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1 Open data and digital morphology

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71

72 Abstract

73

74 Over the past two decades, the development of methods for visualizing and analysing specimens 75 digitally, in three and even four dimensions, has transformed the study of living and fossil organisms. 76 However, the initial promise, that the widespread application of such methods would facilitate 77 access to the underlying digital data, has not been fully achieved. The underlying datasets for many 78 published studies are not readily or freely available, introducing a barrier to verification and 79 reproducibility, and the reuse of data. There is no current agreement or policy on the amount and 80 type of data that should be made available alongside studies that use, and in some cases are wholly 81 reliant on, digital morphology. Here, we propose a set of recommendations for minimum standards 82 and additional best practice for 3D digital data publication, and review the issues around data 83 storage, management and accessibility. 84 85 **Keywords**: 86 digital data, 3D models, phenotype, computed tomography, visualization, functional analysis 87 88 1. Introduction

89 Three-dimensional (3D) digital morphological data are commonly employed by palaeontologists and

90 biologists in research. In palaeontology and anthropology, the widespread application of

91 tomography (especially X-ray computed tomography, CT), laser and structured light scanning and

92 photogrammetry, has revolutionized the study of morphology [1-4]. In biology, optical microscopy,

- 93 magnetic resonance imaging (MRI) and contrast-enhanced CT are important tools for investigating
- 94 soft-tissue anatomy [5-11]. The revolution brought about by these technologies has increased the
- 95 amount and detail of anatomical information recovered from fossil and living organisms,
- 96 transforming the nature of scientific enquiry in related fields (Figure 1). The resulting datasets are
- 97 often reconstructed and presented as 3D digital models, which are themselves sometimes used in
- 98 downstream analyses, including geometric morphometrics [12, 13], finite element analysis [14],
- 99 multibody dynamics analysis [15], and computational fluid dynamics [16], thereby facilitating

100 quantitative tests of functional and evolutionary hypotheses [3]. These types of studies have yielded 101 important advances in our understanding of the anatomy of living and fossil organisms, e.g., [10, 17, 102 18], as well as fundamental aspects of their biology, from feeding mode [19-21] to mobility [22, 23], 103 development [24, 25], and physiology [26-28], as well as developments in taxonomic practise [29, 104 30]. Barriers to data sharing and access to specimens can be eroded because data exist as digital files 105 that can be easily copied and readily distributed, allowing simultaneous analysis by multiple 106 researchers [31]. These attributes should also enhance the verifiability and reproducibility of studies, 107 facilitating the reuse of data and metadata, more in-depth interrogation of any given dataset, and 108 broader-scale comparative analyses through the assembly of large datasets of multiple specimens or 109 taxa.

110

111 However, authors of studies involving 3D digital datasets of biological and palaeontological

112 specimens often do not publish their supporting data, meaning that results and conclusions cannot

easily be verified or replicated, and that this potentially valuable source of novel data cannot be

114 further explored [31]. Ultimately, digital data collected but unpublished are likely to be lost to

science [2, 29]. This also represents a substantial waste of financial and other resources, and places

116 vulnerable original specimens at greater risk of damage or loss, as the same specimens are likely to

117 be reimaged repeatedly to enable different groups of workers to reproduce the data [29, 32].

118 Consequently, the promise of 3D digital data has not yet been fully realized.

119

120 This is not news [2, 29, 31]. However, most national and international funders have imposed 121 regulations on data access and sharing that are forcing researchers and institutions to finally 122 confront this challenge [33]. These regulations range from funder-mandated full release of all data 123 [33], through declarations that the data are available from authors on request, to no release of 124 supporting data [33]. When data are released, they are deposited in a diversity of online databases 125 (e.g. BIRN, Dataverse, Dryad, EOL, Figshare, GigaDB, Github, MorphoBank, MorphoDBase, 126 MorphoMuseuM, MorphoSource, Phenome10K, Zenodo), institutional and funder repositories, 127 physical museums, and research group websites. At least in part, this diversity of approaches reflects 128 uncertainty about the available repositories for data deposition and the cost of storing the 129 comparatively large files associated with digital imaging-based research. Researchers can also be 130 reluctant to share data that remain part of an active research program [34], or to share a subset of 131 data that is part of a larger, unpublished package. There is also a lack of consensus and widespread 132 confusion over issues of data ownership and copyright, and conflict that emerges between

institutional policies asserting copyright ownership (e.g. public museum or even private collections)
and the regulations of funding bodies and publishers with regard to open data. Consequently,
sharing or publishing supporting data is often a low priority and has effectively been considered
optional when not prescribed by a journal. Partial datasets (e.g. low-resolution visualizations or
external surfaces) can be insufficient for reproducibility or even verification. As digital morphology
has evolved, most of us in the research community have failed to achieve what might now be
considered best practise of open data.

