


# Towards Improved Findability of Energy Research Software by Introducing a Metadata-based Registry

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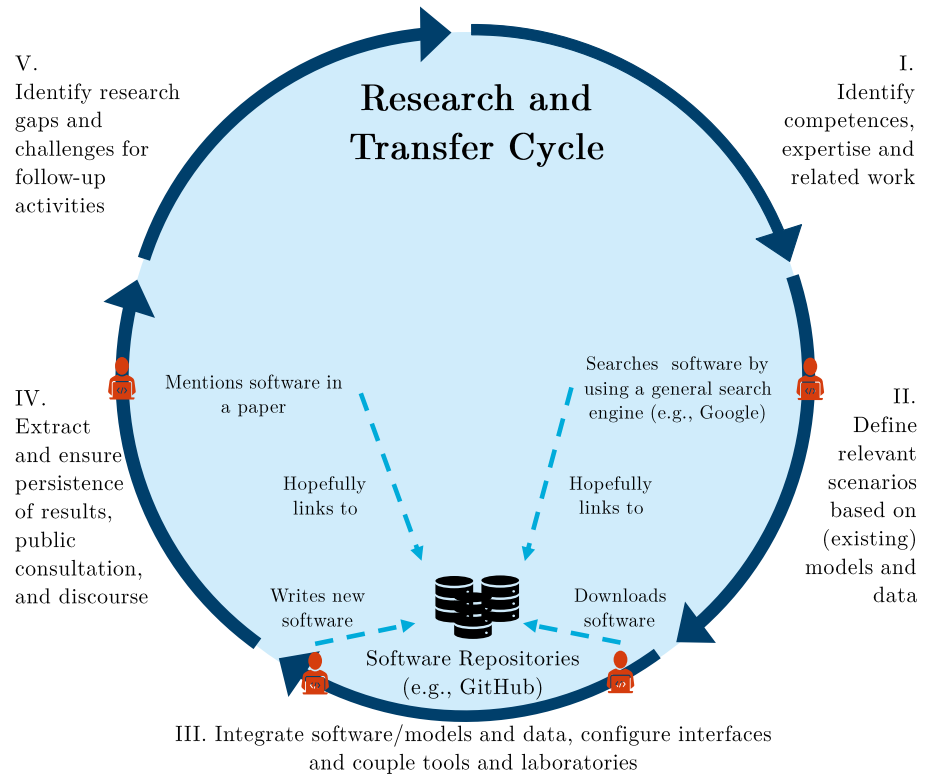
**Abstract.** Research software in the energy domain becomes increasingly important for the analysis, simulation, and optimization of energy systems and supports design decisions in the required transition of energy systems to tackle the climate crisis. To make energy research software (ERS) more findable, it should be described with metadata following the FAIR (findable, accessible, interoperable, and reusable) criteria and be registered in a common registry. To this end, we present a concept for a metadata-based registry for ERS which should enable researchers to easily add new ERS as well as to find new ERS.

## 1 Motivation

2 In energy research, self-designed software is a basic tool for multiple purposes like visualization  
3 of processes and values, e.g., power quality [1], (co-)simulation of smart grids [2], or analysis  
4 of transition paths [3]. Within an exemplified research cycle, this self-designed software is often  
5 a starting point and, therefore, fundamental for producing new research results while it also  
6 presents a result of performed research (see Figure 1).

7 Based on the definition of research software by Hasselbring et al. [4], we define energy research  
8 software (ERS) as software that is employed in the scientific discovery process to understand,  
9 analyze, improve, and/or design energy systems. With respect to the complexity of the software,  
10 ERS ranges from simple scripts and libraries, e.g., for python, up to full software solutions.  
11 Content-wise, it can for example visualize, analyze, and/or generate (artificial) data from energy  
12 (sub)components or grids in laboratories or the real world. Alternatively, it can also represent as  
13 a model particular energy (sub)components, energy (distribution) systems, and transition paths  
14 of energy use, distribution, conversion, and/or generation to analyze the design and/or control in  
15 simulations and optimizations.

16 Multiple models and frameworks in the energy domain have been developed in the last years.  
17 Partly, these have overlapping and similar features. Often new tools are developed without  
18 reusing the already existing ones. Due to the need for interdisciplinary research in the energy  
19 domain as well as the growing number of simulated components ERS will even become more  
20 complex in the upcoming years [5]. Therefore, a lot of time is spent on (re)developing software  
21 instead of doing research slowing down the progress in research.



