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Interoperability and FAIRness through a novel combination of Web technologies

Mark D Wilkinson ^{Corresp.} ¹, **Ruben Verborgh** ², **Luiz Olavo Bonino da Silva Santos** ³, **Tim Clark** ⁴, **Morris A Swertz** ⁵, **Fleur D.L. Kelpin** ⁵, **Alasdair J. G. Gray** ⁶, **Erik A. Schultes** ⁷, **Erik M. van Mulligen** ⁸, **Paolo Ciccarese** ⁹, **Mark Thompson** ⁷, **Rajaram Kaliyaperumal** ⁷, **Jerven T. Bolleman** ¹⁰, **Michel Dumontier** ¹¹

¹ Center for Plant Biotechnology and Genomics - UPM/INIA, Universidad Politécnica de Madrid, Madrid, Spain

² Interuniversity Microelectronics Centre (IMEC), Ghent University, Ghent, Belgium

³ Dutch Techcenter for Life Sciences, Utrecht, The Netherlands

⁴ Department of Neurology, Massachusetts General Hospital, Boston, United States of America

⁵ Genomics Coordination Center and Department of Genetics, University Medical Center Groningen, Groningen, The Netherlands

⁶ Department of Computer Science, School of Mathematical and Computer Sciences, Heriot-Watt University, Edinburgh, United Kingdom

⁷ Department of Human Genetics, Leiden University Medical Center, Leiden, The Netherlands

⁸ Department of Medical Informatics, Erasmus University Medical Center, Rotterdam, The Netherlands

⁹ Elmer Innovation Lab, Harvard Medical School, Boston, United States of America

¹⁰ Swiss-Prot Group, SIB Swiss Institute of Bioinformatics, Centre Medical Universitaire, Geneva, Switzerland

¹¹ Stanford Center for Biomedical Informatics Research, Stanford University School of Medicine, Stanford, California, United States of America

Corresponding Author: Mark D Wilkinson
Email address: markw@illuminae.com

Data in the life sciences are extremely diverse and are stored in a broad spectrum of repositories ranging from those designed for particular data types (such as KEGG for pathway data or UniProt for protein data) to those that are general-purpose (such as FigShare, Zenodo, or EUDat). These data have widely different levels of sensitivity and security considerations. For example, clinical observations about genetic mutations in patients are highly sensitive, while observations of species diversity are generally not. The lack of uniformity in data models from one repository to another, and in the richness and availability of metadata descriptions, makes integration and analysis of these data a manual, time-consuming task with no scalability. Here we explore a set of resource-oriented Web design patterns for data discovery, accessibility, transformation, and integration that can be implemented by any general- or special-purpose repository as a means to assist users in finding and reusing their data holdings. We show that by using off-the-shelf technologies, interoperability can be achieved even to the level of an individual spreadsheet cell. We note that the behaviors of this architecture compare favorably to the desiderata defined by the FAIR Data Principles, and can therefore represent an exemplar implementation of those principles. The proposed interoperability design patterns may be used to improve discovery and integration of both new and legacy data, maximizing the utility of all scholarly outputs.

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3 Interoperability and FAIRness through a novel 4 combination of Web technologies

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6 Authors:

7

8 **Mark D. Wilkinson** - Center for Plant Biotechnology and Genomics, UPM-INIA, Madrid,
9 Spain

10 **Ruben Verborgh** – Ghent University – IMEC, Ghent, Belgium

11 **Luiz Olavo Bonino da Silva Santos** - Dutch Techcentre for Life Sciences, Utrecht, The
12 Netherlands - Vrije Universiteit Amsterdam, Amsterdam, The Netherlands

13 **Tim Clark** - Department of Neurology, Massachusetts General Hospital Boston MA and
14 Harvard Medical School, Boston, MA, USA

15 **Morris A. Swertz** - Genomics Coordination Center and Department of Genetics, University
16 Medical Center Groningen, Groningen, The Netherlands

17 **Fleur D.L. Kelpin** - Genomics Coordination Center and Department of Genetics, University
18 Medical Center Groningen, Groningen, The Netherlands

19 **Alasdair J. G. Gray** - Department of Computer Science, School of Mathematical and
20 Computer Sciences, Heriot-Watt University, Edinburgh, UK

21 **Erik A. Schultes** - Department of Human Genetics, Leiden University Medical Center,
22 Leiden, The Netherlands

23 **Erik M. van Mulligen** - Department of Medical Informatics, Erasmus University Medical
24 Center Rotterdam, The Netherlands

25 **Paolo Ciccarese** - Perkin Elmer Innovation Lab, Cambridge MA and Harvard Medical
26 School, Boston MA, USA

27 **Mark Thompson** - Leiden University Medical Center, Leiden, The Netherlands

28 **Rajaram Kaliyaperumal** - Leiden University Medical Center, Leiden, The Netherlands

29 **Jerven T. Bolleman** - Swiss-Prot group, SIB Swiss Institute of Bioinformatics, Centre
30 Medical Universitaire, Geneva, Switzerland

31 **Michel Dumontier** - Stanford Center for Biomedical Informatics Research, Stanford
32 University, Stanford, California

33

34 Corresponding Author:

35

36 Mark D. Wilkinson

37 *markw@illuminae.com*, +34 622 784 026

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44 Abstract

45 Data in the life sciences are extremely diverse and are stored in a broad spectrum of
46 repositories ranging from those designed for particular data types (such as KEGG for
47 pathway data or UniProt for protein data) to those that are general-purpose (such as
48 FigShare, Zenodo, or EUDat). These data have widely different levels of sensitivity and
49 security considerations. For example, clinical observations about genetic mutations in
50 patients are highly sensitive, while observations of species diversity are generally not. The
51 lack of uniformity in data models from one repository to another, and in the richness and
52 availability of metadata descriptions, makes integration and analysis of these data a manual,
53 time-consuming task with no scalability. Here we explore a set of resource-oriented Web
54 design patterns for data discovery, accessibility, transformation, and integration that can be
55 implemented by any general- or special-purpose repository as a means to assist users in
56 finding and reusing their data holdings. We show that by using off-the-shelf technologies,
57 interoperability can be achieved even to the level of an individual spreadsheet cell. We note
58 that the behaviors of this architecture compare favorably to the desiderata defined by the
59 FAIR Data Principles, and can therefore represent an exemplar implementation of those
60 principles. The proposed interoperability design patterns may be used to improve discovery
61 and integration of both new and legacy data, maximizing the utility of all scholarly outputs.

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65 Introduction

66 Carefully-generated data are the foundation for scientific conclusions, new hypotheses,
67 discourse, disagreement and resolution of these disagreements, all of which drive scientific
68 discovery. Data must therefore be considered, and treated, as first-order scientific output,
69 upon which there may be many downstream derivative works, among these, the familiar
70 research article (Starr et al., 2015). But as the volume and complexity of data continue to
71 grow, a data publication and distribution infrastructure is beginning to emerge that is not *ad*
72 *hoc*, but rather explicitly designed to support discovery, accessibility, (re)coding to
73 standards, integration, machine-guided interpretation, and re-use.

74

75 In this text, we use the word “data” to mean all digital research artefacts, whether they be
76 data (in the traditional sense), research-oriented digital objects such as workflows, or
77 combinations/packages of these (i.e. the concept of a “research object”, (Bechhofer et al.,
78 2013)). Effectively, all digital entities in the research data ecosystem will be considered data
79 by this manuscript. Further, we intend “data” to include both data and metadata, and
80 recognize that the distinction between the two is often user-dependent. Data, of all types,
81 are often published online, where the practice of open data publication is being encouraged
82 by the scholarly community, and increasingly adopted as a requirement of funding agencies
83 (Stein et al., 2015). Such publications utilize either a special-purpose repository (e.g. model-
84 organism or molecular data repositories) or increasingly commonly will utilize general-
85 purpose repositories such as FigShare, Zenodo, Dataverse, or even institutional
86 repositories. Special-purpose repositories generally receive dedicated funding to curate and
87 organize data, and have specific query interfaces and APIs to enable exploration of their
88 content. General-purpose repositories, on the other hand, allow publication of data in
89 arbitrary formats, with little or no curation and often very little structured metadata. Both of
90 these scenarios pose a problem with respect to interoperability. While APIs allow
91 mechanized access to the data holdings of a special-purpose repository, each repository has
92 its own API, thus requiring specialized software to be created for each cross-repository
93 query. Moreover, the ontological basis of the curated annotations are not always
94 transparent (neither to humans nor machines), which thwarts automated integration.
95 General purpose repositories are less likely to have rich APIs, thus often requiring manual
96 discovery and download; however, more importantly, the frequent lack of harmonization of
97 the file types/formats and coding systems in the repository, and lack of curation, results in
98 much of their content being unusable (Roche et al., 2015).

99

100 There are many stakeholders in this endeavour. Scientists themselves, acting as both
101 producers and consumers of these public and private data; public and private research-
102 oriented agencies; journals and professional data publishers both “general purpose” and
103 “special purpose”; research funders who have paid for the underlying research to be
104 conducted; data centres (e.g. the EBI (Cook et al., 2016), and the SIB (SIB Swiss Institute
105 of Bioinformatics Members, 2016)) who curate and host these data on behalf of the research
106 community; research infrastructures such as BBMRI-ERIC (van Ommen et al., 2015) and

107 ELIXIR (Crosswell & Thornton, 2012), and diverse others. All of these stakeholders have
108 distinct needs with respect to the behaviors of the scholarly data infrastructure. Scientists, for
109 example, need to access research datasets in order to initiate integrative analyses, while
110 funding agencies and review panels may be more interested in the metadata associated with
111 a data deposition - for example, the number of views or downloads, and the selected license.
112 Due to the diversity of stakeholders; the size, nature/format, and distribution of data assets;
113 the need to support freedom-of-choice of all stakeholders; respect for privacy;
114 acknowledgment of data ownership; and recognition of the limited resources available to
115 both data producers and data hosts, we see this endeavour as one of the *Grand Challenges*
116 *of eScience*.

