

PROV-man: A PROV-compliant toolkit for provenance management

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Discoveries in modern science can take years and involve the contribution of large amounts of data, many people and various tools. Although good scientific practice dictates that findings should be reproducible, in practice there are very few automated tools that actually support traceability of the scientific method employed, in particular when various experimental environments are involved at different research phases. Data provenance tracking approaches can play a major role in addressing many of these challenges. These approaches propose ways to capture, manage, and use of provenance information to support the traceability of the scientific methods in heterogeneous environments. PROV is a W3C standard that provides a comprehensive model for data and semantics representation with common vocabularies and rich concepts to describe provenance. Nevertheless, it is difficult for domain scientists to easily understand and adopt all the richness provided by PROV. In this paper we describe the design and implementation of the provenance manager *PROV-man*, a PROV-compliant framework that facilitates the tasks of scientists in integrating provenance capabilities into their data analysis tools. *PROV-man* provides functionalities to create and manipulate provenance data in a consistent manner and ensures its permanent storage. It also provides a set of interfaces to serialize and export provenance data into various data formats, serving interoperability. The open architecture of *PROV-man*, consisting of an API and a configurable database, allows for its easy deployment within existing and newly developed software tools. The paper presents examples illustrating the usage of *PROV-man*. The first example illustrates how to create and manipulate provenance data of an online newspaper article using *PROV-man*. The second example demonstrates and evaluates the *PROV-man* implementation in a more complex case for collection of provenance data about biomedical data analysis activities that are carried out using a distributed computing infrastructure.

1 **PROV-man: A PROV-compliant toolkit for provenance management**

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6 **Abstract**

7 Discoveries in modern science can take years and involve the contribution of large amounts of data, many
8 people and various tools. Although good scientific practice dictates that findings should be reproducible, in
9 practice there are very few automated tools that actually support traceability of the scientific method employed,
10 in particular when various experimental environments are involved at different research phases. Data
11 provenance tracking approaches can play a major role in addressing many of these challenges. These
12 approaches propose ways to capture, manage, and use of provenance information to support the traceability of
13 the scientific methods in heterogeneous environments. PROV is a W3C standard that provides a comprehensive
14 model for data and semantics representation with common vocabularies and rich concepts to describe
15 provenance. Nevertheless, it is difficult for domain scientists to easily understand and adopt all the richness
16 provided by PROV. In this paper we describe the design and implementation of the provenance manager
17 *PROV-man*, a PROV-compliant framework that facilitates the tasks of scientists in integrating provenance
18 capabilities into their data analysis tools. *PROV-man* provides functionalities to create and manipulate
19 provenance data in a consistent manner and ensures its permanent storage. It also provides a set of interfaces to
20 serialize and export provenance data into various data formats, serving interoperability. The open architecture
21 of *PROV-man*, consisting of an API and a configurable database, allows for its easy deployment within
22 existing and newly developed software tools. The paper presents examples illustrating the usage of *PROV-*
23 *man*. The first example illustrates how to create and manipulate provenance data of an online newspaper
24 article using *PROV-man*. The second example demonstrates and evaluates the *PROV-man* implementation in a
25 more complex case for collection of provenance data about biomedical data analysis activities that are carried
26 out using a distributed computing infrastructure.

27 **Keywords**

28 Provenance, OPM, PROV, e-science, database design, ER Modeling, RDBMS, open architecture, ORM, Java,
29 Hibernate, Workflow management system.

30 **1 Introduction**

31 Many research laboratories nowadays use (new) technologies for large-scale data acquisition and
32 distributed infrastructures for large-scale and collaborative data analysis. Research can take many
33 years and involve a large number of people, data and tools. In such complex environment, proper
34 methodologies need to be adopted by the scientists to carry out large endeavors in a way to guarantee
35 that all the steps have been correctly performed and that they can be traced back to facilitate
36 reproducibility of scientific results. The proliferation of large data sets and the increasing complexity
37 of the scientific environment pose severe challenges for achieving this in practice.

38 Data provenance mechanisms provide ways to capture, manage, and use provenance information in
39 heterogeneous environments [1]. They refer to the capability of determining the origin and history, or
40 lineage, of a certain piece of data [2]. Therefore, data provenance plays a major role in addressing the
41 emerging challenges in today's and future scientific environments. Additionally, the importance of
42 data provenance is rapidly increasing in a connected digital world where open sources of data are
43 becoming available for everyone [3].

44 In recent years, scientists and researchers from different application domains have increased their
45 efforts in recording and exploiting data provenance facilities. The motivation for introducing
46 mechanisms to manage data provenance in scientific experiments is two-fold. First, data provenance
47 documents the data generation and analysis process by including how data and results were generated,
48 and therefore it provides means to establish credibility and trust in scientific findings. Secondly, it
49 provides useful means for the scientists to better understand the way they perform their experiments
50 and to trace, reproduce and explain the data analysis process.

51 Provenance capture was and still is a crucial component in many developed software tools and
52 applications [4] [5]. Most of these implement provenance in a manner very specific to their
53 application domain or using specific concepts and technologies. Since the emergence of provenance
54 as a standard (OPM [6] in 2007 followed by PROV [7] in 2013), many efforts have attempted to
55 provide implementations of these standards[8]. Nowadays, PROV is being adopted by a large
56 community from the scientific domain, therefore the number of related implementations rapidly
57 increased. However, because the PROV definition is very detailed and complex, most of these
58 implementations cover only part of the complete recommendations, and each focuses on one specific
59 scientific domain. The lack of generic provenance tool means consuming a lot of efforts from
60 experts in the scientific domain, and presenting additional challenges when new updates are
61 introduced to the PROV standard. An exception is the ProvStore [9] and PROV-WF [10], which
62 provide, respectively, a web service to manipulate provenance documents and a runtime provenance
63 that can be queried even during the workflow execution. More clarifications about these development
64 are given in section 2.3.

65 The main issue that remains unsolved for the scientist, even when using all these tools, is: *how can I*
66 *instrument my scientific code to collect provenance data with less efforts and in a comprehensive and*
67 *reliable manner?* Therefore, we felt the need to provide an implementation of PROV-compliant tools
68 that facilitate the capture of provenance data with minimum effort by the developers of scientific
69 applications and services.

70 In this paper we describe the design and implementation of a generic framework that is compliant
71 with the provenance standard PROV, following the latest specification published by the provenance
72 W3C community [7]. The implemented provenance management framework (*PROV-man*) consists of
73 a programming interface (API) and a configurable database that can be used to create and store
74 provenance according to the PROV standard. *PROV-man* deploys permanent back-end storage and
75 follows an open architecture approach, which facilitates its deployment with existing and newly
76 developed software tools. Interoperability and optimization are also considered at both the back-end
77 storage and the core implementation of *PROV-man*.

78 In this paper we first introduce the provenance concepts (section 2), discussing their evolution in the
79 domain of scientific applications, and highlighting the main efforts implementing provenance before
80 and after the release of PROV. Section 3 presents the Implementation details of *PROV-man*, covering
81 the approach, the database model and the API. Section 4 demonstrates the usage and deployment of
82 *PROV-man* framework for provenance data creation and collection on a distributed computing
83 infrastructure. Section 6 raises the implementation challenges and discusses their solutions. Finally,
84 section 6 presents concluding remarks.

85 **2 Provenance: Past and Future**

86 Provenance, as general term, originates from the French *provenir*, "to come from". It refers to the
87 chronology of the ownership, custody or location of a historical object. The term was originally

88 mostly used for works of art, for which a good provenance helps to confirm the date, status, artist,
89 subject, and the past owners of a painting, and can increase its value. Currently, the term Provenance
90 is used in similar ways in a wide range of fields, including archaeology, paleontology, archives,
91 manuscripts, printed books, and e-science [11].

92 In this section we present in more details the evolution of provenance in the context of e-science. The
93 underlying assumption is that scientific research is generally considered to be of good provenance
94 when it is sufficiently documented to allow reproducibility and to facilitate the process of tracking
95 scientific datasets through all transformations, analyses, and interpretations. In the remaining sections
96 of this paper we refer to Provenance in e-science as *data provenance*.

97 **2.1 Early Efforts**

98 At an early stage (before 1990), provenance information was mainly captured using unstructured logs
99 and temporary files stored on the local disks of the machines where the programs are executed [12].
100 Provenance information has also been captured as metadata in information management systems for
101 various applications. For example, DICOM (Digital Imaging and Communication in Medicine [13])
102 is a standard used for medical images that contains detailed information about the origin of medical
103 images. Other examples are Laboratory Information Management Systems (LIMS) [14] and
104 Electronic Laboratory Notebooks (ELN) [15], which have been around since the 90's and provide
105 annotation facilities for workflow metadata and data tracking for experimental data.

106 From 2000, the use of data provenance terms for describing the history and lineage of data has
107 become more prominent in scientific computing systems [12], [16]. In 2005, Yogesh [4] and Bose [5]
108 published surveys and comparisons of the different projects and systems with mechanisms to manage
109 data provenance. These projects cover different applications and disciplines such as Earth sciences
110 [17], finances [18], e-science [19], curated databases [20], grid computing [3], and other projects such
111 as Chimera [21], the Collaboratory for Multi-Scale Chemical Science (CMCS) [1], and Trio [2].