140

141 The academic world has already taken important steps towards overcoming some of these 142 motivational and practical obstacles. Platforms for both archiving and sharing data online are 143 becoming more commonplace, and can handle large file sizes. The standard in molecular biology is 144 Genbank (https://www.ncbi.nlm.nih.gov/genbank/) where sequence data underpinning studies are 145 accessioned before publication. For other data formats, journals and publishers offer a mixed 146 landscape of policies on data publishing that is in need of standardization [35, 36], but many not 147 only mandate data deposition, some are even prepared to bear the associated costs, making data 148 deposition easier and ultimately improving science, both in terms of practice and accessibility. There 149 are also initiatives to integrate data submission with submissions to peer-reviewed journals, 150 requiring, or at least allowing, the submission of data in the article submission process and enabling 151 reviewers to examine supporting data as part of the review process [37]. However, collectively, 152 these initiatives have not been integrated [35] and they have not yet translated into common 153 practice within many subdisciplines in biology, palaeontology and anthropology. 154 155 If a consensus can be established among authors, repositories, journal editors, peer reviewers and 156 funding agencies, there is the prospect of finally realizing the potential of digital morphology in the 157 open-data era. Here, we make recommendations on the nature and extent of essential and 158 recommended best practice datasets that should be made available to support scientific

159 publications using 3D digital datasets across biological sciences (summarised in tables 1 and 2). We

160 review the requirements of associated metadata, discuss the current range of repositories available

161 for such studies, and comment on issues affecting their utility.

162

163 **2.** Publishing tomographic data

164 A range of methods exist for studying 3D specimens through the creation of 2-D image stacks (i.e.

165 tomography), including X-ray CT (encompassing medical CT, micro-CT and synchrotron tomography),

166 MRI, neutron tomography, optical tomography, histological microtomy and physical tomography [1,

167 3, 4, 38-40]. All of these techniques generate datasets consisting of up to several thousand parallel

168 sections or slices (tomograms) through a specimen, with each tomogram represented by an image

169 file. Various techniques exist for the construction of 3D digital models from sets of tomograms [1].

170

171 (a) Data essential for scientific verification

172 The image stack: Image stacks are the starting point for most tomographic studies. These provide 173 immediate insight into internal and external features, and form the basis for any subsequent 174 construction of 3D models. Image stacks exist in a range of non-proprietary file formats, but the 175 most common include DICOM, TIFF, JPEG, PNG, vol, RAW, and BMP [41]. All such files can be opened 176 and viewed in free software such as ImageJ, Drishti, SPIERS, Horos, and 3D slicer [42], and can be 177 converted into different formats, although this can be more difficult with DICOM files which exist in 178 a multitude of sub-formats, not all of which can be handled by all software. For most purposes, TIFFs 179 (16- or 8-bit) provide the best balance of accessibility, file size, and data quality (lossless 180 compression), but any lossless, standard image file-types are sufficient. Most JPEG formats enforce a 181 lossy compression scheme that may degrade over multiple save operations; lossless JPEG formats do 182 exist (JPEG-LS, JPEG 2000), but they are not widely used. These differences underlie the importance 183 of specifying the file standard used [41]. Minimally, image stacks should retain the contrast 184 resolution (bit-depth) and spatial resolution used in the study. In cases where the image stack is 185 derived from K-space filling (e.g. MRI) or a series of angular projections (e.g. X-ray CT), the process of 186 generating the image stack is largely automated and we do not consider it necessary to publish the 187 raw projections.

188

189 Metadata: An image stack alone will not contain all the information necessary to make full use of 190 the data. For example, scale is only preserved if the resolution (e.g. voxel size or slice spacing) is 191 encoded in the files, and for some datasets slice spacing is not constant and requires per-slice 192 documentation. In the case of DICOMs, this information is typically retained within the file or can be 193 added to the file with a header tag editor (e.g. ImageJ). Otherwise, a text file detailing the voxel or 194 pixel size and slice spacing is the minimum necessary information that must accompany publication 195 of any image stacks. Additionally, metadata information should include full details of how the 196 images were acquired (including scan settings) and further information on data copyright, repository 197 and accession of specimens scanned and, if appropriate, comments on preparation or specimen 198 storage for biological specimens; see table 1). This information is necessary to reproduce studies, as

199 well as to evaluate if better quality data could be obtained with a different set of parameters [43].

