Can endocranial volume be estimated accurately from external skull measurements in great-tailed grackles (*Quiscalus mexicanus*)?

Corina J Logan, Christin R Palmstrom

There is an increasing need to validate and collect data approximating brain size on individuals in the field to understand what evolutionary factors drive brain size variation within and across species. We investigated whether we could accurately estimate endocranial volume (a proxy for brain size), as measured by computerized tomography (CT) scans, using external skull measurements and/or by filling skulls with beads and pouring them out into a graduated cylinder for male and female great-tailed grackles. We found that while females had higher correlations than males, estimations of endocranial volume from external skull measurements or beads did not tightly correlate with CT volumes. We found no accuracy in the ability of external skull measures to predict CT volumes because the prediction intervals for most data points overlapped extensively. We conclude that we are unable to detect individual differences in endocranial volume using external skull measurements. These results emphasize the importance of validating and explicitly quantifying the predictive accuracy of brain size proxies for each species and each sex.

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3	tailed grackles (Quiscalus mexicanus)?
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12	
13	Abstract
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15	field to understand what evolutionary factors drive brain size variation within and across species. We
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26

27 Introduction

28 While comparing relative brain sizes (corrected for body size) across species has led to a greater 29 understanding of the evolutionary factors correlated with brain size at a broad taxonomic scale (e.g., 30 Iwaniuk & Nelson 2003, Sakai et al. 2011, Sol et al. 2005), little is known about the within species 31 causes and consequences of variation in brain sizes (see Gonda et al. 2013, Thornton & Lukas 2012). 32 Additionally, the accuracy of brain size proxies, which are frequently used in such comparisons, is not 33 often validated (Healy & Rowe 2007). Therefore, the accuracy of brain size estimations and how they 34 compare to estimations in other species is questionable (Healy & Rowe 2007). Intraspecific brain size 35 comparisons are rare perhaps due to the difficulty of obtaining data on a number of factors for the same 36 individuals (e.g., biometric measurements, reproductive success, dominance rank, position in the social 37 network, and cognitive abilities). Acquiring such data is key for understanding what contributes to the evolution of brain size among individuals, as well as across species (Gonda et al. 2013, Logan & 38 39 Clutton-Brock 2013, Thornton & Lukas 2012).

40 We investigated whether endocranial volume, a proxy for brain size (Iwaniuk & Nelson 2002), 41 can be approximated using measurements of the external skull in great-tailed grackles (Quiscalus 42 mexicanus, JF Gmelin, 1788). Grackles are innovative, having many novel foraging techniques 43 (Overington et al. 2009, 2011), and have successfully expanded their geographical range by exploiting 44 new environments (Peer 2011). However, it is unclear how grackles achieve such feats. Relatively 45 large-brained birds such as New Caledonian crows solve novel problems using enhanced cognitive abilities (e.g., Taylor et al. 2010, Logan et al. 2014), yet grackles have smaller relative brain sizes than 46 47 would be expected for their many innovations (Overington et al. 2009). One of us (Logan) is 48 conducting in depth investigations of great-tailed grackle cognition and innovation in the lab and field 49 to understand whether innovativeness indicates behavioral flexibility and what kinds of cognition

grackles use to solve novel problems. This study system provides a rare opportunity to examine intraspecific differences in brain size, which could have implications for understanding range expansions in the context of innovation and behavioral flexibility. Finding an accurate proxy for endocranial volume would greatly ease the collection of data on brain sizes since external skull measurements could be further validated to account for head measurements that can be taken on live birds, thus allowing for correlations with a number of other factors on which data are gathered on this species.