112 In the domain of e-science, the scientific workflow management systems (WfMS) developers were
113 among the first interested in using and deploying provenance management. This is due to the step-
114 wise design approach used for composing and executing workflows, which enables the capture of
115 data provenance automatically and at fine granularity [22][23]. Examples of WfMS with provenance
116 capabilities include Pegasus [24], Kepler [25], and Taverna [26]. Typically, each of the systems used
117 its custom terminology for defining and capturing data provenance.

118 Around 2006, consensus about provenance concepts and terminology starts to emerge, and
119 community efforts towards standardization become feasible as described below.

120 **2.2 OPM: The Open Provenance Model**

121 As a result of increasing interest in data provenance, in 2006 the International Provenance and
122 Annotation Workshop (IPAW'06) [27] was organized. It involved around 50 participants, interested
123 in the issues of data provenance, process documentation, data derivation, and data annotation. During
124 the IPAW'06 workshop a consensus began to emerge on provenance standardization, hence a series
125 of Provenance Challenges took place [28, 29]. As a result of this community effort, the *Open*
126 *Provenance Model* OPM v1.00 was released in December 2007 [30]. The first OPM workshop, held
127 in June 2008, involved around 20 participants who discussed issues related the OPM specification.
128 This initiative led to a revised specification, referred to as OPM v1.01 [31].

129 OPM is based on three entities (*Artifacts*, *Processes*, and *Agents*) that are linked using *causal*
130 *relationships*, representing their dependency (e.g. *used*, *wasGeneratedBy*, *wasControlledBy*, etc.).

131 OPM defines structures for representing the provenance information as a graph with nodes and edges,
132 and also specifies inference queries. The original intent of OPM has been to define a data model that
133 is open not only from an interoperability viewpoint, but also with respect to the community of its
134 contributors, reviewers and users.

135 Since the release of OPM, various systems have been developed which implement OPM
136 recommendations, or export provenance data using this model. These systems can be classified into
137 two categories:

- 138 1) Specific systems with OPM import/export capabilities (e.g. Kepler/pPOD [32], Taverna
139 Provenance [33], Karma [34], VisTrails [35], and Swift [36]).
- 140 2) Generic OPM-compliant frameworks to manage provenance data (e.g. PLIER [37], *e-BioFlow*
141 [38], *Karma* [39], *OPMProv* [40], Trident workbench [41], and SPADE [42]).

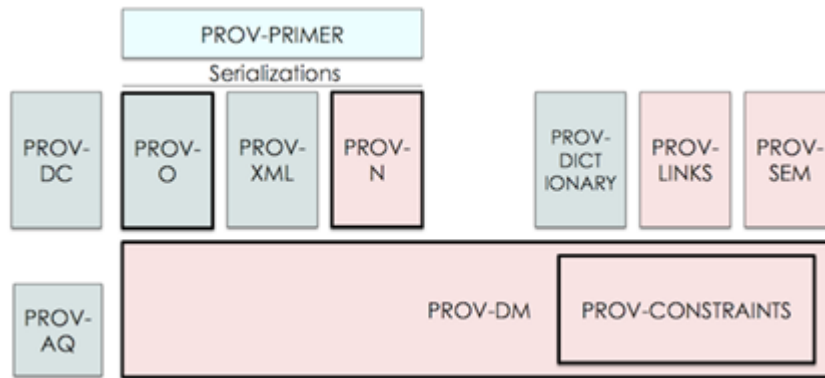
142 Many of these efforts shared a positive experience in using and deploying the OPM standard. In our
143 PLIER implementation [37], briefly described in section 3, we shared the similar positive experience,
144 although we outlined minor difficulties faced when implementing OPM or when making use of
145 provenance data. Some of the outlined difficulties were: (1) the ambiguity of some terms and their
146 usage (e.g. account, profile, and annotations), and (2) the improper design of some concepts (e.g.
147 Time, Properties, and Relations). As a result of these experiences, OPM has been revised and
148 improved since its release in 2007 by means of dedicated workshops, challenge series and community
149 discussions.

150 2.3 PROV: the new release of a Provenance Standard

151 A major revision to OPM has been published in April 2013 as a W3C standard, under the name of
152 PROV [7]. In a nutshell, PROV defines three core data types (*Entity*, *Activity*, and *Agent*); and
153 *Relations* between these data types. *Attributes* can be defined for data and relations, and a *Document*
154 aggregates them all.

155 PROV addresses most of the difficulties faced in OPM and provides a family of documents defining
156 various aspects that are necessary to better achieve the vision of interoperability of provenance
157 information in heterogeneous environments. PROV is conceived from a data modeling point of view
158 and takes into account existing technologies in the field of information representation and data
159 sharing. As such, it provides a set of classes, properties, and restrictions to model provenance
160 information using semantic web technologies such as OWL2 ontologies, XML, and Dublin Core
161 terms.

162 Figure 1 illustrates the organization of PROV components and the dependency between them. *PROV-*
163 *DM* is the core conceptual Data Model that defines a common vocabulary and concepts used to
164 describe provenance, to which a set of constraints apply as defined by *PROV-CONSTRAINTS* [7].
165 Other documents in the PROV family include the PROV OWL2 ontology to define the mapping of
166 the PROV data model to RDF (*PROV-O*); an XML schema for the PROV data model (*PROV-XML*);
167 a mapping between Dublin Core and PROV-O (*PROV-DC*); a declarative specification in terms of
168 first-order logic of the PROV data model (*PROV-SEM*); how to use Web-based mechanisms to
169 locate and retrieve provenance information (*PROV-AQ*); constructs for expressing the provenance of
170 dictionary style data structures (*PROV-DICTIONARY*); extensions to PROV to enable linking
171 provenance information across bundles of provenance descriptions (*PROV-LINKS*); and a human-
172 readable notation for the provenance model (*PROV-N*).



173
174 **Figure 1: Organization of PROV according to [7] showing the core conceptual data model (PROV-DM), the family of**
175 **documents it provides, and their dependencies. Bold bordered boxes denote W3C Recommendations, and regular**
176 **bordered boxes denote Working Group Notes. The colors classify the audience for each document, namely: Users,**
177 **Developers, and Advanced. Source: [7]**

178 The major improvements introduced in PROV, particularly the PROV family of documents, have
179 advanced the provenance standard to a level that attracted a large scientific community and increased
180 the number of efforts in adapting to, and implementing PROV. The latest PROV implementation
181 report, published in April 2013 [8], lists 66 implementations addressing PROV, classified into 5 types,
182 namely: application, framework/API, service, vocabulary, and constraints validator. Most of the
183 published implementations provide tools to convert and export between the different PROV families
184 of documents, mainly to PROV-O, PROV-N, PROV-XML, and PROV-JSON, while others provide
185 generic toolboxes and API frameworks for the management of provenance data. Nowadays, recent
186 developments in the scientific and engineering areas are enhancing their software tools with
187 provenance capabilities; examples include web semantics [43], data visualization [44], decision
188 making [45], scientific documentation [46], security controls [47], workflow systems [48] and many
189 others. The provenance data collection in these developments usually consumes a lot of time and
190 efforts. An out-of-shelf tool to help the developers of these applications collect and format the
191 provenance data according to the PROV standard would alleviate them from this error-prone task and
192 save their time and effort to better focus on the scientific applications.

193 The tools that are most related to our work are presented in [9,10]. Huynh et al. [9] provide ProvStore:
194 a web service to store, browse, visualize, share and manage provenance documents. ProvStore
195 expects the user to have the data already collected in a given format and provides no means to collect
196 the data. Flavio et al. describe in [10] RPOV-wf, a PROV-based database to provide runtime
197 provenance that can be queried even during the workflow execution. The approach collects runtime
198 provenance data from the various WfMS execution engines into the centric database.

199 To our knowledge, to date, none of these implementations provide a generic framework that is open
200 enough to be incorporated and deployed into scientific software tools and systems to facilitate the
201 capture of provenance in full-compliance with PROV.

202 **3 PROV-man: Design and Implementation**

203 This section presents the background of the design of *PROV-man*, which is the framework we
204 developed to facilitate the creation, storage, management and access to provenance data according to
205 the PROV standard recommendations. After presenting some background information, the approach
206 adopted for the data model optimization and the framework implementation are described.

207 **3.1 Background**

208 We have been involved in the design and implementation of a provenance framework for both OPM
209 and PROV. Our former implementation of provenance management was based on OPM and called
210 Provenance Layer Infrastructure for e-Science Resources – PLIER [37]. It was conceived based on an
211 optimal database schema to store provenance for scientific experiments that are performed using grid
212 workflow management systems. PLIER provides an API to record information about the steps of
213 experiments, their order, and the cause-and-effect reflecting linkage of inputs to output results.
214 Additionally, we enhanced PLIER with a set of tools to build, store, retrieve, share, and visualize
215 workflow experiments. PLIER has been extensively used to collect and explore provenance for
216 scientific experiments performed on a grid infrastructure, namely: (1) as an integrated component
217 within the WS-VLAM workflow system [49,50], and (2) as a core component to automatically gather
218 provenance data from existing grid workflow enactments services [51,52]. The results achieved by
219 deploying PLIER for tracing and analyzing the results of experiments motivated us to proceed with
220 the implementation of the provenance framework according to PROV.

