, 2023a]. The U.S. Centers for Disease
100 Control and Prevention’s (CDC) National Environmental Public Health Tracking Program (CDC
101 Tracking Program) tracks exposures and health effects associated with environmental hazards
102 [CDC, 2023a]. The Environmental Public Health Tracking Network (CDC Tracking Network),
103 built by the CDC Tracking Program, is a dynamic, web-based system that brings together health
104 data and environmental data from a network of partners to provide timely, relevant information
105 for better community health (<https://ephtracking.cdc.gov/>). This paper describes recent efforts to
106 augment information provided through the CDC Tracking Network with lower-latency, higher
107 spatiotemporal resolution NASA datasets for high priority environmental hazards. The overall
108 goal is to enable health researchers and practitioners to access critical environmental data needed
109 to understand and respond to health risks and make data-informed decisions.

110 The mission of the CDC Tracking Program is to provide information from a nationwide
111 network of integrated health and environmental data that drives actions to improve community
112 health [CDC, 2023a]. The CDC Tracking Network is a multi-tiered, web-based system with
113 components at national, state, and local levels that unifies health and environmental data from a
114 network of varied sources and makes that information publicly available to its audience of public
115 health researchers, professionals, decision makers, and the public in standardized formats [CDC,

116 2023a]. In collaboration with federal, state, and local partners, priority environmental health
117 issues and key surveillance questions are identified. Existing data are evaluated for their ability
118 to inform these issues and then integrated into the CDC Tracking Network. Data are
119 disseminated through the CDC Tracking Network’s flagship product, the Data Explorer
120 (<https://ephtracking.cdc.gov/DataExplorer/>), as well as dashboards, infographics, and an
121 application programming interface (API). Additionally, when gaps or new data needs are
122 identified, the CDC Tracking Program collaborates with partners
123 (<https://www.cdc.gov/nceh/tracking/partners>) to develop standards for data collection, develop
124 models, expand the utility of non-traditional public health data, or develop new methodologies
125 for using existing data.

126 Near real-time monitoring and forecasting of environmental hazards have the potential to
127 reduce harmful exposures by enabling early warning systems and other timely public health
128 interventions [e.g., WMO, 2023]. Currently, critical near real-time data may not be readily
129 accessible to public health practitioners, particularly those that need to react to public health
130 emergencies, because the data are fragmented across multiple data systems and are not formatted
131 in a manner consistent with typical public health uses (e.g., using administrative boundaries such
132 as county and census tract) [Liu et al, 2021]. Augmenting and improving data pipelines and
133 processes to host these data in a usable and standardized format and disseminating these data via
134 the CDC Tracking Network will assist in making important information about environmental
135 hazards more readily accessible to public health practitioners and partners, including relevant
136 CDC programs, and state, tribal, local, and territorial health departments.

137 The objective of this work is to leverage the NASA Goddard Space Flight Center (GSFC)
138 expertise in processing and interpreting a wide range of environmental satellite data products and

139 to make these data more accessible to public health practitioners through the CDC Tracking
140 Network. Section 2 describes the data needs of the CDC Tracking Program and efforts to identify
141 environmental datasets by NASA GSFC to meet these needs. Section 3 details the data systems
142 methodology and implementation processes to define and establish routine workflows for
143 transforming and transferring data in an automated and timely fashion. Section 4 illustrates
144 applications of the data. Section 5 summarizes limitations of the datasets and offers some
145 guidance on appropriate uses in public health context. Sections 6 and 7 capture outlooks and
146 conclusions, respectively.

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148 **2. Public Health Environmental Data Needs**

149 This section describes the collaborative process between the CDC Tracking Program and
150 NASA GSFC around the use of available environmental hazard data to address current CDC data
151 needs. Priority topic areas include air quality, flooding, extreme weather, wildfires, and climate
152 impacts. These priorities reflect gaps in existing content on the CDC Tracking Network's Data
153 Explorer, feedback from the CDC Tracking Network's Content Work Group focused on air
154 quality, climate, and weather as areas of programmatic emphasis within the CDC National
155 Center for Environmental Health and the CDC Division of Environmental Health Science and
156 Practice. The CDC Tracking Program identified a need for lower latency data (i.e., shorter lag
157 time between data production and availability to users) because of the potential to inform public
158 health emergencies and to have timelier data for decision-making. This paper presents progress
159 on the first two priority topic areas of air quality and flooding.