57 This study is intended as a first step in validating the accuracy of using head measurements taken from live birds to predict their actual brain size, which would require two additional steps: 1) 58 validating the link between external skull measurements and external head measurements, and 2) 59 60 validating the link between endocranial volume and actual brain size. Regarding the latter validation, 61 there is reason to believe that endocranial volume accurately approximates actual brain volume in birds 62 because the meninges, the matter between the skull and the brain, are thinner than in mammals and the 63 shape of the brain case in birds tightly tracks the shape of the brain (Iwaniuk & Nelson 2002). 64 Additionally, it was previously observed in the genus *Quiscalus* that individual variation in endocranial 65 volume is particularly large when compared with variation in other bird species (Quiscalus quiscala; 66 Iwaniuk & Nelson 2002). Assuming great-tailed grackles have similar amounts of variation, our study 67 provides an opportunity to understand how much of this variation in endocranial volume is due to 68 variation in skull morphology (i.e., changes in skull length, width, and/or height specifically).

Great-tailed grackle body sizes are sexually dimorphic (Johnson et al. 2000), therefore we expect sex differences in endocranial volumes and we investigate proxies for each sex independently. We used endocranial volumes calculated from computerized tomography (CT) scans to represent actual endocranial volumes since this measure is the most precise. The complete area of the inside of the skull is accounted for in CT scans, while other methods do not cover the whole endocranial space

(Witmer et al. 2008, Knoll et al. 2012). We compared CT volumes to skull length, width, and height measurements to determine whether the correlation between these two methods and the accuracy of external measures in predicting CT volumes warrants their use as a proxy for endocranial volume. We also evaluated the bead method of generating endocranial volume, where glass beads are poured into the skull and then out into a graduated cylinder, to increase the value of our research by determining whether this widely used method (e.g., Isler et al. 2008, Iwaniuk & Nelson 2002) also accurately predicts actual endocranial volume as estimated by CT scans in this species.

81

82 Methods

83 Specimens

84 We collected data from February through September 2014, and in March 2015, on 40 great-tailed 85 grackle skulls (Supplemental Table S1), 20 female and 20 male (some analyses have 19 males because 86 on one of their skulls the bill was broken off, thus we could not acquire its skull length measurement), 87 obtained from the Museum of Southwestern Biology (n=24, Albuquerque, NM), the Ornithology 88 Division of the University of Kansas (KU) Biodiversity Institute (n=15, Lawrence, KS), and the Santa 89 Barbara Museum of Natural History (n=1, Santa Barbara, CA). Skulls of unknown age were aged by 90 Andy Johnson if they were from the Museum of Southwestern Biology or by us if they were from KU. 91 Skulls were aged using the percentage of ossification to classify each as adult (>7 months old; 100% 92 ossified unless it was collected in February-May because this would mark the start of that individual's 93 first breeding season after having hatched June-August in the previous year) or immature (<7 months 94 old; <100% ossified when collected September-December indicating it had hatched that year; del Hoyo 95 et al. 1992, Winker 2000, Pyle 1997).

- 96
- 97 Collecting endocranial volume measurements

98 Linear measurements: Linear measurements of skulls were collected by placing calipers in locations on 99 the skull that would also be accessible and measurable on a live bird in the field. We recorded skull 100 length from the base of the bill to the back of the skull along the occipital crest (Figure 1A), height 101 from the posterior edge of the foramen magnum (the posterior edge of the neck at the base of the skull 102 on a live bird) to the top of the skull along the frontal region (Figure 1B), and width at the widest part 103 of the brain case along the squamosal bones (Figure 1C). All measurements were taken to the nearest 104 0.1mm. Research on other species has found positive correlations between actual brain mass and brain 105 volume estimated from linear skull measurements by calculating the volume of an ellipsoid (barn 106 swallows: Møller 2010), and endocranial volume and the volume of a cube as estimated using head 107 width (zebra finches: Bonaparte et al. 2011). Therefore, we estimated endocranial volume using a 108 number of volumetric shapes and data transformations to determine which best correlated with 109 endocranial volumes from CT scans. The volumetric shapes included were: cube (Length x Width x Height), sphere $(\frac{4}{3}\pi r^3)$, where $r=\frac{1}{2}L$ or $\frac{1}{2}W$ or $\frac{1}{2}H$), ellipsoid $(\frac{4}{3}\pi abc)$, where $a=\frac{1}{2}L$, $b=\frac{1}{2}W$, $c=\frac{1}{2}H$), and 110 cone/pyramid (3bh, where b=W, h=H). We included log, natural log, and exponential transformations 111 112 of the data, and also allowed polynomial terms.