OPM	PROV
Graph	Document
Artifact	Entity
Process	Activity
Causal Dependencies	Relations
Annotation & Property	Attributes
Account, Profile, OTime	N.A.

221 **Table 1: Relation between OPM and PROV concepts**

222 First, we conducted a study comparing PROV to OPM, based on the provenance specifications as
223 defined for OPM Core Specification (v1.1) and the latest PROV documentation [7]. Table 1
224 illustrates the main OPM concepts with their counterparts in the PROV specification. In more details:

- 225 • The concepts *Graph*, *Artifact*, *Process*, and *Causal Dependency* have been renamed to *Document*,
226 *Entity*, *Activity*, and *Relation*. These new terms are more suitable and representative in the
227 domain of data management.
- 228 • The concepts *Annotation* and *Property* have been refactored and simplified to *Attributes*, which
229 facilitates their use and deployment.
- 230 • The concepts *Account* and *Profile* are not present in PROV¹.

231 Other changes have been also introduced to the structure of the *Relation* and *Activity* concepts in
232 PROV, which make their representations more descriptive (e.g. by adding *Start Time* and *End Time*
233 for the Activity).

234 The main conclusion of our study is that the PROV modeling concepts are more appropriate than
235 their OPM counterparts. Particularly, the relationships concepts in PROV are conceived with rich
236 attributes, which provide comprehensive mechanisms to better describe the semantics of data.

¹ In our deployment of PLIER for collecting provenance data, we did not encounter effective usage for those concepts.

237 **3.2 PROV-man: The Approach**

238 The design and implementation of *PROV-man* follows the PROV recommendations and considers
239 these main design requirements:

- 240 1) To provide permanent storage of provenance data,
- 241 2) To optimize the database model considering data representation and querying,
- 242 3) To implement functions to facilitate access to provenance data,
- 243 4) To support data sharing via a set of utility functions for data conversion to various standard
244 formats,
- 245 5) To allow for easy deployment of the framework in various use cases.

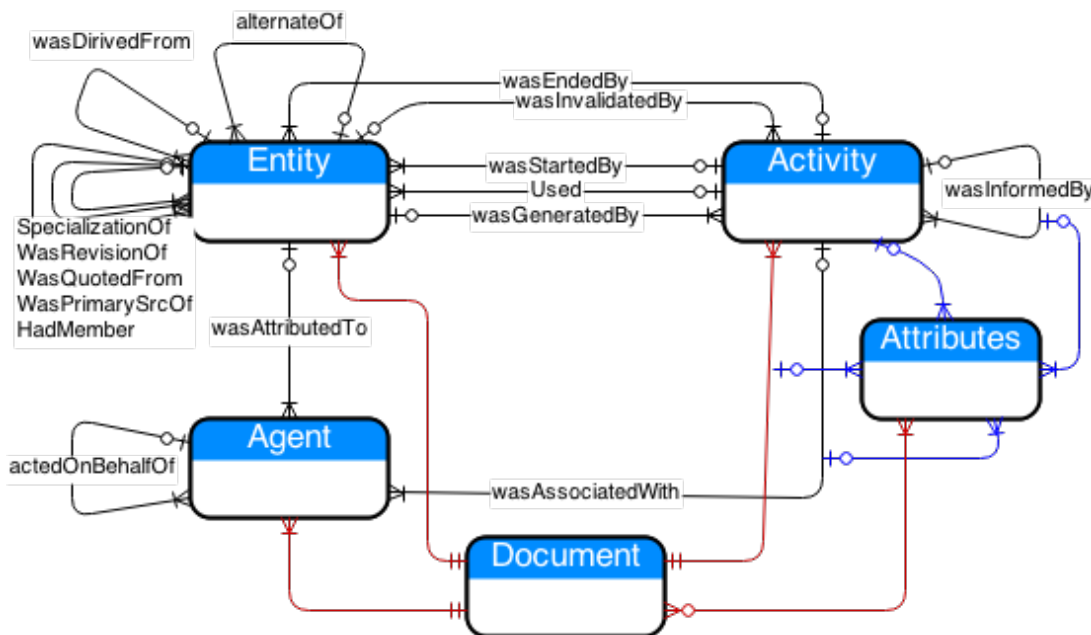
246 The main components of the framework consist of a database implementing the PROV-DM concepts
247 (section 3.3), and an API implementing the set of classes with methods and utility functions
248 (interfaces) to create and manipulate provenance data represented according to this model (section
249 3.4).

250 **3.3 PROV-man Optimized Data Model**

251 Data provenance is described in PROV by the use and production of *Entities* by *Activities*, which may
252 be influenced in various ways by *Agents*. *PROV-DM* is the core conceptual data model that defines a
253 common vocabulary and concepts used to describe provenance. In brief, PROV-DM consists of:

- 254 a) *Core data types* (*Entity*, *Activity*, and *Agent*);
- 255 b) A set of *Relations* between the core data types as defined in PROV (16 in total);
- 256 c) A set of *Attributes* that can be defined for each of the core data types and *Relations*, describing
257 their properties as key-value pairs; and
- 258 d) A *Document* grouping all the above.

259 Figure 2 illustrates a subset of the entity-relationship (ER) diagram of the PROV-DM core data types
260 and their *Relations*. Note that the complete ER diagram would be too complex to display because it
261 would include all optional *Attributes* that can be defined for the core data types and *Relations*.



262 **Figure 2: PROV-DM core data types with their prominent relationships. For readability reasons,**
263 **only a subset of the relationships to the Attributes (highlighted in blue) are presented.**
264

265 *Relations* in PROV-DM are always defined between the three core data types: *Entity*, *Activity*, and
 266 *Agent*. Their richness provides a strong mechanism to describe and express semantics of data. In
 267 addition, *Attributes* allow for further description of the core data types and their relationships. The
 268 strict implementation of this data model, without optimization, would however introduce difficulties
 269 for querying and maintaining the provenance data. For example, to retrieve the *Relations* for a given
 270 *Entity*, separate queries would be required for each of the 13 *Relations* defined for that *Entity*.
 271 Moreover, all the *Relations* have *Attributes*, for which separate tables would be also needed, thus
 272 making the data model even more complex. Thus, there is a need to optimize the data model to
 273 guarantee simplicity and high efficiency when querying the provenance data. The challenge here is to
 274 optimize the number of tables in PROV-DM, while preserving the full semantics and data richness of
 275 those relationships.

276 From a database design perspective, an optimization could be to model all the *Relations* using a
 277 single table. We demonstrate the optimization approach using the example in Figure 3, illustrating
 278 three of the 16 *Relations* defined in PROV-DM. As shown on this example, *Relations* are structurally
 279 similar to each other. For example, the relationships *used* and *wasGeneratedBy* are almost the same,
 280 except for the roles of the cause and effect, which are reversed (*Entity* and *Activity*). In the
 281 *actedOnBehalfOf* relationship, both cause and effect point to objects of the same data type (*Agent*),
 282 with an additional field *Activity* for which the delegation took place.

```

Definition of Relations
used(Identifier, Activity, Entity, Time, Attributes)
wasGeneratedBy(Identifier, Entity, Activity, Time, Attributes)
actedOnBehalfOf(Identifier; Agent, Agent, Activity, Attributes)

Examples of Relations creation
Entity (e1); Entity (e2); Activity (a1); Agent (ag1); Agent (ag2); // given
used ('r1', a1, e1, '23:09:2013 14:04', -); // activity a1 used entity e1 at '23:09:2013 14:04'
wasGeneratedBy ('r2', e2, a1, '24:09:2013', -); // entity e2 wasGeneratedBy activity a1 at '24:09:2013 10:04'
actedOnBehalfOf ('r3', ag2, ag1, a1, -); // agent ag2 actedOnBehalfOf agent ag1 for activity a1
  
```

283 **Figure 3: Examples illustrating three *Relations* expressed using PROV-N notation**

284 Therefore, we have chosen to model all PROV *Relations* using a single table:

285 ***Relation*** (*Identifier, RelationType, Cause, Effect, Time, Activity, Usage, Generation, Entity,*
 286 *Attributes*)

```

Definition of Relations
Relation (Identifier, RelationType, Cause, Effect, Time, Activity, Usage, Generation, Entity, Attributes)

Examples of Relations creation
Entity (e1); Entity (e2); Activity (a1); Agent (ag1); Agent (ag2); //given
Relation('r1', "Used", a1, e1, '23:09:2013 14:04', -, -, -, -);
Relation('r2', "wasGeneratedBy", e2, a1, '24:09:2013 10:04', -, -, -, -);
Relation('r3', "actedOnBehalfOf", ag2, ag1, -, a1, -, -, -);
  