117

118 In January 2014, representatives of a range of stakeholders came together at the request of
119 the Netherlands eScience Center and the Dutch Techcentre for Life Sciences (DTL) at the
120 Lorentz Center in Leiden, the Netherlands, to brainstorm and debate about how to further
121 enhance infrastructures to support a data ecosystem for eScience. From these discussions
122 emerged the notion that the definition and widespread support of a minimal set of
123 community-agreed guiding principles and practices could enable data providers and
124 consumers - machines and humans alike - to more easily find, access, interoperate, and
125 sensibly re-use the vast quantities of information being generated by contemporary data-
126 intensive science. These principles and practices should enable a broad range of integrative
127 and exploratory behaviours, and support a wide range of technology choices and
128 implementations, just as the Internet Protocol (IP) provides a minimal layer that enables the
129 creation of a vast array of data provision, consumption, and visualisation tools on the
130 Internet. The main outcome of the workshop was the definition of the so-called FAIR guiding
131 principles aimed at publishing data in a format that is **Findable**, **Accessible**, **Interoperable**
132 and **Reusable** by both machines and human users. The FAIR Principles underwent a period
133 of public discussion and elaboration, and were recently published (Wilkinson et al., 2016).
134 Briefly, the principles state:

135

136

137 **Findable** - data should be identified using globally unique, resolvable, and persistent
138 identifiers, and should include machine-actionable contextual information that can be
139 indexed to support human and machine discovery of that data.

140

141 **Accessible** - identified data should be accessible, optimally by both humans and
142 machines, using a clearly-defined protocol and, if necessary, with clearly-defined
143 rules for authorization/authentication.

144

145 **Interoperable** - data becomes interoperable when it is machine-actionable, using
146 shared vocabularies and/or ontologies, inside of a syntactically and semantically
147 machine-accessible format.

148

149 **Reusable** - Reusable data will first be compliant with the F, A, and I principles, but
150 further, will be sufficiently well-described with, for example, contextual information, so

151 it can be accurately linked or integrated, like-with-like, with other data sources.
152 Moreover, there should be sufficiently rich provenance information so reused data
153 can be properly cited.

154
155

156 Here we describe a novel interoperability architecture that combines three pre-existing Web
157 technologies and standards to enhance the discovery, integration, and reuse of data in
158 repositories that lack or have incompatible APIs, and/or in formats that normally would not
159 be considered interoperable such as Excel spreadsheets and flat-files. We examine the
160 extent to which the features of this architecture comply with the FAIR Principles, and suggest
161 that this might be considered a “reference implementation” for the FAIR Principles as applied
162 to non-interoperable data formats in any general or special purpose repository.

163

164 **Methods**

165 **Implementation**

166 **Overview of technical decisions and their justification**

167

168 The World Wide Web Consortium’s (W3C) Resource Description Framework (RDF) offers
169 the ability to describe entities, their attributes, and their relationships with explicit semantics
170 in a standardized manner compatible with widely used Web application formats such as
171 JSON and XML. The Linked Data Principles (Berners-Lee, 2006) mandate that data items
172 and schema elements are identified by HTTP-resolvable URIs, so the HTTP protocol can be
173 used to obtain the data. Within an RDF description, using shared public ontology terms for
174 metadata annotations supports search and large scale integration. Given all of these
175 features, we opted to use RDF as the basis of this interoperability infrastructure, as it was
176 designed to share data on the Web.

177

178 Beyond this, there was a general feeling that any implementation that required a novel data
179 discovery/sharing “Platform”, “Bus”, or API, was beyond the minimal design that we had
180 committed to; it would require the invention of a technology that all participants in the data
181 ecosystem would then be required to implement, and this was considered a non-starter.
182 However, there needed to be some form of coalescence around the mechanism for finding
183 and retrieving data. Our initial target-community - that is, the biomedical sciences - have
184 embraced lightweight HTTP interfaces. We propose to continue this direction with an
185 implementation based on REST (Fielding & Taylor, 2002), as several of the FAIR principles
186 map convincingly onto the objectives of the REST architectural style for distributed
187 hypermedia systems, such as having resolvable identifiers for all entities, and a common
188 machine-accessible approach to discovering and retrieving different representations of those
189 entities. The implementation we describe here is largely based on the HTTP GET method,
190 and utilizes rich metadata and hypermedia controls expressed as triples. We use widely-
191 accepted vocabularies not only to describe the data in an interoperable way, but also to

192 describe its nature (e.g. the context of the experiment and how the data was processed) and
193 how to access it. These choices help maximize uptake by our initial target-community,
194 maximize interoperability between resources, and simplify construction of the wide (not pre-
195 defined) range of client behaviors we intend to support.

196

197 Confidential and privacy-sensitive data was also an important consideration, and it was
198 recognized early on that it must be possible, within our implementation, to identify and richly
199 describe data and/or datasets without necessarily allowing direct access to them, or by
200 allowing access through existing regulatory frameworks or security infrastructures. For
201 example, many resources within the International Rare Disease Research Consortium
202 participate in the RD Connect platform (Thompson et al., 2014) which has defined the
203 “disease card” - a metadata object that gives overall information about the individual disease
204 registries, as well as a “disease matrix”. The disease matrix provides aggregate data about
205 what disease variants are in the registry, how many individuals represent each disease, and
206 other high-level descriptive data that allows, for example, researchers to determine if they
207 should approach the registry to request full data access.

208

209 Finally, it was important that the data host/provider is not *necessarily* a participant in making
210 their data interoperable - rather, the interoperability solution should be capable of adapting
211 existing data with or without the source provider’s participation. This ensures that the
212 interoperability objectives can be pursued for projects with limited resourcing, but more
213 importantly, that those with the needs and the resources, should adopt the responsibility for
214 making their data-of-interest interoperable, even if it is not owned by them. This distributes
215 the problem of migrating data to interoperable formats over the maximum number of
216 stakeholders, and ensures that the most crucial resources - those with the most demand for
217 interoperability - become the earliest targets for migration.

218

219 With these considerations in mind, we were inspired by three existing technologies whose
220 features were used in a novel combination to create an interoperability infrastructure for both
221 data and metadata, that is intended to also address the full range of FAIR requirements.
222 Briefly, the selected technologies are:

223

- 224 1) The W3C’s Linked Data Platform (Speicher, Arwe & Malhotra, 2015). We generated
225 a model for hierarchical dataset containers that is inspired by the concept of a LDP
226 Container, and the LDP’s use of the Data Catalogue Vocabulary (DCAT, (Maali,
227 Erickson & Archer, 2014)) for describing datasets, data elements, and distributions of
228 those data elements. We also adopt the DCAT’s use of Simple Knowledge
229 Organization System (SKOS, (Miles & Bechhofer, 18 August, 2009)) Concept
230 Schemes as a way to ontologically describe the content of a dataset or data record.
- 231 2) The RDF Modelling Language (RML, (Dimou et al.)). RML allows us to describe one
232 or more possible RDF representations for any given dataset, and do so in a manner
233 that is, itself, FAIR: every sub-component of an RML model is Findable, Accessible,
234 Interoperable, and Reusable. Moreover, for many common semi-structured data,
235 there are generic tools that utilize RML models to dynamically drive the

236 transformation of data from these opaque representations into interoperable
237 representations (<https://github.com/RMLio/RML-Mapper>).
238 3) Triple Pattern Fragments (TPF - (Verborgh et al., 2016)). A TPF interface is a REST
239 Web API to retrieve RDF data from data sources in any native format. A TPF server
240 accepts URLs that represent triple patterns, and returns RDF triples from its data
241 source that match those patterns. These patterns can be used to obtain entire
242 datasets, slices through datasets, or individual data points even down to a single
243 triple (essentially a single cell in a spreadsheet table). Instead of relying on a
244 standardized contract between servers and clients, a TPF interface is self-describing
245 such that automated clients can discover the interface and its data.
246

247 We will now describe in detail how we have applied key features of these technologies, in
248 combination, to provide a novel data discoverability architecture. We will later demonstrate
249 that this combination of technologies also enables both metadata and data-level
250 interoperability even between opaque objects such as flat-files, allowing the data within
251 these objects to be queried in parallel with other data on the Semantic Web.
252

253 Metadata Interoperability - The “FAIR Accessor” and the Linked Data Platform

254
255 The Linked Data Platform “*defines a set of rules for HTTP operations on Web resources... to*
256 *provide an architecture for read-write Linked Data on the Web*”. All entities and concepts are
257 identified by URLs, with machine-readable metadata describing the function or purpose of
258 each URL and the nature of the resource that will be returned when that URL is resolved.
259

260 Within the LDP specification is the concept of an LDP Container. A basic implementation of
261 LDP containers involves two “kinds” of resources. The first type of resource represents the
262 container - a metadata document that describes the shared features of a collection of
263 resources, and (optionally) the membership of that collection. This is analogous to, for
264 example, a metadata document describing a data repository, where the repository itself has
265 features (ownership, curation policy, etc.) that are independent from the individual data
266 records within that repository (i.e. the members of the collection). The second type of
267 resource describes a member of the contained collection and (optionally) provide ways to
268 access the record itself.
269

270 Our implementation utilizes this container concept described by the LDP, however, it does
271 not require a full implementation of LDP, as we only need read functionality, while LDP
272 defines a read/write interface. In addition, other requirements of LDP would have added
273 complexity without notable benefit. Our implementation, which we refer to as the “FAIR
274 Accessor”, has two resource types, with the following features:
275

276 **Container resource:** This is a composite research object (of any kind - repository,
277 repository-record, database, dataset, data-slice, workflow, etc.). Its representation could
278 include scope or knowledge-domain covered, authorship/ownership of the object, latest
279 update, version number, curation policy, and so forth. This metadata may or may not include

280 URLs representing MetaRecord resources (described below) that comprise the individual
281 elements within the composite object. Notably, the Container URL provides a resolvable
282 identifier independent from the identifier of the dataset being described; in fact, the dataset
283 may not have an identifier, as would be the case, for example, where the container
284 represents a dynamically-generated data-slice. In addition, Containers may be published by
285 anyone - that is, the publisher of a Container may be independent from the publisher of the
286 research object it is describing. This enables one of the objectives of our interoperability
287 layer implementation - that anyone can publish metadata about any research object, thus
288 making those objects more FAIR.

