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## Review

## Towards structured sharing of raw and derived neuroimaging data across existing resources

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## ABSTRACT

Data sharing efforts increasingly contribute to the acceleration of scientific discovery. Neuroimaging data is accumulating in distributed domain-specific databases and there is currently no integrated access mechanism nor an accepted format for the critically important meta-data that is necessary for making use of the combined, available neuroimaging data. In this manuscript, we present work from the Derived Data Working Group, an open-access group sponsored by the Biomedical Informatics Research Network (BIRN) and the International Neuroimaging Coordinating Facility (INCF) focused on practical tools for distributed access to neuroimaging data. The working group develops models and tools facilitating the structured interchange of neuroimaging meta-data and is making progress towards a unified set of tools for such data and meta-data exchange. We report on the key components required for integrated access to raw and derived neuroimaging data as well as associated meta-data and provenance across neuroimaging resources. The components include (1) a structured terminology that provides semantic context to data, (2) a formal data model for neuroimaging with robust tracking of data provenance, (3) a web service-based application programming interface (API) that provides a consistent mechanism to access and query the data model, and (4) a provenance library that can be used for the extraction of provenance data by image analysts and imaging software developers. We believe that the framework and set of tools outlined in this manuscript have great potential for solving many of the issues the neuroimaging community faces when sharing raw and derived neuroimaging data across the various existing database systems for the purpose of accelerating scientific discovery.

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## Q1372 Introduction

73 Acceleration of scientific discovery in neuroimaging and many other  
74 research areas increasingly relies on the availability of large and well-  
75 documented data sets. In fact, many of the major new discoveries in  
76 the genetics of schizophrenia and other psychiatric disorders, multiple  
77 sclerosis, diabetes, obesity, and other metabolic traits have been possi-  
78 ble only through collaborative data sharing (Ripke et al., 2011; Sawcer  
79 et al., 2011; Speliotes et al., 2010). In the area of neuroimaging, such  
80 data sets can be obtained by a) funding large consortia to prospectively  
81 acquire large data sets (Insel et al., 2004), b) harvesting research-ready  
82 data from other sources (Kho et al., 2011; van Erp et al., 2011), and/or c)  
83 data (Biswal et al., 2010) or analysis results (Stein et al., 2012), sharing  
84 between multiple separately funded initiatives that include in-common  
85 measurements. In-common measurements, in the context of neuroim-  
86 aging, refer to imaging protocols that are included in many magnetic  
87 resonance imaging (MRI) related studies such as resting state functional  
88 magnetic resonance imaging (fMRI), structural T1-weighted MRI, and  
89 diffusion tensor imaging (Nooner et al., 2012). Shared and combined  
90 use of in-common measurements is the lowest barrier in the otherwise  
91 complex and often intractable space of combining neuroimaging data  
92 collected under different initiatives; however, acquiring equivalent  
93 data sets at sites with hardware from different vendors requires careful  
94 protocol design (Jack et al., 2008, 2010; Kruggel et al., 2010). Despite ef-  
95 forts from consortia such as the Function and Morphometry test beds of  
96 the Biomedical Informatics Research Network (BIRN) that have published  
97 recommendations for collecting neuroimaging data with the sharing and  
98 combining of data from multiple sites in mind, the task of data sharing  
99 across scanner platforms remains difficult even though the benefits are  
100 both financially and scientifically undeniable (Glover et al., 2012; Poline  
101 et al., 2012). Sharing well-documented, often publicly funded, data sets  
102 for use by the wider research community can be cost-effective as it allows  
103 for 1) increased statistical power through mega-analyses in contrast to  
104 meta-analyses, 2) obtaining new larger data sets to answer questions  
105 not addressed by the original studies, 3) application of newly developed  
106 tools to existing data sets, and 4) replication of research findings via  
107 reanalysis of existing data by other research groups.

