

# Paleomass for R—bracketing body volume of marine vertebrates with 3D models (#87526)

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First submission

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# Paleomass for R—bracketing body volume of marine vertebrates with 3D models

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Body mass is arguably the most important characteristic of an organism, yet it is often not available in biological samples that have been skeletonized, liquid-preserved, or fossilized. The lack of information is especially problematic for fossil species, for which individuals with body mass information are not available anywhere. Multiple methods are available for estimating the body mass of fossil terrestrial vertebrates but those for their marine counterparts are limited. Paleomass is a software tool for estimating the body mass of marine vertebrates from their orthogonal silhouettes through bracketing. It generates a set of two 3D models from these silhouettes, assuming superelliptical body cross-sections with different exponent values. By setting the exponents appropriately, it is possible to bracket the true volume of the animal between those of the two models. The original version phased out together with the language platform it used. A new version is reported here as an open-source package based on the R scripting language. It inherits the underlying principles of the original version but has been completely rewritten with a new architecture. For example, it first produces 3D mesh models of the animal and then measures their volumes and areas with the VCG library, unlike the original version that did not produce a 3D model but instead computed the volume and area segment by segment using parametric equations. The new version also exports 3D models in polygon meshes, allowing later tests by other software. Other improvements include the use of NACA foil sections for hydrofoils such as flippers, and optional interpolation with local regression. The software has a high accuracy, with the mean absolute errors of 1.33% when the silhouettes of the animals are known.

1 Paleomass for R—bracketing body  
2 volume of marine vertebrates with 3D  
3 models

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## 15 Abstract

16 Body mass is arguably the most important characteristic of an organism, yet it is often not  
17 available in biological samples that have been skeletonized, liquid-preserved, or fossilized. The  
18 lack of information is especially problematic for fossil species, for which individuals with body  
19 mass information are not available anywhere. Multiple methods are available for estimating the  
20 body mass of fossil terrestrial vertebrates but those for their marine counterparts are limited.  
21 Paleomass is a software tool for estimating the body mass of marine vertebrates from their  
22 orthogonal silhouettes through bracketing. It generates a set of two 3D models from these  
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24 setting the exponents appropriately, it is possible to bracket the true volume of the animal  
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31 segment using parametric equations. The new version also exports 3D models in polygon  
32 meshes, allowing later tests by other software. Other improvements include the use of NACA  
33 foil sections for hydrofoils such as flippers, and optional interpolation with local regression. The  
34 software has a high accuracy, with the mean absolute errors of 1.33% when the silhouettes of the  
35 animals are known.

## 36 Introduction

37 Body mass is an essential metric to describe aspects of the biology of individual organisms  
38 (Schmidt-Nielsen, 1984). Despite the importance, a body mass record is not always available —  
39 preserved specimens in museum collections often lack body mass information, and fossil  
40 organisms are never found with body mass data. The lack of information is not overly  
41 problematic for extant species for which conspecific individuals are available elsewhere, but  
42 poses a critical hurdle to biological studies of fossil species. Accordingly, paleontologists have  
43 been exploring the possibility of body mass estimation based on what is preserved in fossils.

44 Methods for body mass estimation based on fossils are largely divided into two  
45 categories depending on the underlying principle—one may be called the length-based and the  
46 other volumetric approaches (Hurlburt, 1999; Smith, 2002; Sellers et al., 2012). The length-  
47 based approach first establishes a correlation between the length(s) of one or more  
48 morphological character(s) of the animals in question and their body mass through a linear  
49 regression, based on extant samples for which the body mass is known, and then uses the  
50 regression equation to estimate the body mass of extinct animals for which the length  
51 character(s) are available. Multiple regression with more than one length characters tends to be

52 preferred in clades in which fossils species are nested among abundant extant members, such as  
53 mammals (e.g., Smith, 2002; Mendoza, Janis & Palmqvist, 2006), whereas bivariate regression is  
54 almost exclusively used in clades that have long been extinct with only distantly related  
55 descendants surviving, e.g., non-avian dinosaur clades (Anderson, Hall - Martin & Russell,  
56 1985; Campione & Evans, 2012), probably to avoid overfitting of the model to particular extant  
57 clades that would mislead the outcome.

58 The volumetric approach first estimates the volume of the animal in question and then  
59 converts the value to body mass by assuming an average body density. This approach dates back  
60 at least to 1905, when the body mass of *Brontosaurus* was estimated by measuring the volume of  
61 a cast of a scaled physical model with water displacement and then converting the volume to  
62 mass by assuming the freshwater density (Gregory, 1905). A similar method was used by  
63 (Colbert, 1962) for body mass estimation of broader dinosaurs. As mathematical models became  
64 common, a parametric approach to model the body as a collection of cylinders based on a limited  
65 number of measurements, called Graphic Double Integration, was developed (Jerison, 1973).  
66 What may be considered an extension of this approach, where the body is straightened in a  
67 parametric space and modeled by many cylindrical disks, was later proposed (Seebacher, 2001).  
68 With the arrival of 3D computer technology, methods of incorporating complex 3D computer  
69 models emerged. Such methods include a partly parametric approach as in Paleomass based on  
70 superelliptical cross-sections (Motani, 2001), as well as the minimum convex hull method based  
71 on completely empirical data from laser scanning of mounted skeletons (Sellers et al., 2012).

72 These methods aim to arrive at the best mean estimate of body mass, except Paleomass  
73 which tried to bracket the mass between two values (Fig. 1D bracketed by C and E). The method  
74 was also unique for specifically addressing marine vertebrates, for which a limb-based regression  
75 approach is not suitable because they do not support the body mass with the limbs. Despite the  
76 uniqueness that would allow cross-checking of other methods, the software is no longer available  
77 because its language platform was discontinued. The purpose of the present paper is to report a  
78 completely rewritten and open-source version of Paleomass with a new architecture and  
79 enhancements over the original version.