289

290 **MetaRecord resource:** This is a specific element within a collection (data point, record,
291 study, service, etc.). Its representation should include information regarding licensing and
292 accessibility, access protocols, rich citation information, and other descriptive metadata. It
293 also includes a reference to the container(s) of which it is a member (the Container URL).
294 Finally, the MetaRecord may include further URLs that provide direct access to the data
295 itself, with an explicit reference to the associated data format by its MIME type (e.g.
296 text/html, application/json, application/vnd.ms-excel, text/csv, etc.). As with Container
297 resources, MetaRecords may be published by anyone, and independently of the original
298 data publisher.

299

300 In summary, the FAIR Accessor shares commonalities with the Linked Data Platform, but
301 additionally recommends the inclusion of rich contextual metadata, based on the FAIR
302 Principles, that facilitate discovery and interoperability of repository and record-level
303 information. The FAIR Accessor is read-only, utilizing only HTTP GET together with widely-
304 used semantic frameworks to guide both human and machine exploration. Importantly, the
305 lack of a novel API means that the information is accessible to generic Web-crawling agents,
306 and may also be processed if that agent “understands” the vocabularies used.

307

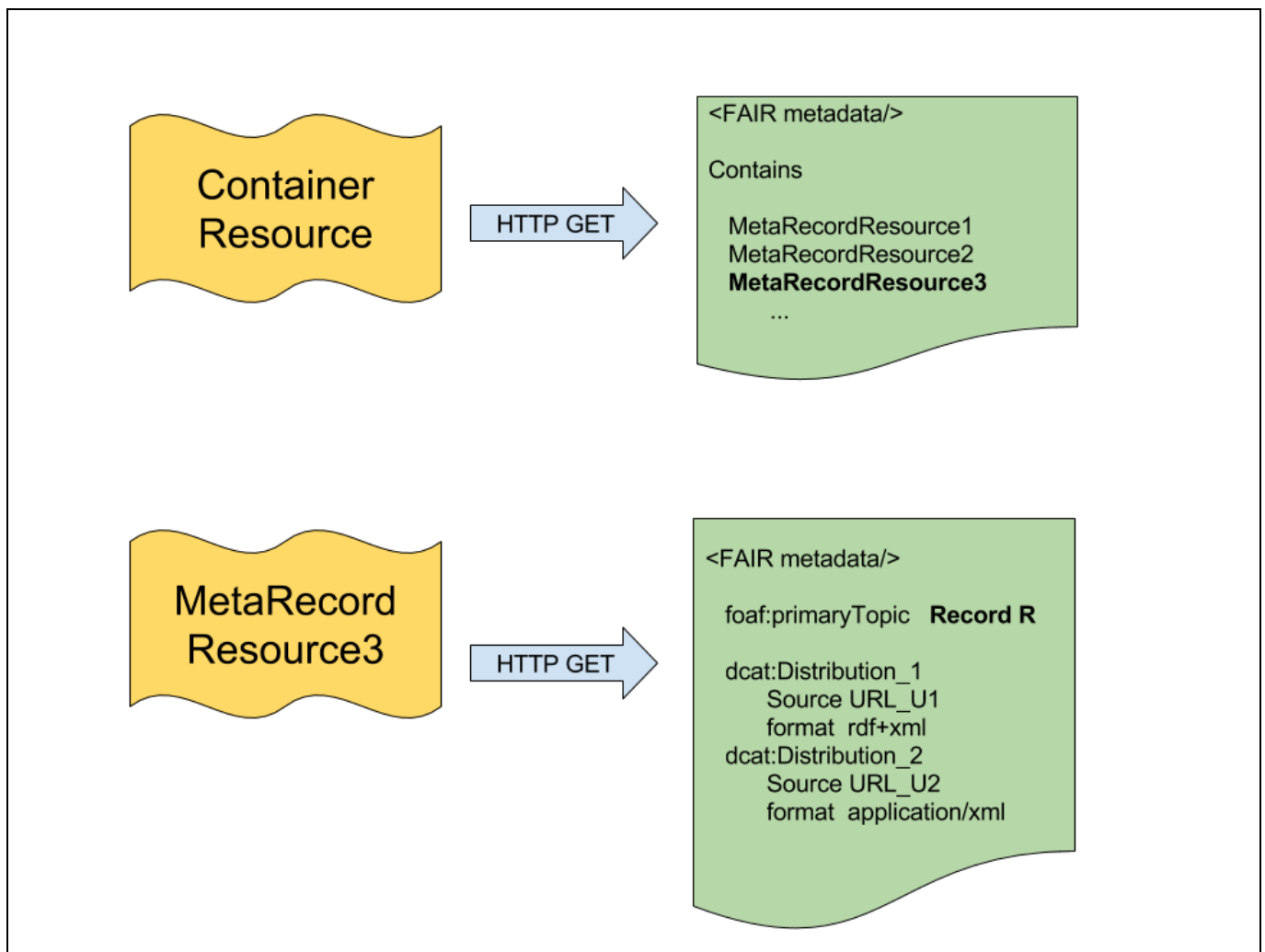


Figure 1 The two layers of the FAIR Accessor. Inspired by the LDP Container, there are two resources in the FAIR Accessor. The first resource is a Container, which provides metadata, following FAIR Principles, about a composite research object, and optionally a list of URLs representing MetaRecords that describe individual components within the collection. The MetaRecord resources resolve to documents containing metadata about an individual data component and, optionally, a set of links structured as DCAT Distributions that lead to various representations of that data.

308

309

310 At the metadata level, therefore, this portion of the interoperability architecture provides a
 311 high degree of FAIRness; however, it does not significantly enhance the FAIRness and
 312 interoperability of the data itself, which was a key goal for this project. We will now describe
 313 the application of two recently-published Web technologies - Triple Pattern Fragments and
 314 RML - to the problem of data-level interoperability. We will show that these two technologies
 315 are capable of transforming non-FAIR data into FAIR data, and will demonstrate how they
 316 can be integrated into the FAIR Accessor to provide a machine-traversable path for
 317 incremental drill-down from high-level repository metadata all the way through to individual
 318 data points within a record.

319

320 **Data Interoperability: Compatible data discovery through RML-based FAIR**
321 **Profiles**

322
323 In our approach to data-level interoperability, we first identified a number of desiderata that
324 the solution should exhibit:

- 325
- 326 ● View-harmonization over dissimilar datatypes, allowing discovery of *potentially*
327 integrable data within non-integrable formats.
 - 328 ● Support for a multitude of source data formats (XML, Excel, CSV, JSON, etc.)
 - 329 ● “Cell-level” discovery and interoperability (referring to a “cell” in a spreadsheet)
 - 330 ● Modular, such that a user can make interoperable only the data component they
331 require
 - 332 ● Reusable, avoiding “one-solution-per-record” and minimizing effort/waste
 - 333 ● Must use standard technologies, and reuse existing vocabularies
 - 334 ● Should not require the participation of the data host (for public data)
- 335

336 The approach we selected was based on the premise that data, in any format, could be
337 metamodeled as a first step towards interoperability; i.e., the salient data-types and
338 relationships within an opaque data “blob” could be described in a machine-readable
339 manner. The metamodels of two data sources could then be compared to determine if their
340 contained data was, in principle, integrable.

341
342 We referred to these metamodels as “FAIR Profiles”, and we further noted that there could
343 be multiple FAIR Profiles for any given data, where the Profiles might differ in structure, or
344 ontological/semantic framework. For example, a data record containing blood pressure
345 information might describe this data facet using the SNOMED vocabulary in one Profile, and
346 the ICD10 vocabulary in another Profile. We acknowledge that these meta-modelling
347 concepts are not novel, and have been suggested by a variety of other projects such as
348 DCAT (called a “DCAT Profile”, though never implemented) and Dublin Core (the DC
349 Application Profile (<http://dublincore.org/documents/profile-guidelines/>), and have been
350 extensively described by the ISO 11179 standard (“metadata registries”: [http://metadata-](http://metadata-standards.org/11179/)
351 [standards.org/11179/](http://metadata-standards.org/11179/)).

352
353 Our investigation into relevant existing technologies and implementations revealed a
354 relatively new, unofficial specification for a generic mapping language called “RDF Mapping
355 Language” (RML (Dimou et al.)). RML is an extension of R2RML (Das, Sundara & Cyganiak,
356 27 September, 2012), a W3C Recommendation for mapping relational databases to RDF,
357 and is described as “*a uniform mapping formalization for data in different format, which*
358 *[enables] reuse and exchange between tools and applied data*” (Dimou et al.). An RML map
359 describes the triple structure of an RDF representation, the semantic types, and the
360 constituent URI structures, that would result from a transformation of non-RDF data into RDF
361 data. RML maps are modular RDF documents where each component is a template,
362 identified by a URI, that describes the schema for a single-resource-centric graph (i.e. a
363 graph with all triples that share the same subject). The “object” position in each of these

364 triple templates may be mapped to a literal, or may be mapped as the value defined by
365 another RML module. These modules therefore assemble into a complete map of an RDF
366 representation of a data source. Finally, RML maps can also be used as templates to guide
367 the data transformation itself, using file-format-specific (but content-agnostic) software such
368 as RML Mapper (<http://github.com/RMLio/RMLMapper>). RML therefore fulfils each of the
369 desiderata for FAIR Profiles, and we have selected this technology as the candidate for their
370 implementation.

371
372 FAIR Profiles enable view harmonization and facilitate search/discovery of compatible but
373 structurally non-integrable data, possibly in distinct repositories. The Profiles of one data
374 resource can be compared to the Profiles of another data resource to identify commonalities
375 at the semantic level (even if the underlying data is semantically opaque) - a key step toward
376 Interoperability. FAIR Profiles created *ab initio* to fully describe a data resource, therefore,
377 have utility independent of any *actuated* transformation of the underlying data. We believe,
378 however, that it is unlikely that repository owners, or third parties, will undertake the effort of
379 creating FAIR Profiles for this purpose. We believe there is an alternative, needs-directed,
380 community-oriented approach to creating FAIR Profiles that distributes the burden of
381 designing these profiles over a broader number of researchers, and in particular, transfers
382 most of the the burden onto those who need the resources (many) rather than those who
383 own the resources (few).

