### WHITE PAPER



New Database Engineering and Archive Construction Technology to Accelerate Bio-Imaging, Biomedical Engineering, and Covid-19 Research

LTS (Linguistic Technology Systems) is founded by Amy Neustein, Ph.D., Series Editor of Speech Technology and Text Mining in Medicine and Health Care (de Gruyter); Editor of Advances in Ubiquitous Computing: Cyber-Physical Systems, Smart Cities, and Ecological Monitoring (Elsevier, 2020); co-author (with Nathaniel Christen) of Cross-Disciplinary Data Integration and Conceptual Space Models for Covid-19 (Elsevier, 2021); and co-editor of Medical Image Processing and Machine Learning (Institution of Engineering and Technology, forthcoming).

### Team

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The "MOSAIC Data-Set Explorer" (MdsX) and "MOSAIC Structured Reporting" (MOSAIC-SR) are tools to help authors develop interactive presentations supplementing academic documents (MOSAIC is an acronym for "Multi-Paradigm Ontologies for Scientific and Technical Publications"). With MdsX, interactive presentations take the form of software applications that provide access to data sets, analytic techniques, or other digitally representable artifacts to document or encapsulate research work. With MOSAIC-SR, authors can implement or reuse code libraries that report on research/experiment methods, workflows, and protocols. Conceptually, MOSAIC-SR is functionally similar to the various domain-specific recommendations collectively gathered into the "Minimum Information for Biological and Biomedical Investigations" (MIBBI) specifications, and indeed one use-case for MOSAIC-SR is that of implementing object models instantiating MIBBI policies. In some contexts, MOSAIC-SR and MIBBI overlap, because elements of scientific workflows are sometimes algorithms implemented within a code package concretizing authors' research.

Mosaic-SR can express both computational workflows that are fully encapsulated by published code as well as real-world protocols concerning laboratory equipment and physical materials or samples under investigation. In the latter guise, MOSAIC-SR code can employ or instantiate standardized terminologies and data structures for describing experiments — such as MIBBI policies or BIOCODER functions. In this case, the role of MOSAIC-SR code is to serve as a serialization/deserialization endpoint for sharing research metadata. Conversely, when workflows are fully implemented within software developed as part of a body of research, MOSAIC-SR can provide a functional interface allowing this code to be embedded in scientific software. For these cases, MOSAIC-SR

provides a framework for modeling how a software component specific to a given research project exposes its functionality to host and/or networked peer applications. There are also scenarios where both scenarios are relevant — the MOSAIC-SR code would simultaneously document real-world experimental protocols and construct a digital interface as part of a workflow which is part digital and part "real-world."

This paper will focus on one specific application of MOSAIC-SR in the context of image analysis and bioimaging — specifically, a "Data Structure Protocol for Image-Analysis Networking" (D-SPIN), which both extends and adds a narrower focus to MOSAIC-SR. Software using the D-SPIN protocol provides a description of image processing capabilities which have been utilized and/or are functionally exposed by code and data associated with a research project. This includes "structured reporting" of research objectives as well as a concrete interface for invoking analyses associated with the research (either new algorithms or techniques used to obtain reported findings). D-SPIN, in turn, is based on CAPTK (the Cancer Imaging and Phenomics Toolkit) and PANDORE (an image-processing environment which includes both data models and interactive software). The PANDORE project encompasses an ontology of "Image Processing Objectives" that provides a structural basis for D-SPIN. For information about how different objectives are merged into workflows, D-SPIN adopts protocols from CAPTK, particularly with respect to implementating image-analysis capabilities as extensions to a core application, and CAPTK's implementation of the Common Workflow Language (CWL). In effect, D-SPIN formalizes the data models and prototypes adopted by PANDORE and CAPTK so as to concretize MOSAIC-SR for the specific domain of image processing and Computer Vision. The following sections will therefore outline D-SPIN features in the context of MOSAIC-SR design principles and objectives.