113

114 <u>CT scans</u>: Skulls were CT scanned at the Pueblo Radiology Medical Group in Santa Barbara, 115 California using a Siemans 16-slice Somatom Sensation 16 (1mm slices, 100Kv, 150MAs, 380mm 116 FOV, soft tissue window, analyzed with bone algorithm on). Endocranial volume (cm³) was calculated 117 using the DICOM viewer OsiriX v5.8.5 (32-bit, Pixmeo SARL, Switzerland; Supplemental Figures S1 118 and S2) for 1mm cross-sectional slices (regular) and for 1mm cross-sectional slices that were taken 119 with the CT scanner bed moved 0.5mm forward (offset), using the average endocranial volume (

 $\frac{regular + offset}{2}$) in analyses. The offset was added to increase the precision of the endocranial volume 120 121 measurements since grackle craniums are small (approximately 20mm in length), resulting in about 20 122 slices per scan (one slice every 1mm). The offset allowed us to measure more area (one slice every 123 0.5mm) by increasing the number of slices to approximately 40 per skull.

124

125 Beads: Endocranial volume was measured by pouring 1mm diameter glass beads (BioSpec Products, 126 catalog number 11079110) into the cranium through the foramen magnum until full. The skull was 127 repeatedly shaken to settle the beads and then filled again until the beads reached the posterior foramen magnum without falling out (Figure 2). The volume was calculated by pouring the beads out of the 128 129 skull and into a graduated cylinder (5ml in 0.1ml graduations, World Precision Instruments, Inc., 130 catalog number CG-0160; note that 1ml=1cm³). In cross-species comparisons, there is mixed evidence 131 about whether pouring the beads into a graduated cylinder introduces error when compared with pouring the beads onto a scale and converting their mass into volume (4% difference: Miller 1997, 0% 132 133 difference: Isler et al. 2008). The skulls in this study were measured with an average error of 2% (i.e., 134 there was a 2% difference in volume between two sets of volume measurements carried out on a subset 135 of 12 skulls), which is small in comparison to the variance between skulls (intra-class correlation 136 coefficient=0.94; Hutcheon et al. 2010). Therefore, the error should not affect the power to detect a 137 correlation with the more precise CT method (intra-class correlation coefficient=0.98) of measuring 138 skulls

139

140 Data availability: The data from skull measurements and intraobserver reliability measurements are

141 available at the KNB Data Repository (Logan & Palmstrom 2015).

142

143 *Statistical analyses*

144 The female and male data (analyzed separately) were normally distributed (Anderson Darling 145 normality test: females: skull height p=0.27, length p=0.30, width p=0.86; males: skull height p=0.35, 146 length p=0.63, width p=0.38). We defined statistical significance as $p \le 0.05$ throughout the paper. Two sets of linear, bead, and CT scan measurements were taken on a subset of skulls on different days 147 148 by Palmstom to quantify the random measurement error (intraobserver reliability). We used intra-class 149 correlation coefficients (ICC) to determine the precision of our estimates using the equation in Figure 2 in Hutcheon and colleagues (2010): $\frac{true \ slope \ \times \ \frac{variance \ (true \ X \ values)}{variance \ (true \ X \ values) + variance \ (random \ error)}} \ (we$ 150 151 assumed that variance in the observed values was equal to the variance in the true values). This ICC is 152 a measure of consistency, not agreement since it does not include rater effects (Auerbach et al. 2004).