## 80 Materials & Methods

### 81 Platform

82 The new Paleomass was written in the R scripting language and run on the R platform (R-Core-  
83 Team, 2020). Apart from the default R packages, it relies on the following packages for parts of  
84 computation: imager (Barthelmé & Tschumperlé, 2019), locfit (Loader, 1999), Morpho  
85 (Schlager, 2017), plot3D (Soetaert, 2022), rgl (Murdoch, 2001), and Rvcg (Schlager, 2017). It is  
86 open-source and provided under GNU General Public License v3.0. A repository for the  
87 package, including the code and a tutorial, is found at: <https://github.com/rmotani/paleomass>.

## 88 Aim

89 Paleomass aims to estimate the body volume of a marine vertebrate with a straight body axis.  
90 The volume is converted to a mass by assuming the average body density that can be specified  
91 by the user. The body surface area is also estimated simultaneously.

## 92 Principle

93 Paleomass aims to bracket the true body volume of a marine vertebrate between those of two 3D  
94 models (Fig. 1). Each of the two models is not the best mean estimate of the true body shape, but  
95 one is expected to have a volume slightly larger than the true body volume (Fig. 1E versus D),  
96 and the other slightly smaller (Fig. 1C). The models are based on the same set of orthogonal  
97 body silhouette images and therefore appear identical in completely dorso-ventral or lateral  
98 views, but have different cross-sectional shapes and differs in coronal view.

99 The cross-sectional shape is based on superellipses (Fig. 1A), which are mathematical  
100 expansions of ellipses. Whereas ellipses are defined as:

$$101 \quad (x/a)^2 + (y/b)^2 = 1$$

102 superellipses are defined by an equation:

$$103 \quad |x/a|^n + |y/b|^n = 1 \quad (1)$$

104 where  $n > 0$ . When  $n=2$ , a superellipse becomes an ellipse (Fig. 1). As  $n$  decreases from 2, the  
105 superellipse approaches a diamond shape as  $n$  approaches 1 and then a cross shape as it  
106 approaches 0. If  $n$  increases beyond 2, the superellipse approaches a rectangle.

107 It is known that a typical body cross-section of a marine vertebrate can be approximated  
108 by a superellipse or a combination of two halves of different superellipses (Motani, 2001). The  
109 true body cross-sections of marine tetrapods are usually found to be bracketed by two  
110 superellipses, one with  $n=2$  and the other with  $n=3$  (but see Validation below for a narrower  
111 range). For fish, the two exponents are  $n=1.5$  and 2.5. Therefore, the true volume of a marine  
112 vertebrate can be bracketed by making two 3D models with these two boundary superelliptical  
113 shapes, depending on the clade (Fig. 1).

## 114 Overall workflow

115 Paleomass first reads in the data from raster images and command line options, based on which it  
116 computes 3D mesh models for the main body and each of the fins and flippers separately. Two  
117 mesh models are made for the main body, with different superelliptical exponents of choice. The  
118 volume and surface area of each mesh model are computed and summed to give two total  
119 estimates, with different main body models. Optionally, these meshes are assembled to make a  
120 complete 3D mesh model. The assembled models and each part model can be saved as 3D  
121 polygon meshes, respectively.



## 122 Coordinate system

123 Modeling and computation take place in a three-dimensional Euclidean coordinate system. The x  
124 axis is set as the bilateral axis with the right side of the body being the positive side. The y axis is  
125 the dorsoventral axis with the dorsal direction being positive, while the z axis is the antero-  
126 posterior axis, which may also be called the body axis hereafter, with the tip of the snout being  
127 the origin and the posterior direction being positive.

## 128 User supplied data

129 The users need to supply the shape and size of the animal to be modeled. First, the shape is  
130 supplied as a set of silhouette raster images, such as JPEG or PNG, one for each  
131 fin/flipper/cephalofoil and a pair for the body (e.g., Fig. 2C). These images need to have the  
132 same pixel size, e.g., if each side of pixel is 0.001 m in one image, then this pixel side length  
133 should be the same in all other images. It is recommended to have at least 3000 pixels along the  
134 body axis of these images (see Validation below), rather than 1000 as originally suggested  
135 (Motani, 2001). The body images are in lateral and dorsoventral views, respectively, with all  
136 fins, flippers and cephalofoils removed. For each fin, flipper, and cephalofoil, a planar view is  
137 required. Second, the length of the body axis as represented in the body images after the removal  
138 of the fin/flipper/cephalofoil is supplied through a command line option, in meters.

139 Paleomass accepts the following types of fins and flippers: pectoral fin/flipper, pelvic  
140 fin/flipper, caudal fin, dorsal fin, second dorsal fin, and anal fin. Not all fins/flippers have to be  
141 present. This versatility allows for different body architectures to be modelled (Fig. 4).

## 142 Computation steps for main body

143 The computation of a 3D model and its volume for the main body follows the steps below.

144 (1) The lateral and dorso-ventral silhouettes of the main body of the animal in question are read  
145 from raster image files (Fig. 3A, B).

146 (2) The outlines of these silhouettes are digitized as coordinates (Fig. 3C, D), which are then  
147 optionally smoothed through interpolation with local regression using the locfit() function  
148 (Loader, 1999). By default, a nearest neighbor parameter of 0.1 and a constant component of 0 is  
149 used for local regression but the former value is user adjustable. The smoothing allows  
150 coordinates to take non-integer values and therefore prevents step-like appearance of the final 3D  
151 model (Fig. 3G, H) that often gives rise to non-manifold edges and triangles that cause errors  
152 later on.

153 (3) The transverse and dorsoventral diameters of the main body are calculated from the  
154 coordinates for each pixel position along the body axis (Fig. 3C, D). There are much less than  
155 3000 lines in Fig. 3C and D for visualization purposes but the actual calculations are done for  
156 each pixel point along the body axis, i.e., there would be 3000 pairs of transverse and  
157 dorsoventral diameters in the input body images have 3000 pixels along the body axis.