### Meta-Procedural Modeling in D-SPIN and Mosaic-SR

Most approaches to modeling research workflows involve some concept of "meta-objects", 1, "tools" (in the terminology of CWL), and "transitions" (in the language of Petri Net theory). In , the analogous concept is that of *meta-procedures*, which are analogous to ordinary computational procedures but add extra sources of information concerning input and output parameters. In general, rather than simply passing an imput value into an executable routine, metaprocedures define steps which can be taken to acquire the proper values when needed. Aside from ordinary runtime values, the most important input sources are methods defined on GUI components; command-line parameters; file contents; and not-yet-evaluated expressions (perhaps encapsulated in scripts or function pointers). A meta-procedure formulation abstracts the acquisition of inputs (or "channels") from the concrete procedure or procedures which are eventually executed. Therefore, a MOSAIC meta-procedure definition has two separate parts: a preamble where input sources are described, and an executive sequence where concrete procedures are indicated. A *meta-evaluator* then operates in accord with these definitions, concretizing the input values and launching the actual procedure(s). For D-SPIN, meta-procedures can be defined using a framework based on BIOCODER, but adopted to the imaging and Computer Vision context.

Image analysis methods are often described in academic literature in terms of methematical formulae and/or characterizations of computational environments (such as Graphical Processing Units); it requires additional construction to translate these overviews into actual computer code. Once Computer Vision innovations are in fact concretely implemented, there is then an additional stage of development requisite for users to actually enact the computations described in the research. Although it is theoretically possible to demonstrate novel methods within fully self-contained autonomous applications, it is more convenient for users if research code is integrated with existing imaging software. The D-SPIN interface can then help connect new code to existing applications, allowing users to access new code's functionaloty through GUI actions, command-line

<sup>&</sup>lt;sup>2</sup>See https://jbioleng.biomedcentral.com/articles/10.1186/1754-1611-4-13



<sup>&</sup>lt;sup>1</sup>See the **VISSION** system: https://pdfs.semanticscholar.org/1ad7/c459dc4f89f87719af1d7a6f30e6f58dff17.pdf.

invocations, or inter-application messaging protocols.

In addition to practically enabling application embedding, D-SPIN models represent research methods and theories, contributing to transparency and reusability according to the MIBBI and FAIR (Findable, Accessible, Interoperable, Reusable) standards.<sup>3</sup> This can be achieved, in part, by implementing data structures conforming to Image Processing Objectives. However, D-SPIN embeds this logic in an Object-Oriented context which allows imaging-specific workflow notations to be paired with specifications outside of imaging processing in the narrowest sense. This allows D-SPIN to be available for hybrid computational objectives representations which are only partially covered by the imaging domain — analogous to the MIAPE-GI (Gel Electrophoresis Informatics) component of MIAPE (Minimum Information About a Proteomics Experiment). The following section will discuss several domains where D-SPIN has been explicitly integrated with code libraries codifying MIBBI-style research protocols.

# D-SPIN in Contexts Supplemental to Image Processing. Image Flow Cytometry

One important use-case for biomedical image processing is to analyze cellular microscopy in conjuncton with cytmetric experiments which investigate cells and cellular-scale entities (such as proteins) indirectly. Conceptually, image analysis and flow cytometry (FCM) analysis are mathematically similar, and some commercial cytometry software has been extended with imageprocessing capabilities. The overlap between cytometric and image analysis has also inspired attempts to merge cytometry standards, such as MIFLOWCYT (the Minimum Information about a Flow Cytometry Experiment policy within MIBBI), with bioimaging standards such as DICOM (Digital Imaging and Communications in Medicine). One such proposal is due to Robert Leif, who argues that "The large overlap between imaging and flow cytometry provides strong evidence that both modalities should be covered by the same standard" and has formalized an XML language (CYTOMETRYML) to serve as that overarching bridge.<sup>4</sup> The D-SPIN project builds off this work by introducing its own FCM/DICOM hybrid, although in an object-oriented rather than XML-based context (discussed further in the next section). As a reference implementation for this D-SPIN extension, the project also provides a pure-C++ cytometry library based on the OPENCYTO and libraries, but eliminating external dependencies such as R and JAVA. The FCM/DICOM bridge is implemented in this context via a D-SPIN supplement to DICOM based on "Semantic DICOM," which is an effort to standardize query processing within PACS (Picture Archiving and Communications Service) workstations and to more effectively integrate **DICOM** with clinical data.