```

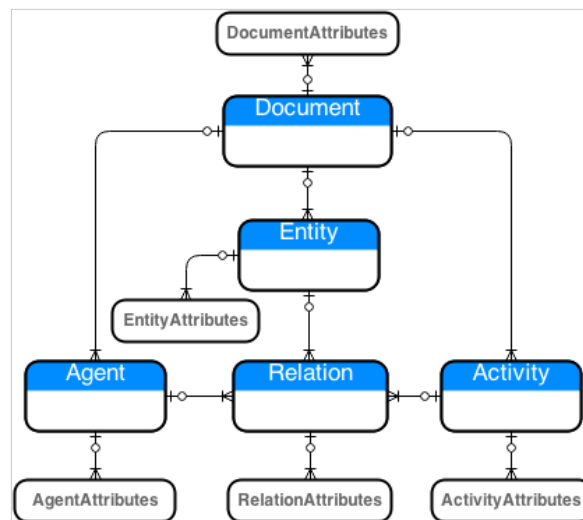
287 **Figure 4: Example of *Relations* from Figure 3 after optimization, using a single relationship that specifies the**
 288 ***RelationType*.**

289 The member *RelationType* plays the role of discriminator and ensures the preservation of the
 290 relationships semantics. Two keys (*Cause* and *Effect*) can point to a foreign key in one of the three

291 other tables (*Entity*, *Activity* and *Agent*). *Time*, *Activity*, *Usage*, *Generation* and *Entity* are optional
292 fields (see more details about these fields in [7]). Figure 4 illustrates how the class hierarchies of the
293 three PROV-DM relationships in Figure 3 are modeled using this optimized model.

294 This optimization approach can be applied to all the sixteen PROV-DM Relations, thus reducing the
295 number of relationships to a single table *Relation*. Consequently, the number of *Attributes* describing
296 the properties of the *Relations* will be also reduced to a single table *RelationAttributes*.

297 Figure 5 depicts the *PROV-man* data model in which the PROV-DM *Relations* are re-arranged in a
298 manner that reduces the model complexity and preserves PROV full semantics. A *Document* is made
299 of a set of *Entities*, *Activities*, and *Agents*; *Relations* may be established between the three core data
300 types; and each of the components can be further described using a set of *Attributes*.



301
302 **Figure 5: Optimized PROV-man data model.**

303 In *PROV-man* we dedicate special attention to the optimization of the underlying database schema, so
304 that it become simpler and more efficient for querying or storing provenance data, in case the scientist
305 needs/prefers direct access to the database. Still, direct access to the database is only suggested for
306 users with advanced database and PROV knowledge.

307 **3.4 PROV-man API implementation**

308 The *PROV-man* API provides an interface to create and manipulate provenance data according to the
309 PROV specifications. It preserves the semantics and richness defined by PROV and makes the
310 *PROV-man* data model transparent to the application developer. *PROV-man* software release and
311 documentation in are available in [53]. Figure 6 depicts the open-architecture of the *PROV-man*
312 framework, providing:

- 313 - A set of *classes* with methods to build and manipulate provenance data according to PROV
314 specifications;
- 315 - A set of *interfaces* implementing utility functions for provenance sharing and interoperation.
- 316 - A back-end database that serves as a main repository for storing provenance data, reflecting the
317 *PROV-man* data model presented in Figure 5; and

318 - Object-relational mapping (ORM) between the Java objects (classes) and the relational database.

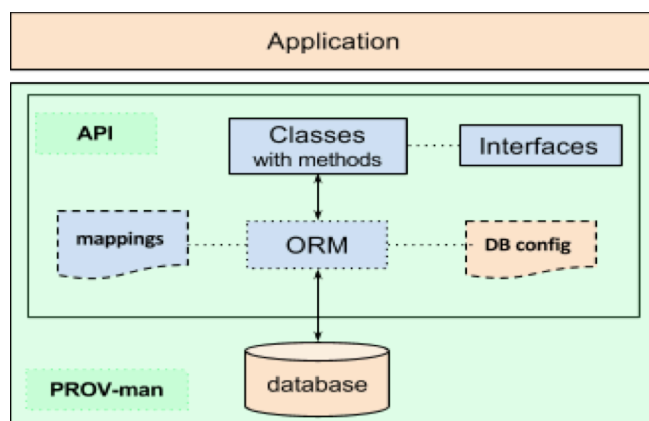


Figure 6: PROV-man architecture consisting of a database and an API. Components highlighted in brown denote the parts that can be controlled

319 The Java programming language has been selected to realize the implementation of the *PROV-man*
320 framework. In addition, ORM technology was used to implement the mapping between the relational
321 *PROV-man* data model and the Java object-oriented programming language. The choices and
322 motivations for selecting the technologies to implement the *PROV-man* framework are the following:

- 323 • A relational DBMS is used as back-end storage, which allows for remote and distributed access,
324 enforces data integrity, and serves as a distributed repository for provenance data. *PROV-man*
325 deploys an XML-configuration file to specify the underlying database with connection and
326 tuning parameters (e.g. database URL, user name and credentials, connection pool parameters,
327 and cache level) .
- 328 • Java was selected for the implementation of the *PROV-man*, due to its portability, platform
329 independency, and richness for modeling the provenance concepts and relationships. Provenance
330 data is created and consolidated as Java objects and then stored into the relational *PROV-man*
331 database.
- 332 • Hibernate [54] is used for the mapping between domain objects and relational database, which
333 permits to select a different DBMS if needed. It provides a smooth mapping between the Java
334 classes reflecting PROV-DM and the *PROV-man* optimized relational data model.

335 The *PROV-man* core API provides a set of 24 classes implementing the PROV-DM core data types,
336 their relationships, and attributes. Figure 7 illustrates an example of methods implemented for the
337 PROV-DM *Activity* class and Figure 8 illustrates methods for the PROV-DM *wasDerivedFrom* relation.
338 Figure 8 also illustrates that the naming of methods and parameter types are enforced accordingly to
339 the specification given by PROV-Constraints.

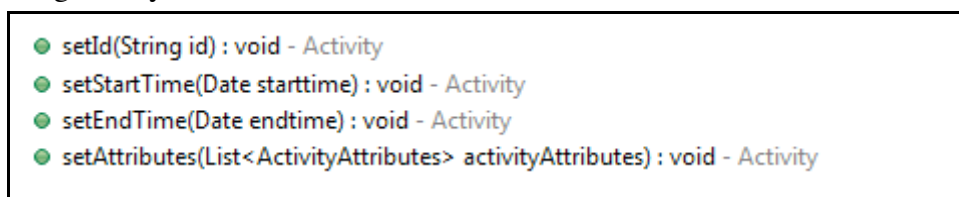


Figure 7: Methods implemented for *Activity*. Each method has parameters and returning value. Similarly, *get* methods exist to retrieve these values.

- setId(String id) : void - Relation
- setgeneratedEntity(Entity influencee) : void - WasDerivedFrom
- setUsedEntity(Entity influencer) : void - WasDerivedFrom
- setActivity(Activity activity) : void - WasDerivedFrom
- setGeneration(WasGeneratedBy generation) : void - WasDerivedFrom
- setUsage(Used usage) : void - WasDerivedFrom
- setAttributes(List<RelationAttributes> attributes) : void - Relation

Figure 8: Methods implemented for *wasDerivedFrom*. Each method has parameters and returning value. The terms in grey indicate whether the method is generic for all *Relation* types or specific to *wasDerivedFrom*

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344 To facilitate the creation of provenance data, *PROV-man* also provides a set of additional methods
345 following a human readable notation. These methods are provided under *PROVmanFactory* and
346 follow a syntax similar to PROV-N. Examples on the usage of the *PROV-man* methods and
347 interfaces are illustrated in section 4.1.

348 Finally, a set of interfaces cover serialization into formats of the PROV family of documents and
349 other formats:

- 350 - *toDB(document)*: maps the provenance *document* from its o-o representation to a relational
351 model, using ORM concepts, and stores it into the *PROV-man* database;
- 352 - *toXML(document, filePath)*: serializes the provenance *document* to the corresponding XML
353 representation, in compliance with the PROV XML schema;
- 354 - *toProvN(document, filePath)*: serializes the provenance *document* to the human-readable
355 notation of PROV-N;
- 356 - *toOWL2(document, filePath)*: serializes the provenance *document* to the corresponding Web
357 Ontology Language (OWL2-RL) representation;
- 358 - *toGraphviz(document, filePath)*: translates the provenance *document* to the Graphviz DOT
359 format [55];
- 360 - *toGraph(document, format, filePath)*: generates a graphical representation of the provenance
361 *document*, according to the specified format (e.g. png, jpg, gif, and pdf). This interface relies
362 on the Graphviz software [55], which supports most of the graphical output formats.

363 These interfaces take a generic and basic serialization approach that can be useful for getting started;
364 they are distributed as examples that possibly need to be customized for a particular application or
365 usage scenario.