153 We used generalized linear models (GLMs) to determine how well linear and bead 154 measurements correlated with volumes from CT scans, while examining whether the age of the 155 individual at death and the year the skull was collected improved the model fit. GLMs were carried out 156 in R v3.1.2 (R Core Team 2014) using the MCMCglmm function (MCMCglmm package; Hadfield 157 2010), while applying the dredge function (MuMIn package; Barton 2012) to select the best-fitting 158 model using the Akaike weight (Akaike 1981, Burnham & Anderson 2002). We considered the bestfitting model to be strongly supported and reliable if its Akaike weight was ≥ 0.9 as suggested by 159 Burnham and Anderson (2002) since this would indicate that the likelihood of the model given the data 160 161 is very high. Female and male data were analyzed in separate models. Full models included 1) 162 endocranial volumes from CT scans as the response variable with the following explanatory variables: 163 volume of a cube or sphere or ellipsoid or cone + age and the interaction with year collected; or 2) 164 endocranial volumes from CT scans as the response variable with the following explanatory variables: skull length + skull width + skull height + age and the interaction with year collected. GLMs were 165

166 conducted on the best-fitting model for each sex to explore whether the adjusted coefficient of 167 determination (adjusted r^2) improved by transforming the endocranial volume proxy (explanatory 168 variable) in the following ways: squared, cubed, quadratic, exponential, square root, log, log base 10, and a polynomial with a degree of two or three. Of these, the model with the highest adjusted r^2 was 169 170 chosen as the final best-fitting model for that sex and included in the results below. Interpretations of correlation strengths were taken from Taylor (1990): ≤ 0.35 =weak correlation, 0.36-0.67=moderate, 171 172 0.68-0.89=high, 0.90-1.00=very high. We set the minimum criteria for a correlation of sufficient strength such that it might be predictive at $r^2 \ge 0.88$, which is equivalent to Pearson's r set to 173 174 alpha=0.05 or $r \ge 0.95$, adjusted for the random measurement error in the response variable (CT) measurements), which has an intra-class correlation coefficient of 0.99 (0.95*0.99=0.94, 0.94²=0.88; 175 176 Hutcheon et al. 2010).

177 Since we want to predict CT volumes from linear measures, we validated whether this was 178 possible by generating prediction intervals with 95% confidence levels. We applied the predict 179 function in the MCMCglmm package to the best-fitting model for each sex and evaluated whether 180 fitted values (predicted CT volumes) had credible intervals small enough such that there was little to no 181 overlap with other fitted values, thus allowing the discrimination of individual differences.

182

183 **Results**

184 Intraobserver reliability

There was very high consistency (precision) between the two sets of CT and bead volume measurements, but no consistency for linear volume (LWH) measurements when sexes were analyzed together and separately (Table 1). There was high (sexes combined) to very high (sexes analyzed separately) consistency when comparing the two sets of skull width measurements, high consistency

189 for skull length (when sexes were separate and combined), and moderate (males and sexes combined)
190 to very high (females) consistency for skull height (Table 1).

191

192 Correlations between methods

193 None of the models using linear measurements to explain variation in CT volumes were likely given the data, as indicated by the low Akaike weights of the best-fitting models (≤ 0.90 ; Table 2; Burnham 194 195 & Anderson 2002). Regardless, we used the best-fitting models to examine these relationships further. 196 The volume of a sphere was the best-fitting shape for both sexes (the radius was based on skull width 197 for males and skull height for females). The best-fitting female model showed a positive relationship 198 between CT volumes and volumes from using the skull height as the radius for a sphere, volumes were 199 larger for immatures than for adults, and volumes slightly decreased over the years collected (Table 2, 200 model 1; Figure 3A). The best-fitting male model showed a positive correlation between CT volumes 201 and volumes using a quadratic polynomial of the skull width as the radius for a sphere, volumes were 202 slightly larger for immatures than for adults, and volumes decreased slightly over the years collected 203 (Table 2, model 2; Figure 3B). Transformations of the explanatory volume variables or substituting volume for individual linear measurements (length, width, height, or some combination of these) did 204 not improve the adjusted r^2 for females. Volumes from CT scans were moderately (males) to highly 205 206 (females) correlated with spherical volumes from linear measurements (Table 2).

None of the models using bead volumes to explain variation in CT volumes were strongly supported given the data as indicated by their low Akaike weights for the best-fitting models (< 0.90; Table 2; Burnham & Anderson 2002). Nonetheless, the best-fitting female model showed that endocranial volumes decreased slightly over time (Table 2, model 3; Figure 4A), while the best-fitting male model included age, with immatures having smaller volumes than adults (Table 2, model 4;

Figure 4B). Volumes from CT scans were highly positively correlated with volumes from the bead method for both sexes (Table 2).