# A Semantic **DICOM** Object Model

As a formal representation of imaging workflows, D-SPIN would reasonably be paired in many contexts with DICOM, insofar as DICOM represents the canonical standard for exchanging medical image data. For its applications within the medical-imaging context, therefore, D-SPIN provides object-oriented accessors to DICOM data such that image-objectives and DICOM object models can interoperate. This object-oriented foundation also provides a basis on which to further integrate clinical data in the form of "Semantic" PACS models. In general, Semantic PACS applications draw clinical data from DICOM headers, though it would be consistent with the Semantic DICOM paradigm to expose this information via sufficiently well-structured supplemental information packages as well. This arrangement then forms the basis of a MOSAIC-Semantic DICOM (MOSAIC-SD) framework which integrates Semantic DICOM with D-SPIN — specifically, which defines a central DICOM object model affixed to both clinical and image-processing object models.

<sup>&</sup>lt;sup>5</sup>See https://www.ncbi.nlm.nih.gov/pmc/articles/PMC5119276/.



<sup>&</sup>lt;sup>3</sup>See https://www.researchgate.net/publication/331775411\_FAIRness\_in\_Biomedical\_Data\_Discovery.

<sup>&</sup>lt;sup>4</sup>See https://spie.org/Publications/Proceedings/Paper/10.1117/12.2295220?SSO=1.

The MOSAIC-SD object system applies not only to DICOM integrated with statements of imageprocessing objectives, but also to other biomedical contexts where image analysis should be integrated with other analytic modalities and also with clinical or epidemiological information. For example, Flow Cytometry overlaps with clinical data tracking because one of FCM's essential investigative roles is to examine patients' immunological response to diseases and/or interventions. In the context of Covid-19, say — with respect to achieving a deeper understanding of how and why SARS-CoV-2 symptoms present differently in different patients — "The starting point will likely be a deep characterization of the immune system in patients with different stage of the disease".6 That is, FCM observations need to be matched with clinical data in order to classify (and consider statistical correlations between) immunology and clinical facts (risk factors, sociodemographics, disease progression, and so forth). This analysis intrinsically assumes that FCM data can be transparently linked with all relevant clinical data, but such integration is difficult even in DICOM, where DICOM headers are specifically for preserving patient data across picture-sharing networks (there is no analogous "header" component in FCS, the Flow Cytometry Standard). In short, D-SPIN extensions to non-imaging domains can promote data integration insofar as D-SPIN Clinical Object Models, based on Semantic DICOM, provide an affixation point for clinical data in analytic contexts which are operationally related to image analysis, not just to image analysis per se.

### Geo-Imaging and Geographic Information Systems

Another area where **D-SPIN** provides structured object models is that of Geographic Information Systems (**GIS**), and specifically **GIS** annotations. There is a direct link between image processing and **GIS** insofar as identifying geotaggable features is one dimension or application of Computer Vision. Effectively manipulating geoimaging data requires mathematical translations between several different coordinate systems, in both two and three dimensions. These coordinate transforms — as well as semantic interpretations of geoimage segments (buildings, land features, roads, etc.) — can serve as the basis for an object model affixing image-processing objectives to **GIS** workflows.

In conventional **GIS** annotation, data structures are linked to both geospatial coordinates and to iconography, or other visual cues, allowing locations of interest to be indicated on maps. The actual geotagged data structures can be derived from any domain; as such any object model can be integrated with **GIS** annotations so long as one can assign spatial interpretations to the phenomena computationally encapsulated by the domain in question. In a medical context, for instance, geotagged data may represent the scope of a vaccination campaign or the extent of an epidemic, along with relevant geographic or civic features (villages, medical clinics, national borders, and so on). The actual map as a visible digital artefact therefore serves as a virtual glue where clinical, geographical, governmental, and **GUI** data are all sutured together. Insofar as geoimaging involved analysis of photographed land features and/or urban environments, image processing information represents a further object model that can be added to the mix. Even when dealing solely with virtual maps, however — rather than with satellite images or other geospatial photography — the analysis and application-level rendering of map features is sufficiently similar to image process that a rigorous model of **GIS** integration belongs properly within **D-SPIN**, specifically within a MOSAIC-GIS extension.

## Markup Serialization and "Grounding"

The data-integration mechanisms discussed in this paper thus far have focused on object models and object-oriented programming techniques. While composing special-purpose object-based libraries specifically tailored to individual data-integration problems is a powerful tool for solbing such problems, in practice data integration initiatives are often organized around standards for

<sup>&</sup>lt;sup>6</sup>See https://onlinelibrary.wiley.com/doi/full/10.1002/cyto.a.24002.



sharing or serializing conformation data structures. As a result, an important aspect of data integration is implementing proper data-serialization technology. In the biomedical and bioimaging context, most serialization operates through **XML**, so discussions of **XML** languages are a good foundation for examining serialization concerns in general.

