# Maggot : An ecosystem for sharing metadata within the web of FAIR Data

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## 22 Abstract

23 Background : Descriptive metadata are crucial for the discovery, reporting and mobilisation of 24 research datasets. Addressing all metadata issues within the Data Management Plan often poses 25 challenges for data producers. Organising and documenting data within data storage entails 26 creating various descriptive metadata. Subsequently, data sharing involves ensuring metadata 27 interoperability in alignment with FAIR principles. Given the tangible nature of these challenges, 28 a real need for management tools has to be addressed to assist data managers to the fullest 29 extent. Moreover, these tools have to meet data producers requirements and be user-friendly as 30 well with minimal training as prerequisites.

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32 **Results** : We developed Maggot which stands for Metadata Aggregation on Data Storage, 33 specifically designed to annotate datasets by generating metadata files to be linked into storage 34 spaces. Maggot enables users to seamlessly generate and attach comprehensible metadata to 35 datasets within a collaborative environment. This approach seamlessly integrates into a data 36 management plan, effectively tackling challenges related to data organisation, documentation, storage, and frictionless FAIR metadata sharing within the collaborative group and beyond.
Furthermore, for enabling metadata crosswalk, metadata generated with Maggot can be
converted for a specific data repository or configured to be exported into a suitable format for data
harvesting by third-party applications.

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42 **Conclusion** : The primary feature of Maggot is to ease metadata capture based on a carefully 43 selected schema and standards. Then, it greatly eases access to data through metadata as 44 requested nowadays in projects funded by public institutions and entities such as Europe 45 Commission. Thus, Maggot can be used on one hand to promote good local versus global data 46 management with open data sharing in mind while respecting FAIR principles, and on the other 47 hand to prepare the future EOSC FAIR Web of Data within the framework of the European Open 48 Science Cloud.

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50 **Keywords** : FAIR, Data management, Metadata, Interoperability, Crosswalk, Controlled

- 51 vocabulary, Ontologies, Thesaurus, Semantic artefacts
- 52

### 53

## 54 Background

55 In the realm of scientific research, metadata plays a key yet often overlooked role. Despite their 56 crucial importance for the discovery, reporting, and mobilisation of research datasets, metadata 57 remains insufficiently known within scientific communities. Yet being data themselves, metadata 58 have to be managed with the same level of rigour as the data produced and consumed by 59 research processes. This lack of awareness persists at a time when the sharing of research data has emerged as a cornerstone of open or at least reproducible science initiatives. As the scientific 60 61 landscape increasingly emphasises transparency and collaboration, understanding the 62 significance of metadata becomes imperative [1, 2].

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64 Producing comprehensive metadata is not a task to be taken lightly. It requires time, effort and expertise. Data producers tasked with generating datasets may be reluctant if they see no tangible 65 return on investment in creating metadata [3]. To overcome this hurdle, proactive efforts are 66 67 required to raise awareness among data producers about the benefits of open data practices [4]. However, crafting metadata poses challenges beyond incentivization. Data Management Plans 68 69 (DMPs), which outline strategies for managing research data throughout their lifecycle, often pose 70 non-trivial questions for data producers. These questions may be time-consuming or complex, 71 particularly when datasets span diverse scientific domains and require input from individuals with 72 varied skill sets. Consequently, collaborative efforts involving domain experts, data managers, 73 and information specialists are essential for navigating the intricacies of DMPs effectively, 74 furthermore when projects are involving multiple partners (e.g. [5]). The sheer diversity of 75 research data and possible dimensions - i.e., the type of characteristics they describe - further 76 complicate metadata management [6]. From omics data to images to experimental data tables, 77 the spectrum of data types is vast and multifaceted. 78

79 Given the complexity of the matter, it is suitable to differentiate between various types and 80 functions of metadata. While not delving into every category, we can simplify them into two main 81 groups: general metadata and specialised metadata. Within the latter category, we encounter 82 structural metadata, which serve to depict the organisation, arrangement, and interconnections 83 within a dataset. For instance, when considering different types of data such as experimental data 84 tables, it becomes obvious that structural metadata are essential for optimising their utility [7]. 85 Conversely, general metadata (descriptive, administrative, rights) apply to all data types 86 generated within studies with similar experimental contexts or even an entire project. The 87 subsequent sections of this article will focus on these general metadata.