158 (4) Based on these coordinates and diameters, a superellipse is drawn per segment (Fig. 3E, F),  
 159 i.e., body mages with 3000 pixels along the body axis will result in 3000 superellipses. Each  
 160 superellipse has 181 vertices around its perimeter so that there is one vertex per every 2° of  
 161 angular displacement around the center, with the first and last vertices overlapping—these two  
 162 vertices will be merged later to make the model watertight, reducing the number of vertices per  
 163 segment to 180. The number of vertices per segment is user adjustable. The exponent for the  
 164 superellipse (n in equation 1) is also set by the user, e.g., 2 for one model (e.g., Fig. 3E) and 2.4  
 165 for the other (Fig. 3F) for marine tetrapods.

166 (5) A tip is added at each of the anterior and posterior ends of the body to help make the model  
 167 watertight at a later stage. These tips are small superelliptical disks with a tiny radius of 10<sup>-4</sup>  
 168 pixels. They do not affect the computation of volume and surface area. The radius of the tip is  
 169 user adjustable.

170 (6) Superellipses from steps 4 and 5 are connected as a 3D mesh (Fig. 3G-J).

171 (7) Small holes at the tips of the body are closed by merging closely located vertices within  
 172 distances of 10<sup>-4</sup> pixels or less, and then the whole mesh is cleaned for duplicate faces and non-  
 173 manifold faces and vertices by `vcgClean()` function (Schlager, 2017). Cleaning may fail if  
 174 smoothing is skipped at step 2, leaving non-manifold edges that would prevent accurate volume  
 175 calculation. Also, sticky non-manifold edges may result from having low-resolution input  
 176 images—having 3000 rather than 1000 pixels are necessary along the body axis would help  
 177 prevent this unintended error.

178 (8) The volume and surface area of the model are measured by `vcgVolume()` and `vcgArea()`,  
 179 respectively (Schlager, 2017). Initially, they are calculated in cubic pixels and square pixels,  
 180 respectively, where pixel size is as in the input images. These values are then converted to m<sup>3</sup>  
 181 and m<sup>2</sup>s using the body axis length provided by the user, in combination with the number of  
 182 pixels along the body axis in the input images.

## 183 Computation steps for fins and flippers

184 The computation of a 3D model and its volume for a fin or flipper follows the steps below.

185 (1) The planar image of a fin is read from a raster image (Fig. 3K).

186 (2) The outline of the image is digitized as coordinates (Fig. 3L), and smoothing through local  
 187 regression is applied as in the main body outline. The default nearest neighbor parameter for  
 188 local regression is 0.1 but Fig. 3O was produced with a value of 0.05.

189 (3) NACA 4-digit foil section is drawn at each pixel point along the span of the fin (Fig. 3M).  
 190 Symmetrical sections without a camber are used. The equation for such a section is given by:

$$191 \quad y = 5t[0.2969x^{0.5} - 0.126x - 0.3516x^2 + 0.2843x^3 - 0.1015x^4]$$

192 where x is the position along the chord given as a fraction between 0 and 1, and t is the thickness  
 193 of the foil relative to the chord in percentages (Ladson & Brooks, 1975). The base value of t is  
 194 set at 10 for the anal and second dorsal fins and 20 for the rest—these values are user adjustable.

195 When using the base thickness to construct a fin, the thickness distribution along the span  
 196 becomes proportional to the chord length distribution and thus results in a strange shape. Most

197 importantly, the part of the fin that is supposed to be thickest along the fin span, e.g., proximal  
198 end of the pectoral fin/flipper, is not always reconstructed with the maximum thickness. To avoid  
199 this, the base thickness is scaled by a thickness envelope calculated with the following steps.  
200 First, the point along the span where the maximum thickness is expected is specified as a fraction  
201 between 0 and 1. For example, this point would be 0 for pectoral fin/flipper and 0.5 for a  
202 symmetric caudal fin. Second, the axis from the thickest point to an end of the fin span is given  
203 new coordinates of 1 to 0, with 1 at the thickest point 0 at the distal tip. Lastly, the square roots  
204 of these values are calculated to form the thickness envelope to scale the raw thickness based on  
205 the chord lengths. For example, at the midpoint between the thickest point and a fin tip, the raw  
206 thickness is multiplied by  $0.5^{0.5}$  to give a scaled thickness. This scaling was not present in the  
207 original Paleomass.  
208 (4) Foil sections from the previous step are connected to produce a 3D mesh (Fig. 3N, O), which  
209 are then cleaned as in the body mesh.  
210 (5) The volume and surface areas are measured as in the body mesh.  
211 (6) The processes above are repeated for all fins/flippers.

## 212 Computation of cephalofoil

213 A simple cephalofoil model is implemented to accommodate hammerhead sharks. A cephalofoil  
214 mesh is built in the same manner as fins and flippers, in that serial NACA foil sections are used.  
215 However, unlike fins and flippers that gradually thin out toward the tip, the two ends of the  
216 cephalofoil, where the eye sockets are located, are thickened.

## 217 Body and fin integration

218 This process is for visualization purposes only at present and does not affect the volume/area  
219 estimation. Paleomass allows adjustment of the position and angle of each fin/flipper relative to  
220 the main body through command line options. Specifically, positioning along the x, y, and z  
221 axes, as well as rotation around these three axes can be adjusted. Rotations are called pitch, yaw,  
222 and roll, around the x, y, and z axes, respectively. Roll is applied first, followed by pitch, and  
223 then yaw.

## 224 Mass calculation

225 Once the volume of each component is estimated, they are summed to give a total volume. If  
226 there is overlap among components, then the overlapping part is counted twice. However, such  
227 overlap is usually limited compared to the body volume and would not cause a significant error  
228 as evident from the validation results given later. It is ideal to find a Boolean union of 3D meshes  
229 but such a function is not yet stably available in R. Future development may allow addition of a  
230 Boolean union procedure.

231 With the total volume estimated, body mass is calculated from the volume by assuming a  
232 mean density of the total body. For marine vertebrates with buoyancy control through an air  
233 bladder or lungs, it is expected that neutral buoyancy is experienced in at least a part of daily life.  
234 The neutral buoyancy near the sea surface would suggest a mean body density of  $1.027 \text{ g/cm}^3$ ,  
235 and that in pure water is approximately  $1 \text{ g/cm}^3$  (Stewart, 2008). By default, Paleomass uses  
236 these two values, although one of them is user adjustable.