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162 2.1 Data selection

163 NASA GSFC produces numerous publicly available data products relevant for air quality
164 (<https://airquality.gsfc.nasa.gov/>) and flooding [NASA, 2023b]. Product applicability was
165 assessed based on geographic coverage and resolution, temporal coverage and resolution,
166 latency, means of data access, file formats, and levels of processing (e.g., Level 1–Level 5
167 [NASA, 2023c]). Broadly, common features of the selected data are that they are available daily
168 over CONUS, remotely accessible through an API, and produced on an ongoing basis. The
169 suitable data products selected are summarized in **Table 1**.

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171 2.2 Air quality

172 The air quality data products selected are the Goddard Earth Observing System
173 Composition Forecast (GEOS-CF, https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)
174 and the Modern-Era Retrospective analysis for Research and Applications, Version 2 (MERRA-
175 2, <https://gmao.gsfc.nasa.gov/reanalysis/MERRA-2/>). Both are global gridded products made
176 publicly available by the NASA GSFC Global Modeling and Assimilation Office (GMAO) on an
177 ongoing basis. GEOS-CF is a global three-dimensional model of atmospheric composition
178 generated using the GEOS Earth System Model coupled with the GEOS-Chem chemical
179 transport model, with initial meteorological conditions constrained by satellite observations
180 [Keller et al., 2021]. Anthropogenic air pollutant emissions are obtained from pre-defined
181 emission inventories [Crippa et al., 2018] and wildfire emissions are derived daily from satellite
182 observations using the Quick Fire Emissions Dataset (QFED) [Darmenov and da Silva, 2015].
183 GEOS-CF produces daily 5-day global forecast at approximately 25 km x 25 km horizontal
184 resolution of surface concentrations of criteria pollutants designated by the U.S. Environmental

185 Protection Agency (EPA) [USEPA, 2023]: particulate matter less than 2.5 microns in diameter
186 (PM_{2.5}), surface ozone (O₃), nitrogen dioxide (NO₂), sulfur dioxide (SO₂), and carbon
187 monoxide (CO) [Keller et al., 2021; Knowland et al., 2022]. GEOS-CF contributes a novel
188 forecasting capability to the suite of air quality information currently made available on the CDC
189 Tracking Network. MERRA-2 is an atmospheric re-analysis product fusing measurements and
190 model results to produce a physically-consistent estimate of the state of the atmosphere and
191 aerosols for the period 1980-present at approximately 50 km x 50 km horizontal resolution with
192 a latency of approximately 1 month behind real-time [Gelaro et al., 2017; Bosilovich et al., 2015;
193 Randles et al., 2016, 2017; Buchard et al., 2016; 2017]. MERRA-2 uniquely contributes a long
194 (+42 year) historical PM_{2.5} record and detailed information about the aerosol constituents (e.g.,
195 black carbon, organic carbon) of total PM_{2.5}.

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197 *2.3 Flooding*

198 The flood product selected is the Moderate Resolution Imaging Spectrometer (MODIS)
199 near real-time (within 3-5 hours after observation) global flood product (MCDWD) [LANCE
200 MCDWD, 2022]. MCDWD is a publicly available satellite-based data product that estimates the
201 presence and surface extent of flood waters (i.e., standing water) over land (**Figure 1**). MCDWD
202 is provided at ~250-meter resolution, with near-global daily coverage spanning 2011-present and
203 is derived from the MODIS Surface Reflectance (MOD09) datasets from the Aqua and Terra
204 satellites [NASA, 2023d, k; Slayback, 2023]. MCDWD is processed and made publicly available
205 within 3 hours of collection by NASA's Land, Atmosphere Near real-time Capability for EOS
206 (LANCE, <https://lance.modaps.eosdis.nasa.gov/>). Detailed descriptions of the MCDWD product,
207 algorithms, product evaluation, data access, and planned improvements are available in Slayback

208 [2023] and online <https://www.earthdata.nasa.gov/global-flood-product>. Near real-time high-
209 resolution (3-meter) flood maps derived from commercial PlanetScope imagery [Policelli, 2022]
210 were considered but excluded because their application is presently more suited for limited areas
211 of interest and routinely generating maps covering CONUS is not yet feasible.

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213 **3. Implementation & Data Systems Methodology**

214 Through an interagency partnership, the goal is to define and establish routine processes for
215 transforming and transferring NASA GSFC data to the CDC Tracking Program and its partners
216 in an automated and timely fashion for dissemination via the CDC Tracking Network. Processes
217 are developed in the CDC Enterprise Data Analytics and Visualization (EDAV) Platform, which
218 is a cloud-based data management and processing ecosystem where users can store, transform,
219 and analyze data. EDAV is built primarily using Microsoft Azure cloud services. Azure's
220 implementation of Databricks, a web-based analytics platform, is used for scripting data
221 ingestion from NASA APIs, further transformation of gridded raw data to geopolitical
222 boundaries, and calculating relevant quantities (e.g., sum, mean, area) using Python and R.
223 Completed data are routed to the Azure Data Lake, where most recent and archival versions of
224 daily data runs are stored. Azure Data Factory is used to orchestrate the initialization of
225 Databricks scripts in the appropriate order and the routing of data between different storage
226 environments (**Figure 2**).

