python*



The Frictionless Data Package: Data Containerization for Addressing Big Data Challenges

Adam Shepherd, Woods Hole Oceanographic Institution | Douglas Fils, Consortium for Ocean Leadership | Danie Kinkade, Woods Hole Oceanographic Institution | Mak Saito, Woods Hole Oceanographic Institution

Big Data Challenges

VERACITY

VARIETY

VOLUME





Abstract

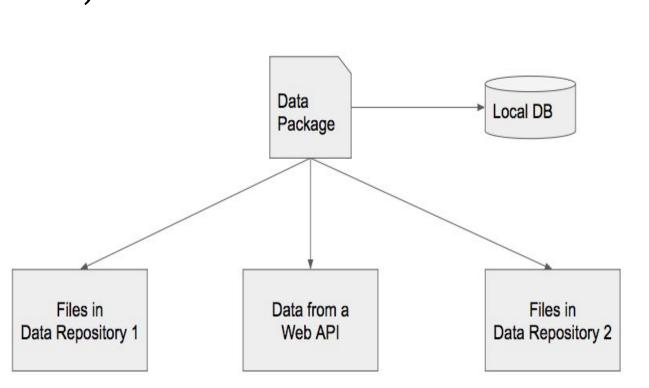
At the Biological and Chemical Oceanography Data Management Office (BCO-DMO) Big Data challenges have been steadily increasing. The sizes of data submissions have grown as instrumentation improves. Complex data types can sometimes be stored across different repositories. This signals a paradigm shift where data and information that is meant to be tightly-coupled and has traditionally been stored under the same roof is now distributed across repositories and data stores. For domain-specific repositories like BCO-DMO, a new mechanism for assembling data, metadata and supporting documentation is needed.

Traditionally, data repositories have relied on a human's involvement throughout discovery and access workflows. This human could assess fitness for purpose by reading loosely coupled, unstructured information from web pages and documentation. Distributed storage was something that could be communicated in text that a human could read and understand. However, as machines play larger roles in the process of discovery and access of data, distributed resources must be described and packaged in ways that fit into machine automated workflows of discovery and access for assessing fitness for purpose by the end-user. Once machines have recommended a data resource as relevant to an investigator's needs, the data should be easy to integrate into that investigator's toolkits for analysis and visualization.

BCO-DMO is exploring the idea of data containerization, or packaging data and related information for easier transport, interpretation, and use. Data containerization reduces not only the friction data repositories experience trying to describe complex data resources, but also for end-users trying to access data with their own toolkits. In researching the landscape of data containerization, the Frictionlessdata Data Package (https://frictionlessdata.io/) provides a number of valuable advantages over similar solutions. This presentation will focus on these advantages and how the Frictionless data Data Package addresses a number of real-world use cases faced for data discovery, access, analysis and visualization in the age of Big Data.

What is Data Containerization?

Problem: Data can be distributed across multiple locations databases, files on a server, files on the web, or available from web APIs, etc.