384
385 Data transformation is a near-daily task for bioinformaticians, however once complete, this
386 effort is largely wasted. It would be more efficient, economical, and collaborative, to capture
387 and reuse the expert knowledge behind those transformations in a FAIR manner. Such
388 knowledge capture must not require any coordinated effort - individual researchers transform
389 data in different ways, at different times, depending on their needs - and preferably should
390 integrate into the researcher's existing work-habits. At present, these transformations - often
391 accomplished by small one-off scripts - are not published, cannot be discovered, and cannot
392 be described.

393
394 We propose that we could use the concept of a FAIR Profile to capture the cognitive effort
395 that is invested in these numerous small data transformations, where the Profile explicitly
396 expresses each individual researcher's perspective of the implicit meaning of the data. To
397 accomplish this, we propose that individuals who create transformation scripts, not only
398 publish those scripts in a publicly-accessible location, but additionally publish small RML
399 models describing the output of that transformation. (additional incentives for doing so will be
400 described in the discussion section). Specifically, we propose the following approach:

- 401
402
- 403 • Data transformers independently publish one or more RML maps, where each map is
404 constrained to describe a single triple pattern that their transformation generates from
405 the underlying data source. We call these single-triple RML maps "Triple
406 Descriptors", and the structure of a Triple Descriptor is shown in Figure 2.
 - 407 • We further propose (but do not demonstrate here) that these Triple Descriptors may
later be aggregated to generate a FAIR Profile containing all triple patterns

408 associated with a given data source, from all providers. This would give us the view-
 409 harmonization we desire for search and discovery, without the centralized effort. It
 410 would, in fact, provide a more comprehensive view-harmonization because it would
 411 also likely be redundant, containing different interpretations/representations of the
 412 same data element based on the perspectives of different researchers.

413

414 RML is fully tolerant to both redundancy, and distribution of its model subcomponents. Any
 415 given data point may be mapped to any number of RML models, and RML models utilize
 416 URIs to identify every model component, allowing individual components to be located
 417 anywhere on the Web.

418

419

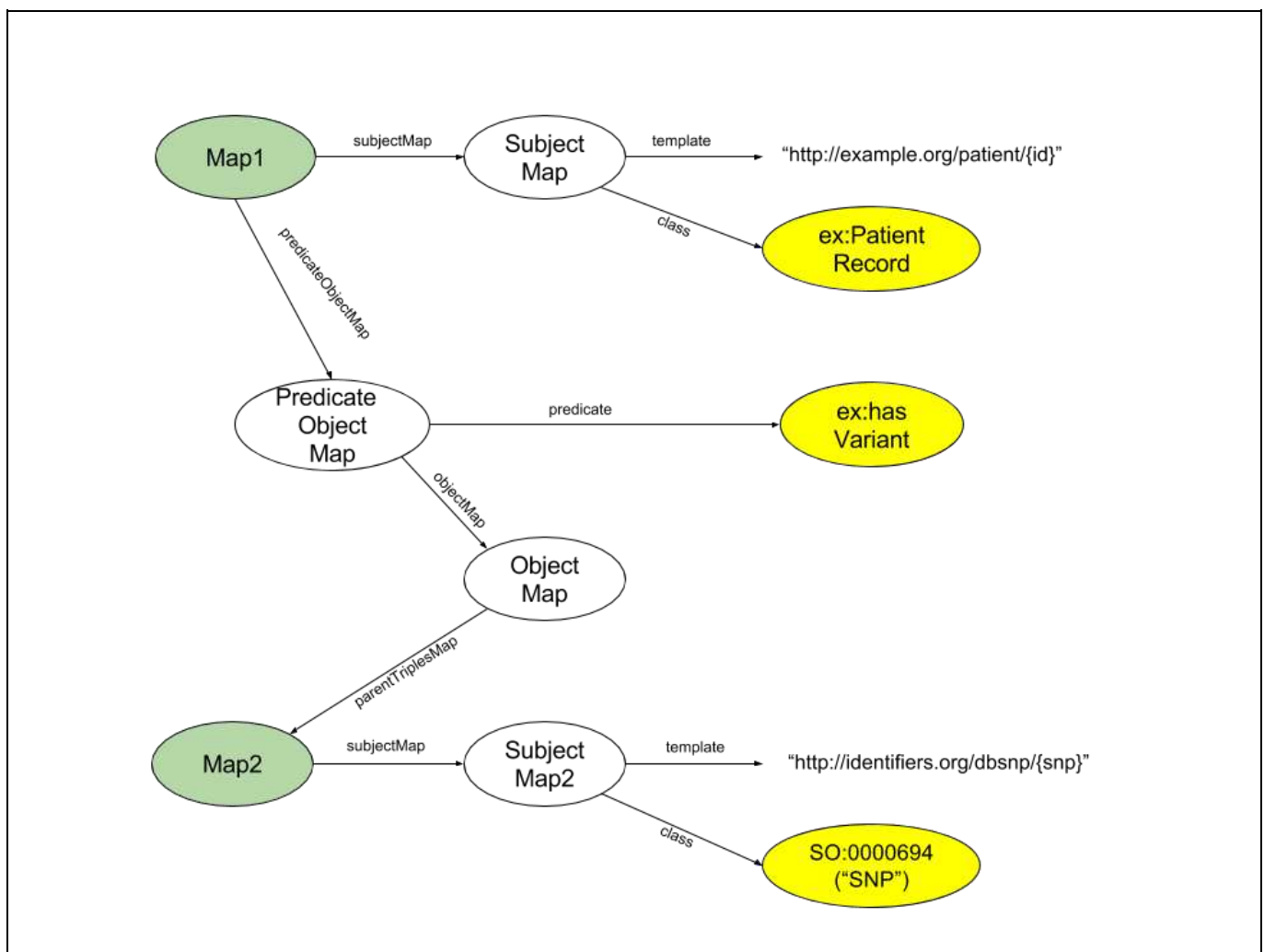


Figure 2: Diagram of the structure of an exemplar Triple Descriptor representing a hypothetical record of a SNP in a patient's genome. In this descriptor, the Subject will have the URL structure *http://example.org/patient/{id}*, and the subject is of type PatientRecord. The predicate is hasVariant, and the object will have URL structure *http://identifiers.org/dbsnp/{snp}* with the rdf:type from the sequence ontology "0000694" (which is the concept of a "SNP"). The two nodes shaded green are of

the same ontological type, showing the iterative nature of RML, and how individual RML Triple Descriptors will be concatenated into full FAIR Profiles. The three nodes shaded yellow are the nodes that define the subject type, predicate and object type of the triple being described.

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FAIR Profiles, therefore, are RML models - authored *ab initio*, and/or aggregated from Triple Descriptors - that describe one or more whole or partial RDF representations for a given data source. Triple Descriptors, and sometimes entire FAIR Profiles, may be re-used to describe other sources if each source shares the aspects described within the model. In this way, it is possible to identify, with considerable precision (i.e. potentially at the level of a spreadsheet column or individual cell) potentially integrable data from two distinct sources, based on the two sources sharing one or more Triple Descriptors in their FAIR Profile.

431 **Data Interoperability: Data transformation with FAIR Projectors and Triple** 432 **Pattern Fragments**

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The ability to identify potentially integrable data within opaque file formats is, itself, a notable achievement compared to the *status quo*. Nevertheless, beyond just discovery of relevant data, our interoperability layer aims to support and facilitate cross-resource data integration and query answering. This requires that the data is not only semantically described, but is also semantically and syntactically transformed into a common structure.

440 Above, we presented a mechanism to describe structure and semantics - Triple Descriptors
441 in RML - what remains lacking, however, is a way to execute data transformations that
442 provide output consistent with a given Triple Descriptor. Although in the previous section we
443 proposed that those who undertake data transformations should publish their transformation
444 script, together with its associated Triple Descriptors, this does not address a critical barrier
445 to interoperability - opaque, non-machine-readable interfaces and API proliferation
446 (Verborgh & Dumontier, 2016). We propose, therefore, that what is required is a universally-
447 applicable way of retrieving data from any transformation script (or any data source), without
448 inventing a new API. We now describe our suggestion for how to achieve this behavior, and
449 we refer to such transformation tools as “FAIR Projectors”.

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Triple Pattern Fragments (TPF) defines a REST interface through which clients can request triples based on a triple pattern [S,P,O] where any component of that pattern is either a constant or a variable. In response, a TPF server returns pages with all triples from its data source that match the incoming pattern. We use the TPF interface for FAIR Projectors, and therefore all Projectors share a common URL pattern, defined by the Triple Pattern Fragments specification (Verborgh et al., 2016). In addition, we require that the semantics of the output triple patterns are defined by (one or more) Triple Descriptors, thus allowing a client to select the appropriate FAIR Projector for its needs.