227

228 *3.1 Air quality*

229 The air quality data products are queried remotely, spatially transformed from model grid to
230 U.S. counties or census tracts, depending on the dataset, and converted to measures tailored to

231 environmental public health audiences. An automated daily query pulls GEOS-CF via Open-
232 source Project for a Network Data Access Protocol (OPeNDAP), retrieving global 5-day
233 forecasts for PM2.5, O3, SO2, NO2, and CO. Forecast results are extracted for the surface
234 (model level 1) for CONUS and transformed from the GEOS-CF model grid to U.S. counties
235 based on boundaries derived from Census TIGER 2000 [NOAA, 2023] (see **Section 6** for
236 planned updates). Daily statistics (24-hour mean [PM2.5, SO2], 24-hour maximum [SO2, NO2],
237 and maximum 8-hour rolling mean [CO, O3]) are calculated from hourly values for each county.
238 An automated monthly query pulls MERRA-2 M2T1NXAER from the Goddard Earth Sciences
239 Data and Information Services Center (GES DISC) [GMAO, 2015]. Surface (model level 1)
240 aerosol constituents, SO2, and CO concentrations are extracted for CONUS. PM2.5
241 concentrations in micrograms per cubic meter are estimated by summing the aerosol constituents
242 following [Buchard et al., 2017]

243

$$\begin{aligned} 244 \quad \text{PM2.5} = & \text{PM2.5}[\text{DU}] + \text{PM2.5}[\text{SS}] + \text{PM2.5}[\text{OC}] + \text{PM2.5}[\text{BC}] + \\ 245 \quad & (132.14/96.06) * \text{PM2.5}[\text{SO4}] \end{aligned}$$

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247 where DU is dust, SS is sea salt, OC is organic carbon, BC is black carbon, SO4 is sulfate, and
248 the factor of 132.14/96.06 is applied to convert sulfate ion (molar mass of 96.06 g mol⁻¹)
249 concentration output by MERRA-2 to ammonium sulfate (132.14 g mol⁻¹), assuming that sulfate
250 is primarily present as neutralized ammonium sulfate. MERRA-2 SO2 surface concentrations are
251 converted from kilograms per cubic meter (kg m⁻³) to parts per billion (ppb) using the following
252 equation:

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$$\text{SO}_2 \text{ (ppb)} = (\text{T(k)} * \text{SO}_2 \text{ (kg m}^{-3}\text{)} * 8.314 * 10^7) / (1.013 * 64.066)$$

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256 Coordinated Universal Times are converted to the local time of each county centroid, then daily
257 statistics (24-hour mean [PM2.5 (and constituents), SO₂], 24-hour maximum [SO₂, NO₂], and
258 maximum 8-hour rolling mean [CO]) are calculated from hourly values for each county. The Air
259 Quality Index (AQI) and AQI category (reflecting level of health concern) associated with each
260 daily concentration for both GEOS-CF and MERRA-2 estimates are determined as specified in
261 the U.S. EPA technical documentation [USEPA, 2018]. Concentrations higher than the AQI
262 scale (“beyond the AQI”) are classified as if they are in the highest AQI category [USEPA,
263 2018].

264 County-level daily AQI category estimates for GEOS-CF and MERRA-2 air pollutant
265 estimates are mapped on the CDC Tracking Network’s Data Explorer (**Figure 3**). Measures
266 incorporating GEOS-CF data are housed within the “Forecasted Air Quality” indicator and those
267 derived from MERRA-2 can be found in the “Current and Historical Air Quality” indicator, both
268 of which are in the “Air Quality” content area. The “Forecasted Air Quality” indicator relies
269 entirely on GEOS-CF data that updates each morning at approximately 7 am U.S. Eastern
270 Standard Time (EST) and includes a 1-day hindcast and 4-day forecast (including current day).
271 MERRA-2 is incorporated in a suite of daily air quality measures that also includes monitor data
272 from the U.S. EPA’s AirNow platform (<https://www.airnow.gov/>) [CDC, 2023b]. These
273 measures include data from the previous several months, though only AirNow data is available
274 for the present month due to the 1-month latency of MERRA-2. Users can view the underlying
275 concentration used to derive the AQI category by hovering the cursor over a county of interest.