214 None of the correlations between CT volumes and linear measures met our minimum criteria (r² \geq 0.88) for a strong enough relationship such that it might predict endocranial volumes from linear 215 216 measurements of skulls. Since we want to predict CT volumes from linear measures, we determined 217 whether this was possible by generating prediction intervals for the best-fitting female and male 218 models for the linear measurements (models 1 and 2) and bead method (models 3 and 4; Table 2). We 219 found that the lower and upper limits of the 95% confidence intervals of the predicted values for both 220 sexes show extensive overlap such that individual differences would not be able to be resolved if a 221 new, unvalidated data point was obtained (Table 3).

222

223 *Comparing method means*

Endocranial volume means were significantly different from each other when comparing across methods (mean±standard deviation: females: CT 2.29cm³±0.20, sphere 32459.1mm³±4344.7, bead 2.60ml±0.28; males: CT 2.54cm³±0.15, sphere 59292.1mm³±2360.4, bead 2.91ml±0.21; Welch two sample t-test: females: CT x Sphere t=31, p<0.0001, df=19; sphere x bead t=-31, p<0.0001, df=19; bead x CT t=4, p=0.0003, df=34; males: CT x sphere t=96, p<0.0001, df=19; sphere x bead t=-96, p<0.0001, df=19; bead x CT t=6, p<0.0001, df=35).

230

231 Discussion

While female great-tailed grackle endocranial volumes from linear measurements were highly correlated with volumes from CT scans, which we consider a more accurate proxy for brain size than bead volume (Witmer et al. 2008, Knoll et al. 2012), the correlation did not meet our criteria of having

235 a coefficient of determination (r^2) equal to or greater than 0.88 - a level of correlation that might be 236 strong enough to allow for the resolution of individual differences in endocranial volumes. This 237 correlation was only moderate in males, which is likely due to the sexual dimorphism in this species. 238 Our sample includes individuals from a range of populations in which the extent of sexual dimorphism 239 might vary. If, in some populations, there has been strong selection for males to increase in body size, 240 then the skeletal measures will not reflect brain size accurately because, in many instances, skeletal 241 size changes faster than brain size as has been found in primates (Montgomery 2011). Perhaps 242 additional biometric measurements would explain more of the variation in their endocranial volumes 243 from CT scans, however we only had access to skulls for most of the specimens and therefore could 244 not test this hypothesis.

245 We were more interested in whether a given value of an external skull measurement could accurately predict actual endocranial volume from CT scans, rather than setting a subjective criterion 246 about how high r² should be, especially given the extensive debate around the latter approach (e.g., 247 Legates & McCabe Jr. 1999, Müller & Büttner 1994). In particular, r² "[...] describes the proportion of 248 249 the *total* variance in the observed data that can be explained by the model" (Legates & McCabe Jr. 250 1999, p. 233, emphasis added) and thus does not allow one to investigate differences in the variance of 251 individual data points. Our predictive analyses showed that prediction intervals for new data points 252 overlapped to such a degree (within 95% credible intervals) that it was not possible to distinguish 253 among individuals, as we would need to when collecting linear measurements on new individuals in 254 the field. We must conclude that external skull measurements are not accurate enough to estimate endocranial volume in great-tailed grackles. 255

Predictive analyses are crucial for determining the accuracy of predicting individual data points by a particular method and should be applied extensively in future research, rather than relying solely on correlation coefficients (r) or coefficients of determination (r^2). The omission of such an analysis

leaves data uninterpretable for its purported use of discerning intraspecific differences in a morphological feature. Additionally, we caution against using a proxy validated in one species as evidence that the same proxy will apply to other species (e.g., great tits: Dreyer 2012). Until intraspecific validations of brain size proxies using skull or head measurements have been validated across species, we cannot assume that what works (or not) for one species will work (or not) for another.