- 200 Minimally, these data should be provided in a simple text file (e.g. .txt or .vgi) associated with the
- 201 dataset, regardless of whether the information is provided in any study based on the data.
- 202

203 3D models: Typically, tomographic studies involve the reconstruction of 3D models from image 204 stacks, in some cases after image segmentation or other preparation (see below). 3D models are 205 normally triangle-mesh geometries generated via isosurfacing (usually known as surface models) [1]. 206 Publication of the 3D models resulting from isosurfacing allows for the interactive examination of 207 specimen morphology in three dimensions; a wide range of free software is available for this task [1, 208 3], although no ideal general-purpose file-format exists for complex models (see below). 3D models 209 may have been modified after initial isosurface-construction, for example through smoothing, island 210 removal or hole-filling. Consequently, the most appropriate model to publish to enable verification is 211 the final model (or models) on which the results of the study are based, or which is used in downstream analyses.

212

213

214 The 3D models generated using tomographic data are available in a range of different file formats [1, 215 44]. The choice of file type may be influenced by various factors including file size and whether 216 colour/texture information is required; it is essential that openly accessible, standard formats are 217 used (e.g. STL, PLY or OBJ), but there is no single 'ideal' file format. The Stereolithography (STL) 218 format is the most widely used standard for publishing 3D triangle meshes derived from 219 tomographic techniques, and it is simple and supported by the vast majority of 3D visualization 220 programs, including freely available software [1]. STL files are also compatible with most modern 3D 221 printers, offering potential for wider applications in specimen conservation, public outreach or 222 teaching [3, 45]. However, STL files cannot store data on colour, texture, or scale. Where these are 223 an essential part of the study, an alternative format such as PLY, OBJ with MTL, or VAXML [1, 41, 44] 224 will be required. These formats are also recommended for meshes with a high number of triangles, 225 which can result in very large file sizes in the STL format. 226

227 (b) Additional data required for best practice

228 Prepared datasets. While some tomographic datasets are reconstructed as 3D models without any

229 modification or mark-up, this is unusual. Most datasets are subjected at least to segmentation, the

- 230 semi-automated or manual differentiation of voxels (3D pixels) into distinct regions-of-interest
- 231 (using, for example, 'label fields' in Avizo, or 'masks' in SPIERS). Some datasets also require semi-

232 automated or manual modification of the data (e.g. through brightness modifications) to better 233 separate specimen from background (we term this 'editing'). These processes involve a degree of 234 subjective interpretation; this is especially true for palaeontological datasets, which are often very 235 noisy and can require extensive manual intervention to extract maximal information from the 236 original data. Thus, publication of the original tomographic dataset and final 3D model may not be 237 sufficient to enable other researchers to assess the association between the two. Segmenting and/or 238 editing a tomographic dataset can be very time-consuming and therefore difficult to reproduce in 239 practice; without access to prepared datasets, most secondary users would not be able to fully 240 interrogate the data underlying a 3D model. In such instances, prepared datasets should be 241 released. No standard file-format exists, but labels and masks can be released in the native formats 242 by the software used to generate them, or as binary image stacks, which can then be readily 243 reconstructed as a 3D model in a variety of software packages [1, 44]. 244 245 Development of back-projection algorithms can improve signal to noise ratio in generated image 246 stacks and, hence, recent open data mandates at synchrotron facilities require archiving of the

radiograph projections, not the resulting slice data [46]. Thus, it may be sensible for authors to
archive the raw projection libraries themselves. This is especially important where access to the
same specimen may be problematic, or as a precaution in case unique specimens are damaged, lost
or destroyed.

251

Image registration: For physically destructive and optical tomography, tomograms need to be registered (aligned relatively and absolutely in the X, Y, and Z planes, either manually or semiautomatically) prior to any reconstruction of 3D models. This adds a potentially subjective step that may have a bearing on downstream analyses, and so we recommend publishing both the original (unregistered) and registered image stacks as best practice.

257

258 **3.** Publishing 3D data from surface-based methods

Alternative surface-based methods exist for digitizing only the exterior features of specimens in 3D, most notably laser or structured light scanning [47] and photogrammetry [1, 48, 49]. For photogrammetry, data begin as 2-D photographs, whereas in surface-scanning techniques, the 3D shape is usually directly captured as 3D point clouds, with or without texture capture (colour) for each point. In photogrammetry, a 3D polygonal mesh with texture data is generated and warped onto the 3D surface (typically automatically), giving each triangle a colour value. Scanning

265 methodologies may directly visualize point clouds, or may generate and visualize a 3D triangle mesh, 266 with or without texture mapped onto triangles or vertices.