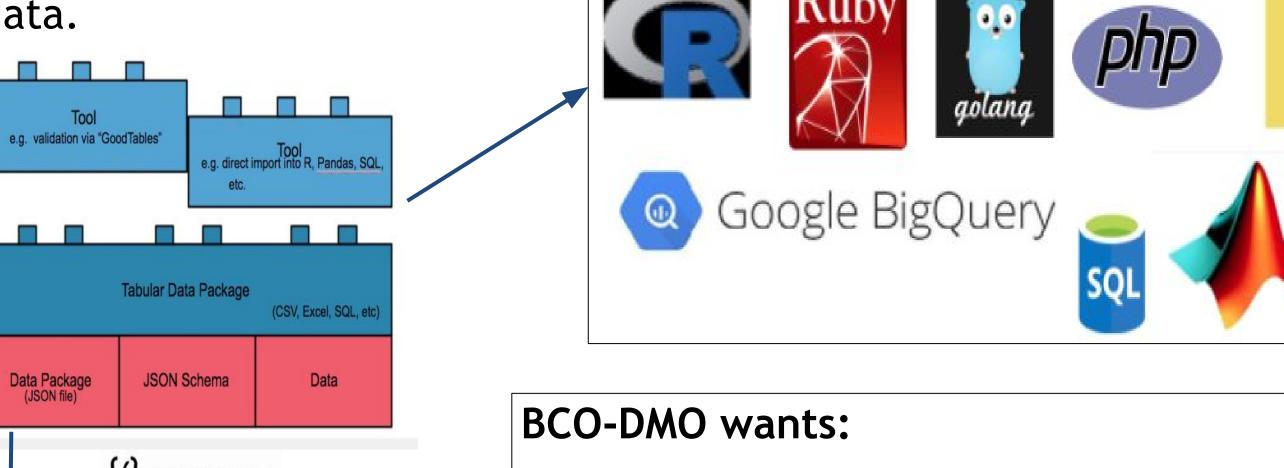
Problem: Data can be large so that traditional packages (TAR, ZIP, Baglt) are inefficient transports.

Q: How do we package data to handle these various locations and sizes?

The Frictionless Data Package is a set of extendible, lightweight formats for packaging data and metadata.

√ JSON Schema

e.g. validation via "GoodTables"



pandas

Simpler, expedited data ingest for submitters

 Data transformation that captures provenance Continuous integration testing of data holdings

datapackage.json can be added to traditional packaging formats (BagIt, TAR, ZIP) for _ describing local files

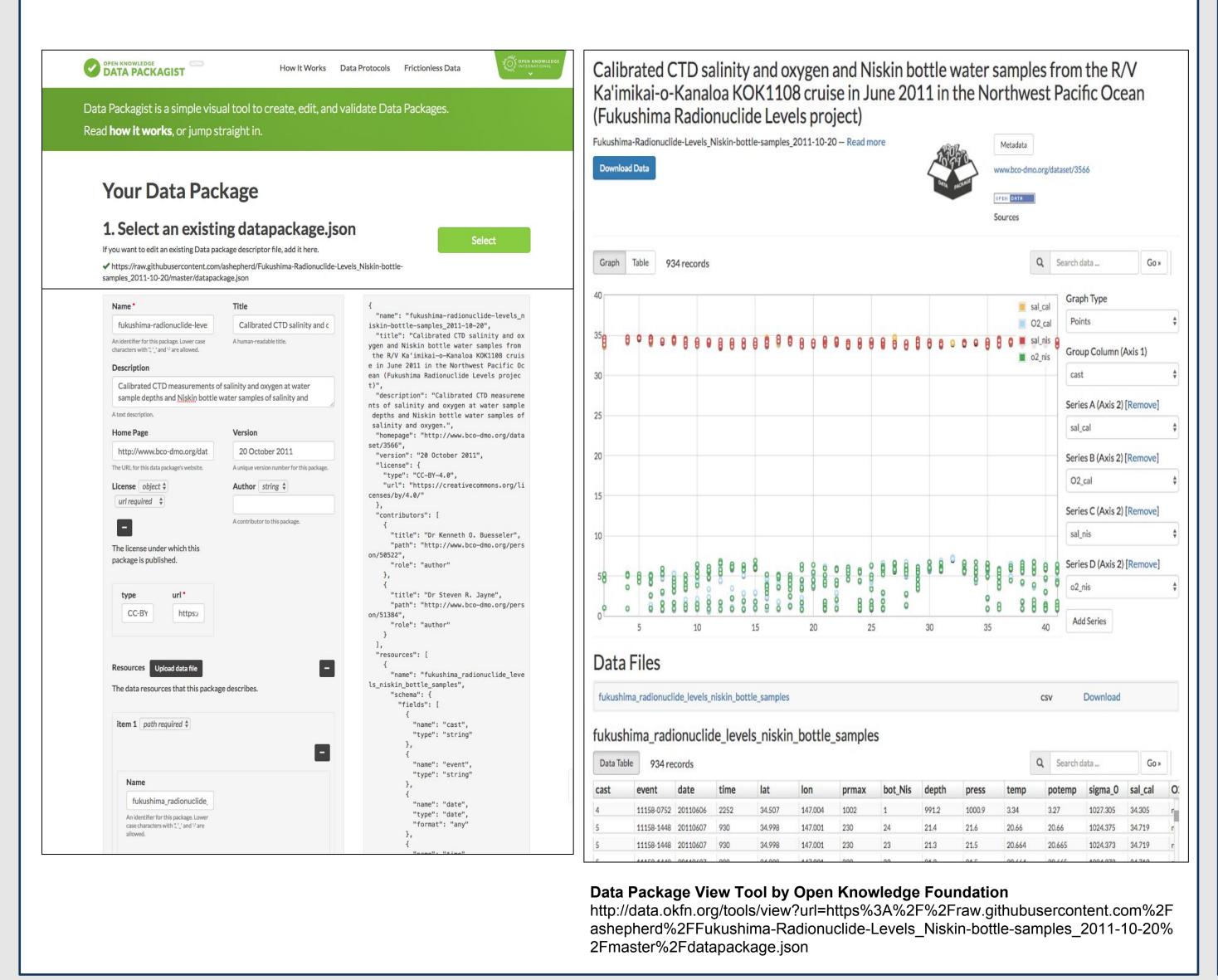
- can point to data resources that aren't local to the package or file system - for data accessible by URL, can be the only file needed to be passed in transport

