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Real-time Data Services for the Geosciences (CHORDS)

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8 While modern sensing and communication technologies are enabling the observations of
9 geophysical processes at unprecedented spatiotemporal resolutions, the development of
10 these technologies is significantly outpacing their actual use across the geosciences. This
11 is particularly true of real-time data systems, which are now permitting the streaming and
12 analysis of data at the instant of their measurement. Though the use of real-time scientific
13 data are limited, their importance is ever increasing, particularly in mission critical
14 scenarios where informed decisions must be made rapidly.

15

16 Beyond applications tied to disaster resilience (earthquake prediction, flood forecasting,
17 etc.), now more than ever there is potential to leverage real-time data to fundamentally
18 change how scientific experiments are conducted. For example, in many geoscientific
19 experiments, faulty sensors are often only detected too late, forcing experiments to be

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20 repeated. In settings where mobile sensor nodes are used, or where sampling frequencies
21 need to be adjusted to capture events of interest, few tools are available to adaptively
22 guide the experimental process. This often results in missed observations and wasted
23 experimental investments but can be remedied rapidly by enabling means analyze and
24 respond to streaming data.

25

26 While real-time data stand to enable a paradigm-shift in geoscientific experimentation,
27 they rarely, if ever, form the first step in a geoscientific workflow. The vast majority of
28 existing data platforms are inherently tuned to non real-time applications, where data are
29 often stored in large databases for retrospective analysis and visualization. The few
30 existing real-time data platforms, however, are either proprietary, feed into mission-
31 specific tools, or are otherwise not available to broader stakeholders within the
32 geosciences. While the complexity of these platforms presents a major barrier to the
33 broader adoption of real-time data systems, there are also a number of technical
34 challenges that must be addressed before the use of real-time data becomes commonplace
35 across the geosciences.

36

37 **Existing real-time data platforms**

38 While interoperability standards such as the *Open Geospatial Consortium*¹ (*OGC*) *Sensor*
39 *Web Enablement* (*SWE*) specifications (Nittel, Labrinidis, & Stefanidis, 2008), have
40 created interfaces and metadata encodings to fuse real-time sensor streams into
41 information infrastructures, a common set of tools to couple these streams with
42 workflows and models has yet to be developed. To that end, pioneering efforts are
43 underway by groups such as 52°North² to develop standards and reference
44 implementations for real-time data within the field of Geoinformatics (Andres, Jirka, &
45 Utech, 2014; Jirka, Bröring, Kjeld, Maidens, & Wytzisk, 2012; Reed, Botts, & Davidson,
46 2007). Beyond SWE-based initiatives, a number of other platforms have also been
47 developed to address the emergence of real-time data within the geosciences.

¹ www.opengeospatial.org/standards

² <http://52north.org>

48 UNIDATA's Local Data Manager (Davis & Rew, 1994) provides an event driven
49 infrastructure to manage streaming data. While it has served the purpose of specific
50 projects for many years, the system can be difficult even for an experienced user to install
51 and maintain. Since LDM queues data, the system is not suited for environments in which
52 the stability of networks cannot be assured, which may often be the case with data
53 originating from real-world sensor networks. Its queuing process may also lead to
54 situations where the latest real-time data are not accessible until the queue buffers are
55 flushed, thus causing a backlog of data that prevent timely use.

56

57 Other recent real-time efforts have been undertaken through the *DataTurbine* (Tilak,
58 Hubbard, Miller, & Fountain, 2007) and Antelope³ initiatives. *DataTurbine* is based on a
59 ring-buffer architecture and is implemented in Java as an open-source, server-side
60 platform for the transport and management of real-time data originating from
61 heterogeneous sensors. While powerful, the ring-buffer architecture does not actively
62 support real-time database operations or coupled model-sensor applications. Furthermore,
63 local server resources can limit the size of the ring-buffer, making it possible to drop
64 incoming data. Cloud-based functionalities and OGC standard support are yet to be
65 implemented as features. Significant overhead exists on the part of users, as *DataTurbine*
66 has to be individually ported to field-specific data loggers and instruments. While these
67 examples may appear specific to one platform, they are echoed by all the other real-time
68 data systems as well. The complexities associated with the deployment and operation of
69 existing real-time data platforms present an overhead too large for most research groups
70 to take on, thus significantly limiting the broader adoption of real-time data across the
71 geosciences. The emergence of commercial data platforms under the Internet of Things
72 (IoT) is beginning to provide easier to use alternatives, but these platforms are not
73 directly tailored to the demands imposed by geoscientific applications (Gubbi, Buyya,
74 Marusic, & Palaniswami, 2013; Palattella et al., 2013).