276 All AQI category and concentration data are also available in tabular format using the data
277 download tool on the Data Explorer or via the CDC Tracking Network API.

278

279 *3.2 Flooding*

280 The MCDWD flood product is used to generate the “Total Area Flooded” and “Total
281 Population Affected by Flooding” measures on the CDC Tracking Network’s Data Explorer.
282 Measures incorporating MCDWD data are housed within the “Current and Historical Flooding”
283 indicators in the “Precipitation & Flooding” content area. The data are queried remotely,
284 mosaicked, clipped, and spatially summarized from raster data to U.S. counties, U.S. territory
285 county equivalents, and census tracts. A daily automated query retrieves MCDWD from the
286 NASA LANCE system (<https://lance.modaps.eosdis.nasa.gov/>). The query is made at 3 am EST
287 to retrieve the most-recent 2-day composite. The data are provided by LANCE in 10 degree x 10
288 degree tiles, with variables for two types of flood conditions: recurring flooding (available in
289 2024, see Section 6) and unusual flooding. The total area flooded is estimated by determining the
290 proportion of the MCDWD tiles where flood has been identified relative to the total raster
291 overlap with each jurisdiction. This proportion is then multiplied by the total area of the
292 jurisdiction. Separate calculations are made to determine the quantity of recurring and unusually
293 flooded land. To estimate the total number of people affected by flooding within each county and
294 census tract, the sum of the population of all the affected census blocks within each county and
295 tract is calculated. A block is considered affected if a flooded tile is identified anywhere within
296 its boundary. Similarly, separate calculations are made to determine the total population affected
297 by recurring and unusual flooding.

298

299 4. Applied Uses

300 Data integrated into the CDC Tracking Network are broadly intended to inform both
301 research and early warning activities, including (1) describing temporal and spatial trends in
302 disease and potential environmental exposures, (2) identifying populations most affected, (3)
303 generating hypotheses about associations between health and environmental exposures, and (4)
304 developing, guiding, and assessing environmental public health policies and interventions aimed
305 at reducing or eliminating health outcomes associated with environmental factors. Numerous
306 contextual data are available on the Data Explorer that can be viewed contemporaneously to help
307 drive data-informed decision-making, such as biomonitoring data, disease burdens, community
308 characteristics, and environmental justice indicators. Hypothetical use cases are provided below
309 for illustrative purposes.

- 310 ● **Target public health surveillance:** In the CDC Tracking Network’s Data Explorer, the
311 “Total Area Flooded” measure derived from MCDWD can provide situational awareness
312 and identify where to target public health surveillance, community services,
313 environmental sampling, or additional imagery acquisition (e.g., drone, commercial
314 satellite). The maps can assist in identifying inundated areas, as well as dry areas where
315 emergency response assets can be staged. The Data Explorer also has points of interest,
316 such as nursing homes and day care centers, that can be viewed alongside a map of
317 “Total Area Flooded” to better inform public health activities. Environmental justice
318 measures, such as populations that are linguistically isolated, can also be layered on the
319 Data Explorer map to help prioritize assistance.
- 320 ● **Reduce personal exposure:** In summer 2023, a historic Canadian wildfire smoke event
321 left millions of residents in the northeastern U.S. under air quality advisories [NASA,

322 2023e]. During a similar major air quality, an individual with asthma might consult the
323 “Forecasted Air Quality” indicator derived from GEOS-CF on the CDC Tracking
324 Network’s Data Explorer alongside AirNow and other sources of information to inform
325 the timing and confidence around decisions of when (or if) to perform strenuous exercise
326 outdoors.

327 ● **Examine trends in patient-centered outcomes:** A longitudinal study of wildfire smoke
328 impacts on patient-centered outcomes for adult patients with asthma might examine the
329 “Current and Historical Air Quality” indicators derived from MERRA-2 on the CDC
330 Tracking Network’s Data Explorer for insight on spatiotemporal trends in smoke
331 exposure, cumulative exposures, timing and duration of smoke events, as well as
332 statistical anomalies. In the Data Explorer, measures of PM_{2.5} and its constituents can be
333 viewed alongside indicators of social vulnerability, occupation, and environmental justice
334 indices to illuminate the interplay of wildfire smoke exposure and social determinants of
335 health. The study results might then inform improved interventions such as refined public
336 health messaging to populations who are at higher risk for asthma-related emergency
337 department visits or hospitalizations during wildfire events.