267

268 (a) Data essential for verification:

269 3D models: The production of the initial 3D surface from photographs or surface-scans is largely 270 automated. The most critical data are the final 3D surface file(s) (which may be fused from the 271 original component meshes), e.g. in STL, PLY or OBJ format(s) [41]. In cases where the surface 272 texture (i.e. colour information) is directly relevant to the outcomes of a study, the published 3D 273 models must retain this information (i.e. should be provided in PLY or OBJ formats). Surface models 274 are not normally segmented into multiple geometric objects, so single-file models in PLY or STL

- 275 format are practical.
- 276

277 *Metadata*: A text file of metadata should be provided that documents details of the imaging settings 278 and techniques used to generate the 3D model (Table 1). Preparation of 3D meshes may involve a 279 range of operations, including trimming irrelevant data, realigning or reorienting components of the 280 mesh, fusion into a single mesh, smoothing, hole-filling, and/or manual manipulation of the location 281 of individual point coordinates or surfaces. These operations should be detailed in the metadata file. 282 Where such operations are non-trivial and/or involve interpretation, those data (photographs, raw 283 point clouds) are an essential provision, in open and widely accessible formats, where possible.

284

285 (b) Additional data required for best practice

286 Models including texture information: Colour data from the surface can provide useful information 287 to help interpret the specimen (e.g. taphonomic preservation). As best practice, this should be 288 included if available, in PLY or OBJ format.

289

290 Original capture data: The photographs or data captured by the scanner or the 3D data generated 291 by the photogrammetry software allow verification of the processes used to generate the model and 292 should be included as best practice. For 3D scanning, in some cases it may only be feasible to release 293 the raw data in proprietary formats but, where possible, widely compatible (e.g. STL) surfaces should 294 be exported. For methods that involve the digital alignment of different aspects of a specimen, or 295 significant manual intervention in the model construction, the unfused data should be released as 296 the accuracy of the original alignment may be of variable quality.

4. Downstream analyses (morphometric and functional analyses)

It is important to consider not only the generation of 3D models, but also the data that may be produced in the course of downstream analyses to which these data are subjected. Common types of analysis include: (1) Size and shape analyses through topological and landmark-based techniques such as geometric morphometrics; and (2) assessment of the functional performance of specimens through computer modelling approaches, such as finite element analysis (FEA), multibody dynamics analysis (MDA), or computational fluid dynamics (CFD). These studies are often based on 3D models with the data subsequently analysed in specialist software packages [1].

306

307 (a) Data essential for verification

308 Morphometric data: For morphometric approaches, the original landmark coordinates, or the rules 309 defining landmark location should be provided as these constitute the raw data for the 310 morphometric analyses. For 2D landmark data, a .tps file or similar format links landmarks to their 311 constituent images. Where 3D landmark data points are collected via a 3D digitizer, it is common 312 practice to tabulate the specimen number of the digitized specimen. Where the analyses are based 313 on 3D surfaces or digital models, it is desirable that the models (surface or volume) used in the 314 analysis should be published in an accessible format (following the guidelines outlined above). 315 316 Downstream functional data: Functional analyses typically convert 3D digital datasets into

317 proprietary formats for specific methodologies, such as FEA, CFD and MDA. Free software packages

318 do exist, but typically industry standard commercial packages are employed. These have the

319 advantage of reliability and standardized algorithms underpinning the computational analysis.

320

321 Project files or metadata: Specialist software has the disadvantage that it outputs data in proprietary 322 file formats that may not be widely accessible to many potential users. For morphometrics, a text 323 file detailing any corrections or transformations applied to the data and an explanation of the 324 analyses should be published. If the morphometric analysis is conducted in the R environment, an 325 annotated .R script is a convenient solution. For 3D functional analyses, the (usually proprietary) files 326 containing the analysis set-up and parameters, either with or without the results files, are required 327 for model verification. This addition enables a user with access to the appropriate software to 328 replicate the analyses. Full metadata should be provided with details of processing techniques used 329 to generate the final model, as well as a description of any parameters specified by the user in the 330 analysis (Table 1).

331

332 (b) Data required for best practice.