75

76 **Challenges**

³ <http://www.brnt.com>

77 A workshop was held in the summer of 2013 as part of the U.S. NSF's *EarthCube*
78 *Initiative*⁴, entitled "*Integrating Real-time Data into the EarthCube Framework.*" The
79 *EarthCube* program seeks to build a common framework for the analysis, aggregation,
80 and coupling of geoscientific data and models. The primary consensus of the workshop⁵,
81 as provided by over 75 participants spanning a broad set of geoscientific disciplines,
82 revealed that while *EarthCube* will provide an unprecedented framework for
83 disseminating historical data sources, the use of real-time data raises an additional set of
84 complex challenges, which must be addressed explicitly. Furthermore, it was agreed that
85 these challenges are not being addressed by existing real-time data tools.

86

87 Complexity of deployment is perhaps the biggest barrier to the adoption of real-time data.
88 A key aspect of managing in-situ and dynamic sensor data in real-time is providing
89 efficient discovery, access and processing of sensor observations. Ideally, scientists
90 should not have to be concerned with heterogeneous formats, sensors and sources of data.
91 Rather, easy-to-use systems must be developed to permit scientists to focus on analysis
92 and experimentation rather than complex system maintenance. To that end, a number of
93 core challenges should be addressed to facilitate the adoption of real-time data:

- 94 • Continued community discussion is required to build consensus around features
95 and the real-world uses of real-time data platforms.
- 96 • Installation and configuration of these systems should be seamless and as easy as
97 possible. This may be accomplished by cloud-hosted infrastructure that features
98 pre-configured instances of the platform, thus reducing the need for complex,
99 local user maintenance.
- 100 • Real-time data systems should provide standard interoperability interfaces to
101 sensor data to minimize the custom software required for management,
102 visualization and analysis of different types of sensor observations. These
103 platforms should also adhere in as much as possible to common data and metadata
104 formats that adhere to standards (such as the OGC's *Sensor Web*).

⁴ <http://earthcube.org>

⁵ <https://www.eol.ucar.edu/news-and-events/workshops/earthcube-realtime-data-workshop>

- 105 • Platforms should also provide a system to archive, navigate and distribute non
106 real-time data streams via the Internet.

