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 Open data and digital morphology

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Open data and digital morphology

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71

Abstract

72

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

Keywords:

85

86   digital data, 3D models, phenotype, computed tomography, visualization, functional analysis

87

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

113 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

115 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

133   institutional policies asserting copyright ownership (e.g. public museum or even private collections)

134   and the regulations of funding bodies and publishers with regard to open data. Consequently,

135   sharing or publishing supporting data is often a low priority and has effectively been considered

136   optional when not prescribed by a journal. Partial datasets (e.g. low-resolution visualizations or

137   external surfaces) can be insufficient for reproducibility or even verification. As digital morphology

138   has evolved, most of us in the research community have failed to achieve what might now be

139   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

212   downstream analyses.

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

247   radiograph projections, not the resulting slice data [46]. Thus, it may be sensible for authors to

248   archive the raw projection libraries themselves. This is especially important where access to the

249   same specimen may be problematic, or as a precaution in case unique specimens are damaged, lost

250   or destroyed.

251

252   *Image registration:* For physically destructive and optical tomography, tomograms need to be

253   registered (aligned relatively and absolutely in the X, Y, and Z planes, either manually or semi-

254   automatically) prior to any reconstruction of 3D models. This adds a potentially subjective step that

255   may have a bearing on downstream analyses, and so we recommend publishing both the original

256   (unregistered) and registered image stacks as best practice.

257

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

259   Alternative surface-based methods exist for digitizing only the exterior features of specimens in 3D,

260   most notably laser or structured light scanning [47] and photogrammetry [1, 48, 49]. For

261   photogrammetry, data begin as 2-D photographs, whereas in surface-scanning techniques, the 3D

262   shape is usually directly captured as 3D point clouds, with or without texture capture (colour) for

263   each point. In photogrammetry, a 3D polygonal mesh with texture data is generated and warped

264   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.

297

298   **4. Downstream analyses (morphometric and functional analyses)**

299   It is important to consider not only the generation of 3D models, but also the data that may be

300   produced in the course of downstream analyses to which these data are subjected. Common types

301   of analysis include: (1) Size and shape analyses through topological and landmark-based techniques

302   such as geometric morphometrics; and (2) assessment of the functional performance of specimens

303   through computer modelling approaches, such as finite element analysis (FEA), multibody dynamics

304   analysis (MDA), or computational fluid dynamics (CFD). These studies are often based on 3D models

305   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 Abaqus FE software.

344

345 **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.

362

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**

388 Digital repositories should have the same qualities as repositories of physical specimens, in that they

389 should ensure the long-term persistence and preservation of datasets in their published form,

390 provide expert curation, stable identifiers for submitted datasets, and facilitate public access to

391 data without unnecessary restrictions. However, by their very nature, they should also ensure

392 that the data are discoverable online, provided with unique, permanent and citable reference codes

393 (e.g. DOIs), associated with relevant metadata (e.g. .readme text file), and have links to relevant

394 publications and funding bodies [2, 29].

395

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

445   In our enthusiasm for recycling 3D digital data and easing reproducibility of morphological studies

446   based on them, the environmental costs of storage should be considered. Most datasets will be

447   accessed infrequently and so there is no need or justification for their storage on spinning disks.

448   Many repositories make use of automated tape storage which is stable and comparatively low in

449   direct costs for the same reasons that make it environmentally low-cost. However, in such cases

450   data will not be available instantly on demand and access will instead have to be requested.

451

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

569

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725

726

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_2KI) 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. (*l*) MicroCT scan reconstruction of

752 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*)

754 Finite element (FE) model of the skull of *Allosaurus fragilis* (reproduced with permission from

755 Lautenschalger & Rahman in press). (*n*) Results of CFD simulation of water flow around a 3D model

756 of the cinctan echinoderm *Protocinctus mansillaensis*. All images obtained from authors unless

757 stated otherwise.

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