366 **4 PROV-man Usage Examples**

367 The open architecture of the *PROV-man* framework, illustrated in Figure 6, allows for its flexible
368 integration into existing and newly developed software tools. The application layer can consist of
369 existing software (e.g. workflow systems or some data analysis tool) that deploys and integrates
370 *PROV-man* into its core implementation to store the fine-grained provenance details. *PROV-man* can
371 be used to build provenance extraction tools, for example, to gather provenance data from logs or
372 other information sources available for an application or system. *PROV-man* could be also deployed
373 in scenarios where multiple provenance tools/applications share the same *PROV-man* database by
374 using the same database configuration.

375 Below we present two usage examples: a simple case that illustrates the use of the set of methods and
376 interfaces provided by *PROV-man* in a stand-alone program, and a more complex case, which
377 demonstrates the deployment of *PROV-man* into a science gateway.

378 4.1 Simple Example: online newspaper article

379 Here we present and discuss the implementation of an online newspaper article described in the
380 PROV-PRIMER [56]. The newspaper publishes an article with a chart about crime statistics based
381 on existing data, with values composed (aggregated) by geographical regions. Different namespace

```
01: Document document = new Document();
02: Entity e1 = new Entity(); e1.setId("exg:DataSet1");
03: Entity e2 = new Entity(); e2.setId("exc:RegionList1");
04: document.getEntities().add(e1); document.getEntities().add(e2);
05: Entity e3 . . .
06: Activity act = new Activity(); act.setId("exc:Compose1");
07: document.getActivities().add(act); . . .
08: Activity act3 = new Activity(); . . .
09: ActivityAttributes Attr = new ActivityAttributes();
10: Attr.setId("Status"); Attr.setValue("Planned");
11: act.getAttributes().add(Attr);
12: document.getActivities().add(act3);
13: Agent agent = new Agent(); agent.setId("exc:derek");
14: document.getAgents().add(agent);
15: Agent agent2 = new Agent(); . . . . .
16: WasAssociatedWith waw = new WasAssociatedWith();
17: waw.setId("waw"); waw.setActivity(act2);
18: waw.setAgent(agent); waw.setPlan(e1);
19: document.getRelations().add(waw);
20: ActedOnBehalfOf abo = PROVmanFactory.ActedOnBehalfOf("abo", agent, agent2);
21: WasAttributedTo wat = PROVmanFactory.WasAttributedTo("wat", e4, agent);
22: document.getRelations().add(abo); document.getRelations().add(wat);
23: Used used = PROVmanFactory.Used("used", act, e1, "prov:role", "exc:dataToCompose");
24: document.getRelations().add(used); Used used2 . . .
25: WasGeneratedBy wgb= PROVmanFactory.WasGeneratedBy("wgb", e3, act, "prov:Role", "exc:composedData");
26: document.getRelations().add(wgb);
27: WasDerivedFrom wdf = PROVmanFactory.WasDerivedFrom("wdf", e4, e3, "prov:type", "prov:Revision");
28: document.getRelations().add(wdf);
29: PROVman.toDB(document);
30: PROVman.toXML(document, "/home/PROVman/doc/xml");
31: PROVman.toGraphviz(document, "/home/PROVman/doc/dot");
32: PROVman.toGraph(document, "png", "/home/PROVman/doc/png");
```

create provenance data objects

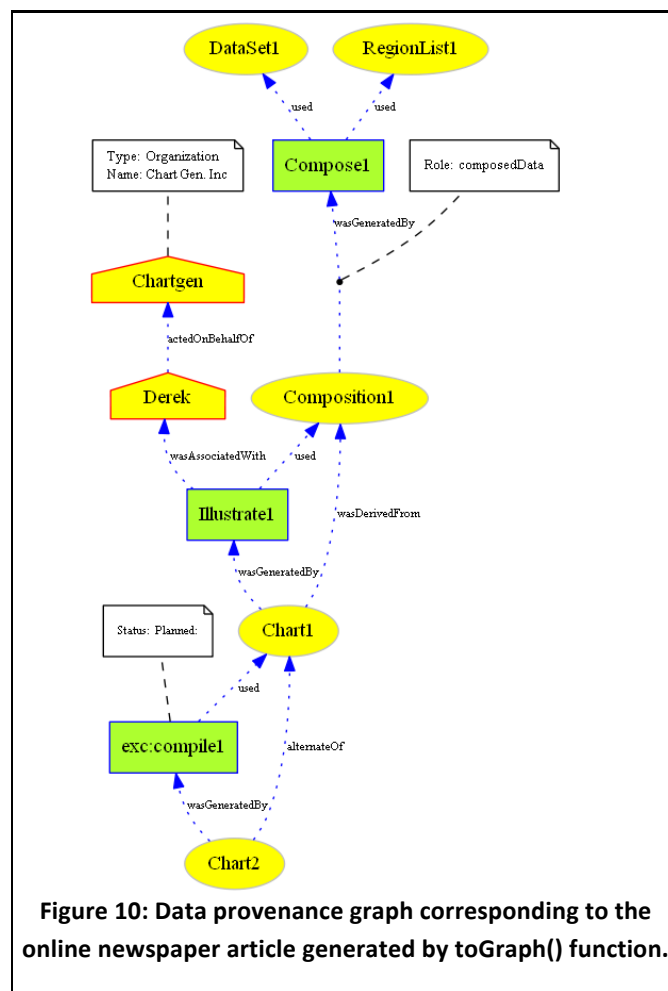
establish the link between data objects using relationships

Use of PROVmanFactory to simplify the creation of provenance data using syntax similar to PROV-N

Figure 9: Java sample code illustrating the use of PROV-man for creating and manipulating provenance data.

382 prefixes are used to identify the source creating the data and to distinguish between identifiers with
383 the same name used in these sources (e.g. **exb**, **exn**, **exc**, and **exg**). Figure 9 shows part of the Java code
384 to create data provenance. The complete code and the implementation details of this example are
385 available at the *PROV-man* release page [53].

386 Figure 9 also illustrates calls to the *PROV-man* interfaces for interoperability and data sharing (lines
387 29-32). The corresponding data provenance graph generated by the *toGraph()* function for the on-line
388 newspaper article is depicted in Figure 10.



389 4.2 Provenance of a science gateway

390 Here we demonstrate the deployment of *PROV-man* within an existing system, namely the AMC
 391 Neuroscience Gateway (NSG) [57]. This section briefly introduce the approach used to collect
 392 provenance using *PROV-man*. More details about the usage of the collected provenance data and the
 393 potential for their exploration can be found on [58].

394 The Neuroscience Gateway (NSG) is deployed at the Academic Medical Center (AMC) of the
 395 University of Amsterdam (UvA), The Netherlands. Its design is based on the WS-PGRADE/gUSE
 396 [59] scientific workflow management portal and framework, which supports various distributed
 397 computing infrastructures (DCIs). The gateway simplifies the usage of the Dutch e-Science Grid [60]
 398 for biomedical researchers by providing services such as community grid certificate and automatic
 399 file transport between the data servers and the grid resources. Workflows implemented using the WS-
 400 PGRADE/gUSE framework are the core of this platform. The workflows implement the data analysis
 401 tools for different applications (e.g. neuroscience and DNA sequencing). The users of the
 402 Neuroscience gateway are biomedical researches who perform data analysis tasks (coined
 403 *experiments*) by running these workflows on their data sets. Finally, the workflows are executed on
 404 the grid infrastructure by the WS-PGRADE/gUSE execution service, which does not have
 405 provenance capabilities yet.

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A provenance data collector was developed to gather provenance information about the scientific experiments performed using the Neuroscience gateway. For each workflow execution it collects data related to the jobs, their inputs and output results, users in charge of the experiments, and dependency relationships among these data. The collector follows a similar approach to our previous implementations [51, 52], deploying *PROV-man* to gather provenance information and organize it according to the experiment context. Figure 11 illustrates two use case scenarios of the provenance collector, namely gUSE/WS-PGRADE and Neuroscience gateway where detailed information about executed workflows are gathered from gUse and NSG databases, as well as from the log files generated by the jobs executed on the DCIs.