333 Project and results files: Analytical techniques used to investigate the function and biomechanical 334 performance of 3D modelled taxa will produce a range of additional digital data, which should also 335 be made available in order to replicate studies. In the case of FEA, programs use volumetric meshes 336 consisting of a finite number of elements. For MDA and CFD, formats such as the parasolid standard 337 are often essential to perform the analyses. Further parameters and boundary conditions are then 338 defined in specialist software (e.g. Abaqus, Ansys, Strand 7, Adams, Opensim, Gaitsym, COMSOL). 339 Ideally, both the model set-up as well as the result files would be published alongside a study (e.g. 340 [50]). For commercial packages, viewing software is sometimes available which allows the display of 341 models and results files, but no additional analyses. Some industry software packages have text 342 editor readable files that list and detail the location and nature of boundary conditions, e.g. .inp files 343 for Abagus FE software.

344

5. Data repositories

346 Researchers have a responsibility to ensure that all of the data necessary to reproduce a published 347 study are made available. As explained above, for 3D digital datasets these data may include original 348 2D images, prepared/segmented 3D images, 3D geometries, and relevant metadata. These datasets 349 can be, in toto, very large by today's standards; over 100 GB per specimen is possible in some 350 scenarios, and there may be some instances where single publications utilize huge numbers of 351 specimens, the storage of which is in itself a project. Publishers and other institutions hosting 352 repositories must manage and facilitate access to the data they host, with these obligations 353 persisting into the future, ideally indefinitely. Museums and other institutions holding original 354 specimens often consider digital data as an intrinsic aspect of the specimen, and request researchers 355 to deposit these data with them. Many have active programmes of 2D and 3D digital curation and 356 normally make data freely available for research purposes. Data access for commercial use is a 357 source of much needed income, and commercial reuse of data released for research purposes is a 358 genuine concern. However, most museums do not yet have systems, policies, or resources in place 359 for the long-term curation and distribution of digital morphological data [31]. This is not surprising 360 given the paradigm shift in the concept of the accessioned specimen brought about by digital 361 morphology, expanding from the physical specimen to a diversity of avatars.

363 Digimorph.org pioneered the curation of digital morphological data, and there are now a number of 364 general and specialist repositories facilitating the publication and dissemination of supporting data 365 at a variety of scales (Table 3). Many journals have agreements with such repositories and will cover 366 charges, even for relatively large datasets. In addition, many funding agencies are building in 367 facilities to cover costs of long-term data storage, and many institutions have developed their own 368 data repositories to manage research data generated by their own researchers. Out-moded 369 promises to make data "available on request" should give way to permanent URL links to 3D image 370 data in biology, anthropology, and palaeontology (cf. [36]).

371

372 (a) Available data repositories

373 A range of repositories are available that cater for 3D digital datasets arising from research in 374 biological sciences (Table 3). These can vary greatly in terms of the size and types of data they are 375 willing to accept, as well as the cost of storage. In some cases, the choice of repository may be 376 prescribed by the funding body or journal, but this decision will most often be made by the 377 researcher. Modern facilities for publically sharing datasets include national data centres (typically 378 supported by a research funding body; e.g. RCUK data centres), multidisciplinary (e.g. Dryad; 379 [datadryad.org]; figshare [figshare.com], MorphoMuseuM [morphomuseum.com], MorphoSource 380 [morphosource.org], Phenome10K [phenome10k.org], and Zenodo [zenodo.org] or discipline-381 specific (e.g. XROMM [xromm.org]) repositories, and institutional repositories for data produced in-382 house (e.g. Bristol University's Research Data Repository [data.bris.ac.uk/data], Natural History 383 Museum London's Data Portal [http://data.nhm.ac.uk]). It is not entirely clear that all of these are 384 sustainable in the long term. Traditional repositories of physical specimens can also store and 385 disseminate data, and many are moving towards online access to their digital collections.

386

387 (b) Necessary standards for data repositories

Digital repositories should have the same qualities as repositories of physical specimens, in that they should ensure the long-term persistence and preservation of datasets in their published form, provide expert curation, stable identifiers for submitted datasets, and facilitate public access to data without unnecessary restrictions. However, by their very nature, they should also ensure that the data are discoverable online, provided with unique, permanent and citable reference codes (e.g. DOIs), associated with relevant metadata (e.g. .readme text file), and have links to relevant publications and funding bodies [2, 29].