107

108 **A reference implementation**

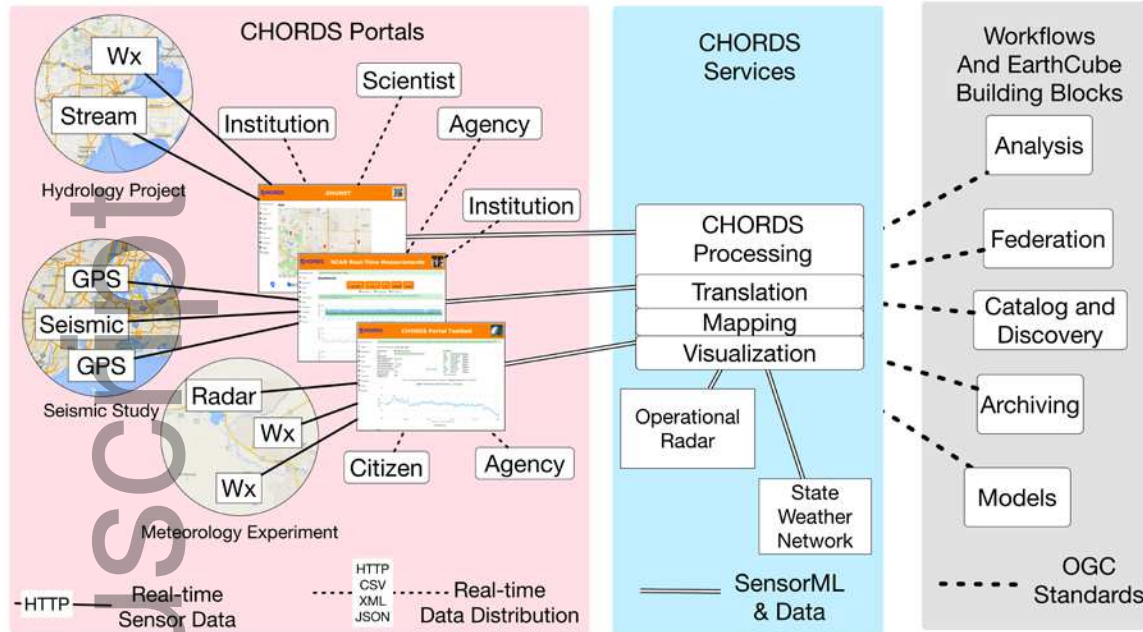
109 Presently, a working group is spearheading the use of real-time data within *EarthCube*
110 under the *Cloud-Hosted Real-time Data Services for the Geosciences (CHORDS)* project⁶.

111 While the primary goal of CHORDS is to drive a community discussion around the
112 adoption of real-time data, reference architecture is also being developed to serve as an
113 example for future implementations of real-time data systems. A number of use cases are
114 being evaluated within this platform to showcase the potential of real-time data toward
115 improving scientific experiments. Examples include, but are not limited to, the analysis
116 and visualization of measurements collected by scientific aircraft, real-time seismic
117 sensor networks for the detection of tornadoes, GPS-based volcano monitoring, and data
118 streaming services for a new generation of affordable 3D-printed weather stations.

119

120 One particular use case involves the coupling of real-time, distributed meteorological and
121 hydrologic data. The use case is intended to illustrate the study of extreme events, such as
122 flooding, where hydrologic models are forced by meteorological inputs. In such cases it
123 is vital to couple precipitation data with local flow conditions to forecast flooding. This
124 application couples complex raster data, time series, and metadata, which must be
125 reconciled within the same framework. The CHORDS reference architecture (Figure 1) is
126 explicitly developed with ease-of-use in mind, permitting even small research teams to
127 have a turnkey path toward using real-time data. Three main layers comprise the
128 architecture: 1) the CHORDS Portals, which are the entry and distribution points for all
129 real-time data, 2) CHORDS Services, which provide optional, value-added features, and
130 3) powerful standards to interface with workflows and EarthCube building blocks.

⁶ <http://chords.earthcube.org>



131

132 **Figure 1: CHORDS Architecture:** Sensors push real-time data to CHORDS Portals,
 133 which provide easy web-services access to the data streams. Portals can optionally
 134 interface CHORDS Services, which provide additional functionality and interoperability
 135 with popular standards and EarthCube services.