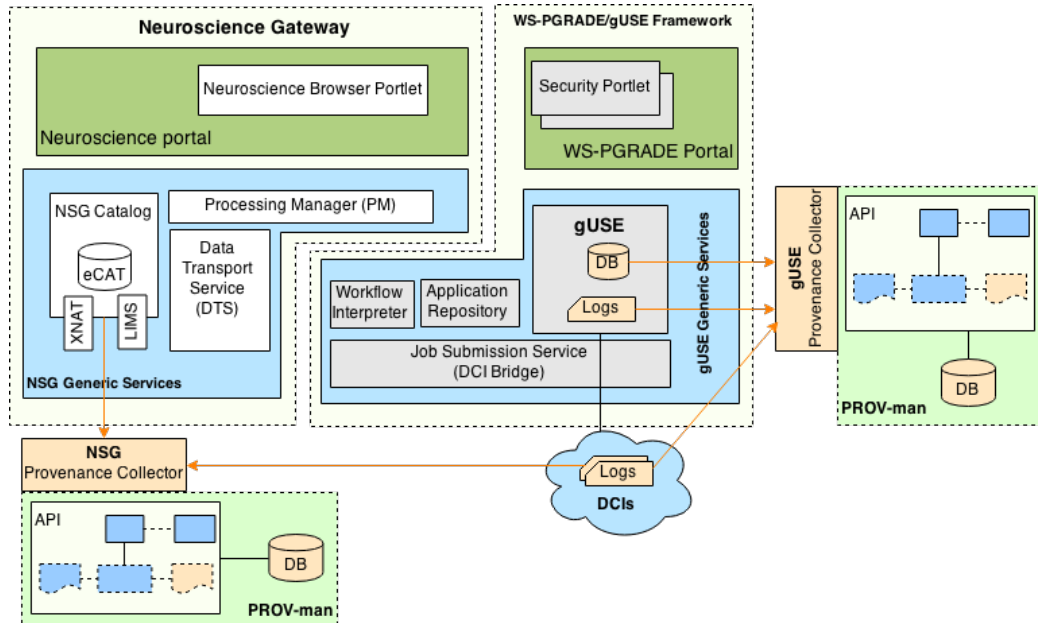


Figure 11: Architecture of the provenance data collector for the Neuroscience gateway. Only components related to provenance are depicted.

The mapping of workflows execution data to *PROV* concepts is straightforward for both use cases. Each workflow/experiment maps to a *Document* in the *PROV-man* database, jobs are mapped to *Activities*, input/output data to *Entities* and users are mapped to *Agents*. The most important *Relations* linking the input data to the output results in each experiment are *used* and *wasGeneratedBy*. Descriptive details documenting the properties of the core data types and relationships are mapped into the *PROV-man* database as *Attributes*, such as format, location, and size of input/output data; hostname of computing nodes where the jobs are executed; operating system on the computing nodes; the version of the software tools; etc.

Two main challenges were faced during the data collection and organization using *PROV-man*. The first relates to accessing the log files on the DCIs (Dutch Grid in our case), where the logs are only kept for a short period of time after the job execution. We therefore configured the provenance collector to be triggered as soon a workflow terminates execution. For this reason, for most workflows executed in the past it was not possible to collect details such as start and end time of jobs and computing nodes on which they run. Job start and end time are mapped as direct members of an *Activity*; however, the final status of a job had to be mapped as an *Attribute*.

433 The second challenge was to reconstruct the full dependencies between data and jobs in a workflow
434 from the various scattered information sources of gUse and grid job logs. In particular, various
435 operations are needed to correctly link all jobs to their proper input and output data in the context of
436 the workflow. The full dependencies were made possible by identifying the jobs that consume the
437 output generated by other jobs.

438 To completely avoid both challenges it would be more appropriate to instrument WS-PGRADE/gUse
439 directly to collect such data, following the approach presented in section 4.1.

440 Enhancing the Neuroscience Gateway with provenance capabilities enabled the automatic collection
441 of provenance information, whenever the scientists used the gateway to analyze and process their data.
442 Currently, the provenance data is used by administrators to generate experiments reports, to draw
443 their execution graphs, and to provide statistics about the executed experiments, used data analysis
444 tools, users in charges, experiments failure/success ratio, execution time, etc. Further exploration of
445 experiment provenance with interactive tools for end users is under development.

446 5 Discussion

447 The design and implementation of *PROV-man* can be discussed from different perspectives:
448 technology choices, data model optimization, performance, experiences in adopting the PROV
449 recommendations, and how the *PROV-man* approach fulfills the design requirements.

450 *Technology choices:* The choice of a relational DBMS as a back-end for the provenance framework,
451 in combination with Java and Hibernate, guarantees flexibility and openness of the system for
452 selecting the back-end storage. Currently Hibernate supports almost all the RDBMSs, including
453 ORACLE, DB2, MS SQL, MySQL, PostgreSQL, Sybase, Informix, and HSQL. The selection of Java
454 programming language limits the deployment of *PROV-man* into the core of existing software tools
455 (e.g. workflow systems) that are implemented in another language. In such cases, external data
456 collectors can be implemented using *PROV-man*, such as presented in section 4.2. To re-implement
457 *PROV-man* using another programming language, the developer has to select a proper ORM
458 technology, which requires re-designing part the proposed *PROV-man* data model to comply with the
459 chosen technology while keeping the optimizations proposed here. Another solution would be to
460 provide *PROV-man* as a service.

461 *Data model optimization:* By using Hibernate ORM constructs, all the PROV relationships could be
462 properly modeled as one *Relation*. We also tested other ORM technologies (namely, Castor JDO [61]
463 and datanucleus [62]), but it was not possible to reach such an optimized data model with them. In
464 our case, each *Relation* contains two foreign keys pointing to the primary keys in the associated core
465 data types; therefore, strict ER modeling would require different tables for each of the PROV
466 *Relations*. Using Hibernate, we were able to use a foreign key in the *Relation* table (*Cause* and *Effect*)
467 to reference to a primary key in more than one table, based on the type of the relationship (*Entity*,
468 *Activity*, *Agent*).

469 *Performance:* The deployment of PROV-man within the Neuroscience Gateway, presented in section
470 4.2, didn't present any performance issues while collecting provenance data related to more than
471 5000 experiments executed under WS-PGRADE/gUSE framework. The data collection was
472 performed after all experiments are finished or terminated, in such a scenario, the process takes few
473 milliseconds to a second per experiment. However, we didn't test the data collection in cases, where
474 the data is progressively collected during experiments execution, in such a scenario we assume that
475 some performance issues may occur in distributed environments involving large number of
476 experiments executed simultaneously.

477 *Experiences in adopting PROV:* With regard to the implementation of the PROV specifications, for
478 our application we noticed that minor modifications could enhance the readability of the standardized
479 provenance data. *Types* and *Roles of Agents* and *Relationships* are currently specified as key-value
480 pairs using *Attributes*; however, they are important elements for provenance of scientific experiments
481 and could be better modeled as direct members of these entities. This would make the PROV data
482 model more comprehensive. Similarly, a field *Status* could be added as a member to the *Activity* data
483 type, to indicate its final status (e.g. Done, Failed, Planned).

484 *Design Requirements:* With regard to the approach followed by *PROV-man*, we have shown in
485 section 4.2 the flexibility of the *PROV-man* framework and its easy deployment within an existing
486 application. However, it required detailed knowledge about the WS-PGRADE/gUSE framework to
487 identify the pieces of provenance data to be collected and linked according to their proper context.
488 The NSG case also illustrates the compliance of *PROV-man* with the design requirements, defined in
489 section 3.2, in terms of permanent storage of provenance data and support for data sharing using
490 utility functions.

491 **6 Conclusion**

492 In this paper we described the design and implementation of the *PROV-man* framework for
493 management of provenance data. *PROV-man* implements the provenance standard in compliance with
494 the PROV-Constraints and according to the PROV specifications [7]. It has been released as a library
495 that can be directly used from Java applications. To our knowledge, this work is the first to describe a
496 framework to facilitate the capture and storage of PROV-compliant provenance data from generic
497 scientific applications

498 *PROV-man* provides methods to create and manipulate provenance data in a consistent manner and
499 ensures the permanent storage of provenance data into a relational database that can be configured
500 and tuned for each application. A set of basic interfaces are provided to serialize and export the
501 provenance data to various data formats. These interfaces can be enhanced with new methods,
502 whenever needed, to better serve the interoperation with emerging applications and eventually, to
503 provide data representation for the PROV family of documents (e.g. PROV-DC, and PROV-LINKS).
504 The open architecture of *PROV-man*, consisting of an API and a configurable database, allows for its
505 straightforward deployment within other software tools to enable or enhance their provenance
506 capabilities. By deploying *PROV-man*, applications can more easily benefit from the advantages of
507 the PROV standard for provenance interoperability.

508 For example, collaboration project is planned with the developers of WS-PGRADE/gUSE [59] and
509 WSVLAM [63] workflow management systems to implement provenance into their core software
510 using *PROV-man*. The granularity of the provenance data to be collected has to be specified, and, a
511 mapping needs to be defined between workflow and *PROV* concepts. The deployment of *PROV-man*
512 within the workflow management systems will enable the automatic collection of provenance
513 information in interoperable format, whenever scientists use the platform to analyze and process their
514 data.