338

339 **5. Limitations**

340 *5.1 Air Quality*

341 The GEOS-CF 5-day air quality forecast and MERRA-2 historical re-analysis provide regional
342 estimates of air quality. Greater caution is warranted when interpreting results in geographic
343 areas with strong pollution gradients (e.g., urban areas), mountainous terrain, locations near
344 emission point sources, and locations on the coast. On average, GEOS-CF and MERRA-2 have

345 greater skill in the eastern U.S. and in rural areas [Randles et al., 2016; Buchard et al., 2017].
346 MERRA-2 has less skill during winter and spring and in environments dominated by nitrate
347 aerosol (e.g., intense agricultural areas) because nitrate is not explicitly accounted for in the
348 GOCART aerosol mechanism [Randles et al., 2016]. MERRA-2 is constrained by aerosol optical
349 depth (AOD), and there is inherent uncertainty in how aerosols are vertically distributed in the
350 model. For GEOS-CF, forecast skill is generally greater for the nearer-term forecasts (i.e., Day 1,
351 Day 2) [Keller et al., 2021]. For both GEOS-CF and MERRA-2, relative changes and trends are
352 more robust than absolute changes. The air quality measures displayed in the Data Explorer
353 represent outdoor ambient air pollution concentrations and do not directly reflect human
354 exposures. The relationships between outdoor ambient concentrations, indoor air pollution, and
355 individual exposures are active subjects of ongoing research, and likely vary depending upon
356 pollutant, behavioral patterns, microenvironments, and building ventilation [Özkaynak et al.,
357 2009; Patel et al., 2020; Singer et al., 2020]. The NASA air quality data products disseminated
358 via the CDC Tracking Network are research-grade products and should not be used to assess
359 National Ambient Air Quality Standards (NAAQS) compliance or to evaluate progress toward
360 attaining compliance.

361

362 *5.2 Flooding*

363 MCDWD provides estimates of surface flood water extent at the time of satellite
364 overpass between roughly 70N–70S. Greater caution is warranted in interpreting results for
365 MODIS data under shadows caused by terrain and clouds, which can strongly resemble water in
366 the spectral bands (red and near infra-red) used to generate the product [Slayback, 2023]. Flash
367 floods have a low likelihood of being observed due to their rapid appearance and disappearance,

368 unless this coincides with the twice-daily observation times of the Terra and Aqua satellites at
369 approximately 10:30 am and 1:30 pm local solar time [Gosset et al., 2023]. Flooding in urban
370 areas is also difficult for the 250-meter resolution MCDWD product to detect, due to the
371 mismatch in sensor resolution to on-the-ground flooded areas. In addition to shadows, dark
372 volcanic rock and recently melted snow can result in false positives, and vegetation cover can
373 result in false negatives in MCDWD. The MODIS instruments aboard Terra and Aqua have
374 flown in space for over 20 years and these missions have been scheduled for decommissioning in
375 the 2025-2026 timeframe. MCDWD is transitioning to ingest data from the Visible Infrared
376 Imaging Radiometer Suite (VIIRS) aboard the joint NASA/NOAA Suomi-NPP and NOAA-20
377 satellites.

378

379 **6. Future Planned Work**

380 Planned air quality-related updates include: improving the computational efficiency and
381 flexibility of the Python workflow used to spatially aggregate MERRA-2 and GEOS-CF data;
382 adding coverage over Alaska, Hawaii, Puerto Rico, and Virgin Islands; and updating the county
383 and census boundary shapes to Census TIGER 2020. The update to Census TIGER 2020
384 boundaries is anticipated to mostly impact Alaska and Bedford, Virginia.

385 Planned flooding updates include: MCDWD recurring flood classification and exploring
386 a second, complimentary flood product. The NASA MCDWD team plan to update the product to
387 include a flood classification for “recurring flood” (water occurring where it has occurred in the
388 past with some regularity, but is not permanent water) in early 2024. Maps on the CDC Tracking
389 Network’s Data Explorer will be updated with new data when available. The HydroSAR flood
390 product is being explored to potentially augment MCDWD daily maps. HydroSAR HyP3-

391 watermap is a 30-meter surface water extent product with a 12-day revisit period based on
392 European Space Agency's (ESA) Sentinel-1 C-band synthetic aperture radar (SAR) [ASF, 2023].
393 The HydroSAR project focuses on cloud-based SAR data processing for rapid response and
394 mapping of hydrological hazards. HydroSAR maps are retrieved by querying Alaska Satellite
395 Facility's (ASF) Hybrid Pluggable Processing Pipeline (HyP3) service [Hogenson et al., 2020].
396 HydroSAR's revisit time is expected to be reduced to near-daily coverage with the planned
397 launches of ESA's Sentinel-1C and NASA-Indian Space Research Organization (ISRO)
398 Synthetic Aperture Radar (NISAR) in 2024. For public health surveillance and practice, the
399 unique contributions of HydroSAR would be an order of magnitude increase in spatial resolution
400 (MCDWD 250-meter vs. HydroSAR 30-meter), and SAR technology has all-weather and day-
401 and-night imaging capabilities.