460 A FAIR Projector, therefore, is a Web resource that is associated with *both* a particular data
461 source, and particular Triple Descriptor(s). Calling HTTP GET on the URL of the FAIR
462 Projector produces RDF triples from the data source that match the format defined by that
463 Projector's Triple Descriptor. The originating data source behind a Projector may be a
464 database, a data transformation script, an analytical web service, another FAIR Projector, or
465 any other data-source.
466

467 **Linking the Components: FAIR Projectors and the FAIR Accessor**

468
469 At this point, we have a means for obtaining triples with a specific structure - TPF Servers -
470 and we have a means of describing the structure and semantics of those triples - Triple
471 Descriptors. Together these two elements define a FAIR Projector. However, we still lack a
472 formal mechanism for linking these two components, such that the discovery of a Triple
473 Descriptor with the desired semantics, also provides its associated TPF Server (Projector)
474 URL.
475

476 We propose that this association can be easily accomplished, without defining any novel API
477 or standard, if the output of a FAIR Projector is considered a type of DCAT Distribution. In
478 this way, it may be included as another distribution component of the MetaRecord metadata
479 from a FAIR Accessor, where the URL of the Projector, and its Triple Descriptor, are
480 metadata elements of that Distribution. This is diagrammed in Figure 3, where Distribution_3
481 and Distribution_4 include Triple Pattern Fragment URLs served by a FAIR Projector, and
482 also include the Triple Descriptor RML model that describes the structure and semantics of
483 the data that will be produced by calling that URL.
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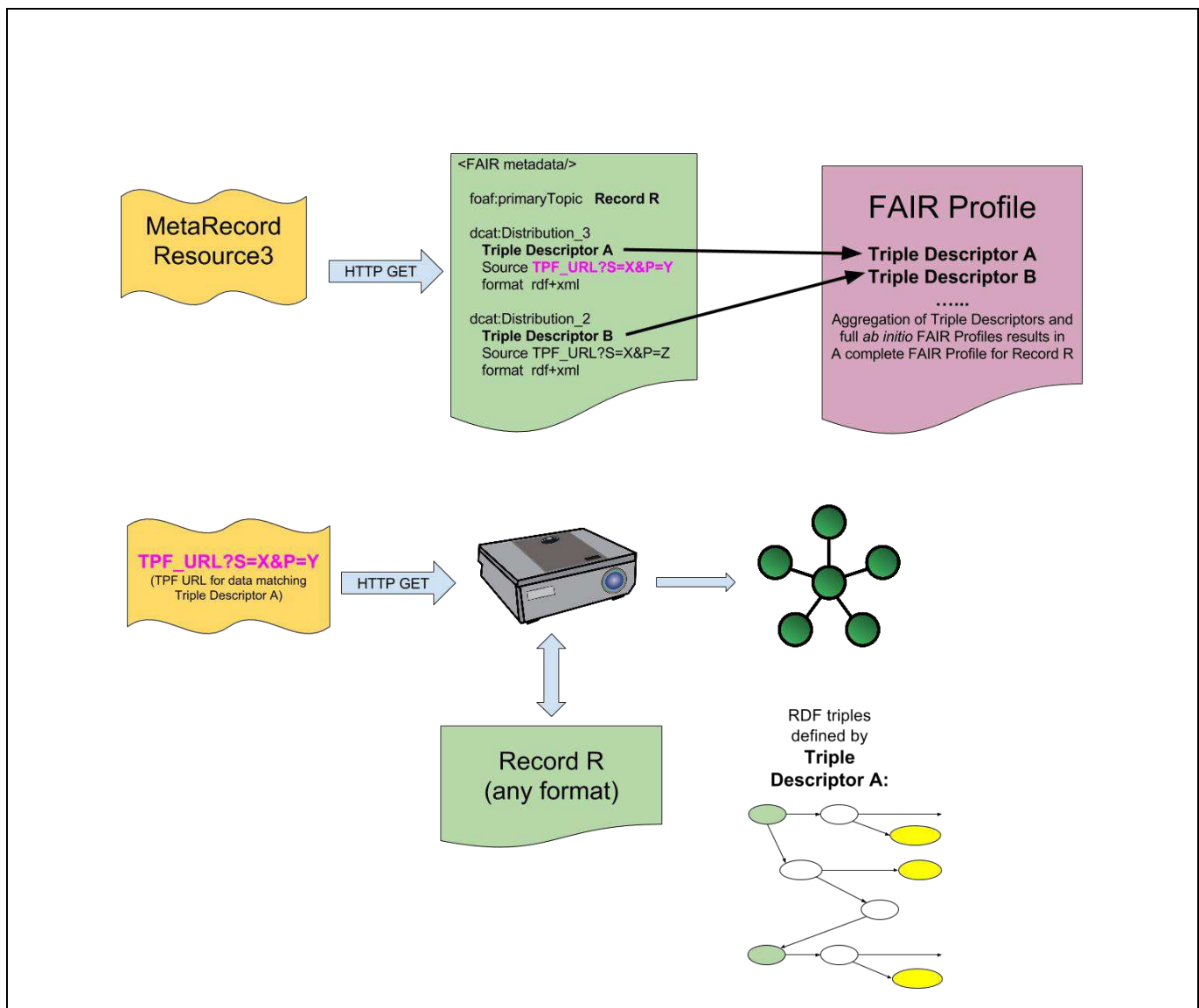


Figure 3. Integration of FAIR Projectors into the FAIR Accessor. Resolving the MetaRecord resource returns a metadata document containing multiple DCAT Distributions for a given record, as in Figure 1. When a FAIR Projector is available, additional DCAT Distributions are included in this metadata document. These Distributions contain a URL (purple text) representing a Projector, and a Triple Descriptor that describes, in RML, the structure and semantics of the Triple(s) that will be obtained from that Projector resource if it is resolved. These Triple Descriptors may be aggregated into FAIR Profiles, based on the Record that they are associated with (Record R, in the figure) to give a full mapping of all available representations of the data present in Record R.

488

489

490 Results

491

492 To demonstrate the interoperability layer, we will explore an example involving UniProt. In
 493 this example, we create a FAIR Accessor for a dataset that consists of a specific “slice” of

494 the Protein records within the UniProt database - that is, the set of proteins in *Aspergillus*
495 *nidulans* (taxon 16245) that are annotated as being involved in RNA Metabolism (GO
496 0006396). We first demonstrate the functionality of the two layers of the FAIR Accessor. We
497 then demonstrate a FAIR Projector, and show how its metadata integrates into the FAIR
498 Accessor. In this example, the Projector modifies the ontological framework of the UniProt
499 data such that the ontological terms used by UniProt are replaced by the terms specified in
500 EDAM. We will demonstrate that this transformation is specified, in a machine-readable way,
501 by the FAIR Triple Descriptor that accompanies each Projector's metadata.
502

503 The two-step FAIR Accessor

504
505 The example FAIR Accessor serves the results of the following query against the UniProt
506 SPARQL endpoint:

```
507 PREFIX up:<http://purl.uniprot.org/core/>  
508 PREFIX taxon:<http://purl.uniprot.org/taxonomy/>  
509 PREFIX rdf:<http://www.w3.org/1999/02/22-rdf-syntax-ns#>  
510 PREFIX rdfs:<http://www.w3.org/2000/01/rdf-schema#>  
511 SELECT distinct ?id  
512  
513 WHERE  
514 {  
515     ?protein a up:Protein .  
516     ?protein up:organism ?organism .  
517     ?organism rdfs:subClassOf taxon:162425 .  
518     ?protein up:classifiedWith ?go .  
519     ?go rdfs:subClassOf* <http://purl.obolibrary.org/obo/GO_0006396> .  
520  
521     bind(replace(str(?protein), "http://purl.uniprot.org/uniprot/", "",  
522 "i") as ?id)  
523 }  
524  
525
```

526 Accessor output is retrieved from the Container Resource URL:

```
527  
528 http://linkeddata.systems/Accessors/UniProtAccessor  
529
```

530 The result of calling GET on the Container Resource URL is visualized in Figure 4, where
531 Tabulator (Tim Berners-lee et al., 2006) is used to render the output as HTML for human-
532 readability.
533
534

UniProt Slice FAIR Accessor - Aspergillus RNA Processing proteins	creator	wilkinsonlab.info/
	language	eng
	license	cc by nd4.0
	title	UniProt Slice FAIR Accessor - Aspergillus RNA Processing proteins
	authored By	0000 0002 9699 485X
	entities	412
	term has Principal Investigator	Dr. Mark Wilkinson
	type	Dataset Basic Container Collection
	contact Point description	Wilkinson.rdf Takes a SPARQL query of the UniProt database specific to proteins and their GO annotations related to RNA Processing proteins in Aspergillus and makes it a FAIR Accessor source. The precise query is: <pre>PREFIX up:<http://purl.uniprot.org/core/> PREFIX taxon:<http://purl.uniprot.org/taxonomy/> PREFIX rdf:<http://www.w3.org/1999/02/22-rdf-syntax-ns#> PREFIX rdfs:<http://www.w3.org/2000/01/rdf-schema#> SELECT distinct ?id WHERE { ?protein a up:Protein . ?protein up:organism ?organism . ?organism rdfs:subClassOf taxon:162425 . ?protein up:classifiedWith ?go . ?go rdfs:subClassOf* <http://purl.obolibrary.org/obo/GO_0006396> . bind(replace(str(?protein), "http://purl.uniprot.org/uniprot/", "", "i") as ?id) }</pre>
	identifier keyword	Uni Prot Accessor Aspergillus nidulans Aspergillus Proteins RNA Processing
	landing Page	uniprot.org/
	language	en
	publisher	wilkinsonlab.info/
	theme	RNA Processing conceptscheme.rdf
	contains	C8UZX9 C8UZY5 C8V0B4 C8V0M2 C8V0I17

Figure 4. A representative portion of the output from resolving the Container Resource of the FAIR Accessor, rendered into HTML by the Tabulator Firefox plugin. The three columns show the label of the Subject node of all RDF Triples (left), the label of the URI in the predicate position of each Triple (middle), and the value of the Object position (right), where blue text indicates that the value is a Resource, and black text indicates that the value is a literal.

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536
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539

Of particular note are the following metadata elements:

<http://purl.org/dc/elements/1.1/license>

<https://creativecommons.org/licenses/by-nd/4.0/>

http://purl.org/pav/authoredBy	http://orcid.org/0000-0002-9699-485X
http://rdfs.org/ns/void#entities	411
a	http://purl.org/dc/dcmitype/Dataset http://www.w3.org/ns/ldp#BasicContainer http://www.w3.org/ns/prov#Collection
http://www.w3.org/ns/dcat#contactPoint	http://biordf.org/DataFairPort/MiscRDF/Wilkinson.rdf
http://www.w3.org/ns/dcat#keyword	"Aspergillus nidulans", "Aspergillus", "Proteins", "RNA Processing";
http://www.w3.org/ns/dcat#theme	http://linkeddata.systems/ConceptSchemes/RNA_Processing_conceptscheme.rdf
http://www.w3.org/ns/ldp#contains	http://linkeddata.systems/cgi-bin/Accessors/UniProtAccessor/C8UZ9 http://linkeddata.systems/cgi-bin/Accessors/UniProtAccessor/C8UZ5 ...