**Figure 1:** An exemplified research cycle of energy research with software usage

22 Different approaches to formulate FAIR criteria for research software show that metadata and  
 23 repositories for these metadata, e.g., software registries, are key elements for FAIR research  
 24 software [4], [6], [7].

25 Especially the findability of ERS can be increased by describing it with useful metadata and  
 26 including it into a registry. Table 1 gives some examples for possible metadata elements with  
 27 metadata for the co-simulation framework mosaik<sup>1</sup> as example instance. Together, good metadata  
 28 and a registry are a first step for increasing the reuse of ERS and improving the research process  
 29 in energy research.

30 For this goal, we propose a metadata-based registry for ERS based on a good metadata scheme.  
 31 Our contribution are:

- 32 • We outline related work in the fields of metadata schemes for research software, metadata  
 33 schemes in the field of energy research, and ontologies in the energy domain in Section 2.
- 34 • We introduce our concept of metadata-based registry for ERS in Section 3.
- 35 • We give an outlook of the further required work in Section 4.

1. <http://mosaik.offis.de/>, accessed 12.12.2022

**Table 1:** Examples for possible metadata elements based on the FAIR criteria with the software mosaik as example instance

Findability	Name	mosaik <sup>1</sup>
	Identifier	-
	Version	3.0
Accessibility	Link to repository	<a href="https://gitlab.com/mosaik/mosaik">https://gitlab.com/mosaik/mosaik</a>
Interoperability	Programming language	python
	Input/output formats	depends on used models
	Dependencies	-
Reusability	License	LGPL v3.0
	Link to documentation	<a href="https://mosaik.readthedocs.io/en/latest/quickstart.html">https://mosaik.readthedocs.io/en/latest/quickstart.html</a>

## 36 2 Related Work

37 Within this section, we give an overview on existing metadata schemes. First, we present current  
 38 metadata schemes for research software in general or in other specific domains in [Section 2.1](#).  
 39 Second, we introduce metadata schemes in the field of energy research and engineering in  
 40 [Section 2.2](#). These are often designed for research data but can be used as foundation for a  
 41 metadata scheme for ERS.

42 Since value vocabularies are an important part of metadata schemes, the third part ([Section 2.3](#))  
 43 outlines ontologies in the energy domain.

### 44 2.1 Metadata for Research Software

45 This section gives an overview of existing approaches for metadata schemes for research software.  
 46 While some publications directly focus on metadata, others only introduce software ontologies,  
 47 which can be used as metadata vocabulary for research software.

48 CodeMeta<sup>2</sup> is a community driven metadata standard for research software based on [schema.org](https://schema.org)<sup>3</sup>.  
 49 Various crosswalks to other metadata schemes already exist. CodeMeta contains multiple  
 50 elements, some focusing on technical details like file size or supported operating systems and  
 51 others including administrative information like license. The metadata standard does not have  
 52 mandatory elements. It supports the use of URIs for authors and contributors as well as for  
 53 licenses. The content specific metadata are limited to an application category and keywords.

54 For geosciences, Gil et al. introduced an ontology to describe research software, OntoSoft, with  
 55 six categories: identify, understand, execute, do research, get support, and update [8]. They  
 56 also developed an automated extraction tool for metadata and provided a registry for software  
 57 metadata<sup>4</sup>. Garijo et al. expanded this approach by developing the Software Description  
 58 Ontology<sup>5</sup> [9] with additional description for input and output data based on the Scientific  
 59 Variables Ontology<sup>6</sup>. Also, they aligned their approach with CodeMeta and enabled publishing  
 60 the metadata into an open knowledge graph including links to additional instances in the semantic

2. <https://codemeta.github.io/>, accessed 12.12.2022

3. <https://schema.org/>, accessed 12.12.2022

4. <https://csdms.ontosoft.org/#home>, accessed 12.12.2022

5. <https://w3id.org/okn/o/sd>, accessed 12.12.2022

6. <https://scientificvariablesontology.org/svo/>, accessed 12.12.2022

**Table 2:** Overview of metadata scheme for research software

✓: fulfilled                      ✗: not fulfilled  
 (✓): (partly) fulfilled

	Metadata scheme (s) vs. ontology (o)	Domain	Mandatory elements	Support for URIs	Use of domain ontologies as value vocabulary
CodeMeta <sup>2</sup>	s	General	✗	(✓)	✗
OntoSoft [8]	o	Geoscience	✗	✗	✗
Software Description Ontology [9]	o	Geoscience	✗	✓	✓
Software Ontology [10]	o	Bioinformatics	✗	✓	✓
biotoolsXSD [11]	s	Bioinformatics	✓	(✓)	✓

61 web like wikidata<sup>7</sup>. Additionally, they developed software to support researchers in the metadata  
 62 creation and to find models<sup>8</sup>.