402

403 **7. Conclusions**

404 Through a partnership between the CDC Tracking Program and NASA GSFC that was
405 established in 2022, critical environmental data products are delivered into the CDC Tracking
406 Network's nationwide network of integrated health and environmental data in order to drive
407 actions to improve community health. Near real-time monitoring and forecasting of
408 environmental hazards have the potential to reduce harmful exposures by enabling early warning
409 systems and other timely public health interventions. Currently, near real-time NASA data may
410 not be readily accessible to public health practitioners, particularly those that need to react to
411 public health emergencies, because the data are fragmented across multiple non-interoperable
412 data systems and are not formatted in a manner consistent with typical public health uses.
413 Developing data pipelines and processes to host these data in a usable and standardized format

414 and disseminating these data via the CDC Tracking Network assists in making important
415 information about environmental health hazards more readily accessible to public health
416 practitioners, decision makers, and partners.

417 Based on CDC and partner priorities, NASA GEOS-CF daily 5-day air quality forecasts,
418 MERRA-2 historical daily PM_{2.5} concentrations, and MCDWD daily near real-time maps of
419 flooding are integrated into the CDC Tracking Network's publicly accessible Data Explorer
420 (<https://ephtracking.cdc.gov/DataExplorer/>). These data products are mature, well-established
421 resources produced on an ongoing basis and can be remotely accessed with automated queries to
422 deliver timely CONUS-wide maps of environmental hazards. With the forthcoming
423 decommissioning of the Aqua/Terra satellites, MCDWD will transition to ingesting
424 VIIRS/Suomi NPP observations to ensure continuity. Augmenting MCDWD flood maps with
425 other satellite-based flooding data products is being explored. The air quality and flooding data
426 products newly integrated into the CDC Tracking Network are broadly intended to promote
427 improve community health through data-informed decisions by informing trends, identifying
428 potentially affected populations, and informing interventions that support better patient-centered
429 outcomes.

430

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441

442 **Conflict of Interest**

443 The authors declare no conflicts of interest relevant to this study.

444

445 **Data Availability Statement**

446 All data sources in this manuscript are free and publicly available. NASA GEOS-CF five-day
447 forecasts of air quality pollutants are accessed via OpenDAP [Keller et al., 2021; GMAO, 2023].
448 NASA MERRA-2 historical aerosol concentrations are accessed via GES DISC [GMAO, 2015].
449 The NASA MCDWD daily near real-time flood product is accessed via LANCE [LANCE
450 MCDWD, 2022]. Open-source software is used to query the data, spatially transform the data,
451 and compute measures consumable to a public health audience (e.g., AQI). The code is written in
452 R [R Foundation, 2022] and Python [Python Software Foundation, 2021]. The air quality and
453 flooding measures are available as CONUS-wide maps and as downloadable data layers on the
454 CDC Environmental Public Health Tracking Network Data Explorer [CDC, 2023a].

455

456 **Author Contributions**

- 457 • **Conceptualization:** H. Amos, N. Skaff, A. Werner, S. Schollaert Uz
- 458 • **Formal analysis:** F. Policelli, K. Patel, P. Abue, E. Macorps
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462

463 **Disclaimer**

464 The findings and conclusions in this report are those of the authors and do not necessarily
465 represent the official position of the Centers for Disease Control and Prevention or National
466 Aeronautics and Space Administration.

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641 **Figures and Tables**

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644

645 **Table 1.** NASA Goddard research-grade products for air quality and flooding implemented into
 646 the CDC Tracking Program’s workflow.