265 The bead method was highly correlated with CT volumes in both sexes, however, it also did not meet our minimum criteria ($r^2 \ge 0.88$), and prediction intervals extensively overlapped for individual 266 267 data points. Great-tailed grackles and common grackles are among the species with the largest ranges 268 in endocranial volumes (as measured using the bead method) when compared with most other species 269 in Iwaniuk & Nelson's (2002) study on 81 bird species. Both grackle species had large standard 270 deviations when compared with mean volumes (common grackles: mean \pm SD=2.59ml \pm 0.37, n=10, 271 Iwaniuk & Nelson 2002; great-tailed grackles: female 2.60ml \pm 0.28, n=20, male 2.91ml \pm 0.21, n=20, 272 this study). It appears that grackle endocranial volumes are more variable than those of many other 273 species. This is not likely due to variation in skull morphology since we did not find a perfect 274 correlation between endocranial volume and external skull measurements. Therefore, we caution 275 against using external skull measurements to estimate endocranial volume without proper validation.

To infer differences in brain size among individuals of the same species, and of the same sex, there must be a high degree of accuracy to have the ability to detect actual individual differences (Legates & McCabe Jr. 1999, Logan & Clutton-Brock 2013). Our results highlight the need to validate brain size and/or endocranial volume proxies and their predictive power for each species under investigation, and for each sex if they are sexually dimorphic. It is unfortunate that there is not an easier, more accurate way to approximate brain size in the field where we have the potential to

understand how evolutionary factors drive brain size variation within species. However, this study accentuates the importance of knowing how accurate brain size measures and proxies are when including such data in analyses.

285

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1

Skull length, height, and width

Measuring skull length (A), skull height (B), replicating the height that can be measured on live birds, and skull width (C) at the widest part of the braincase



2

Bead method

Skull holes are plugged with cotton and then the cranium is filled with glass beads



3

Plots of the volumes of spheres and volumes calculated from CT scans for females and males

Correlations between CT volumes and the volume of a sphere as calculated from linear measurements for female (A) and male (B) adults (small circles) and immatures (large circles), with the year the skull was collected represented by a red-blue spectrum (earlier years are redder and recent years are bluer). Note that regression lines only reflect the relationship between spherical volume and CT volume and do not correct for age or year (factors in the best-fitting model for both sexes) as in the GLMs. Skulls were aged as described in the methods.



4

Plots of the bead volumes and CT volumes for females and males

Correlations between CT volumes and bead volumes for female (A) and male (B) adults (small circles) and immatures (large circles), with the year the skull was collected represented by a red-blue spectrum (earlier years are redder and recent years are bluer). Note that regression lines only reflect the relationship between bead volume and CT volume and do not correct for age (in the best-fitting male model) or year (in the best-fitting female model) as in the GLMs. Skulls were aged as described in the methods.



Table 1(on next page)

Intraobserver reliability

Intraobserver reliability summary data and results (measurements: mean±standard deviation; Both: data from both sexes combined; ICC: intra-class correlation)

2

		Measurement 2					ICC				
Variable	Female	Male	Both	Female	n	Male	n	Both	Female	Male	Both
Volume CT	2.30 ± 0.15	2.56 ± 0.06	2.43 ± 0.17	2.38 ± 0.17	4	2.48 ± 0.08	4	2.38 ± 0.16	0.99	0.99	0.95
Volume LWH	13814 ± 141	15605 ± 127	14650 ± 141	13779 ± 123	5	15320 ± 78	7	14672 ± 121	0.009	0.003	0.003
	5	6	7	3		5		8			
Volume bead	2.72 ± 0.15	2.93 ± 0.15	2.88 ± 0.21	2.68 ± 0.19	5	2.91 ± 0.13	7	2.84 ± 0.21	0.92	0.94	0.94
Skull Length	29.04 ± 0.98	30.06 ± 0.81	29.72 ± 0.99	29.40 ± 1.27	5	30.24 ± 0.7	7	29.94 ± 1.00	0.83	0.76	0.82
						8					
Skull Width	23.83 ± 0.86	23.99 ± 0.36	23.83 ± 0.67	23.39 ± 0.90	5	24.01 ± 0.3	7	23.81 ± 0.69	0.92	0.92	0.88
						9					
Skull Height	20.24 ± 0.87	20.89 ± 1.44	20.65 ± 1.18	19.96 ± 0.48	5	21.09 ± 0.5	7	20.68 ± 0.76	0.91	0.51	0.65
						4					