237 The total body density of vertebrates has been controversial. Colbert (1962) used  $0.9$   
238  $\text{g/cm}^3$  based on a value from *Alligator mississippiensis*—however, alligators are capable of being  
239 suspended in freshwater, i.e., their mean density could become about  $1 \text{ g/cm}^3$  depending on how  
240 much air is in the lung. Sellers et al. (2012) used a value of  $0.896 \text{ g/cm}^3$ , which they calculated  
241 based on a dataset from a frozen horse reported by Buchner et al. (1997). However, a  
242 reexamination of this dataset suggests that the value should be  $0.915 \text{ g/cm}^3$ — $0.896$  would be  
243 derived instead if the limbs from only one side of the body are included in calculation. This last  
244 value of  $0.915$  is almost identical to the density of ice, so freezing of the specimen may have  
245 biased the data, i.e., the true value may be close to  $1$  without freezing. The total body density of  
246 sharks in Florida varied between  $1.02$  and  $1.07 \text{ g/cm}^3$  (Baldrige, 1970), whereas the same in rat  
247 varied between  $1.03$  and  $1.09 \text{ g/cm}^3$  depending on the body fat content (Dahms & Glass, 1982).  
248 Overall, freshwater or seawater density still remains the most reasonable value, especially for  
249 marine vertebrates.

## 250 Validation

251 The accuracy of the software was tested in two ways. First, its accuracy under the best condition  
252 was tested by geometric objects of known volume and area, a sphere and prolate spheroid.  
253 Second, its ability to bracket the true volume and surface area of actual aquatic vertebrates was  
254 tested. In both tests, Paleomass was used with interpolation with local regression enabled.

## 255 Geometric objects

256 The first test followed the steps below. A circle and an ellipse were drawn in CorelDraw  
257 and exported as raster images, respectively, so that the long axis of the object varies from  $100$  to  
258  $10000$  pixels, with an increment of  $100$  between  $100$  and  $1000$ , and  $1000$  between  $1000$  and  
259  $10000$ . Then, the volumes of spheres and prolate spheroids based on these images were estimated  
260 by using each image as both the lateral and dorso-ventral views for the body in Paleomass, per  
261 run. The estimated values were then compared to the true values from parametric equations  
262 describing the volume and surface area of spheres and prolate spheroids. The result shows that  
263 the error is less than  $0.5\%$  in both volume and surface area estimation as long as the resolution of  
264 the input image is high, with at least about  $800$  pixels along the long axis (Fig. 5). However, to  
265 stably obtain best results, it is recommended to have  $3000$  or more pixels along the body axis

266 (Fig. 5B). Such a high resolution is also beneficial in minimizing unintended production of non-  
267 manifold edges as mentioned earlier.

### 268 3D models of actual animals

269 The second test is based on 3D mesh models of 25 marine vertebrate species, digitized  
270 from actual animals. Only those 3D models that were produced in association with universities  
271 (Kano et al., 2013; Irschick et al., 2021) were used. The species include 20 osteichthyes, 3  
272 chondrichthyes, and 2 cetaceans—the list of species used is given in supplementary information.  
273 The uneven distribution across clades reflects biased data availability that cannot be easily  
274 amended.

275 First, the true volume and surface area of each animal were recorded, after making its 3D  
276 model watertight in Meshlab (Cignoni et al., 2008). This involved removal of duplicate vertices  
277 and faces, followed by an iteration of a sequence comprising deletion of non-manifold edges and  
278 self-intersections and filling of the resulting holes. If error-causing borders remained after the  
279 iteration, the borders were removed and the iteration sequence was reinitiated. Second,  
280 Paleomass estimates of the volume and surface area were calculated based on the lateral and  
281 dorso-ventral images of the model, which were captured under orthographic projection in  
282 Meshlab (Fig. 2A), together with images from angles that reveal the planar views of individual  
283 fins/flippers (Fig. 2B). Attention was paid **not to** change the magnification between image  
284 captures. These images were edited in CorelDraw to separate fins/flippers from the body and  
285 then each part was saved as a raster image (Fig. 2C). The image resolution was set so that there  
286 are 3000 pixels along the long axis of the main body. Paleomass estimates were made for  
287 superelliptical exponents ( $n$  in equation 1) from 1.5 to 3.0 by an increment of 0.1. Finally, the  
288 Paleomass estimates were compared to the true values to test if the latter were bracketed by any  
289 pair of the Paleomass estimates.

290 The results are summarized in Fig. 6. **In all cases, the true volumes of the marine**  
291 **vertebrates were found to be bracketed between Paleomass estimates with superelliptical**  
292 **exponent values of 1.6 and 2.4.** Within this range, cohorts are recognized based on how round  
293 the ventral half of the body **coronal** sections are—some species have rounded ventral halves that  
294 appear U-shaped (e.g., Fig. 6E) and found toward the right side of the plot, whereas others have  
295 sharper ventral halves appearing closer to a V-shape (e.g., Fig. 6C) and located toward the left  
296 side. In sharks, this is upside down, i.e., it is the shape of the dorsal halves that may be rounded  
297 or  $\Lambda$ -shaped (Fig. 6A). Most osteichthyes in the data have intermediate ventral halves between  
298 V- and U-shape (e.g., Fig. 6D) and consequently found in a moderate exponent range of 1.8 to  
299 2.1. However, unusual forms are found outside of this typical range—those with flattened ventral  
300 sides, such as pufferfish and eels, are in the range of 2.2 to 2.4, whereas those with exceptionally  
301 compressed cross-sections with V-shaped ventral halves, such as flatfish and small herring, are  
302 in the range of 1.6 to 1.7. Cetaceans, with their **coronal** sections rounded ventrally, have a high  
303 optimal exponent range of 2.0 to 2.3. Sharks tend to have  $\Lambda$ -shaped dorsal halves but this is

304 partly compensated for by the flat ventral halves, resulting in a moderate exponent range of 1.8  
305 to 2.0. The optimal exponent range for the surface area was between 1.6 and 2, when excluding  
306 unusual forms such as flatfish and pufferfish. These optimal ranges mostly overlap the  
307 previously suggested ranges (Motani, 2001) while being narrower and better defined.