396 The specific license used by the repository should be considered. Many facilities currently use the 397 CC-BY-NC licence, which disallows re-use for commercial activities. This may be desirable where 398 there are concerns over activities such as selling 3D prints of museum specimens with no benefit to 399 the institutions charged with maintaining those collections. Authors may prefer to choose the CC-BY 400 license, which is among the most open creative commons licenses available and has become the 401 standard for open access publication of journal articles. This license lets others distribute, edit and 402 build upon the original data, even commercially, as long as they credit the original creator. The CC-0 403 license (Dryad default) goes further and allows copyright-owners to waive all rights. CC-BY-ND is less 404 attractive, since it allows sharing but does not allow the end user to publish derivatives of the data.

405

406 3D digital datasets associated with published studies should be verifiable and fully traceable from 407 production to publication, and later republication. One option is digital watermarking, which 408 provides a means of achieving verification of the authenticity and integrity of data, is imperceptible 409 to the human eye, but also durable in both digital and printed forms, surviving most image edits, file 410 format conversions, data compression, filtering, and partial data removal, smoothing. Another 411 option would be to require users to register with the repository before data can be downloaded and 412 used, a practice already imposed by some repositories (e.g. Dryad, Morphosource). Registration is 413 usually free and open to everyone, but allows the repository to track data access.

414

415 (c) Costs

416 When publishing large (e.g. > 10 GB) 3D digital datasets, it is vital to consider the financial costs, 417 which are typically proportional to the amount of data being stored. Some repositories do not 418 currently charge for accessions (e.g. Morphosource) but, for some, accession charges are not 419 insignificant. The popular online digital repository Dryad [datadryad.org] currently charges \$120 per 420 data package of 20 GB plus \$50 for each additional 10 GB. Datasets based on synchrotron 421 tomography supporting a single publication can easily run to 100 GB for a relatively small number of 422 scans of individual specimens (e.g. [51]), and it is possible to envisage future projects, especially 423 synthetic papers and large-scale comparative analyses, generating datasets that are orders of 424 magnitude greater in size. Publishing such datasets can quickly become prohibitively expensive; 425 many journals offer to fully or partially cover the costs of depositing digital datasets, but do not have 426 a clear policy for datasets that are 100s GB to TB in size. Applications for research funding are 427 increasingly budgeting for data storage costs, but this does not assist projects making use of pre-428 existing data, or those where funds for data publication are not available.

429

430 One way of minimizing costs is by reducing the total size of data published without compromising 431 the quality. Cropping of redundant space around a volume representing the specimen is an obvious 432 first step. Lossless compression of individual image files is an excellent route to reduce data storage 433 for image stacks in certain formats. For example, LZW compression, both lossless and fully 434 reversible, can provide upwards of 40% reduction in file size on 8-bit TIFFs with no evident effect on 435 data quality, but is often not routinely applied. The PNG image format provides a similar level of 436 lossless compression. As noted above, the JPEG image format enforces lossy compression that 437 degrades data, and should not be used despite appealingly high compression ratios. Placing files into 438 ZIP archives (e.g. one ZIP file per image stack) also reduces disk space through lossless compression 439 and is more convenient for downloading. However, ZIP and .VOL archives are less secure for long-440 term storage, since, if the single file containing a dataset becomes corrupted, the entire dataset will 441 be lost. Corruption of single files within a large dataset is less serious, and at least some repositories 442 have procedures in place to detect and remediate bitrot [32]. We recommend that unarchived 443 copies of the original data are stored and made available where possible.

444

In our enthusiasm for recycling 3D digital data and easing reproducibility of morphological studies based on them, the environmental costs of storage should be considered. Most datasets will be accessed infrequently and so there is no need or justification for their storage on spinning disks. Many repositories make use of automated tape storage which is stable and comparatively low in direct costs for the same reasons that make it environmentally low-cost. However, in such cases data will not be available instantly on demand and access will instead have to be requested.

452 6. Rescuing legacy data and constraints on data use

453 An increase in the availability and ease of use of data repositories raises the prospect of making data 454 available from previously published studies where the data were not released at the time of 455 publication. Digital datasets can be uploaded to online data repositories and linked to past 456 publications. At present there are no policies or mechanisms we are aware of among journals and 457 publishing houses to link archival publications to newly deposited data. However, there is no 458 material technical barrier to salvaging legacy data in this way. Publishers are likely to welcome such 459 an initiative since it would obviously improve data visibility, facilitate reproducibility, and likely 460 rejuvenate old publications in terms of access, citations and, ultimately, their marketability. 461