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- License information is provided as an HTML + RDFa document, following one of the primary standard license forms published by Creative Commons. This allows the license to be unambiguously interpreted by both machines and people prior to accessing any data elements, an important feature that will be discussed later.
- Authorship is provided by name, using the Academic Research Project Funding Ontology (ARPFO), but is also unambiguously provided by a link to the author's ORCID, using the Provenance Authoring and Versioning (PAV) ontology.
- The repository descriptor is typed as being a Dublin Core Dataset, a Linked Data Platform container, and a Provenance Collection, allowing it to be interpreted by a variety of client agents, and conforming to several best-practices, such as the Healthcare and Life Science Dataset Description guidelines (Dumontier et al., 2016))
- Contact information is provided in a machine-readable manner via the Foaf record of the author, and the DCAT ontology "contactPoint" property.
- Human readable keywords, using DCAT, are mirrored and/or enhanced by a machine-readable RDF document which is the value of the DCAT "theme" property. This RDF document follows the structure determined by the SKOS ontology, and lists the ontological terms that describe the repository for machine-processing.
- Finally, individual records within the dataset are represented as the value of the Linked Data Platform "contains" property, and provided as a possibly paginated list of URLs (a discussion of machine-actionable pagination will not be included here). These URLs are the MetaRecord Resource URLs shown in Figure 1.

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Following the flow in Figure 1, the next step in the FAIR Accessor is to resolve a MetaRecord Resource URL. For clarity, we will first show the metadata document that is returned if there are no FAIR Projectors for that dataset. In the subsequent section, we will show how FAIR Projectors enhance this basic metadata with additional features.

Calling HTTP GET on a MetaRecord Resource URL returns a document with the structure shown in Figure 5.

UniProt Protein C8UZX9	bibliographic Citation	The UniProt Consortium (2015). UniProt: a hub for protein information. Nucleic Acids Res. 43: D204-D212
	creator	UniProt Consortium
	language	eng
	license	cc by nd3.0
	title	UniProt Protein C8UZX9
	in Dataset	Uni Prot Accessor/
	contact Point	contact
	description	KRR1 small subunit processome componentKRR-R motif-containing protein 1
	distribution	C8UZX9.rdf C8UZX9.html
	identifier	C8UZX9
	keyword	Annotation Aspergillus nidulans Aspergillus Functional Annotation GO Gene Ontology Proteins RNA Processing
	landing Page	uniprot.org
	language	en
	publisher	uniprot.org
	page	sparql uniprot.org/ C8UZX9
	primary topic	C8UZX9
C8UZX9	...	
C8UZX9.rdf	format	application/rdf+xml
	type	Dataset Dataset Distribution
	download URL	C8UZX9.rdf
C8UZX9.html	format	text/html
	type	Dataset Distribution
	download URL	C8UZX9.html

Figure 5. A representative portion of the output from resolving the MetaRecord Resource of the FAIR Accessor for record C8UZX9, rendered into HTML by the Tabulator Firefox plugin. The columns have the same meaning as in Figure 4.

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Many properties in this metadata document are similar to those at the higher level of the FAIR Accessor, however, the primary topic of this document is the original UniProt record.

577 Therefore, the values of these facets now reflect the authorship and contact information for
 578 the record itself. We do, however, recognize that MetaRecords are themselves scholarly
 579 works and should be properly cited. The MetaRecord includes the “in dataset” predicate,
 580 which refers back to the first level of the FAIR Accessor, thus this provides one avenue for
 581 capturing the provenance information for the MetaRecord. If additional provenance detail is
 582 required, we propose (but do not describe further here) that this information could be
 583 contained in a separate named graph, in a manner akin to that used by
 584 NanoPublications(Kuhn et al., 2016).

585
 586 The important distinctive property in this document is the “distribution” property, from the
 587 DCAT ontology. For clarity, an abbreviated document in Turtle format is shown in Figure 6,
 588 containing only the “distribution” elements and their values.

589
 590

```
@prefix dcat: <http://www.w3.org/ns/dcat#>.
@prefix Uni: <./>.
@prefix n0: <http://purl.org/dc/elements/1.1/>.
@prefix void: <http://rdfs.org/ns/void#>.

Uni:C8UZX9
  dcat:distribution
    <http://www.uniprot.org/uniprot/C8UZX9.rdf>,
    <http://www.uniprot.org/uniprot/C8UZX9.html> .

<http://www.uniprot.org/uniprot/C8UZX9.rdf>
  n0:format
    "application/rdf+xml";
  a      n0:Dataset, void:Dataset, dcat:Distribution;
  dcat:downloadURL
    <http://www.uniprot.org/uniprot/C8UZX9.rdf>.

<http://www.uniprot.org/uniprot/C8UZX9.html>
  n0:format
    "text/html";
  a      n0:Dataset, dcat:Distribution;
  dcat:downloadURL
    <http://www.uniprot.org/uniprot/C8UZX9.html>.
```

Figure 6. Turtle representation of the subset of triples from the MetaRecord metadata pertaining to the two DCAT Distributions. Each distribution specifies an available representation (media type), and a URL from which that representation can be downloaded.

591

592

593 There are two DCAT Distributions in this document. The first is described as being in format
 594 “application/rdf+xml”, with its associated download URL. The second is described as being
 595 in format “text/html”, again with the correct URL for that representation. Both are typed as

596 Distributions from the DCAT ontology. These distributions are published by UniProt
597 themselves, and the UniProt URLs are used. The additional metadata in the FAIR Accessor
598 explicitly describes the keywords that relate to that record (both machine and human-
599 readable), access policy, license, and format, allowing machines to more accurately
600 determine the utility of this record prior to retrieving it.

601

602 Several things are important to note before moving to a discussion of FAIR Projectors. First,
603 the two levels of the FAIR Accessor are not interdependent. The Container layer can
604 describe relevant information about the scope and nature of a repository, but might not
605 provide any further links to MetaRecords. Similarly, whether or not to provide a distribution
606 within a MetaRecord is entirely at the discretion of the data owner. For sensitive data, an
607 owner may chose to simply provide (even limited) metadata, but not provide any direct link to
608 the data itself, and this is perfectly conformant with the FAIR guidelines. Further, when
609 publishing a single data record, it is not obligatory to publish the Container level of the FAIR
610 Accessor; one could simply provide the MetaRecord document describing that data file,
611 together with an optional link to that file as a Distribution.

612

613 **The FAIR Projector**

614

615 FAIR Projectors can be used for many purposes, including (but not limited to) transformation
616 of a data source from non-Linked Data to Linked Data, transformation of a Linked Data
617 source into a different Linked Data structure or ontological framework, load-
618 management/query-management, or as a means to explicitly describe the ontological
619 structure of an underlying data source in a searchable manner. In this demonstration, the
620 FAIR Projector transforms the semantics of the native RDF provided by UniProt into a
621 different ontological framework (EDAM).

622

623 The address of this FAIR Projector's TPF interface is:

624

625 `http://linkeddata.systems:3001/fragments`

626

627 The TPF API requires a subject and/or predicate and/or object node to be specified as
628 parameters; a request for the all-variable pattern will (currently) return nothing. How can a
629 software agent know what parameters are valid, and what will be returned from such a call?

630

631 In this interoperability infrastructure, we propose that Projectors should be considered as
632 DCAT Distributions, and thus TPF URLs, with appropriate parameters added and bound, are
633 included in the distribution section of the MetaRecord metadata. An example is shown in
634 Figure 6, again rendered using Tabulator.

635

636

UniProt Protein C8UZX9	bibliographic Citation	The UniProt Consortium (2015). UniProt: a hub for protein information. Nucleic Acids Res. 43: D204-D212
	creator	UniProt Consortium
	language	eng
	license	cc by nd3.0
	title	UniProt Protein C8UZX9
	in Dataset	Uni Prot Accessor/
	contact Point	contact
	description	KRR1 small subunit processome componentKRR-R motif-containing protein 1
	distribution	http://linkeddata.systems:3001/fragments?subject=http%3A%2F%2Fidentifiers%2Eorg%2FuniProt%2FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2FclassifiedWith%2Fidentifiers%2Eorg%2FuniProt%2FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2ForganismC8UZX9.rdf http://linkeddata.systems:3001/fragments?subject=http%3A%2F%2Fidentifiers%2Eorg%2FuniProt%2FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2ForganismC8UZX9.html C8UZX9 C8UZX9
	identifier	C8UZX9
	keyword	Annotation Aspergillus nidulans Aspergillus Functional Annotation GO Gene Ontology Proteins RNA Processing
	landing Page	uniprot.org
	language	en
	publisher	uniprot.org
	page	sparql
		uniprot.org/ C8UZX9
	primary topic	C8UZX9

Figure 7. A portion of the output from resolving the MetaRecord Resource of the FAIR Accessor for record C8UZX9, rendered into HTML by the Tabulator Firefox plugin. The columns have the same meaning as in Figure 4. Comparing the structure of this document to that in Figure 5 shows that there are now four values for the “distribution” predicate. An RDF and HTML representation, as in Figure 5, and two additional distributions with URLs conforming to the TPF design pattern (highlighted).

637

638

639 Note that, in addition to the two distributions C8V1J1.html and C8V1J1.rdf that were seen in

640 Figure 5, there are now two additional distributions that include both a subject and predicate

641 parameter in their URLs. These are the URLs for two FAIR Projections of that data.

642

643 Again, looking at an abbreviated and simplified Turtle document for clarity (Figure 8) we can

644 see the metadata structure of one of these two new distributions.