63 In the domain of bioinformatics, the EDAM ontology describes general concepts. The Software  
 64 Ontology (SWO) was developed to extend the EDAM ontology to describe software in this  
 65 research field [10]. The SWO includes licenses, programming languages, and data formats  
 66 as taxonomies. In contrast to OntoSoft, the use of the taxonomies improves the usability for  
 67 semantic web applications and linking. Also for bioinformatics, Ison et al. developed the  
 68 metadata scheme biotoolsXSD for the software registry bio.tools<sup>9</sup> [11]. The metadata scheme  
 69 is expressed as an XML scheme containing 55 elements of which 10 are mandatory. The use  
 70 of the EDAM ontology as value vocabulary is required for some elements like function, input,  
 71 and output. The metadata scheme also contains software specific elements like programming  
 72 language, license, and operating system for which the use of an ontology is not required.

73 Table 2 gives an overview of the diverse outlined metadata schemes and ontologies. While some  
 74 metadata schemes are less extensive, like CodeMeta<sup>2</sup>, others try to include value vocabularies to  
 75 improve interoperability by using semantic web technologies. Some schemes include detailed  
 76 domain knowledge based on domain ontologies like biotoolsXSD [11] or the Software Description  
 77 Ontology [9].

## 78 2.2 Metadata in Engineering and the Energy Domain

79 In this section, metadata schemes for data and software in the engineering domain and especially  
 80 in the energy domain are introduced, starting from the broad engineering perspective.

81 Schembera and Iglezakis developed the metadata scheme EngMeta for data in computational  
 82 engineering which includes existing elements for technical and general descriptive information  
 83 from DataCite, CodeMeta, and other relevant metadata schemes. They added additional elements  
 84 for domain specific information. Controlled vocabularies and restrictions are available for  
 85 multiple elements. Schembera and Iglezakis also presented a tool for automatic metadata

7. [https://www.wikidata.org/wiki/Wikidata:Main\\_Page](https://www.wikidata.org/wiki/Wikidata:Main_Page), accessed 12.12.2022

8. <http://models.mint.isi.edu>, accessed 12.12.2022

9. <http://bio.tools>, accessed 12.12.2022

**Table 3:** Overview of metadata scheme for energy research software

✓: fulfilled                      ✗: not fulfilled  
 (✓): (partly) fulfilled

	Focus on models (m) vs. data (d)	Formalized metadata scheme	Based on existing scheme	Use of value vocabularies
EngMeta [12]	d	✓	✓	✓
Catalog of energy co-simulation components [13]	m	(✓)	✗	✗
openmod <sup>11</sup>	m	✗	✗	✗
Open Energy Metadata <sup>12</sup>	d	✓	✗	(✓)
Open Energy Platform factsheets on models <sup>13</sup>	m	✗	✗	✗

86 extraction and included the metadata scheme in a data repository based on Dataverse<sup>10</sup>. They  
 87 validated their approach against common recommendations for good metadata schemes. [12]

88 Schwarz and Lehnhoff described a catalog of energy co-simulation components. They used a  
 89 semantic media wiki to collect information on components and the Functional Mockup Interface  
 90 (FMI) to add descriptions on the simulation interfaces. The elements of the catalog are usable as  
 91 metadata scheme but are neither formalized nor described in more detail. [13]

92 The open energy modeling initiative (openmod) includes a list of energy models in their wiki<sup>11</sup>.  
 93 For each model, administrative and descriptive metadata are listed, like license, link to a code  
 94 repository, model class. The descriptive elements include detailed information on the models.  
 95 The metadata scheme is not formalized and controlled vocabularies are neither used for the  
 96 elements nor for the values.

97 The open energy platform introduces the open energy metadata<sup>12</sup> for data. The metadata scheme  
 98 is designed for energy data and contains multiple elements. For a lot of elements, the use of  
 99 controlled vocabularies is required, e.g., for language and license. The use of an energy ontology  
 100 is not required for the description of the data but is planned as extension. The documentation of  
 101 the metadata does not link to any existing schemes used for the design of the metadata scheme.  
 102 Additionally, the open energy platform<sup>13</sup> includes information on models and frameworks. The  
 103 metadata elements are similar to the ones of the openmod wiki and also not formalized.