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89 How can we collect such metadata while ensuring that they ultimately meet the requisite criteria 90 for interoperability? Indeed, standardisation is the key towards interoperability and consistency in 91 metadata practices. Sustainable metadata have to adhere to established standards and be 92 described using controlled vocabularies endorsed broadly by the scientific community [8]. 93 However, the responsibility for metadata creation predominantly falls upon data producers who 94 possess intimate knowledge of the data intricacies. This is challenging as data producers may 95 lack familiarity with metadata standards and best practices, and so reinforces the importance of 96 roles within the data management ecosystem. Data managers and data stewards, equipped with 97 expertise in applying FAIR (Findable, Accessible, Interoperable, Reusable) principles [9] and 98 metadata standards, play a key role in guiding metadata creation and management processes. 99 Conversely, scientists, serving as data producers, possess deep domain knowledge essential for 100 contextualising and enriching metadata. Recognizing the complementary of these roles, 101 collaborative partnerships between data managers and scientists are indispensable for ensuring 102 the effective and sustainable management of research data [10].

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104 Attempting to find a one-size-fits-all data warehouse capable of accommodating every data type 105 proves to be a futile endeavour. Invariably, some data types remain unsupported or inadequately 106 represented within existing data repositories. To address this challenge, we propose a pragmatic 107 approach that involves managing data directly from storage spaces and then depositing them 108 when the time comes into repositories tailored to requirements of each data type. To this end, we 109 have developed Maggot (Metadata Aggregation on Data Storage), a specialised tool for 110 aggregating metadata into data storage. It is specifically designed for annotating datasets by 111 generating metadata files to attach to the data storage. It has been designed to help data 112 managers in solving as many of the aforementioned challenges as feasible, all while catering to 113 the needs of data producers with minimal training on their part.

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### 116 **Design Considerations**

117 The development and implementation of Maggot followed a structured approach, involving 118 multiple steps and actors (**Fig. 1**). These steps encompass in particular the identification of 119 metadata fields, terms, vocabularies, and standards. Inspiration for this approach was drawn from

120 an online document detailing the implementation of a descriptive metadata plan [11]. An approach

121 with some similarities has been adopted with the FAIR-DS application [12], dedicated to 122 nucleotide sequences.

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124 Within any research team, collaboration between the data manager and data producers is 125 essential to select and customise the minimum metadata required and an associated metadata 126 schema suitable for the specific scientific domain. While this process may present challenges, it 127 is crucial to meticulously construct and adapt the schema to align with existing data. Rather than 128 mandating data conformity, the schema should be flexible to accommodate pre-existing data. 129 Thus, the adoption of a schema, such as the one implemented within the Harvard Dataverse 130 software [13] (https://dataverse.harvard.edu), followed by an iterative and progressive 131 adjustment, is the approach embraced by Maggot. Indeed, Harvard Dataverse itself is built upon 132 the standard DDI (Document, Discover and Interoperate) metadata schema 133 (https://ddialliance.org), which has been expanded to accommodate its requirements. The 134 advantage of the DDI schema is that it encompasses a wealth of general information for 135 describing datasets of any type.

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137 In the same way, Maggot advocates an iterative and progressive approach regarding the 138 management of controlled vocabulary. Recognizing the impossibility of achieving exhaustiveness 139 in the initial stages, Maggot facilitates a process of continuous improvement. This involves starting 140 with a simple vocabulary dictionary sourced locally and consolidating community-used vocabulary 141 within the related scientific domain. Subsequently, consideration could be given to the creation of 142 a thesaurus (or at least a controlled vocabulary), with or without mapping to existing ontologies. 143 Maggot is seamlessly based on the SKOSMOS web application [14] (https://skosmos.org) to 144 query thesauri directly, streamlining the process. Furthermore, ontologies can be chosen 145 progressively by selecting those which are truly relevant for the collective and by drawing up an 146 understandable landscape of the context in which they fit. In the same way for thesauri, Maggot 147 offers the opportunity to enrich metadata using ontologies seamlessly accessible through the 148 OntoPortal web applications such as BioPortal [15] (https://bioportal.bioontology.org) or 149 AgroPortal [16] (https://agroportal.lirmm.fr).