308 The accuracy of Paleomass estimates was computed in the following manner. Paleomass  
309 provides a range of estimates rather than a single mean estimate, while the latter would be  
310 required to compute accuracy. Therefore, the mean of the upper and lower bounds of the  
311 estimated volume range was used as a single estimate of the volume to facilitate error  
312 calculation. With this treatment, the mean and maximum absolute estimation errors are 1.33 and  
313 3.15% across 25 species, using the cohort-specific superelliptical exponent ranges of 2.0-2.3 for  
314 cetaceans, 1.8-2.0 for sharks, 1.8-2.1 for typical fish, 2.2-2.4 for U-shaped fish, and 1.6-1.7 for  
315 V-shaped fish. When applying the more inclusive exponent range of 1.6-2.4 to all species, the  
316 errors increase to 4.61 and 7.21%, respectively. For the surface area, the mean and maximum  
317 absolute error are 2.64 and 10.5%, respectively, when using the same inclusive range of 1.6-2.4.

## 318 Discussion

319 The validation results suggest that Paleomass successfully brackets the true volume and surface  
320 area of marine vertebrates when the body silhouettes are known—the software has high  
321 accuracy, with a mean absolute error of 1.33%. At the same time, there are limitations to the  
322 software package. Paleomass is designed for marine vertebrates with straight body axis and  
323 cannot manage lateral concavities in body shape or dorso-ventral concavities in fins/flippers.  
324 Also, as stated earlier, the software lacks the capability for Boolean union of body part meshes  
325 until such becomes stably available in R. Finally, the accuracy of body mass estimates depends  
326 on that of the body outline images, as well as the choice of superelliptical exponent and mean  
327 body density.

328 The accuracy of body outline information merits a discussion. There is a paucity of body  
329 outline information in the fossil record in general: some fossil species, such as the ichthyosaur  
330 *Stenopterygius* and *Aegirosaurus* (Motani, 2005; Deltset et al., 2022), are occasionally preserved  
331 with body outlines but the majority of species lack such information. For species without body  
332 outlines preserved, outlines are often drawn around the skeleton, usually without strict accuracy  
333 control. Therefore, the accuracy of body mass estimation for those fossil vertebrates would be  
334 lower than that of the software itself because of additional errors introduced while body outlines  
335 are reconstructed around the skeleton. One way to remedy this problem may be to employ the  
336 minimum hull approach of Sellers et al. (2012), where the body mass is estimated by multiplying  
337 the minimum skeletal hull volume by an empirical ratio between such volumes and the actual  
338 volumes in extant mammals. In the present case, the volume of an animal may be estimated from  
339 a set of orthogonal minimum skeletal hull silhouettes, provided that the ratio between Paleomass  
340 estimates from such silhouettes and the true volume is known. However, derivation of such a  
341 ratio would require a broad taxonomic sample of CT scan data that records both the skeletons

342 and body surface of individuals. At present, most publicly available CT scans of marine  
343 vertebrates are based on liquid preserved and sometimes eviscerated individuals that do not  
344 retain the original body outlines (e.g., MorphoSource.org, Kamminga et al., 2017), making it  
345 difficult to obtain sufficient data. This possibility may be pursued in the future as more data are  
346 added to public data repositories.

347 At present, it is difficult to assess how much loss of accuracy would result from body  
348 outline reconstruction errors based on fossils. However, even if the error level increases by ten-  
349 to twenty-fold compared to that from the Paleomass software alone, the total error level would  
350 still be comparable to those of other body mass estimation methods for fossil vertebrates. For  
351 example, the minimum hull approach had 11-20% errors when applied to primates (Brassey &  
352 Sellers, 2014), whereas the mean absolute error is 26.35% in the regression-based body mass  
353 estimation of terrestrial vertebrates, with the maximum absolute error being about 300% based  
354 on the data in Campione & Evans (2012).

355 Paleomass fills the niche left by other body mass estimation methods. It is applicable to  
356 animals for which limb-based regression methods are not suitable, as noted earlier for marine  
357 vertebrates. It also enables body mass estimation from flattened fossils, which would supply the  
358 body outline images but not a 3D skeletal model necessary for minimum hull construction—  
359 again, marine vertebrate fossils tend to be flattened. Application to flattened fossils would  
360 depend on the availability of two conspecific individuals with almost identical sizes, exposing  
361 the body from two different angles, as in *Stenopterygius* reconstructed by Motani (2001).  
362 Overall, Paleomass is a viable alternative to existing body mass estimation methods for fossil  
363 vertebrates.

## 364 Conclusions

365 Paleomass allows estimation of body volume and surface areas of marine vertebrates with  
366 straight body axis through bracketing with 3D models with superelliptical cross-sections. The 3D  
367 models are built based on orthogonal silhouettes of the animal in question, which are supplied by  
368 the user as raster images. The volumes are converted to body mass by assuming a total body  
369 density, which may be the seawater density ( $1.027 \text{ g/cm}^3$ ) for forms that use the lungs or air  
370 bladders to control buoyancy. Optimal superelliptical values for bracketing are 2.0 and 2.4 for  
371 cetaceans, 1.8 and 2.0 for sharks, and 1.8 and 2.1 for most bony fish, although the values may be  
372 higher or lower for unusual forms, such as pufferfish and flatfish. When using proper exponent  
373 ranges, the errors in volume estimation are about 1.33% on average. The software is open access  
374 under GNU General Public License v3.0. at <https://github.com/rmotani/paleomass>.

## 375 Acknowledgements

376 I thank the following individuals for assessing the Paleomass package prior to public release:  
377 Benjamin Faulkner, Kiersten Formoso, Yu Qiao, Nicholas Thurber, and Kazuko Yoshizawa.