136

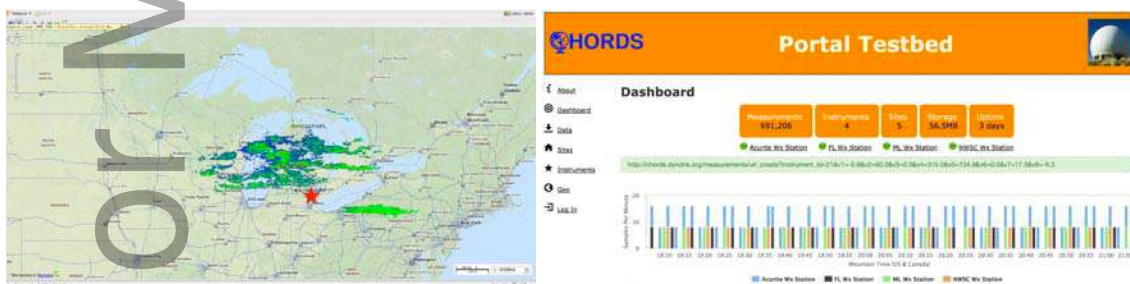
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138 A CHORDS portal can be launched as a pre-configured instance on a commercial cloud
 139 platform, allowing users to deploy it with minimal setup overhead. Each CHORDS user
 140 owns and manages their own CHORDS instance and interfaces it with their data streams.
 141 A preconfigured web server on the instance hosts a user interface, which is used to define
 142 data streams that will be ingested by the instance. This interface is used to generate
 143 simple URL schemes, which can be loaded directly into data sources (data loggers,
 144 instruments, algorithms, etc). A corresponding ingester is generated for each URL
 145 scheme, which translates the incoming sensor data into a common CHORDS format that
 146 is then hosted by the portal for external distribution. This permits users to keep their data
 147 sources relatively unaltered, having only to push a simple HTTP/REST post when new
 148 measurements are made. Data can be written to and read from the CHORDS portal via a
 149 set of standard encodings, such as JSON and XML. Data can even be pushed into
 150 CHORDS using simple CSV or binary formats to limit programming of field-deployed

151 devices. The support for other popular data formats is continually expanding, with plans
152 to incorporate formats such as netCDF in the near future.

153

154 While CHORDS portals provide a rapid way to ingest and share data from multiple real-
155 time sensor networks, their functionality can be vastly expanded by interfacing with the
156 CHORDS services layer. This layer is hosted by EarthCube's CHORDS team and
157 provides a central registry of all deployed portal instances. It serves as a repository of
158 metadata and expands the portal functionality with additional higher-level features such
159 as visualization, mapping, and basic resampling or filtering algorithms. It even provides
160 access to some popular real-time feeds, such as radar data or operational weather
161 networks. A GIS framework (*GeoServer*⁷) is built into the services layer to facilitate the
162 visualization, retrieval and discovery of data based on geographic regions of interest
163 (Figure 2). The services layer interoperates with the larger family of evolving web-based
164 OGC data services and standards, a feature that is continual maintained and updated by
165 the CHORDS team to support a growing set of external services and workflows, such as
166 those offered by EarthCube.



167

168 **Figure 2:** Example use case, where data from a hydrologic sensor node (red star) is
169 overlaid on radar data to predict local precipitation and forecast flooding conditions.

170

171 A major advantage of CHORDS will be that the end user can work in whatever
172 environment is most effective for them. No specific programming languages are forced
173 onto data producers or end users, as the only requirement is the ability to process
174 HTTP/RESTful requests. This permits the seamless integration of CHORDS services into

⁷ <http://geoserver.org/>

175 most existing instrumentation, models, and visualizations. Once configured, research
176 teams can then easily incorporate a suite of algorithms into their real-time workflows.
177 These workflows could include systems ranging from highly integrated command and
178 control systems, data assimilation into models, field project control centers, standalone
179 applications, web visualizations, or spreadsheets.

180

181 While the project is still in its infancy, initial use-case assessments are very favorable.
182 CHORDS does not aim to be a one-size-fits-all solution for real-time data, nor is its
183 present implementation an operational real-time data platform. A number of limitations
184 currently exist, which will be addressed in the future based on community feedback. All
185 of the current use cases are based on low latency requirements. The current
186 implementation does not support photo or video data, which may be relevant to studies
187 that require real-time image analysis. While existing system could readily support data
188 rates at 10-60Hz per feed, data rates at higher magnitudes, especially for spatial data,
189 would require further testing and improvements. Model integration has also not been
190 tested yet, but use cases are underway to investigate how to best couple CHORDS with
191 publically hosted modeling services. For example, work is underway to connect the real-
192 time hydrometeorological application with hydrologic models for flood forecasting. Bi-
193 directional communications are currently not supported, which means that CHORDS can
194 receive data from remotely-deployed instruments but not control them. More advanced
195 OGC SWE functionalities, such as Sensor Planning Services, are also planned for
196 implementation to enable remote tasking of a field sensors, which will enable adaptive
197 sampling of geoscientific phenomena (Andres et al., 2014). Given the infancy of the
198 project, there are many more features that will be required to make CHORDS a fully
199 hardened real-time data platform. This will require the need for built-in security and
200 encryption, which will be vital in protecting field-deployed scientific assets and servers.
201 CHORDS is also not a storage repository, data discovery or cataloging service, as those
202 features are expected to be addressed by existing domain-community repositories and
203 services. . Rather, its goal is to serve as a reference for community feedback, which will
204 ultimately lead to consensus on architectures for real-time data.