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151 In our view, creating metadata should not compel data producers to engage in training in topics 152 or concepts outside their expertise, such as FAIR principles, the semantic web, or metadata 153 schema, unless they choose to do so. The data manager needs to recognize that data producers 154 may not possess extensive knowledge of data management practices, especially in the realm of 155 open science and FAIR data. Therefore, pedagogy should be prioritised by refraining from 156 overwhelming data producers with technical details specific to their field. Instead, the data 157 manager should focus on raising awareness and encouraging data producers to improve the 158 quality and reusability of their data [3]. This includes providing guidance on relevant metadata 159 and controlled vocabulary within their scientific domain, as well as training on best practices such 160 as the use of permanent identifiers like DOI, ORCID, RoR, ePICs, Handle and other well 161 recognized identifying systems. Furthermore, data producers should be informed about the 162 selection of appropriate licences, such as CC-BY licences (https://creativecommons.org). One 163 of the features of Maggot is precisely to allow the data manager to document each of the terms, 164 providing examples and links for additional information if necessary. This contextual online assistance is thus accessible during entry to allow data producers to fill in each field in the mostrelevant way, ensuring optimal support throughout the process.

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168 For downstream metadata management, Maggot provides functionality enabling transformation 169 of metadata in two ways: to targeted data repositories with prerequisites as defined by the Core 170 Trust Seal [17] (https://www.coretrustseal.org), or to an export format suitable for harvesting data 171 by third-party applications via an application programming interface (API). These functionalities 172 are built upon a metadata crossing approach based on the mapping defined upstream by the data 173 manager. They ensure compatibility with systems allowing content indexing, thus aligning with 174 the FAIR principles. Adopting this approach improves organisations' data-management practices 175 by effectively using metadata throughout the data lifecycle and facilitates data linkage. It 176 enhances data interoperability and reusability, optimising the value derived from data assets. It 177 also provides practical accessibility through the Web of FAIR Data - i.e. data which meet FAIR 178 principles - by increasing data linkage possibilities, as envisioned by international consortiums 179 like the European Open Science Cloud (EOSC, https://open-science-cloud.ec.europa.eu).

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181 Finally, let us emphasise that a software application rarely meets a need alone but is part of a 182 more global approach, involving several roles. Namely, the data manager is the person who 183 defines the data policy, i.e., its implementation and governance, while data stewards are 184 responsible for data quality, and therefore have a role in data curation. Unavoidably, the 185 development and dissemination of metadata, involving a metadata-centric culture, underline the 186 need for ongoing training initiatives and data stewards are pivotal in educating data producers. 187 Despite advancements in tool intuitiveness and automation, the growing complexity and scale of 188 data across the Web of Data call for an increase in the number of skilled data stewards within 189 research organisations to ensure maximum data reusability [18].

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## 192 **Results**

193 Maggot (Metadata Aggregation on Data Storage) was developed to address the need for a multi-194 purpose data management tool capable of supporting the wide data diversity range within a 195 collective. Its primary objectives are to provide visibility into the collective's data assets, facilitate 196 better data description and increase early adoption of FAIR principles. It contributes to guarantee 197 the sustainability of data reusability, especially for those produced by fixed-term personnel 198 (limited-term contract, doctoral students and postdoctoral fellows). It helps to raise awareness 199 among newcomers and students about the importance of robust data description practices which 200 is crucial for fostering a culture of data management excellence [3].