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453

454 Figure 1. How superellipses of different exponent values are used to bracket the true volume of a  
455 marine vertebrate. (A) variations of superelliptical shapes, with numbers being the exponents  
456 used to produce respective shapes. (B) a skinny dolphin model with an n value of 1.5 based on  
457 the silhouettes from D. (C) Same with an exponent of 2.0. (D) 3D model of Tursiops truncatus  
458 (model 61 from digitallife3d.org). (E) A fat model with an exponent of 2.5. (F) Same with an  
459 exponent of 3.0.

460

461 Figure 2. Shape input images from *Sphyrna lewini*. (A) Orthogonal views of the target animal,  
462 with the overall outlines traced in red. (B) Planar views of fins that are angled in A, with fins in  
463 question outlined in red. (C) Input images for Paleomass based on A and B, where fins are  
464 separated from the main body. Scale bar in 10cm. A resulting Paleomass mode is found in Fig.  
465 4A.

466

467 Figure 3. Computation process of main body and fin/flipper 3D meshes with examples from  
468 *Cephalorhynchus heavisidii*. (A) Lateral silhouette image input. (B) Dorso-ventral silhouette  
469 image input. (C) Coordinates around A in dots, with dorso-ventral diameters in lines, down-  
470 sampled to one in every ten coordinates for visualization purposes. (D) Same as C but based on  
471 B. (E) Serial superelliptical sections based on diameters from C and D, with an exponent of 2,  
472 downsampled at the same rate as in C. (F) Same as E but with an exponent of 3. (G) 3D mesh  
473 combining all superelliptical slices as in E but without downsampling. (H) Same as G but based  
474 on F. (I) Same as G but with interpolation with local regression with a nearest neighbor  
475 parameter of 0.1. (J) Same as H but with interpolation with local regression. (K) Planar silhouette  
476 image input. (L) Coordinates around A in dots, with chords in lines. Downsampled to one in  
477 every five slices for visualization purposes. (M) Serial foil section based on NACA 0020,  
478 downsampled at the same rate as in L. (N) 3D mesh that connected serial foil sections as in C but  
479 without downsampling. (O) Same as C but with interpolation with a nearest neighbor parameter  
480 of 0.05. M-O are slightly tilted for visualization purposes and thus appear narrower than K-L.  
481

482 Figure 4. Range of body designs modelled by Paleomass. (A) *Sphyrna lewini*. (B) *Rhincodon*  
483 *typus*. (C) *Tursiops truncatus*. (D) *Stenopterygius quadriscissus*. (E) *Chaohusaurus*  
484 *chaoxianensis*. (F) *Plesiopterys guilelmiimperatoris*. (G) *Latolabrax japonicus*. (H) *Eopsetta*  
485 *grigorjewi*. (I) *Anguilla marmorata*.

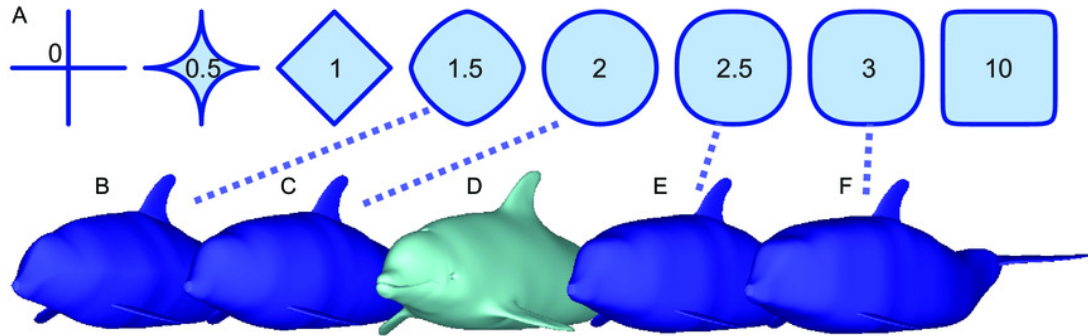
486  
487 Figure 5. Errors from volume and surface area estimates for a sphere and prolate spheroid  
488 depending on the input image resolution. (A) Errors from the sphere. (B) Errors from a prolate  
489 spheroid whose major axis is five times the minor axis. Blue lines are for the surface area and  
490 black for the volume. The independent is the number of pixels along the long axis of the  
491 geometry, i.e., pixels per diameter.

492  
493 Figure 6. Optimal superelliptical exponents for 25 species of extant marine vertebrates, with  
494 coronal views of five species. Horizontal bars show the range of optimal superelliptical  
495 exponents for individual species. Coronal views are given for the following species. (A)  
496 *Mustelus manazo*. (B) *Phocoena phocoena*. (C) *Clupea pallasii*. (D) *Auxis thazard*. (E) *Salvelinus*  
497 *leucomaenis*. Species with V-shaped ventral halves of the coronal views, e.g., C, tend to have  
498 lower exponent values than those with U-shaped ventral halves, such as E. Squares associated  
499 with coronal views are each 1cm.

# Figure 1

How superellipses of different exponent values are used to bracket the true volume of a marine vertebrate