104 Table 3 summarizes the most relevant metadata schemes in the engineering and energy domain.  
 105 While EngMeta [12] presents a good broad metadata scheme in the engineering domain following  
 106 best practices, a formalized metadata scheme for ERS is still missing. However, the open energy  
 107 platform<sup>13</sup>, the openmod wiki<sup>11</sup>, and the work of Schwarz and Lehnhoff [13] present good  
 108 starting points for developing a formalized metadata scheme for ERS.

10. <https://dataverse.org/>, accessed 12.12.2022

11. [https://wiki.openmod-initiative.org/wiki/Open\\_Models](https://wiki.openmod-initiative.org/wiki/Open_Models), accessed 12.12.2022

12. <https://github.com/OpenEnergyPlatform/oemetadata>, accessed 12.12.2022

13. <https://openenergy-platform.org/factsheets/models/>, accessed 12.12.2022

### 109 2.3 Domain Ontologies for Energy Research

110 Domain ontologies are necessary to improve the interoperability of the metadata by being used  
111 as value vocabularies. Wierling et al. [14] give a broad overview of ontologies in the energy  
112 domain. In the following, we only give a short overview and refer to Wierling et al. for further  
113 information.

114 Cuenca et al. introduced an approach to unify different existing ontologies in the energy do-  
115 main [15]. They focused especially on ontologies for energy management applications in smart  
116 grids. Their ontology is called OEMA (Ontology for Energy Management Applications) and  
117 is formulated in OWL2 (Web Ontology Language 2<sup>14</sup>). It reuses existing ontologies and adds  
118 additional concepts. It has eight parts: infrastructure, energy and equipment, geographical,  
119 external factors, person and organization, energy saving, smart grid stakeholders, and units.  
120 The authors extended their work by introducing the Domain Analysis-Based Global Energy  
121 Ontology (DABGEO) [16]. To increase the reusability compared to OEMA, this ontology is  
122 constructed in a way that makes it easy to extend with application specific vocabulary. Using  
123 OWL2 DABGEO includes five parts: energy equipment, infrastructure, energy performance,  
124 energy external factors, and smart grid stakeholders.

125 Lefrançois presented the Smart Energy-Aware Systems (SEAS) ontology. It is a modular ontology  
126 published in Turtle<sup>15</sup>. The SEAS ontology contains the following modules: DeviceOntology,  
127 ForecastingOntology, OptimizationOntology, TradingOntology, and SmartMeterOntology. In  
128 the design process, he tried to follow the current best practices in ontology engineering. [17]

129 Booshehri et al. introduced the open energy ontology (OEO) as an ontology for energy systems  
130 analysis. The OEO is developed using OWL (Web Ontology Language<sup>16</sup>) and consists of  
131 oeo-modal, oeo-social, oeo-physical, and oeo-shared. The OEO includes concepts from other  
132 ontologies as the Financial Industry Business Ontology and the Unit Ontology. It can be used to  
133 annotate scenarios, factsheets, and data used in energy systems analysis. The ontology will be  
134 further developed on Github making it possible for everyone to contribute. [18]

135 Oppermann et al. introduced an ontology for EnArgus<sup>17</sup>, the funding information system for  
136 energy research in Germany. The ontology aims to map the whole domain of energy research  
137 and is used to improve the search in a project database. The ontology is not yet publicly available  
138 but the authors see its publication as a future work. Also, the authors admitted that they followed  
139 less strict ontology engineering rules than the developers of the OEO. [19]

140 Fernández-Izquierdo et al. gave a good overview on ontologies in the energy domain and intro-  
141 duced an ontology for demand response to improve interoperability between demand response  
142 stakeholders. The ontology described in OWL is based on OpenADR<sup>18</sup>, an open exchange model  
143 and global smart grid standard. The ontology is available on Github<sup>19</sup>. [20]

144 The Common Information Model (CIM) is a group of IEC standards and a domain ontology

14. <https://w3.org/TR/owl2-overview/>, accessed 12.12.2022

15. <https://w3.org/TR/turtle/>, accessed 12.12.2022

16. <https://w3.org/TR/owl-semantic/>, accessed 12.12.2022

17. <https://enargus.de/>, accessed 12.12.2022

18. <https://openadr.org/>, accessed 12.12.2022

19. <https://github.com/albaizq/OpenADRontology>, accessed 21.07.2021

145 which can be used to describe multiple aspects in energy systems. The standard is mainly applied  
146 in the electric utilities sector and has a high industry focus. The description is given in the Unified  
147 Modeling Language (UML) and a conversion to OWL is possible by using the CIMTool<sup>20</sup>. [21]  
148 There exist different ontologies for the energy (research) domain. It seems promising that one or  
149 a combination of multiple ontologies, for example of the OEO [18] and CIM [21], can be used  
150 as value vocabulary for describing ERS.