205

206 **Discussion and Conclusions**

207 Given resource constraints of existing experiments, real-time data has the potential to
208 play a pivotal role in the future discovery of geoscientific processes. This will be
209 achieved by responding to data as soon as they are collected to detect faulty
210 instrumentation and adaptively allocate in-situ measurement resources. Furthermore,
211 many geoscientific data streams have the potential to change how information is
212 consumed by non-scientific stakeholders (during disaster events, for example). Given the
213 complexity of existing platforms however, much work remains to be done on simplifying
214 the use of real-time data platforms, so that scientists may focus on experimentation,
215 rather than platform maintenance. Over the coming years, the CHORDS initiative will
216 seek to carve out a vision and reference implementation of real-time data. During this
217 process, community engagement will be the most critical mechanism toward making real-
218 time data in the geosciences a reality.

219

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223

224 **References**

225 Andres, V., Jirka, S., & Utech, M. (2014). *OGC Best Practice: OGC Sensor Observation*
226 *Service 2.0 Hydrology Profile (OGC 14-004r1)*. Wayland, MA, USA.

227 Davis, G., & Rew, R. (1994). The Unidata LDM: programs and protocols for flexible
228 processing of data products. *International Conference on Interactive Information*
229 *and Processing Systems*.

230 Gubbi, J., Buyya, R., Marusic, S., & Palaniswami, M. (2013). Internet of Things (IoT): A
231 vision, architectural elements, and future directions. *Future Generation Computer*
232 *Systems*, 29(7), 1645–1660. doi:10.1016/j.future.2013.01.010

233 Jirka, S., Bröring, A., Kjeld, P., Maidens, J., & Wytzisk, A. (2012). A Lightweight
234 Approach for the Sensor Observation Service to Share Environmental Data across
235 Europe. *Transactions in GIS*, 16(3), 293–312. doi:10.1111/j.1467-
236 9671.2012.01324.x