Data product	Resolution	Strengths and limitations for public health uses
GEOS-CF air quality forecast	Daily 5-day forecast 25 km x 25 km grid, global	+ daily forecast of criteria pollutants - coarse resolution
MERRA-2 aerosols	Daily, 1980-present 50 km x 50 km, global	+ 40-year record, daily resolution - coarse resolution, 1-month lag
MODIS NRT global daily flood product (MCDWD)	Daily, 2002-present 250 m, global	+ low latency, daily coverage - cannot see through clouds

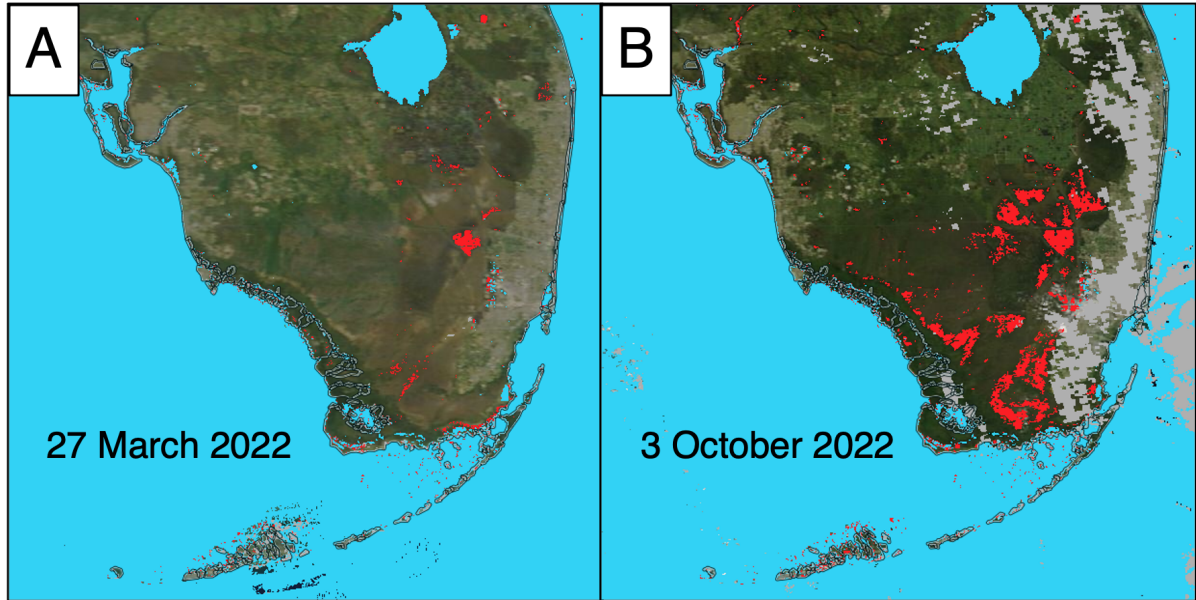
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653 **Figure 1.** Flooding before (A) and after (B) Hurricane Ian as mapped by the NASA MODIS near
 654 real-time global flood product (MCDWD). Flooded pixels are red, gray pixels are clouds. Ian
 655 made landfall on 28 Sep 2022 near Fort Meyers, Florida as a Category 4 hurricane. (A) 27 March
 656 2022 is shown to illustrate a representative baseline of conditions before hurricane season. (B)
 657 Not all the flood water detected on 3 October 2022 is from Hurricane Ian. Flood water can
 658 persist for weeks or longer.

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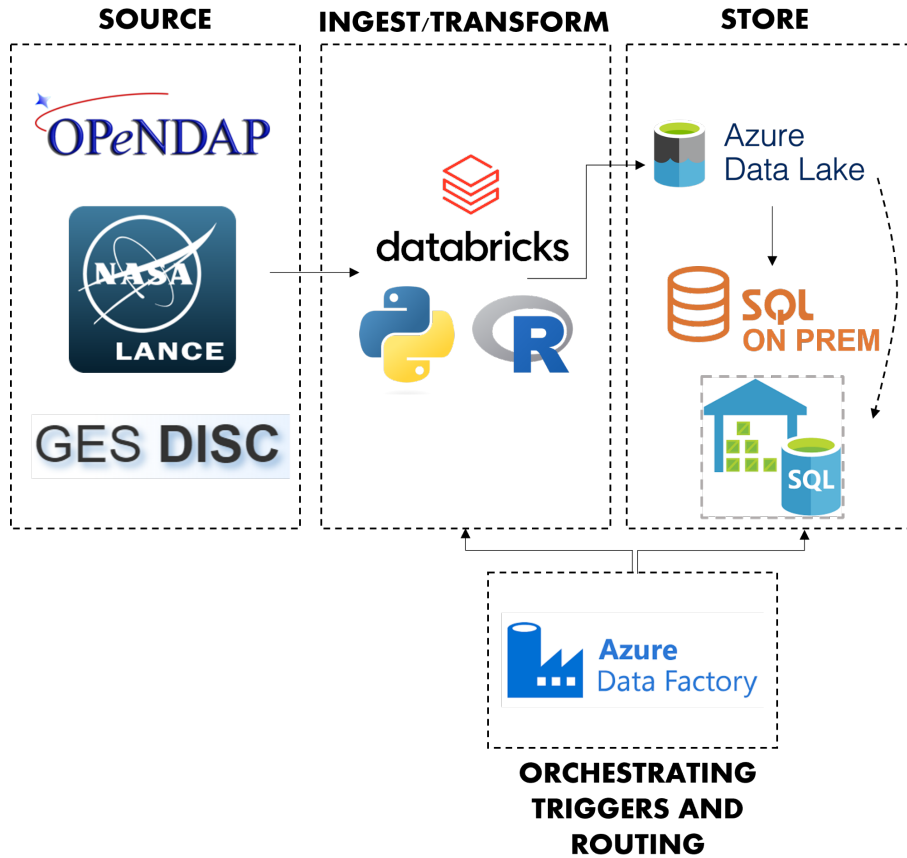
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670 **Figure 2:** Schematic of the CDC Tracking Network’s extract, transform, and load (ETL)