645

646

```

Uni:C8UZX9
  dcat:distribution
<http://linkeddata.systems:3001/fragments?subject=http%3A%2F%2Fidentifiers%2Eorg%2FuniProt%2FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2FclassifiedWith> .

<http://linkeddata.systems:3001/fragments?subject=http%3A%2F%2Fidentifiers%2Eorg%2FuniProt%2FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2FclassifiedWith>

```

```

n0:format
  "application/rdf+xml", "application/x-turtle", "text/html";
a    FAI:Projector, n0:Dataset, void:Dataset, dcat:Distribution;
dcat:downloadURL
<http://linkeddata.systems:3001/fragments?subject=http%3A%2F%2Fidentifiers%2Eorg%2Funiprot%2
FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2FclassifiedWith> .

loc:Source3C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rml:hasMapping
    loc:Mappings3C0D4EAA-8497-11E6-99DD-D5545D07C3DD;
  rml:referenceFormulation
    ql:TriplePatternFragments;
  rml:source
<http://linkeddata.systems:3001/fragments?subject=http%3A%2F%2Fidentifiers%2Eorg%2Funiprot%2
FC8UZX9&predicate=http%3A%2F%2Fpurl%2Euniprot%2Eorg%2Fcore%2FclassifiedWith> .

loc:Mappings3C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rml:logicalSource
    loc:Source3C0D4EAA-8497-11E6-99DD-D5545D07C3DD;
  rr:predicateObjectMap
    loc:POMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD;
  rr:subjectMap
    loc:SubjectMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD .

```

Figure 8. Turtle representation of the subset of triples from the MetaRecord metadata pertaining to one of the FAIR Projector DCAT Distributions of the MetaRecord shown in Figure 6. The text is color-coded to assist in visual exploration of the RDF. The DCAT Distribution blocks of the two Projector distributions (black bold) have multiple media-type representations (red), and are connected to an RML logicalSource (purple), which itself is linked by the hasMapping predicate to a Mappings (blue) block of RML that semantically describes the subject, predicate, and object (green and orange) of the Triple Descriptor for that Projector. The full RML model is shown separately in Figure 9.

647

648 Following the Triple Pattern Fragments behavior, requesting the downloadURL with HTTP
649 GET will trigger the Projector to generate all triples where the subject is UniProt record
650 C8UZX9, and the predicate is “classifiedWith” from the UniProt Core ontology. Those triples
651 will match the semantics and structure defined in the Mappings (blue) block. The
652 interpretation of the Dublin Core “format” predicate in this context is noteworthy, as its value
653 is only loosely defined by Dublin Core. A Projector is a RESTful resource that will respond to
654 HTTP content-negotiation to select the representation of the requested resource. The values
655 of the “format” predicate in this example should be interpreted as a list of the possible
656 formats available, for example, to be used as valid values for the HTTP Accept Header when
657 calling that resource. In this case, there are three available representations - Turtle, HTML,
658 and RDF/XML.

659

660 The schematic structure of the Mapping RML is visualized in Figure 2, with the actual output
661 from the Accessor shown in Figure 9, color-coded to assist visual exploration. The RML
662 describes a Triple where the subject will be of type `edam:data_0896` (“Protein record”), the

663 predicate will be “classifiedWith” from the UniProt Core ontology, and the object will be
 664 of type `edam:data_1176` (“GO Concept ID”).

665

666

```

loc:Mappings3C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rml:logicalSource
    loc:Source3C0D4EAA-8497-11E6-99DD-D5545D07C3DD;
  rr:predicateObjectMap
    loc:POMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD;
  rr:subjectMap
    loc:SubjectMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD.

loc:SubjectMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rr:class ed:data_0896; rr:template "http://identifiers.org/uniprot/{ID}".

loc:POMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rr:objectMap
    loc:ObjectMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD;
  rr:predicate
    core:classifiedWith.

loc:ObjectMap3C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rr:parentTriplesMap loc:SubjectMap23C0D4EAA-8497-11E6-99DD-D5545D07C3DD.

loc:SubjectMap23C0D4EAA-8497-11E6-99DD-D5545D07C3DD
  rr:class ed:data_1176; rr:template "http://purl.obolibrary.org/obo/{GO}".
  
```

Figure 9. Turtle representation of a Triple Descriptor within the MetaRecord metadata shown in Figures 6 and 7. The text is color-coded to assist in visual exploration of the RDF. The RDF structure shown here is represented schematically in Figure 2. The black bold text shows the locations where the semantic type of the projected subject, predicate, and object are stored.

667

668 The triples that are returned by calling HTTP GET on that Projector URL are:

669

```

670 @prefix uni: <http://identifiers.org/uniprot/>.
671 @prefix obo: <http://purl.obolibrary.org/obo/>.
672 uni:C8UZX9 core:classifiedWith obo:GO_0000447, obo:GO_0000462
673 .
  
```

673

674

675 This is accompanied by a block of hypermedia controls (not shown) using the Hydra
 676 vocabulary (Lanthaler & Gütl; Das, Sundara & Cyganiak, 27 September, 2012) that provide
 677 machine-readable instructions for how to navigate the remainder of that dataset - for
 678 example, how to get the entire row, or the entire column for the current data-point.

679

680 Though the subject and object are not explicitly typed in the output from this call to the
 681 Projector, further exploration of the Projector's output, via those TPF's hypermedia controls,
 682 would reveal that the Subject and Object are in fact typed according to the EDAM ontology
 683 (Ison et al., 2013), as declared in the RML meta-descriptor. Thus, this FAIR Projector
 684 transformed data from UniProt Core semantic types, to the equivalent data, now represented
 685 within the EDAM semantic framework, as shown in Figure 10. Also note that the URI
 686 structure for the UniProt entity has been changed from the UniProt URI scheme to the more
 687 interoperable Identifiers.org scheme.

688
 689 This example was chosen because, in UniProt and the Gene Ontology consortium's
 690 representation, Gene Ontology terms do not have a richer classification than "owl:Class".
 691 With respect to interoperability, this is problematic, as the lack of rich semantic typing
 692 prevents them from being used for automated discovery of resources that could potentially
 693 consume them, or use them for integrative, cross-domain queries.

694
 695

In UniProt	<pre> http://purl.uniprot.org/uniprot/C8UZX9 a http://purl.uniprot.org/core/Protein ; http://purl.uniprot.org/core/classifiedWith http://purl.obolibrary.org/obo/GO_0000462 . http://purl.obolibrary.org/obo/GO_0000462 a http://www.w3.org/2002/07/owl#Class </pre>
After Projection	<pre> http://identifiers.org/uniprot/C8UZX9 a http://edamontology.org/data_0896 ; http://purl.uniprot.org/core/classifiedWith http://purl.obolibrary.org/obo/GO_0000462 . http://purl.obolibrary.org/obo/GO_0000462 a http://edamontology.org/data_1176 </pre>

Figure 10: Data before and after FAIR Projection. Bolded segments show how the URI structure and the semantics of the data were modified, according to the mapping defined in the Triple Descriptor (data_0896 = "Protein report" and data_1176 = "GO Concept ID"). URI structure transformations may be useful for integrative queries against datasets that utilize the Identifiers.org URI scheme such as OpenLifeData (González et al., 2014). Semantic transformations allow integrative queries across datasets that utilize diverse and redundant ontologies for describing their data, and in this example, may also be used to add semantics where there were none before.

696

697 Discussion

698

699 Interoperability is hard. It was immediately evident that, of the four FAIR principles,
700 Interoperability was going to be the most challenging. Here we have designed a novel
701 infrastructure with the primary objective of interoperability for both metadata and data, but
702 with an eye to all four of the FAIR Principles. We wished to provide discoverable and
703 interoperable access to a wide range of underlying data sources - even those in
704 computationally opaque formats - as well as supporting wide array of both academic and
705 commercial end-user applications above these data sources. In addition, we imposed
706 constraints on our selection of technologies; in particular, that the implementation should re-
707 use existing technologies as much as possible, and should support multiple and
708 unpredictable end-uses. Moreover, it was accepted from the outset that the tradeoff between
709 simplicity and power was one that could not be avoided. While other interoperability projects
710 such as caBIO (Covitz et al., 2003) and TAPIR (De Giovanni et al., 2010) created rich APIs
711 or query languages, enabling extremely powerful cross-resource data exploration and
712 integration, this was done at the expense of broad-scale uptake and/or with the explicit and
713 unavoidable participation of the individual providers. Thus, with the goal of maximizing global
714 uptake and adoption of this interoperability infrastructure, and democratizing the cost of
715 implementation over the entire stakeholder community - both users and providers - we opted
716 for lightweight, weakly integrative, REST-based solutions, that nevertheless lend themselves
717 to significant degrees of mechanization in both discovery and integration.

718

719 We now look more closely at how this interoperability infrastructure meets the expectations
720 within the FAIR Principles.

721

722 **FAIR facet(s) addressed by the Container Resource:**

723

- 724 ● **Findable** - The container has a distinct globally unique and resolvable
725 identifier, allowing it to be discovered and explicitly, unambiguously cited. This
726 is important because, in many cases, the dataset being described does not
727 natively possess an identifier, as in our example above where the dataset
728 represented the results of a query. In addition, the container's metadata
729 describes the research object, allowing humans and machines to evaluate the
730 potential utility of that object for their task.

730

- 731 ● **Accessible** - the Container URL resolves to a metadata record using
732 standard HTTP GET. In addition to describing the nature of the research
733 object, the metadata record should include information regarding licensing,
734 access restrictions, and/or the access protocol for the research object.
735 Importantly, the container metadata exists independently of the research
736 object it describes, where FAIR Accessibility requires metadata to be
737 persistently available even if the data itself is not.

737

- 738 ● **Interoperable** - The metadata is provided in RDF - a globally-applicable
739 syntax for data and knowledge sharing. In addition, the metadata uses
740 shared, widely-adopted public ontologies and vocabularies to facilitate
interoperability at the metadata level.