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Maggot enables users to achieve these objectives through various features and functionalities which can be divided into three parts: creation, sharing, dissemination (**Fig. 2**). Maggot supports by default the common metadata standards of Harvard Dataverse (based on DDI) serving as a starting point which can be extended/specialised to suit individual contexts. It then offers scalability and flexibility to enrich the core metadata to adapt any experimental context. Controlled vocabulary management is another key aspect of Maggot, offering options ranging from simple dictionaries to ontologies as discussed above. The tool includes enrichment functionalities of
 existing resources (e.g., dictionary editing, adding additional ontologies), ensuring that users can
 effectively manage and utilise controlled vocabularies relevant to their data. To this end, it offers
 great flexibility in configuration, allowing organisations to tailor the tool to their specific needs and
 requirements (Fig. 3).

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214 Implementing a data management plan (DMP) entails certain prerequisites, including data 215 externalisation to preserve them outside users' disk space. This ensures data is secured in one 216 location and serves as an initial backup, which becomes particularly crucial when fixed-term 217 personnel are involved in data production. Consequently, considerations arise regarding the 218 organisation of storage spaces, such as harmonising folder and file naming conventions, setting 219 up folder structure, and using README files to provide essential information. Maggot precisely 220 answers these DMP challenging questions related to organising, documenting, storing and 221 sharing data from various sources. In our approach the data storage space becomes a local 222 reference data repository, mitigating the risk of data duplication or divergence to another medium. 223 Then, only metadata need to be added to this centralised space, streamlining data management 224 processes, and enhancing efficiency. Indeed, once the metadata file is generated in JSON format, 225 it has to be placed in the storage space reserved for this purpose alongside the corresponding 226 dataset. This metadata file can be seen as a README file adapted to machines (Additional file 227 1), but still readable by humans. In contrast, with an internal structure, it offers better coherence 228 and consistency of information than a simple README file with a completely free and therefore 229 unstructured text format. In this way, the storage space becomes a data asset which can therefore 230 be efficiently leveraged using metadata. Indeed, all the JSON metadata files are scanned and 231 parsed according to a fixed time interval (30 min) then loaded into a database. This allows users 232 to guery datasets based on metadata filters. The search form, in a compact shape, is almost the 233 same as the entry form. Matching datasets are returned as a list, and for each of them a provided 234 link helps to access the detailed metadata.

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236 Metadata entry can be initiated at the outset of a project or study, without requiring the completion 237 of all data acquisition or processing, or waiting until the data need to be published. Indeed, the 238 ability to reload a metadata file facilitates gradual and iterative metadata addition across the 239 project, thereby spanning the research data lifecycle to the greatest extent possible. Maggot 240 supports the input of both descriptive and administrative metadata for any type of data, including 241 datasets, images, sequences, and more, with customizable field definitions to suit diverse user 242 requirements. Moreover, Maggot emphasises ease of use and adaptability. It offers guided 243 assistance through drop-down menus and vocabulary lists featuring autocompletion, greatly 244 speeding up the process of filling in numerous descriptive metadata. Crucially, Maggot does not 245 restrict the choice of data repository, ensuring compatibility with currently supported platforms 246 knowing that others may be supported in the future (e.g. Dryad [19], RO-Crate [20]). This also 247 does not prejudge the use of metadata. It is entirely possible, for example, to set up an internal 248 metadata harvesting process to automatically fill in another data source (e.g., FAIRDOM-SEEK 249 data management platform [21]). It is essential to highlight that opting for Maggot to generate 250 metadata does not confine the data to an isolated silo. In case one day the Maggot tool was no 251 longer supported, all metadata will persist in disk space in a format accessible to both humans

252 and machines. This ensures that future applications/services are able to continue to use legacy 253 metadata and therefore warranty data reuse. For this purpose, Maggot enables data scientists or 254 data repositories to harvest data. The OAI-PMH (Open Archives Initiative Protocol for Metadata 255 Harvesting, https://www.openarchives.org) allows for listing all datasets based on the DublinCore 256 schema (https://www.dublincore.org), while the metadata of each dataset can be harvested in 257 JSON-LD format (JSON for Linking Data, https://json-ld.org), mainly adhering to the schema.org 258 standard (https://schema.org). This aspect is particularly critical for linking metadata within the 259 realm of linked data, thereby ensuring their interoperability. For instance, we plan in the near 260 future to support DCAT-based harvesting (https://www.w3.org/TR/vocab-dcat-3/).

