An Asset Management Approach to Continuous Integration of

Heterogeneous Biomedical Data

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A Software as A Service ased Approach to Digital Asset Management for Complex Big-Data.

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BigData Landscape



- Increasingly need to combine multiple data in cross cutting analytic methods
 - E.G. in biomedical: genetics, multiple imaging modalities, proteomics, and clinical elements,
- Must integrate into a formal, standard, clean, consistent, accessible, and linked representation
- * "data wrangling" is often the most resource intensive activity in data analysis
 - 50% time overheads reported



Current Tools...

- Shared file systems with data organized in directory hierarchies and with metadata coded into "meaningful" file names
- Idiosyncratic methods used to capture and unify pertinent metadata such as phenotype, experiment details, preparation methods, and quality control flags.
- Manifests in spreadsheets, often the preferred means of describing and tracking data



E.G. A Physical Sciences Network Approach to Understanding Cancer

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H130715101142 -Study participant Sample number

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H130715101142 Study participant Sample number

H13071510-1142 Study participant Sample number







H13071512042 Human sample Year Month Day Study participant Sample number







H130715101142 Study participant Sample number









Data Management

Why don't we have tools for managing data sets of cancer and kidneys that are as good as the tools we have for managing data Flets bot cate and kids?

organization



Digital Asset Management

- *management tasks and decisions surrounding the ingestion, annotation, cataloguing, storage, retrieval and distribution of digital assets"
- streamline free-form "creative" processes rather than enforce predefined business processes.
- Many commercial DAM offerings, but not well suited to biomedical data
 - Complex and diverse data types
 - Specialized data ingest requirements
 - Data size (big data)



Biomedical Digital Asset Management

- Digital Asset Management approach applied to to biomedical data.
- "pay-as-you-go" approach provides continuous integration throughout the research and discovery lifecycle.
 - influenced by the concept of Dataspaces,
 - incremental refinement promotes flexible, use-case driven data integration,
- Focus on usability
 - Low barrier to entry
 - Mimic cloud based tools and services investigators are already familiar with



Design Requirements (1)

- Integrate early and often
 - Learn from data spaces and MAD
 - ETL and query mediation approach can require too much upfront schema structure
- BDAM implications
 - relaxed consistency: data evolves, rather than entering the system fully formed
 - incremental refinement: content and schema, throughout discovery
 - schema introspection: interfaces must be able to inspect the catalog's schema and present interfaces for the user to query and manipulate



Design Requirements (2)

- Ioose coupling
 - in which core components are operated in a software-as-a-service (SaaS) platform while user data may reside in local storage
- fine-grain access control
 - restrict access to metadata about specific assets attributes of the any asset, or collection of metadata
- multi-tenancy
 - to allow each scientific application to operate at its own pace and with its own content and access policies.



BDAM System Architecture





Data Catalog

- Records individual data assets along with meaningful metadata
- Can be browsed or searched to find assets matching certain criteria.
- Catalog schema can be queried and modified
- RESTful web services APIs
 - functions for retrieve and edit the metadata schema;
 - create, destroy, update and retrieve whole metadata records;
 - updating or retrieving metadata properties for specific records;
 - queries by metadata criteria and associations to other contextual records.
- Evaluated two catalog implementations
 - Triple based tagging model
 - Table based model
 - Both implemented as web services layer on Postgres



Tagfiler

- Compromise evolving understanding and benefits of strict schema
 - Subject, tag, value triples (like RDF)
 - type of a metadata record is determined by the properties it has (i.e. "duck-typed")
- Assets identified based on patterns constraining arbitrary sets of attributes.
 - tuned for arbitrary graph-query patterns
- Tuned for sparse data
 - Decompositional Storage Model (DSM) to store triples in property-specific tables and to generate complex joining queries when searching.



Use case example: imaging data from a slide scanner

Managing data from slide scanner:



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Managing data from slide scanner:





https://bdam.example.org/tagfiler/catalog/42/subject /slideref=@(/year=2014;experimentref=@(/id=123)) (id;year;fileurl;slideref)





<u>https://bdam.example.org/tagfiler/catalog/42/</u>subject /slideref=@(/year=2014;experimentref=@(/id=123)) (id;year;fileurl;slideref)

select catalog 42





https://bdam.example.org/tagfiler/catalog/42/subject /<u>slideref=@(/year=2014;experimentref=@(/id=123))</u> (id;year;fileurl;slideref)

- select catalog 42
- find subjects tagged with "slideref" referencing other subjects that have a year tag with value 2014 and "experimentref" referencing other subjects with an identifier equal to 123.





https://bdam.example.org/tagfiler/catalog/42/subject /slideref=@(/year=2014;experimentref=@(/id=123)) (id;year;fileurl;slideref)

- select catalog 42
- find subjects tagged with "slideref" referencing other subjects that have a year tag with value 2014 and "experimentref" referencing other subjects with an identifier equal to 123.
- return properties (id, year, fileurl, and slideref).



ERMrest

Table of typed entities with type-specific properties

- Scientists can think in tables
 - Capture entities and relations
- supports schema evolution and introspection by being schema agnostic.
 - Similar in phylosiphy to SQLShare (Howe)
- Compact URI naming scheme to traverse linked
 - A URI denoting one set of typed entities can be extended with the name of another linked entity type to denote a set of related entities of that other type.
 - Either URI may also be extended with filter expressions to denote a subset of entities of the same type.



ERMrest Query

For same model:

https://bdam.example.org/ermrest/catalog/42/enti ty/experiment/id=123/slide/year=2014/scan

can also follow references in either direction:

https://bdam.example.org/ermrest/catalog/42/entity/sc an/id=456/slide/experiment.



IOBox Ingest pipeline

- File scanning stage to generate a manifest
- Extraction of basic statistics
- Format specific extraction
 - HDF5, NetCDF, DICOM, NIfTI, Excel, Olympus SVI, Aperio SVS, and Hamamatsu NDPI, OME-XML and OME-TIFF, and SAM, VCF, and CSV files.
- User-defined rules to harvest additional metadata
- Entry into catalog





Data Movement

Pure cloud model

- Dropbox as a transfer service
- Hybrid cloud model
 - Globus online in cloud combined with local storage services





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Summary



- Continuous evolution and integration is necessary to address the realities of how data is used
- Data-wrangling and data management process with little tooling to support it
- Digital asset management techniques can be effectively applied to biomedical data