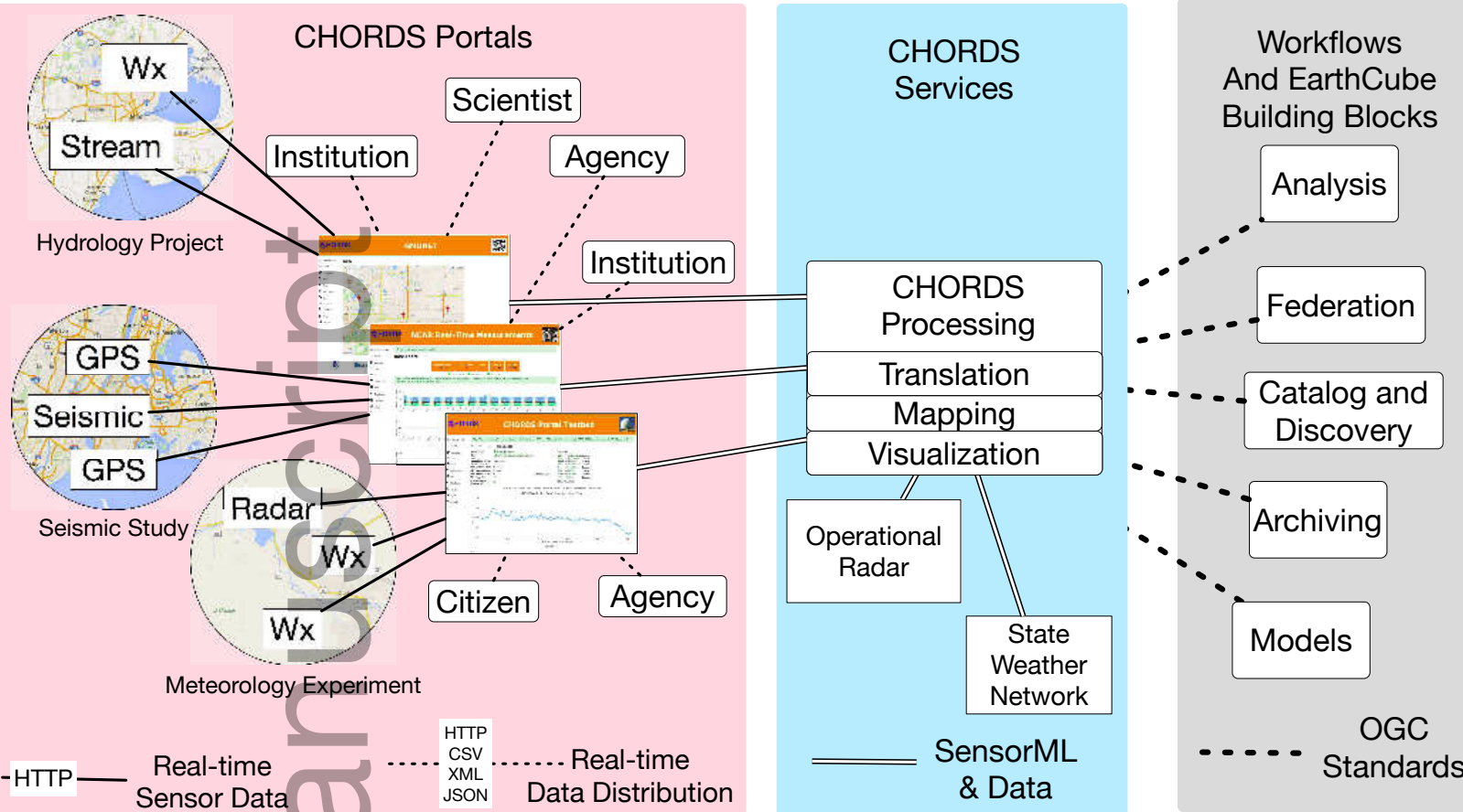
237 Nittel, S., Labrinidis, A., & Stefanidis, A. (Eds.). (2008). *GeoSensor Networks* (Vol.
238 4540). Berlin, Heidelberg: Springer Berlin Heidelberg. doi:10.1007/978-3-540-
239 79996-2

240 Palattella, M. R., Accettura, N., Vilajosana, X., Watteyne, T., Grieco, L. A., Boggia, G.,
241 & Dohler, M. (2013). Standardized Protocol Stack for the Internet of (Important)
242 Things. *IEEE Communications Surveys & Tutorials*, 15(3), 1389–1406.
243 doi:10.1109/SURV.2012.111412.00158

244 Reed, C., Botts, M., & Davidson, J. (2007). Ogc® sensor web enablement:overview and
245 high level achhitecture. In *2007 IEEE Autotestcon* (pp. 372–380). IEEE.
246 doi:10.1109/AUTEST.2007.4374243

247 Tilak, S., Hubbard, P., Miller, M., & Fountain, T. (2007). The Ring Buffer Network Bus
248 (RBNB) DataTurbine Streaming Data Middleware for Environmental Observing
249 Systems. In *Third IEEE International Conference on e-Science and Grid Computing*
250 (*e-Science 2007*) (pp. 125–133). IEEE. doi:10.1109/E-SCIENCE.2007.73

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CHORDS Portal Testbed



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