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520 **7 References**

- 521 1. J. Myers, C. Pancerella, C. Lansing, K. Schuchardt, and B. Didier, "Multi-Scale Science, Supporting
522 Emerging Practice with Semantically Derived Provenance," in ISWC workshop on Semantic Web
523 Technologies for Searching and Retrieving Scientific Data, 2003.
- 524 2. J. Widom, "Trio: A System for Integrated Management of Data, Accuracy, and Lineage," in CIDR, 2005.
- 525 3. J. Zhao, C. A. Goble, R. Stevens, and S. Bechhofer, "Semantically Linking and Browsing Provenance Logs
526 for Escience," in ICSNW, 2004.
- 527 4. Yogesh L. Simmhan, Beth Plale, and Dennis Gannon. A survey of data provenance in e-science. SIGMOD
528 Rec., 34(3):31–36, 2005.
- 529 5. R. Bose and J. Frew, "Lineage retrieval for scientific data processing: a survey," in ACM Comput. Surv., vol.
530 37, 2005.
- 531 6. L. Moreau, et al. The Open Provenance Model Core Specification (v1.1). Future Generation Computer
532 Systems, vol. 27(6) pp.743-756, June 2011.
- 533 7. PROV-Overview: <http://www.w3.org/TR/2013/NOTE-prov-overview-20130430/>
- 534 8. PROV Implementation Report: <http://www.w3.org/TR/prov-implementations>
- 535 9. Huynh, Trung Dong and Moreau, Luc (2014) ProvStore: a public provenance repository. In Proceedings
536 of *5th International Provenance and Annotation Workshop (IPAW'14)*, Cologne, Germany, 09 - 13 Jun
537 2014.
- 538 10. Flavio Costa, Vítor Silva, Daniel de Oliveira, Kary A. C. S. Ocaña, Eduardo S. Ogasawara, Jonas Dias, Marta
539 Mattoso: Capturing and querying workflow runtime provenance with PROV: a practical
540 approach. EDBT/ICDT Workshops 2013: 282-289
- 541 11. Provenance Wikipedia: <http://en.wikipedia.org/wiki/Provenance>
- 542 12. Peter Buneman, Sanjeev Khanna, and Wang chiew Tan. Why and where: A characterization of data
543 provenance. In In ICDT, pages 316–330. Springer, 2001.
- 544 13. DICOM - Digital Imaging and Communications in Medicine: <http://dicom.nema.org>
- 545 14. LIMS: http://en.wikipedia.org/wiki/Laboratory_information_management_system
- 546 15. Michael H. Elliott, "Electronic Laboratory Notebooks Enter Mainstream Informatics," Scientific
547 Computing, November 2008
- 548 16. J. Lyle and A. Martin. Trusted computing and provenance: better together. In Proceedings of TAPP 2010,
549 Berkeley, CA, USA, 2010. USENIX Association.
- 550 17. J. Frew and R. Bose, "Earth System Science Workbench: A Data Management Infrastructure for Earth
551 Science Products," in SSDBM, 2001.
- 552 18. Tinga Provenance Service: <http://www.tingatech.com>
- 553 19. M. Greenwood, C. Goble, R. Stevens, J. Zhao, M. Addis, D. Marvin, L. Moreau, and T. Oinn, "Provenance
554 of e-Science Experiments - experience from Bioinformatics," in Proceedings of the UK OST e-Science 2nd
555 AHM, 2003.
- 556 20. Peter Buneman, Adriane Chapman, and James Cheney. Provenance management in curated databases.
557 In SIGMOD '06: Proceedings of the 2006 ACM SIGMOD International conference on Management of data,
558 pages 539–550, New York, NY, USA, 2006. ACM.
- 559 21. I. T. Foster, J.-S. Vöckler, M. Wilde, and Y. Zhao, "Chimera: A Virtual Data System for Representing,
560 Querying, and Automating Data Derivation," in SSDBM, 2002.
- 561 22. Davidson, S.B., Freire, J.: Provenance and scientific workflows: challenges and opportunities. In: SIGMOD
562 Conference, pp. 1345–1350 (2008)
- 563 23. Gil, Y., Deelman, E., Ellisman, M., Fahringer, T., Fox, G., Gannon, D., Goble, C., Livny, M., Moreau, L.,
564 Myers, J.: Examining the challenges of scientific workflows. IEEE Computer 40(12), 26–34 (2007)

- 565 24. Jihie Kim, Ewa Deelman, Yolanda Gil, Gaurang Mehta, and Varun Ratnakar. Provenance trails in the
566 wings-pegasus system. *Concurr. Comput. : Pract. Exper.*, 20(5):587–597, 2008.
- 567 25. Bertram Ludascher, Ilkay Altintas, Chad Berkley, Dan Higgins, Efrat Jaeger, Matthew Jones, Edward A.
568 Lee, Jing Tao, and Yang Zhao. Scientific workflow management and the kepler system: Research articles.
569 *Concurr. Comput. : Pract. Exper.*, 18(10):1039–1065, 2006.
- 570 26. P. Missier, K. Belhajjame, J. Zhao, and C. Goble, Data lineage model for Taverna workflows with
571 lightweight annotation requirements, In *Proc. of the International Provenance and Annotation*
572 *Workshop (IPAW)*, 2008.
- 573 27. Moreau, L., Foster, I. (eds.): *IPAW 2006*. LNCS, vol. 4145. Springer, Heidelberg (2006)
- 574 28. The Provenance Challenge Wiki (June 2006), <http://twiki.ipaw.info/bin/view/Challenge>
- 575 29. Miles, S.: Technical summary of the second provenance challenge workshop, King’s College (July 2007),
576 <http://twiki.ipaw.info/bin/view/Challenge/SecondWorkshopMinutes>
- 577 30. The Open Provenance Model Luc Moreau, University of Southampton, Juliana Freire, University of Utah,
578 Joe Futrelle, NCSA, Robert E. McGrath, NCSA Jim Myers, NCSA, Patrick Paulson, PNNL December 18,
579 2007
- 580 31. Luc Moreau, Juliana Freire, Joe Futrelle, Robert E. McGrath, Jim Myers, and Patrick Paulson. The Open
581 Provenance Model: An Overview. J. Freire, D. Koop, and L. Moreau (Eds.): *IPAW 2008*, LNCS 5272, pp.
582 323–326, 2008. © Springer-Verlag Berlin Heidelberg 2008
- 583 32. Shawn Bowers, Timothy McPhillips, Sean Riddle, Manish Kumar Anand, Bertram Ludäscher. Kepler/pPOD:
584 Scientific Workflow and Provenance Support for Assembling the Tree of Life. *Lecture Notes in Computer*
585 *Science Volume 5272*, 2008, pp 70-77
- 586 33. Paolo Missier, Satya Sahoo, Jun Zhao, Carole Goble, Amit Sheth. Janus: from workflows to semantic
587 provenance and linked open data: *Lecture Notes in Computer Science*, Vol. 6378/2010 (2010), pp. 129-
588 141 Key: citeulike:10019128
- 589 34. Y. Simmhan, B. Plale, and D. Gannon, Karma2: Provenance Management for Data Driven Workflows,
590 *International Journal of Web Services Research*, 5(2):1-22, 2008.
- 591 35. C. Silva, J. Freire, and S. Callahan, Provenance for Visualizations: Reproducibility and Beyond, *IEEE*
592 *Computing in Science and Engineering*, 9(5):82-29, 2007.
- 593 36. Y. Zhao, M. Hategan, B. Cliord, I. Foster, G. vonLaszewski, I. Raicu, T. Stef-Praun, and M. Wilde, Swift:
594 Fast, Reliable, Loosely Coupled Parallel Computation, In *Proc. of the International Workshop on Scientific*
595 *Workflows (SWF)*, pages 199-206, 2007.
- 596 37. PLIER - Provenance Layer Infrastructure for e-Science Resources:
597 <http://twiki.ipaw.info/bin/view/OPM/Plier>
- 598 38. I. Wassink, Matthijs Ooms, P. Neerincx, G. van der Veer, Han Rauwerda, Jack A. M. Leunissen, T. M. Breit,
599 A. Nijholt, P. van der Vet. (2010) *e-BioFlow: improving practical use of workflow systems in bioinformatics*.
600 In: *Information Technology in Bio- and Medical Informatics, ITBAM 2010*, Sept 1-2, 2010, Bilbao, Spain.
- 601 39. Karma provenance collection toolkit: http://d2i.indiana.edu/provenance_karma
- 602 40. Chunhyeok Lim , Shiyong Lu , Artem Chebotko , Farshad Fotouhi, Storing, reasoning, and querying OPM-
603 compliant scientific workflow provenance using relational databases, *Future Generation Computer*
604 *Systems*, v.27 n.6, p.781-789, June, 2011.
- 605 41. Yogesh Simmhan, Roger Barga .Analysis of approaches for supporting the Open Provenance Model: A case
606 study of the Trident workflow workbench Published in: *Journal Future Generation Computer Systems*
607 *archive Volume 27 Issue 6*, June, 2011. Pages 790-796
- 608 42. Ashish Gehani and Dawood Tariq, SPADE: Support for Provenance Auditing in Distributed Environments,
609 *13th ACM/IFIP/USENIX International Conference on Middleware*, 2012.
- 610 43. Rinke Hoekstra and Paul Groth. Linkitup: Link discovery for research data. In *Discovery Informatics: AI*
611 *Takes a Science-Centered View on Big Data*, AAAI Fall Symposium Series, 2013.