462 Obtaining digital characterizations of morphology can be time-consuming and expensive, and 463 researchers rarely exhaust their data with the first publication. Funders and publishers are 464 increasingly removing choice over whether to release supporting data, and so it can seem unfair that 465 the researchers who generated datasets have to subsequently compete to exploit them further. This 466 can be particularly difficult for lone early career researchers potentially competing with large 467 experienced research groups [34]. One potential solution to this would be the introduction of time-468 limited embargos, which can already be facilitated by some data repositories. However, such 469 embargos violate the most basic tenet of open data, that of removing barriers to assessing the 470 reproducibility of research [52]. After the point of publication, it is also effectively impossible to 471 police the release of supporting data and, consequently, we see no alternative to the release of data 472 with publication. A possible compromise may be borrowed from the Bermuda [53], Fort Lauderdale 473 [54], and Toronto [55] agreements of the genomics community. These mandate data release at the 474 time they are obtained but, more germane to morphologists, these agreements provide 475 safeguarding for data generators through published, time-limited, statements of intent of how they 476 propose to exploit the data [55]. Other researchers are free to exploit the data for other purposes, 477 and for any purpose after the stated period of limitation of the statement of intent [56]. Third party 478 users with overlapping research interests are expected to proceed respectfully and in dialogue with 479 the data generators to identify a mutually agreeable publication schedule [55]. Invariably, much 480 more is at stake in such projects, and though these informal agreements are rarely violated, they are 481 generally well-policed by the peer review process [56], and by the reputational damage suffered by 482 those who choose not to observe these agreements.

483

484 Practice in the genomics community underscores the point that there is more to gain from open 485 data than the warm glow of altruism [55, 57]. Not only has it led to greater and more rapid scientific 486 advance [52, 55], it can lead to material personal gain, through the proposals for collaborative 487 exploitation of published data, both to achieve stated research objectives, and to achieve new 488 objectives that would not be possible without unforeseen collaborators [55, 57]. Citation and access-489 tracking of published datasets provides credit to the authors [32]. Attribution of authorship is 490 mandated under CC-BY licenses and is in any case integral to the academic culture. Many journals 491 already mandate citation of published datasets, not (or not merely) the publications describing 492 research based upon them; this must become common practice. Further mechanisms of 493 encouraging researchers to share their data should only add to this motivation, such as explicitly 494 evaluating the open sharing of data as part of CVs in hiring, promotion or other award processes.

495

496 Nevertheless, data can be associated with ethical sensitivities that may require the withholding, or
497 restriction on public distribution, of data (e.g. anthropology or medical science [58, 59]). In such
498 instances, the issues that apply should be clearly defined so that beyond these boundaries
499 researchers and publishers can follow an ethos of open data publication. Mechanisms already exist
500 to cope with these constraints while still making data available, such as data anonymization and
501 vetted access [55].

502

503 **7. Outstanding challenges**

504 While the principle of open data has been mandated by the majority of funders [33], publishers, 505 physical repositories and researchers are all scrambling to meet the resulting challenges. Above all, 506 the competing interests over ownership of digital data need to be resolved between: (i) funders who 507 pay for research, (ii) researchers who collect specimens and create the digital datasets, (iii) research 508 facilities where data are collected, (iv) museums who have a duty of care for the physical specimens, 509 and (v) research publishers. Funders, researchers, and publishers may have converged on an ethos 510 of open data. However, the institutions that are responsible for the physical specimens have not 511 obviously been invited to engage in the development of open data policy, and yet it is museums that 512 will have to change most in terms of their policies on the nature of what they consider intrinsic 513 aspects of the physical specimens that they hold in their care. One solution for museums might be to 514 comply with research funders' requirements, and waive copyright over digital representations of 515 their collections, along with its associated income stream. Another solution would be for these 516 institutions, which are those best-placed to inform policy on the curation, storage and distribution of 517 data, to develop digital collections with the stability to match that of their physical inventory. 518 Indeed, with the development of cybertypes [29, 30], this may be an inevitable future aspect of the 519 world's leading museums. However, if this readily realisable vision of data repository quality, 520 stability, and credibility, is to be achieved, it will require the funders who have mandated data 521 deposition to cover the costs of establishing and maintaining such facilities, through block grants, 522 not through piecemeal funding to researchers. If such change is to be achieved, it must not only 523 happen in wealthier countries, but worldwide and, thus, more amply provisioned funders should 524 provide further means to help other countries improve their data-sharing capacities. 525 526 Data access is not only important post-publication, to aid reproducibility, but during peer review, so

527 that the results of a study and their interpretations can be verified prior to publication. Providing