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Table 2(on next page)

Model results

Model results for the best-fitting female and male models

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Method	Sex	Model	Akaike weight	Adjusted r ²	р	y =
CT- sphere	Female	1	0.60	0.80	<0.0001	0.00002 x VolumeSphere + 1.10 x Age – 0.007 x Year + 1.44
	Male	2	0.26	0.39	0.02	0.47 x VolumeSphere + 0.19 x VolumeSphere ² + 0.12 x Age – 0.003 x Year + 5.25
CT- bead	Female	3	0.74	0.77	< 0.0001	0.37 x VolumeBead - 0.0005 x Year + 11.24
beau	Male	4	0.45	0.68	<0.0001	0.66 x VolumeBead – 0.09 x Age + 0.66

3 4

Table 3(on next page)

Prediction analysis results

Predicted CT volume (fitted value) and the predicted intervals in which these new data points would occur with 95% credible intervals based on inputs from linear measures or the bead method in the best-fitting female and male models for each method

Linear Measurements							Bead Method						
	Males		Females			Males			Females				
Fitted value	Lower	Upper	Fitted value	Lower	Upper	Fitted value	Lower	Upper	Fitted value	Lower	Upper		
2.72	2.37	3.07	2.23	2.03	2.42	2.61	2.42	2.80	2.30	2.07	2.48		
2.41	2.17	2.73	2.08	1.86	2.29	2.43	2.23	2.61	2.21	1.99	2.42		
2.40	2.12	2.69	1.87	1.61	2.07	2.37	2.18	2.55	1.90	1.63	2.11		
2.60	2.33	2.92	1.99	1.78	2.22	2.70	2.52	2.90	2.02	1.78	2.26		
2.42	2.10	2.70	2.20	2.02	2.39	2.50	2.34	2.68	2.05	1.81	2.30		
2.52	2.22	2.83	2.30	2.10	2.51	2.55	2.35	2.75	2.42	2.22	2.62		
2.77	2.45	3.07	2.43	2.23	2.60	2.76	2.54	2.94	2.39	2.16	2.59		
2.67	2.41	2.94	2.39	2.20	2.58	2.70	2.50	2.89	2.28	2.07	2.52		
2.35	2.06	2.64	2.32	2.14	2.53	2.37	2.17	2.56	2.31	2.11	2.54		
2.55	2.30	2.84	2.51	2.30	2.71	2.50	2.31	2.67	2.57	2.31	2.81		
2.46	2.21	2.72	2.51	2.30	2.71	2.43	2.25	2.63	2.43	2.20	2.64		
2.64	2.40	2.92	2.39	2.19	2.59	2.63	2.45	2.82	2.32	2.10	2.53		
2.53	2.24	2.77	2.48	2.27	2.70	2.44	2.26	2.63	2.46	2.20	2.70		
2.58	2.31	2.85	2.40	2.21	2.60	2.37	2.18	2.57	2.32	2.11	2.52		
2.55	2.31	2.85	2.32	2.13	2.52	2.47	2.28	2.67	2.31	2.11	2.56		
2.64	2.39	2.94	2.41	2.23	2.60	2.76	2.57	2.97	2.43	2.20	2.64		
2.54	2.29	2.82	2.36	2.16	2.55	2.61	2.40	2.80	2.43	2.22	2.66		
2.55	2.24	2.80	2.42	2.21	2.60	2.67	2.49	2.87	2.43	2.20	2.65		
2.52	2.26	2.80	2.32	2.13	2.52	2.48	2.27	2.68	2.36	2.15	2.57		
2.51	2.24	2.80	2.02	1.82	2.13	2.57	2.38	2.74	1.98	1.74	2.18		

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