671 workflow.

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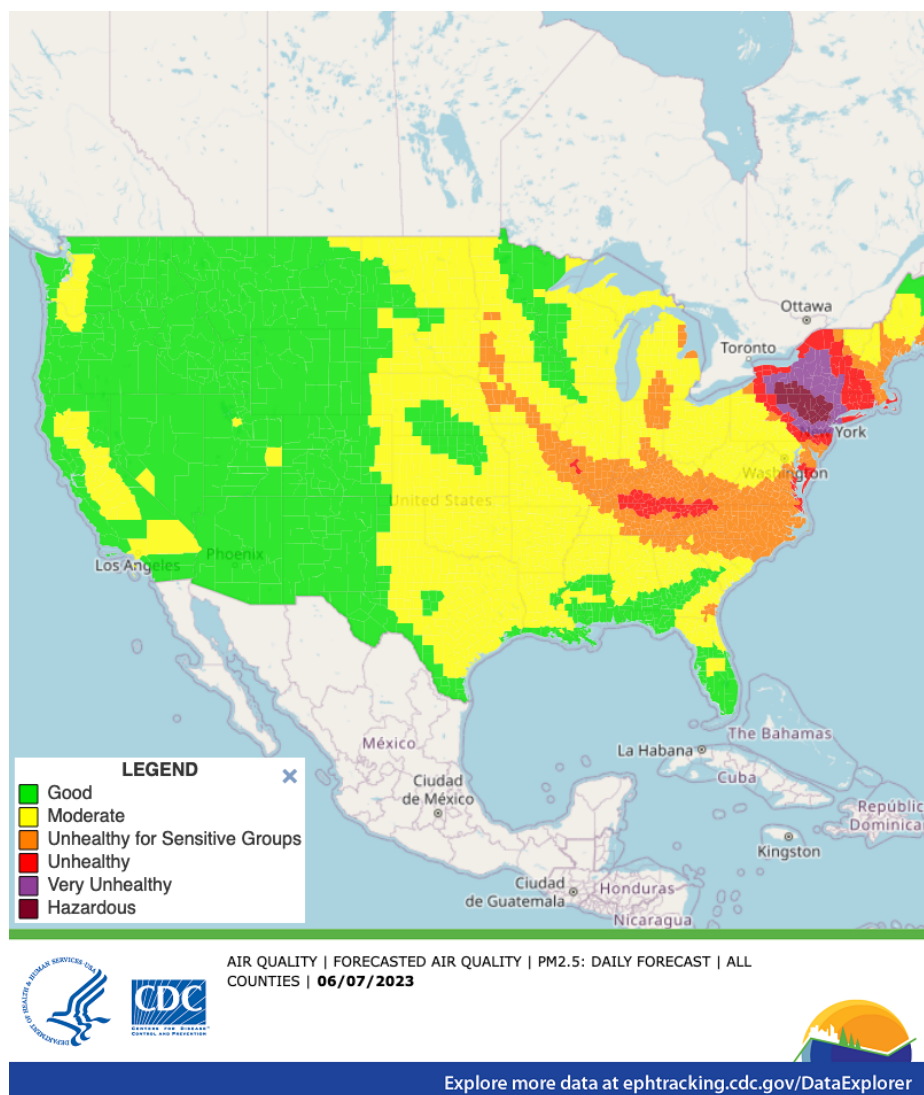
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679 **Figure 3.** Example of a map of GEOS-CF forecasted PM2.5 as seen on the CDC Tracking

680 Network's publicly available Data Explorer. On this day, 6 June 2023, heavy smoke from

681 Canadian wildfires impacted major population centers in the eastern U.S.

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