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262 Maggot also provides a solution to data fragmentation. Indeed, data is often scattered across 263 various platforms, databases, and file formats, making it challenging to locate and access. 264 Moreover, non-standardized metadata and inconsistent data organisation hinder effective data 265 discovery and reuse. Therefore, Maggot allows data producers to specify resources, i.e., data in 266 the broader sense, whether external or internal, to centralise all links towards data (Fig. 2). 267 External resources must be specified by an URL with preference for a permanent identifier (e.g., 268 DOI) but also any URL pointing to data whether they comply with the FAIR principle or not. 269 Furthermore, in the case of local data management, it is wise to indicate in which space the data 270 is located if it is not located in the same place as metadata (e.g., NAS unit or data cloud). Maggot 271 can thus become a data hub by gathering all references to several data sources in one place at 272 hand.

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274 While the primary focus is managing metadata linked to data stored within a given collective, 275 Maggot also facilitates data openness through metadata, especially in projects funded by public 276 institutions. By setting up metadata schemas that facilitate crosswalks with established data 277 repositories, users can seamlessly push metadata along with the corresponding data without the 278 need for additional data entry. This promotes data openness and accessibility in accordance with 279 international standards and community norms. Such functionality empowers organisations to 280 share their data with external stakeholders while ensuring consistency and interoperability. 281 Maggot thus offers a comprehensive and open solution for metadata management, catering to 282 the diverse requirements of organisations and promoting best practices in data description and 283 dissemination.

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### 286 Implementation and Documentation

The deployment of Maggot requires two infrastructure components: a dedicated server for the web application and a designated data storage space. Regarding the server, it must be capable of running an operating system compatible with Linux. In addition, it should support containerization using Docker. This latter aspect offers a simplified approach to installation and administration, but also ease of use and flexibility. Regarding data storage, any technology is suitable. Data storage can be local (e.g., NAS unit) or remote (e.g., data cloud). Successful tests have been performed by implementing a server on our institutional data center and data storage on another data center. Access to the storage space can easily be done using the rclone tool
 (https://rclone.org), a real Swiss army knife for disk sharing.

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297 web-based PHP Maggot is а application as а front-end usina MongoDB 298 (https://www.mongodb.com) to index all scanned metadata from disk storage every 30 minutes. 299 Moreover, Maggot mobilises various vocabularies (thesauri and ontologies), most of which are in 300 remote resources. So the utilisation of APIs plays a significant role, particularly for integrating 301 these vocabularies. This extensive use of APIs facilitates real-time imports, thus reducing the 302 need for pre-updating information.

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Documentation is available via https://inrae.github.io/pgd-mmdt/ and from within the app. It includes technical information on how to configure Maggot, but also a quick overview of how to use it. For data managers, it explains in detail how to construct the terminology with the associated vocabularies.

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## 310 Conclusion

311 Maggot is specifically designed to annotate datasets by generating metadata files to be linked 312 into storage spaces, to tackle challenges related to data organisation, documentation, storage, 313 and frictionless FAIR metadata sharing within the collaborative group and beyond. Indeed, 314 Maggot meets the Open Data requirements beyond the simple provision of data with unlimited 315 access. This essentially implies: i) to ensure search and access to metadata that define data 316 access and usage conditions, and ii) to foster metadata and data interoperability to break down 317 silos, highlighting the necessity of embracing FAIR principles even when complete openness is 318 not achievable.

By covering as much of the research data lifecycle as possible, Maggot ensures effective and sustainable research data management and significantly simplifies the adoption of FAIR principles thereby empowering organisations to elevate the value and usability of their own data assets. Moreover, its ability via crosswalk approaches to distribute metadata based on standard schemas while being machine-readable, expands the toolbox needed to prepare the future EOSC FAIR Web of Data within the framework of the European Open Science Cloud.