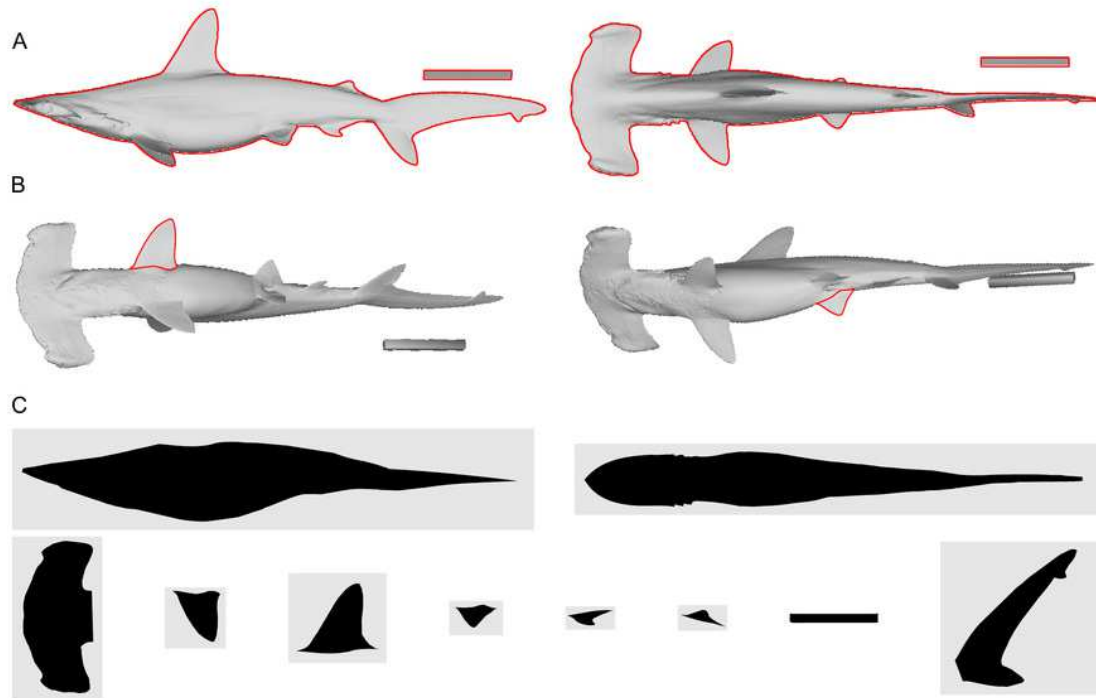
(A) variations of superelliptical shapes, with numbers being the exponents used to produce respective shapes. (B) a skinny dolphin model with an  $n$  value of 1.5 based on the silhouettes from D. (C) Same with an exponent of 2.0. (D) 3D model of *Tursiops truncatus* (model 61 from [digitallife3d.org](http://digitallife3d.org)). (E) A fat model with an exponent of 2.5. (F) Same with an exponent of 3.0.



## Figure 2

Shape input images from *Sphyrna lewini*

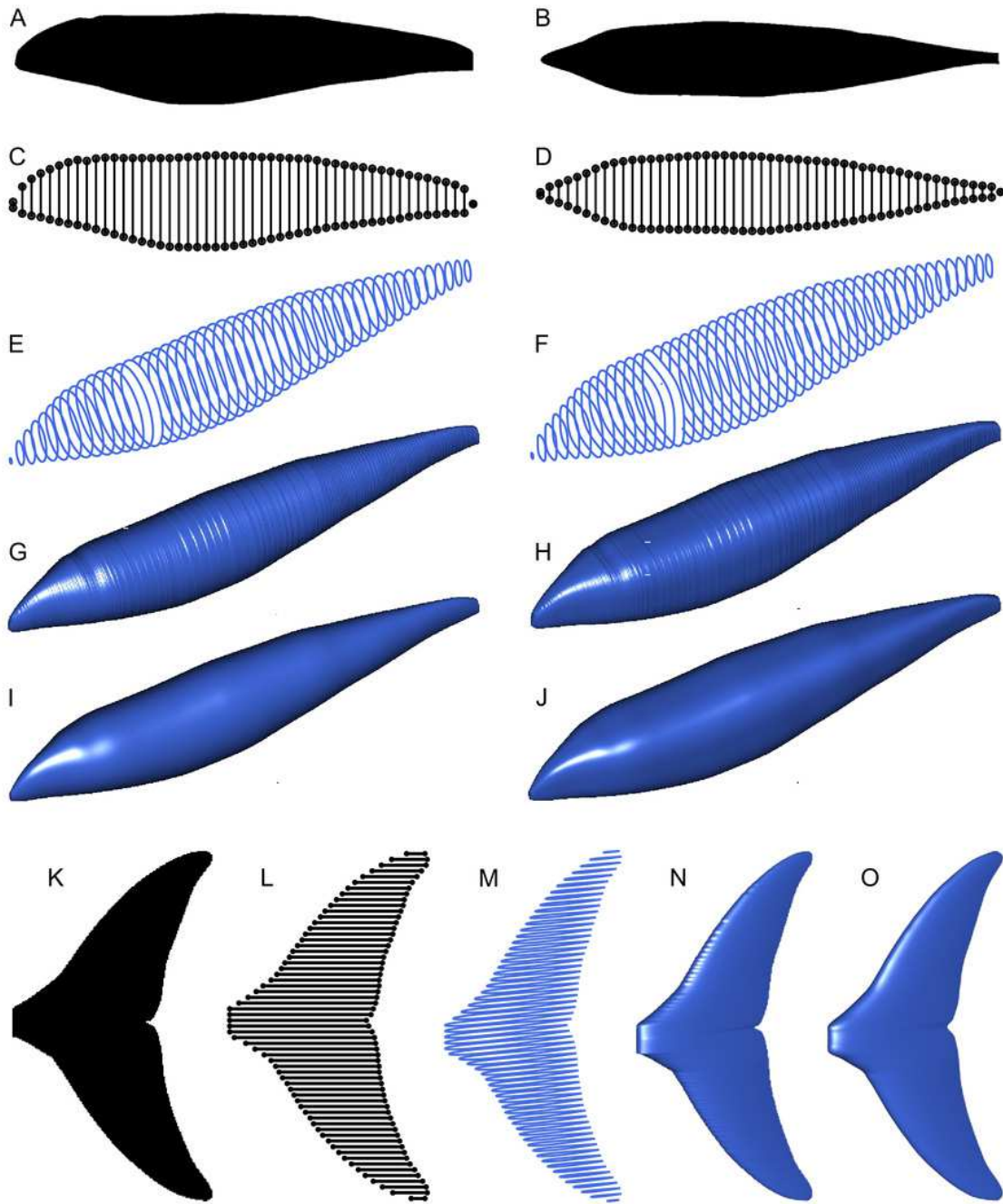
Figure 2. Shape input images from *Sphyrna lewini*. (A) Orthogonal views of the target animal, with the overall outlines traced in red. (B) Planar views of fins that are angled in A, with fins in question outlined in red. (C) Input images for Paleomass based on A and B, where fins are separated from the main body. Scale bar in 10cm. A resulting Paleomass mode is found in Fig. 4A