151 Overall, there exist good approaches for metadata schemes for research software in other domains  
152 like life science and geoscience, as shown in Section 2.1, which can be used as inspiration. The  
153 existing approaches for ERS often lack formalization, are not based on existing approaches  
154 and/or do not use value vocabularies like shown in Section 2.2. The last part of the related  
155 work showed that there exist multiple ontologies which can be used as value vocabularies for a  
156 metadata scheme for ERS.

### 157 3 Concept for a Metadata-based Registry for ERS

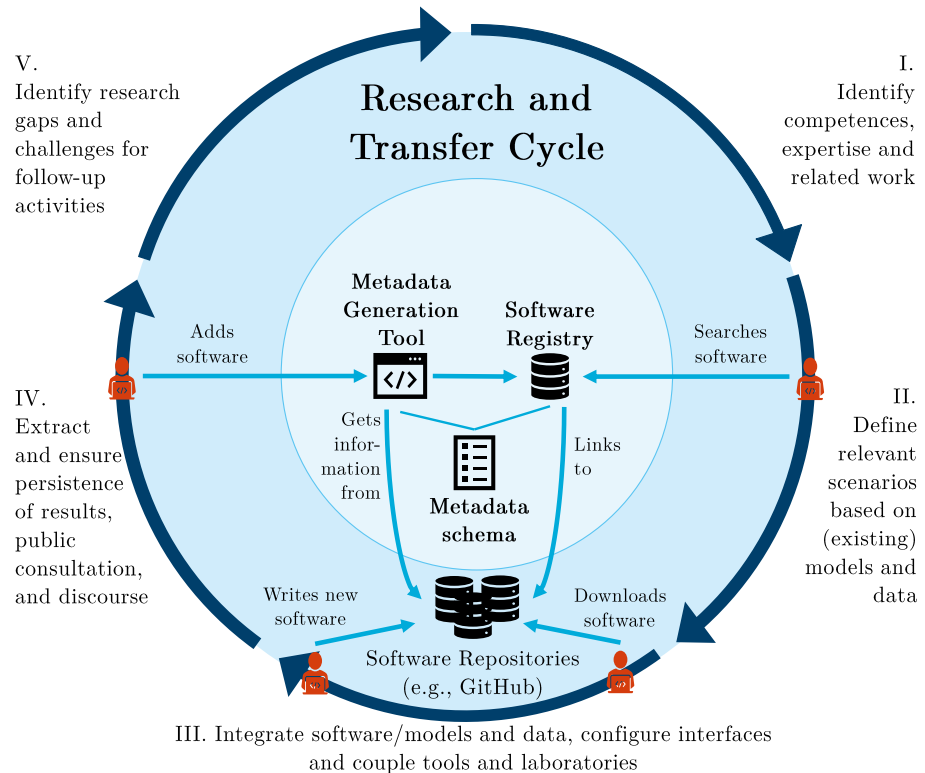
158 We introduce our concept for a Metadata-based Registry for ERS in Figure 2. As first step in  
159 the research cycle, researchers should be able to use the software registry in our concept to  
160 find relevant ERS for their research problem. All listed software should link to the responding  
161 software repositories (e.g., GitHub). We describe the registry in more detail in Section 3.3. After  
162 downloading the software researchers will use it as well as write additional software for their  
163 research. The new or extended software should be publish on any existing software repository.  
164 Then, researchers should be able to add the software to the software registry by using the metadata  
165 generation tool. It should extract as many information as possible from the software repository  
166 and, therefore, helps the researchers with creating metadata. We further describe the concept of  
167 the metadata generation tool in Section 3.2. Both the metadata generation tool and the registry  
168 are based on a common metadata scheme for ERS. Therefore, we first give more details on that  
169 in Section 3.1.