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## 327 Availability of Source Code and Requirements

- 328 Project name: Maggot
- Project homepage: https://pmb-bordeaux.fr/maggot/
- Project code repository: https://github.com/inrae/pgd-mmdt
- Documentation: https://inrae.github.io/pgd-mmdt/
- Operating system(s): Platform independent
- Programming languages: PHP, python, javascript
- Licence: GNU GPL v3

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- Maggot, RRID: SCR\_025261
- Biotools: https://bio.tools/maggot
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## 339 Abbreviations

340 API: Application Programming Interface; DDI: Document, Discover and Interoperate; DMP: Data

- 341 Management Plan; EOSC: European Open Science Cloud; FAIR: Findable, Accessible,
- 342 Interoperable, Reusable; JSON: JavaScript Object Notation; JSON-LD: JSON for Linking Data;
- OAI-PMH: Open Archives Initiative Protocol for Metadata Harvesting; NAS: Network AttachedStorage;
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## 347 **Competing interest**

- 348 The authors declare that they have no competing interests.
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# 361 Authors' Contributions

Conceptualization: D.J, F.E, PC.; funding acquisition: D.J, F.E.; methodology: D.J., F.E, R.D.;
software: D.J, F.E.; writing—original draft: D.J., R.D.; writing—review and editing: All authors. All
authors read and approved the final manuscript.

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371 for his fruitful feedback in the implementation of Maggot within his research infrastructure.

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## 374 Additional files

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Additional file 1 : Examples of metadata files along with corresponding definitions files within
 an <u>Excel workbook</u>.

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- 380 Figures

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383 Figure 1: Development and implementation of Maggot tool structured into three steps, namely before, 384 during and after metadata capture: i) Upstream, collaboration between the data manager and the data 385 producers was essential to select and customise a flexible metadata schema adapted to the scientific 386 domain as well as the identification of terms and vocabularies (dictionaries, thesauri, ontologies). Therefore, 387 Maggot proposes the Dataverse schema serving as a fundamental model, itself based on the standard DDI 388 metadata schema. ii) For metadata entry, data producers must be trained on good practices such as the 389 proper use of permanent identifiers or the choice of licences. iii) Downstream, data can easily be pushed 390 into a support data repository without any addition or can be harvested based on a dedicated protocol (OAI-391 PMH). JSON-LD format is also supported for linking metadata within the realm of linked data, thereby 392 ensuring their interoperability. The complementarity of roles between data manager and data producers 393 ensures effective and sustainable research data management.

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#### 395

396 Figure 2: Main functionalities of Maggot split into three parts: creation, sharing, dissemination. First, 397 producing a document with metadata sets of data within a collective of people, thus allowing i) to answer 398 certain questions of the Data Management Plan (DMP) concerning data organisation, documentation, 399 storage and sharing in the data storage space, ii) to meet certain data and metadata requirements, listed 400 for example by Open Research Europe in accordance with FAIR principles. Next, searching for datasets 401 by their metadata. Indeed, the descriptive metadata thus produced can be associated with the 402 corresponding data directly in the storage space and then it is possible to perform a search on the metadata 403 to find one or more sets of data. Only descriptive metadata is accessible by default. Finally, publishing the 404 metadata of the datasets as well as their data files in a European-approved repository, with the possibility 405 either to directly harvest the metadata via the OAI-PMH protocol, or to export the associated metadata with 406 their semantic context for full interoperability.

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**Figure 3** : Maggot tool flexibility in configuration. Maggot allows users to choose all the metadata describing their data with two levels of definition files. The first level concerns the definition of metadata similar to a descriptive metadata plan. This category is more akin to configuration files, and constitutes the heart of the configuration around which everything else is based. The input and search interfaces are completely generated from these definition files, thus defining each of the fields, their input type and the associated controlled vocabulary. The second level concerns the definitions of the mapping to a differently structured metadata schema (metadata crosswalk, i.e a specification for mapping one metadata standard to another), used either for metadata export to a remote repository (e.g. Dataverse, Zenodo) or for metadata harvesting
 (e.g. JSON-LD, OAI-PMH).

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