## Figure 3

Computation process of main body and fin/flipper 3D meshes with examples from *Cephalorhynchus heavisidii*

(A) Lateral silhouette image input. (B) Dorso-ventral silhouette image input. (C) Coordinates around A in dots, with dorso-ventral diameters in lines, down-sampled to one in every ten coordinates for visualization purposes. (D) Same as C but based on B. (E) Serial superelliptical sections based on diameters from C and D, with an exponent of 2, downsampled at the same rate as in C. (F) Same as E but with an exponent of 3. (G) 3D mesh combining all superelliptical slices as in E but without downsampling. (H) Same as G but based on F. (I) Same as G but with interpolation with local regression with a nearest neighbor parameter of 0.1. (J) Same as H but with interpolation with local regression. (K) Planar silhouette image input. (L) Coordinates around A in dots, with chords in lines. Downsampled to one in every five slices for visualization purposes. (M) Serial foil section based on NACA 0020, downsampled at the same rate as in L. (N) 3D mesh that connected serial foil sections as in C but without downsampling. (O) Same as C but with interpolation with a nearest neighbor parameter of 0.05. M-O are slightly tilted for visualization purposes and thus appear narrower than K-L.

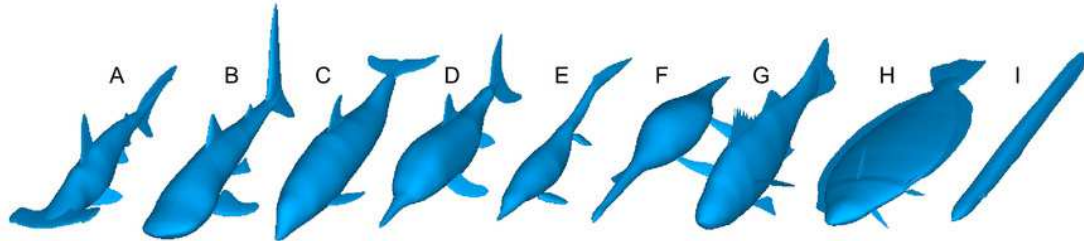




## Figure 4

Range of body designs modelled by Paleomass.

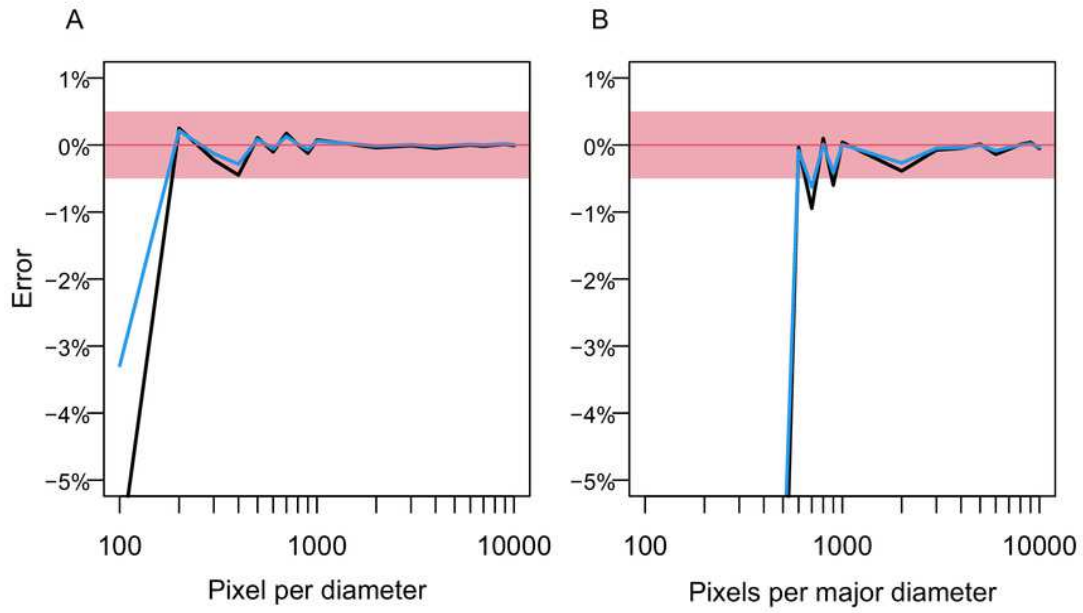
(A) *Sphyrna lewini*. (B) *Rhincodon typus*. (C) *Tursiops truncatus*. (D) *Stenopterygius quadriscissus*. (E) *Chaohusaurus chaoxianensis*. (F) *Plesiopterys guilelmiimperatoris*. (G) *Latolabrax japonicus*. (H) *Eopsetta grigorjewi*. (I) *Anguilla marmorata*.



## Figure 5

Errors from volume and surface area estimates for a sphere and prolate spheroid depending on the input image resolution.

(A) Errors from the sphere. (B) Errors from a prolate spheroid whose major axis is five times the minor axis. Blue lines are for the surface area and black for the volume. The independent is the number of pixels along the long axis of the geometry, i.e., pixels per diameter.



## Figure 6

Optimal superelliptical exponents for 25 species of extant marine vertebrates, with coronal views of five species

Horizontal bars indicate the range of optimal superelliptical exponents for individual species. Coronal views are given for the following species. (A) *Mustelus manazo*. (B) *Phocoena phocoena*. (C) *Clupea pallasii*. (D) *Auxis thazard*. (E) *Salvelinus leucomaenis*. Species with V-shaped ventral halves of the coronal views, e.g., C, tend to have lower exponent values than those with U-shaped ventral halves, such as E. Squares associated with coronal views are each 1cm.