Simpler, expedited data ingest for submitters

DataPackagist - A web service for creating Data Packages.

https://github.com/frictionlessdata/datapackagist

Data Submitters: login, describe submission, get immediate feedback



What about the work of a data manager?

niskin-bottle-samples.cs

datapackage.json

Data Package Pipelines - Framework for processing data packages in pipelines of modular components.

https://github.com/frictionlessdata/datapackage-pipelines

- A pipeline has a list of processing steps, and it generates a single data package as its output.
- A pipeline is defined in a declarative way, not in code, stored in a file named pipeline-spec.yaml.
- Data Package Pipelines define some common processors, custom processors

can be created.

itle: Transform an Excel File uploaded by a submitter to CS parameters: url: ./datapackage.json run: add_resource url: ./data-uploaded-by-submitter.xlsx sheet: 1 encoding: 'utf-8' run: stream remote resources cache: True run: dump.to_path out-path: project_123/dataset_456/v01 pipeline-spec.yaml project_123 dataset_456 ▼ data

Data Managers can extend the datapackage.json to add semantic markup.

"format": "csv", "dialect": { "delimiter": "ر

doubleQuote": false

The resulting datapackage.json can then be used to populate repository metadata catalog.

url: ./project_123/dataset_456/v01/datapackage.json created: '2017-07-10T12:48:35-04:00' '@graph': '@type': 'odo:Dataset' '@id': 'http://lod.bco-dmo.org/id/affiliation/191' '@type': 'xsd:boolean' '@value': true '@type': 'xsd:boolean '@type': 'odo:Person foaf:title: 'Dr' '@type': 'odo:Organization' '@language': 'en-US'

Continuous Integration Testing for Data

Goodtables.io - Continuous data validation as a service.

https://github.com/frictionlessdata/goodtables.io

Because Data Package Pipelines are declarative, pipeline-spec.yaml files are provenance records.

- Pipelines can be re-run to verify that the workflow is reproducible
- Data Packages can be validated
 - Is the datapackage.json well-formed?
 - Does it meet the JSON schema specification?
- Tabular Data Packages have deeper validation
- Column header checking vs. datapackage.json defined fields
- Methods for handling missing data values
- Regex processing of a data cell for conforming to a pattern