108 In the last ten years, large neuroimaging data sets have become  
109 publicly available, although, there are significant differences in the  
110 requirements for data access. These data sets are in domain-specific  
111 repositories. Some examples of completely open-access neuroimaging  
112 repositories include XNAT Central (<https://central.xnat.org>) which in-  
113 cludes over 3000 subjects stored in the XNAT database (Marcus et al.,  
114 2007), the BIRN data repository ([www.birncommunity.org/resources/](http://www.birncommunity.org/resources/data)  
115 [data](http://www.birncommunity.org/resources/data)) which includes large cohorts of both mouse and human imag-  
116 ing data stored in the BIRN Human Imaging Database (Florescu et  
117 al., 1996; Ozyurt et al., 2010) and elsewhere, the 1000 Functional  
118 Connectomes project ([www.nitrc.org/projects/fcon\\_1000/](http://www.nitrc.org/projects/fcon_1000/)) which,  
119 at the present time, contains over 1000 subjects, and the relatively  
120 new OpenfMRI repository ([www.openfmri.org](http://www.openfmri.org)) which contains

imaging data from over 200 subjects. The Neuroimaging Informatics 121  
122 Tools and Resources Clearinghouse (NITRC, [www.nitrc.org](http://www.nitrc.org)) also hosts  
123 neuroimaging data, in addition to neuroimaging processing and analysis  
124 tools (Buccigrossi et al., 2008). Examples of neuroimaging repositories  
125 that require some form of permission to download data (e.g. prior IRB  
126 approval or simply an application to the host site), include the  
127 Alzheimer's Disease Neuroimaging Initiative (ADNI; <http://adni.loni.ucla.edu/>)  
128 which contains imaging data from over 800 subjects, 128  
129 and the National Database for Autism Research (NDAR; [ndar.nih.gov](http://ndar.nih.gov))  
130 which contains data from over 6000 subjects. It is clear from this short  
131 (and by no means exhaustive) list of available neuroimaging reposito-  
132 ries that data is accumulating in distributed domain-specific databases,  
133 rather than in a small number of central repositories. In addition, there  
134 is no integrated access mechanism, even for open-access resources, nor  
135 an accepted format for the critically important meta-data, necessary for  
136 making use of combined neuroimaging data. The Neuroscience Infor-  
137 mation Framework (NIF; [www.neuinfo.org](http://www.neuinfo.org)) (Gupta et al., 2008) pro-  
138 vides integrated access to many neuroscience-related databases as  
139 well as other resources; researchers can identify imaging data sets for  
140 download from certain resources that have been mapped for the NIF in-  
141 terface, for example. Developing the meta-data formats and standards  
142 needed to understand the imaging data sets, or to capture the details  
143 of how the data were collected and processed, is outside the scope of  
144 NIF and other database mediators. Integrated access to existing re-  
145 sources, many already identified by NIF, when combined with such  
146 meta-data documentation, would provide a full-service shop for queries  
147 and download of publicly available data across projects.

148 A critical barrier in enabling structured sharing of raw and derived  
149 neuroimaging data across existing resources is the lack of a standard  
150 meta-data model and a set of informatics tools that enables the sharing  
151 of meta-data, including provenance, associated with neuroimaging data  
152 (Teeters et al., 2008). Meta-data are descriptive elements associated  
153 with data that provide additional clarity regarding acquisition param-  
154 eters, experimental conditions, analysis procedures, and any other forma-  
155 tion about the experiment or analyses that helps one understand and use  
156 the data. The benefits of neuroimage data sharing were introduced more  
157 than a decade ago (Van Horn and Gazzaniga, 2002; Van Horn et al., 2001),  
158 several successful data sharing projects exist (Biswal et al., 2010; Nooner  
159 et al., 2012; Weiner et al., 2012), and many of the technical, legal, and  
160 social issues of data sharing have been discussed (Mennes et al., 2012;  
161 Milham, 2012; Poline et al., 2012), but there currently is no standard for-  
162 mat nor a set of lightweight tools that allow small laboratories or individ-  
163 ual investigators to share imaging and meta-data in a structured way, nor  
164 a set of tools that allows for queries across existing databases or data shar-  
165 ing efforts (Poline et al., 2012). These problems are especially acute when  
166 attempting to construct large data sets from data available through online  
167 repositories, each of which uses different structures of storing meta-data.

168 The options for making data available are limited to putting raw data  
169 sets online, or putting raw and processed data sets online together with  
170 descriptions of the derived data in text documents (or to be gleaned

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