528 tomographic or 3D data at the point of journal submission is, in our experience, a comparatively rare 529 phenomenon that the publishing infrastructure is not currently well set up to facilitate. Publishers 530 must develop a more homogenous policy on open data [35], along with procedures to ensure data 531 sources are acknowledged and linked electronically to the derivative publications [52]. It is also 532 important that systems are developed to ease the submission of such data, and facilitate secure, 533 anonymised distribution of data to reviewers. Dryad offers an integrated submission system where 534 publishers can coordinate submission of a manuscript with submission of data, which can then be 535 accessed securely by referees and editors. For non-integrated journals, an interim solution may be 536 to host data at a temporary, hidden-URL that can be forwarded to the reviewers via the journal. 537 Authors may be cautious about sharing such data ahead of an article being accepted for publication, 538 and there should be a clear policy governing the restrictions of use for reviewers.

539

540 8. Conclusions

541 Data sharing is essential in order for the benefits of 3D digital data to be fully realized by the 542 scientific community, as well as for the maximum benefit to be gained from the public and private 543 funding that allows these data to be collected. Not only are the benefits of 3D digital data not 544 currently being fully realized, but failure to publish supporting data is rendering many studies based 545 on 3D digital data at least difficult to reproduce. We have presented a series of proposals for open 546 3D digital data. These outline the minimal standards of verifiability that studies should meet before 547 they are published. We also present more ambitious standards that we hope can be assumed as 548 normal best practice (Table 1). We have all been guilty of failing to meet these standards in the past 549 because of technical and other limitations; however, technology has changed and so must we. There 550 are costs associated with releasing data, both real and in-kind, but these are insignificant in 551 proportion to the real costs of regenerating the data, and the reputational costs to individuals, 552 institutions, journals and editors, of publishing research predicated upon inaccessible data. 553 554 Ethics statement: No data were harmed in the formulation of this contribution. 555 Data accessibility statement: There are no data associated with this manuscript. 556 **Competing interests statement:** The authors declare there are no competing interests. 557 Authors' contributions statement: The project was conceived by TGD, IAR, SL, JAC, EJR, and PCJD, all

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727	Figure and table captions
728	
729	Table 1. Summary table of recommendations for types of data files that should be published in
730	support of published articles.
731	
732	Table 2. Summary of the principles of open data for digital morphology.
733	
734	Table 3. Summary of main online repositories for 3D digital morphological data.
735	
736	Figure 1. Examples of digital data and downstream uses. (a) Medical CT image of the dentary of the
737	holotype of Tyrannosaurus rex CM 9380. (b) Reconstructed MicroCT dataset of vascular injected
738	green iguana (Iguana iguana) skull [OUVC 10677]. (c) Slice through braincase region of microCT
739	scanned Iodine-potassium iodide (I ₂ KI) stained contrast-enhanced grey squirrel (Sciurus carolinensis)
740	skull. (d) MRI scan midline slice of neonatal white rhino (Ceratotherium simum). (e) Synchrotron
741	Radiation X-ray Tomographic Microscopy (SRXTM) partial reconstruction of putative red alga from
742	the Ediacaran Weng-an Biota, South China. (f) Digital reconstruction of Offacolus kingi, a chelicerate
743	from the Silurian of Hertfordshire, UK, reconstructed via serial grinding and optical microscope
744	photography; Inset: digital segmentation of microphotograph. (g) image of stl (stereolithography)
745	file of skull of foetal Tammar wallaby Macropus eugenii. (h) Optical projection tomography of mouse
746	hindlimb at embryonic stage E19, stained with Alcian blue and Alizarin red and imaged using visible
747	light and fluorescent light to image cartilage and bone respectively (image courtesy of Karen Roddy).
748	(i) Photogrammetry reconstruction of guineafowl trackway. (j) Surface scan of human subject, with
749	subject-specific skeleton and muscle volumes segmented from MRI scan data and resulting
750	multibody dynamics analysis (MDA) model of same subject. (k) SIMM (Software for Interactive
751	Musculoskeletal Modelling) model of Tyrannosaurus rex hindlimb. (/) MicroCT scan reconstruction of

- the skull of the common buzzard, Buteo buteo, detailing landmarks and semilandmarks used for
- 753 geometric morphometrics (GMM) analysis (reproduced with permission from Bright et al. [60]). (m)
- Finite element (FE) model of the skull of *Allosaurus fragilis* (reproduced with permission from
- T55 Lautenschalger & Rahman in press). (*n*) Results of CFD simulation of water flow around a 3D model
- of the cinctan echinoderm *Protocinctus mansillaensis*. All images obtained from authors unless
- 757 stated otherwise.
- 758