To be sure, standardizing a data format by stipulating how XML files may encode the data is simpler than defining an analogous specification in terms of executable computer code. For sake of discussion, data structures defining a novel image-analysis method can serve as a case-study in standardization-through-serialization: one way to document the shape of any relevant data is to explain how an XML document will be structured insofar as it encodes data accordingly. Potentially, such specification can be a single XML DTD file, or an XML sample, providing a convenient reference point for developers to grasp the underlying data model.

However, the structure of XML document does not, in itself, present a clear picture of how the information which the document represents is semantically organized. Even though XML is processed by computer programs, it is not even evident from an XML document or schema which XML elements (if any) correspond to data types recognized by applications which read and/or write the corresponding XML code. For example, the XML portion of OME-TIFF (the principal Open Microscopy Environment imaging format) includes an explicit Image element (which gathers up all significant image metadata); an application reading OME-TIFF files might therefore introduce a single datatype — analogous to itk::Image from the Insight Toolkit imaging libraries — bundling the data in that part of the OME XML. In this case there will be a one-to-one correspondance between XML structure and application-level data types, at least for that one Image node. On the other hand, software reading OME-TIFF information may not manipulate images directly, but rather pull out other kinds of metadata, such as an experimenter's name or description of the microscopic setup. In this case, the application may not have an explicit "object" representing the image itself, but it may still read information about the image from child nodes of the XML Image element.

In short, applications can read or use data from an XML document in different ways, so the document's structure does not itself provide a clear picture of how code which reads the XML is organized. This uncertainty is significant insofar as one wishes to use XML specifications as an indirect strategy for documenting parameters and features of the data structures which are serialized via the relevant XML language. In effect, XML serialization operates on two levels: on the one hand, the specific XML document provides an encoding of data conforming to a given structure; but, at a more abstract level, one can model the relationship between the surface-level XML node structure and the application-level data structures thereby serialized. In MOSAIC, such "meta-serialization" — a term used to suggest the idea of providing meta-data *about* a serialization — is formulated through a notion of "grounding", which involves adding supplemental markup clarifying how documents instantiating a serialization format (such as XML) relate to the coding protocols and data types of software which reads or creates these documents.

For maximum expressivity, MOSAIC introduces a meta-serialization system that can be applied to language more flexible than XML — notably TAGML — as well as to XML proper. In short, MOSAIC provides parsers for an extended version of TAGML which includes an addition "grounding" layer. Grounding, in this context, means describing how elements in the markip — nodes, attributes, and character sequences — are "grounded" in application-level types, data fields, and other programming constructs. MOSAIC provides parsers for this "Grounded TAGML" (GTAGML) language as well as converters which can output serialized data as conventional XML, so that GTAGML specifications can be used in contexts (MIFLOWCYT, for instance) where ordinary XML is expected and serves as a basis of standardization. MOSAIC also provides code libraries for extracting data from GTAGML documents.

 $<sup>^{\</sup>textbf{7}} See \ http://www.balisage.net/Proceedings/vol21/print/HaentjensDekker01/BalisageVol21-HaentjensDekker01.html.}$ 



By design, GTAGML schemata operate on two levels: on the one hand, they constrain the organization of GTAGML files themselves; but they also stipulate how GTAGML document structure relates to the type systems of code libraries serializing and deserializing GTAGML files. To model this second, type-oriented metadata, GTAGML introduces a hypergraph-based type model organized around "infosets," which are structured overview of application-level data types. To fulle utilize GTAGML features, programmers would then compose "infoset classes" as wrappers around ordinary data types (e.g., C++ classes) whose instances are serialized via nodes or character strings within GTAGML documents. Infoset classes provide hypergraph "views" onto type-instances, and act as a bridge between data types and GTAGML schema. In particular, infoset classes (rather than the types they encapsulate) are the basis for schematizing the relation between GTAGML document elements and type instances and the data they contain.

The hypergraph-based modeling incorporated into **GTAGML** reuses much of the code associated with **CONCEPTSDB**, which is a new hypergraph database being developed alongside MOSAIC. More information about **CONCEPTSDB** can be found on the guthub project page for Linguistic Technology Systems (https://github.com/Mosaic-DigammaDB/LingTechSys).