#### 170 3.1 A Metadata Scheme for ERS

171 A metadata scheme for ERS should be usable for all different types of ERS to increase their  
172 findability. It is the foundation for the other two artifacts. A metadata scheme comprises elements  
173 describing the categories for the metadata, of guidelines for creating according metadata, of a  
174 syntax and of constrains for the metadata, e.g., ontologies as value vocabularies [22].

175 The metadata scheme should be developed as an application profile using existing methods  
176 like the approach of Curado Malta and Baptista [23]. It should be based on a domain model  
177 containing desired elements with short descriptions, examples, and their relations. These elements  
178 and their description should only be formulated in English to be usable by the international  
179 research community. The development of this metadata scheme for ERS should be based on an  
180 environmental scan [24] to incorporate existing approaches, especially the one in semantic-web  
181 compatible formats, like the general work on metadata for research software, e.g., CodeMeta<sup>2</sup>,  
182 and the domain-specific developments like the metadata schemes of the Open Energy Platform<sup>13</sup>.

20. <https://wiki.cimtool.org/>, accessed 12.12.2022



**Figure 2:** A metadata-based registry support the exemplified research cycle of energy research with software usage

183 For each element of the metadata scheme, it should be specified if it is mandatory or optional and  
 184 if constraints exist, e.g., if the use of a controlled value vocabulary is required or not. Especially  
 185 for the domain specific elements, value vocabularies should be used. It should be carefully  
 186 analyzed if one or multiple ontologies, like OEO [18], can be integrated as value vocabulary.  
 187 The metadata scheme should be used by the registry and the metadata generation tool, as described  
 188 in the following sections.

### 189 3.2 A Metadata Generation Tool

190 A metadata generation tool should support all researchers to create high-quality metadata for  
 191 their ERS. It should lower the entrance barrier for creating metadata for all researchers in the  
 192 energy domain without the need for a deeper understanding of the underlying technologies.

193 If the ERS is already in an open software repository like GitLab or GitHub, the tool should  
 194 start by extracting as many metadata as possible from the repository. Mao et al. developed the  
 195 framework SoMEF which already can extract a lot of metadata out of readme files [25]. The  
 196 metadata will be mapped to the metadata scheme for ERS as described in Section 3.1. The tool  
 197 should allow the user to add additional metadata as well as correct the automatic extracted ones.  
 198 It should support the use of controlled value vocabularies as defined in the specification of the  
 199 metadata scheme and should allow to integrate additional ontologies to include many links to  
 200 other semantic web sources.



### 201 3.3 A Metadata Registry for ERS

202 A registry for ERS should help researchers to find the right software based on multiple search  
203 criteria, e.g., researchers wanting to do a grid simulation can look for a library or framework  
204 compatible with the models in a certain programming language they already use. For the  
205 development of such a registry, existing approaches from other domains like bio.tools<sup>9</sup> should  
206 be considered. The registry should be based on the new developed metadata scheme for ERS.

## 207 4 Conclusion and Outlook

208 There already exist a lot of ERS and even more ERS will be needed in the future. ERS is often  
209 developed without reusing existing software. One reason for that is that the existing software is  
210 not simple to find. The overall findability of ERS can be improved when ERS is registered in  
211 a registry with relevant metadata. Therefore, we introduced the concept of a **metadata-based**  
212 **registry for ERS** to enable this process.

213 We presented the state of the art in three parts. First, we described general approaches on metadata  
214 for research software as well as approaches focusing on research software in other domains.  
215 Second, we outlined metadata approaches in the energy domain where most current approaches  
216 focus on research data. The approaches already focusing on research software are not formalized  
217 and, therefore, do not meet our requirements. Third, we gave an overview on existing ontologies  
218 in the energy domain which can be serve as value vocabularies for a metadata scheme.

219 Based on the presented state of the art, we defined our concept consisting of three main artifacts.  
220 The metadata scheme for ERS presents the foundation and should be developed following best  
221 practices in metadata scheme development. It should include mandatory elements and value  
222 vocabularies. The metadata generation tool should assist researchers in creating metadata by  
223 extracting as many metadata as possible from different sources, e.g., GitLab or GitHub. The  
224 registry should be the place where ERS can be found. It should store metadata of ERS and should  
225 use advanced search technologies.

226 As future work, the different proposed artifacts will be developed and tested. Therefore, first a  
227 detailed requirement analysis will be performed as basis for the domain model of the metadata  
228 scheme.

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## 233 6 Roles and contributions

234 **Stephan Ferenz:** Conceptualization, Writing – original draft

235 **Astrid Nieße:** Conceptualization, Writing – review & editing

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