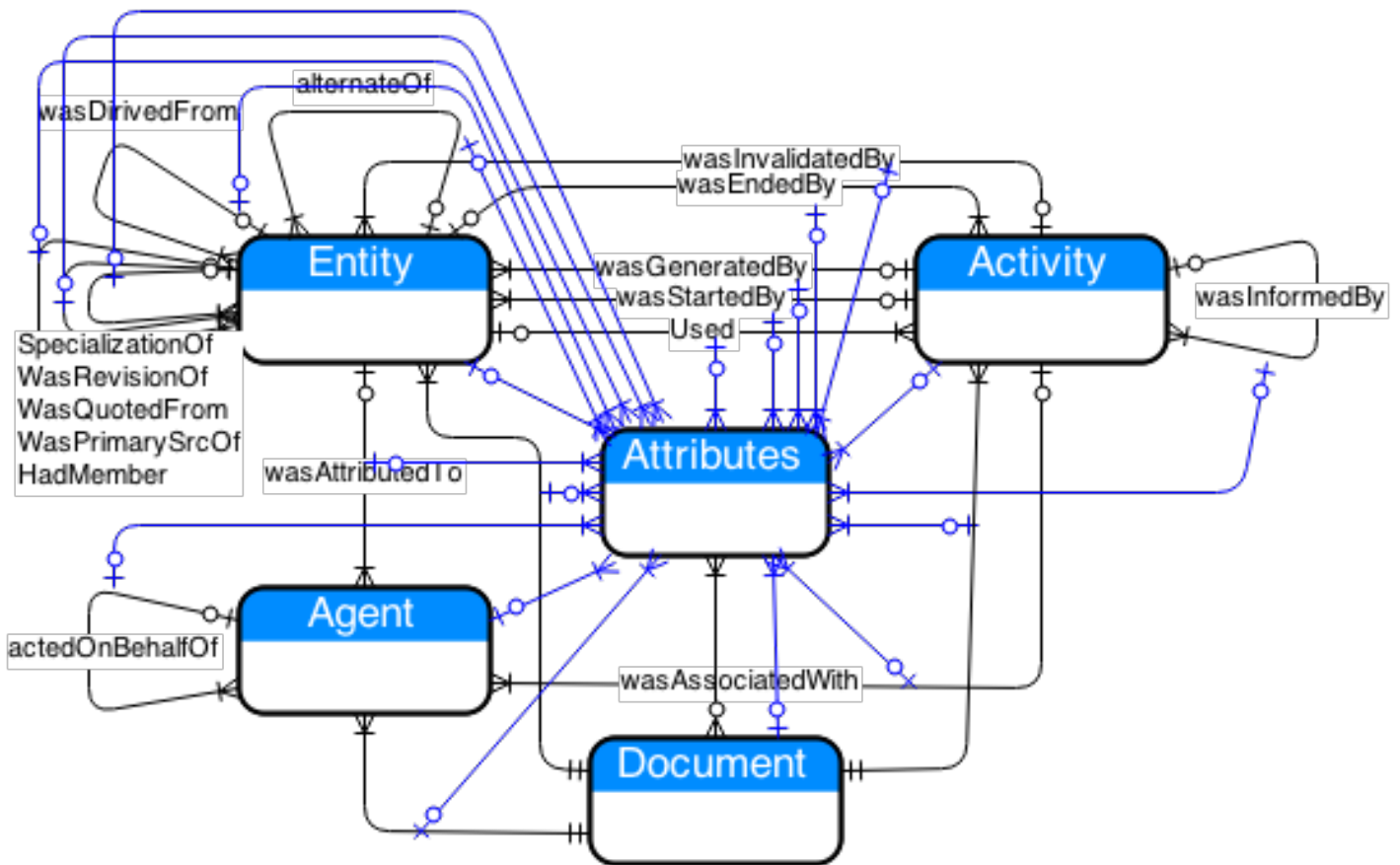
- 612 44. Hoekstra, R. and Groth, P. PROV-O-Viz - Understanding the Role of Activities in Provenance. In
613 Proceedings of *5th International Provenance and Annotation Workshop (IPAW'14)*, Cologne,
614 Germany, 09 - 13 Jun 2014.
- 615 45. Amir Sezavar Keshavarz, Trung Dong Huynh and Luc Moreau. Provenance for Online Decision Making. In
616 Proceedings of *5th International Provenance and Annotation Workshop (IPAW'14)*, Cologne,
617 Germany, 09 - 13 Jun 2014.
- 618 46. Adiando Wibisono, Peter Bloem, Gerben De Vries, Paul Groth, Adam Belloum and M. Generating Scientific
619 Documentation for Computational Experiments Using Provenance. In Proceedings of *5th International
620 Provenance and Annotation Workshop (IPAW'14)*, Cologne, Germany, 09 - 13 Jun 2014.
- 621 47. Luiz Gadelha and Marta Mattoso. Applying Provenance to Protect Attribution in Distributed
622 Computational Scientific Experiments. In Proceedings of *5th International Provenance and Annotation
623 Workshop (IPAW'14)*, Cologne, Germany, 09 - 13 Jun 2014.
- 624 48. Wellington Oliveira, Daniel de Oliveira, Vanessa Braganholo. Experiencing PROV-Wf for Provenance
625 Interoperability in SWfMSs. In Proceedings of *5th International Provenance and Annotation Workshop
626 (IPAW'14)*, Cologne, Germany, 09 - 13 Jun 2014.
- 627 49. Michael Gerhards, Sascha Skorupa, Volker Sander, Adam Belloum, Dmitry Vasunin, A. Benabdelkader.
628 Hist/PLIER: A two-fold Provenance Approach for Grid-enabled Scientific Workflows using WS-VLAM. In
629 *the 12th IEEE/ACM International Conference on Grid Computing, 22-23 September 2011*, Lyon, France,
630 2011. ICGC 2011.
- 631 50. Michael Gerhards, Sascha Skorupa, Volker Sander, Adam Belloum, Dmitry Vasunin, A. Benabdelkader.
632 Provenance Opportunities for WS-VLAM: An Exploration of an e-Science and an e-Business Approach.
633 Submitted to *the 6th Workshop on Workflows in Support of Large-Scale Science, November 12-18, 2011*,
634 Seattle, 2011. - WSLSS 2011
- 635 51. A. Benabdelkader, M. Santcroos, S. Madougou, A. H. van Kampen, S. Olabarriaga. A Provenance approach
636 to trace scientific experiments on a grid infrastructure. In *the 7th IEEE International Conference on e-
637 Science, 05-08 December 2011*, Stockholm, Sweden, 2011: 134-141. - e-science 2011
- 638 52. Souley Madougou, Shayan Shahand, Mark Santcroos, Barbera D. C. van Schaik, Ammar Benabdelkader,
639 Antoine H. C. van Kampen, Sílvia Delgado Olabarriaga: Characterizing workflow-based activity on a
640 production e-infrastructure using provenance data. *Future Generation Comp. Syst.* 29(8): 1931-1942
641 (2013) - FGCS 2013
- 642 53. PROV-man software release: <http://www.sharp-sys.nl/PROV-man.html>
- 643 54. G. King, C. Bauer, *"Java Persistence with Hibernate (Second ed.)"*, "Manning Publications, pp. 880, ISBN
644 1932394885, November 2006.
- 645 55. Graph Visualization Software – Graphviz: www.graphviz.org
- 646 56. PROV Model Primer: <http://www.w3.org/TR/prov-primer/>
- 647 57. Shahand S, Benabdelkader A, Jaghoori MM, al Mourabit M, Huguet J, Caan MWA, van Kampen AHC,
648 Olabarriaga SD. A data-centric neuroscience gateway: design, implementation, and experiences. *Journal
649 of Concurrency and Computation: Practice and Experience*, 27 (2):pp. 489-506, 2015
- 650 58. Benabdelkader et al, Collection of provenance data from grid workflow execution using WS-
651 PGRADE/gUse. (initiative <https://groups.google.com/forum/#!forum/prov4guse>)
- 652 59. Kacsuk et al., "WS-PGRADE/gUSE Generic DCI Gateway Frame-work for a Large Variety of User
653 Communities," *Journal of Grid Computing*, vol. 10, no. 4, pp. 601–630, 2012
- 654 60. The SURFsara website, <https://www.surfsara.nl>
- 655 61. Castor 1.3.1 - release and documentation. <http://castor.codehaus.org>
- 656 62. datanucleus open project: <http://www.datanucleus.org>

657 63. V. Korkhov, D. Vasyunin, A. Wibisono V. Guevara-Masis, A. Belloum “WS-VLAM: Towards a Scalable
 658 Workflow System on the Grid” Workshop on workflows in Support of Large-Scale Science (WORKS 07); In
 659 conjunction with HPDC 2007; Monterey Bay, June 2007.
 660 64. COMMIT Project: <http://www.commit-nl.nl>
 661 65. SCI-BUS - SCientific gateway Based User Support: <http://www.sci-bus.eu>

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663 **5 Supplementary Material:**

664 **5.1 PROV Data Model: Complete ER Schema**



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Figure 12: PROV-DM core data types with their complete set of relationships.