- 741
- **Reusable** - the metadata includes citation information related to the authorship of the container and/or its contents, and license information related to the reuse of the data, by whom, and for what purpose.
- 742
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744

745 **Other features of the Container Resource**

- **Privacy protection** - The container metadata provides access to a rich description of the content of a resource, without exposing any data within that resource. While a provider may choose to include MetaRecord URLs within this container, they are not required to do so if, for example, the data is highly sensitive, or no longer easily accessible; however, the contact information provided within the container allows potential users of that data to inquire as to the possibility of gaining access in some other way. As such, this container facilitates a high degree of FAIRness, while still providing a high degree of privacy protection.
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756 **FAIR Facet(s) Addressed by the MetaRecord:**

- **Findable** - The MetaRecord URL is a globally-unique and resolvable identifier for a data entity, regardless of whether or not it natively possesses an identifier. The metadata it resolves to allows both humans and machines to interrogate the nature of a data element before deciding to access it.
 - **Accessible** - the metadata provided by accessing the MetaRecord URL describes the accessibility protocol and license information for that record, and describes all available formats.
 - **Interoperable** - as with the Container metadata, the use of shared ontologies and RDF ensures that the metadata is interoperable.
 - **Reusable** - the MetaRecord metadata should carry record-level citation information to ensure proper attribution if the data is used. We further propose, but do not demonstrate, that authorship of the MetaRecord itself could be carried in a second named-graph, in a manner similar to that adopted by the NanoPublication community.
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772 **Other features of the MetaRecord**

- **Privacy protection** - the MetaRecord provides for rich descriptive information about a specific member of a collection, where the granularity of that description is entirely under the control of the data owner. As such, the MetaRecord can provide a high degree of FAIRness at the level of an individual record, without necessarily exposing any identifiable information. In addition, the provider may choose to stop at this level of FAIRness, and not include further URLs giving access to the data itself.
 - **Symmetry of traversal** - Since we predict that clients will, in the future, query over indexes of FAIR metadata searching for dataset or records of interest, it is not possible to predict the position at which a client or their agent will enter your FAIR Accessor. While the container metadata provides links to individual MetaRecords, the MetaRecord similarly provides a reference back “upwards”
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785 to its container. Thus a client can access repository-level metadata (e.g.
786 curation policy, ownership, linking policy) for any given data element it
787 discovers. This became particularly relevant as a result of the European Court
788 of Justice decision
789 ([http://curia.europa.eu/jcms/upload/docs/application/pdf/2016-](http://curia.europa.eu/jcms/upload/docs/application/pdf/2016-09/cp160092en.pdf)
790 [09/cp160092en.pdf](http://curia.europa.eu/jcms/upload/docs/application/pdf/2016-09/cp160092en.pdf)) that puts the burden of proof on those who create
791 hyperlinks to ensure the document they link to is not, itself, in violation of
792 copyright.

- 793 ● **High granularity of access control** - individual elements of a collection may
794 have distinct access constraints or licenses. For example, individual patients
795 within a study may have provided different consent. MetaRecords allow each
796 element within a collection to possess, and publish, its own access policy,
797 access protocol, license, and/or usage-constraints, thus providing fine-
798 grained control of the access/use of individual elements within a repository.
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FAIR Facet(s) Addressed by the Triple Descriptors and FAIR Projectors:

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- 812 ● **Findable** - Triple Descriptors, in isolation or when aggregated into FAIR
813 Profiles, provide one or more semantic interpretations of data elements. By
814 indexing these descriptors, it would become possible to search over datasets
815 for those that contain data-types of interest. Moreover, FAIR Projectors, as a
816 result of the TPF URI structure, create a unique URL for every data-point
817 within a record. This has striking consequences with respect to scholarly
818 communication. For example, it becomes possible to unambiguously refer-to,
819 and therefore “discuss” and/or annotate, individual spreadsheet cells from any
820 data repository.
- 821 ● **Accessible** - Using the TPF design patterns, all data retrieval is
822 accomplished in exactly the same way - via HTTP GET. The response
823 includes machine-readable instructions that guide further exploration of the
824 data without the need to define an API. FAIR Projectors also give the data
825 owner high granularity access control; rather than publishing their entire
826 dataset, they can select to publish only certain components of that dataset,
827 and/or can put different access controls on different data elements, for
828 example, down to the level of an individual spreadsheet cell.
- 829 ● **Interoperable** - FAIR Projectors provide a standardized way to export any
830 type of underlying data in a machine-readable structure, using widely used,
831 public shared vocabularies. Data linkages that were initially implicit in the
832 datastore, identifiers for example, become explicit when converted into URIs,
833 resulting in qualified linkages between formerly opaque data deposits.
- 834 ● **Reusable** - All data points now possess unique identifiers, which allows them
835 to be explicitly connected to their citation and license information (i.e. the
836 MetaRecord). In this way, every data point, even when encountered in
837 isolation, provides a path to trace-back to its reusability metadata.
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830

Other features of FAIR Projection

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- 832 ● **Native formats are preserved** - As in many research domains,
833 bioinformatics has created a large number of data/file formats. Many of
834 these, especially those that hold “big data”, are specially formatted flat-files
835 that focus on size-efficient representation of data, at the expense of general
836 machine-accessibility. The analytical tooling that exists in this domain,
837 therefore, is capable of consuming these various formats. While the FAIR
838 Data community has never advocated for wholesale Interoperable
839 representations of these kinds of data - which would be inefficient, wasteful,
840 and lacking in utility given that no tooling exists to consume such
841 representations - the FAIR Projector provides a middle-ground. Projection
842 allows software to query the core content of a file in a repository prior to
843 downloading it; for example, to determine if it contains data about an entity or
844 identifier of interest. FAIR Projectors, therefore, enable efficient efficient
845 discovery of data of-interest, without requiring wasteful transformation of all
846 data content into a FAIR format.
- 847 ● **Semantic conversion of existing Triplestores** - It is customary to re-cast
848 the semantic types of entities within triplestores using customized SPARQL
849 BIND or CONSTRUCT clauses. FAIR Projectors provide a standardized,
850 SPARQL-free, and discoverable way to accomplish the same task. This
851 further harmonizes data, and simplifies interoperability.
- 852 ● **Standardized interface to (some) Web Services** - Many Web Services in
853 the biomedical domain have a single input parameter, generally representing
854 an identifier for some biochemical entity. FAIR Projectors can easily replace
855 these myriad Web Services with a common TPF interface, thus dramatically
856 enhancing discoverability, machine-readability, and interoperability between
857 these currently widely disparate services.

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858 Incentives - why will this happen?

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Looking forward, there is every indication that FAIRness will be a requirement of funding agencies and/or journals. As such, infrastructures such as the one described in this exemplar will almost certainly become a natural part of scholarly data publishing in the future. We indicated earlier, however, that we also believe that the creation of FAIR layers over pre-existing data will become a natural part of the daily data transformation activities of the global bioinformatics community. We suggest that this will happen because, though we utilize RML in this demonstration only for its modelling properties, there exists tooling for a wide variety of common file formats such as CSV and Excel that allow RML models to drive the data transformation itself. As such, we predict that those who need to transform data will begin to create, publish, and use RML models together with these generic transformation tools to enact their data transformations, rather than continuing to write one-off scripts. This may be incentivized even more by creating repositories of RML models that can be reused by those needing to do data transformations. Though the infrastructure for capturing these user-driven transformation events and formalizing them into FAIR Projectors does not yet

873 exist, it does not appear on its surface to be a complex problem. Thus, we expect that such
874 infrastructure should appear soon after FAIRness becomes a scholarly publishing
875 requirement.

876

877 Indeed, several communities of data providers are currently planning to use this, or related
878 FAIR implementations, to assist their communities to find, access, and reuse their valuable
879 data holdings. For example, the Biobanking and Rare disease communities will be given
880 end-user tools that utilize/generate such FAIR infrastructures to: guide discovery by
881 researchers; help both biobankers and researchers to re-code their data to standard
882 ontologies building on the SORTA system (Pang et al., 2015); assist to extend the
883 MOLGENIS/BiobankConnect system (Pang et al., 2016); add FAIR interfaces to the BBMRI
884 and RD-connect national and European biobank data and sample catalogues. There are
885 also a core group of FAIR infrastructure authors who are creating large-scale indexing and
886 discovery systems that will facilitate the automated identification and retrieval of relevant
887 information, from any repository, in response to end-user queries, portending a day when
888 currently unused - "lost" - data deposits once again provide return-on-investment through
889 their discovery and reuse.

890

891 Conclusions

892

893 There is a growing movement of governing bodies and funding organizations towards a
894 requirement for open data publishing, following the FAIR Principles. It is, therefore, useful to
895 have an exemplar "reference implementation" that demonstrates the kinds of behaviours that
896 are expected from FAIR resources.

897

898 Of the four FAIR Principles, Interoperability is arguably the most difficult FAIR facet to
899 achieve, and has been the topic of decades of informatics research. Several new standards
900 and frameworks have appeared in recent months that addressed various aspects of the
901 Interoperability problem. Here, we apply these in a novel combination, and show that the
902 result is capable of providing interoperability between formerly incompatible data formats
903 published anywhere on the Web. In addition, we note that the other three aspects of FAIR -
904 Findability, Accessibility, and Reusability - are easily addressed by the resulting
905 infrastructure. The outcome, therefore, provides machine-discoverable access to richly
906 described data resources in any format, in any repository, with the possibility of
907 interoperability of the contained data down to the level of an individual "cell". No new
908 standards or APIs were required; rather, we rely on RESTful behaviours, with all entities
909 being resolvable resources that allow hypermedia-driven "drill-down" from the level of a
910 repository descriptor, all the way to an individual data point in the record.

911

912 Such an interoperability layer may be created and published by anyone, for any data source,
913 without necessitating an interaction with the data owner. Moreover, the majority of the
914 interoperability layer we describe may be achieved through dynamically generated files from
915 software, or even (for the Accessor portion) through static, manually-edited files deposited in

916 any public repository. As such, knowledge of how to build or deploy Web infrastructure is not
917 required to achieve a large portion of these FAIR behaviors.

918

919 The trade-off between power and simplicity was considered acceptable, as a means to
920 hopefully encourage wide adoption. The modularity of the solution was also important
921 because, in a manner akin to crowdsourcing, we anticipate that the implementation will
922 spread through the community on a needs-driven basis, with the most critical resource
923 components being targeted early - the result of individual researchers requiring interoperable
924 access to datasets/subsets of interest to them. The interoperability design patterns
925 presented here provide a structured way for these individuals to contribute and share their
926 individual effort - effort they would have invested anyway - in a collaborative manner, piece-
927 by-piece building a much larger interoperable and FAIR data infrastructure to benefit the
928 global community.

929

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931

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