

Sexual Dimorphism in the Human Corpus Callosum: An MRI Study Using the OASIS Brain Database

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A number of studies have reported that, “relative to brain size,” the midsagittal corpus callosum cross-sectional area (CCA) in females is on average larger than in males. However, others suggest that these may be spurious differences created in the CCA-to-brain-size ratio because brain size tends to be larger in males. To help resolve this controversy, we measured the CCA on all 316 magnetic resonance imaging (MRI) scans of normal subjects (18–94 years) in the OASIS (Open Access Series of Imaging Studies) cross-sectional dataset, and used multiple regression analysis to statistically control for the confounding effects of brain size and age to test the null hypothesis that the average CCA is not different between genders. An additional analysis was performed on a subset of 74 young adults (37 males and 37 females; 18–29 years) matched closely to brain size. Our null hypothesis was rejected in both analyses. In the entire sample ($n = 316$), controlling for brain size and age, the average CCA was significantly ($P < 0.03$) larger in females. The difference favoring females was more pronounced in the young adults cohort ($P < 0.0005$). These results provide strong additional evidence that the CCA is larger in females after correcting for the confounding effect of brain size.

Keywords: automatic segmentation, connectivity, gender differences, multiple regression analysis, neuroimaging

Introduction

De Lacoste-Utamsing and Holloway (1982) first suggested that on average, “relative to brain size,” the corpus callosum (CC) midsagittal cross-sectional area (CCA) in females may be larger than in males. Their conclusions were based on measurements of CCA in 14 postmortem brains (5 females and 9 males). They were able to replicate their original findings in an independent sample of 16 brains (8 females and 8 males) in a later study (Holloway and de Lacoste 1986). Subsequently, the findings were replicated on 3 independent autopsy samples (total $n = 119$) (Holloway et al. 1993). The latter work also reviewed 25 studies on CC sexual dimorphism published before 1991 and noted that a large majority of the studies that claimed no significant sexual dimorphism were, in fact, consistent with their own findings if they considered the “relative” size of CC. A later study that typifies the observation by Holloway et al. is that of Oka et al. (1999) who in a sample of 67 adults (34 females and 33 males) did not find a statistically significant difference between males and females in the CCA on magnetic resonance imaging (MRI) scans. Mean (\pm SD) CCA was determined to be $657 \pm 80 \text{ mm}^2$ in females and $655 \pm 85 \text{ mm}^2$ in males. However, the authors did not compute or compare the ratio of the CCA with brain size between groups. Given the larger average brain size in

males, it is quite likely that Oka et al. would have found a significant group difference in favor of females in the relative size measure.

On the other hand, Bishop and Wahlsten (1997) performed a meta-analysis of 49 studies published before 1994, which included most of the studies reviewed by Holloway et al. (1993), but came to the seemingly opposite conclusion that there is no evidence suggesting a significant sex difference in the size of the CC! In spite of this apparent contradiction between the conclusions reached by Holloway et al. and Bishop and Wahlsten while reviewing overlapping literature, a closer examination of these works reveals that the questions being considered were, in fact, subtly different.

The difference is highlighted in a study (Smith 2005) that elucidates the distinction between the concepts: relative to brain size and “statistical control for brain size as a confounding effect.” Smith performed a meta-analysis of 21 studies published before 2003 and concluded that there is strong support for the idea that on average the relative (relative to some measure of brain size) size of the CC in females is larger than in males. However, he points out that in order to “statistically control” the confounding effect of brain size when comparing the CCA between groups the proper approach is to use statistical procedures such as analysis of covariance, multiple regression, or partial correlations. In this paper, we used multiple regression analysis to control the confounding effects of brain size and age. Therefore, any differences observed can be said to be in the average CCA after statistically controlling brain size.

Regardless of whether gender differences are found in relative size of the CCA or in the CCA itself after statistically controlling for brain size, it has been contended that any observed difference between groups is not gender specific but may be due to differences in brain size (Peters 1988; Going and Dixon 1990; Jäncke et al. 1997; Bermudez and Zatorre 2001; Leonard et al. 2008; Tepest et al. 2010; Bruner et al. 2012). It is suggested that smaller brains may have relatively larger CC regardless of gender. And since the average brain size in females is smaller than in males, the observations may be due to comparing groups with different average brain sizes and not due to gender. Hitherto this point has not been refuted. In the current paper, noting that there is a substantial overlap between brain size distributions in males and females, we compared the CCA between 2 groups of young adult females and males that had been closely matched for brain size. Therefore, any observed difference between CCA cannot be attributed to groups having different average brain sizes. As far as we are aware, none of the previous studies have used subjects matched for brain size in investigating the current question.

Matching for brain size between groups and using what is considered to be an appropriate statistical control for brain size, however, does not fully explain the considerable disparity in the results and conclusions that have been hitherto reported in the literature on sexual dimorphism of the CC. In general, failure to detect a statistically significant difference in measurements between groups does not imply that no true difference exists. It may merely mean that on the basis of sample size and measurement errors the statistical test used was insufficiently powered to detect a true difference (Elster et al. 1990). This is especially true in the study of CCA because, firstly, as many authors have noted, the size and shape of CC vary considerably among individuals (Byne et al. 1988; Peters 1988; Clarke et al. 1989; Elster et al. 1990; Allen et al. 1991; Smith 2005), which can completely mask group differences. Therefore, a large sample size is required to demonstrate a significant gender difference. In this study, we used a sample size of 316 subjects from the publicly available Open Access Series of Imaging Studies (OASIS) MRI database (Marcus et al. 2007), which we believe to be the largest used to date to study sexual dimorphism in CC.

Secondly, measurement of CCA is particularly error-prone for several reasons, which add to the inherent variability of CCA, making it even more difficult to detect a true difference if any indeed exists. Studies based on postmortem brains usually include small samples and could suffer from measurement errors due to fixation and deformation of the brains. An advantage of MRI is that brains can be measured in vivo in larger samples, but it is more difficult to estimate the brain size using MRI. Early MRI studies of gender differences in CC suffered from the fact that only a single thick midsagittal slice was available (Oppenheim et al. 1987; Byne et al. 1988; Clarke et al. 1989; Weis et al. 1989; Elster et al. 1990; Allen et al. 1991; Rauch and Jinkins 1994; Constant and Ruther 1996; Oka et al. 1999). Error would thus be introduced due to partial volume effects, and the fact that the single slice almost always differs from the true midsagittal plane (MSP) but cannot be corrected. It has been shown that variability in appearance of CC attributable to differences in the orientation of the MSP can result in large measurement errors (Rauch and Jinkins 1996). Also the brain size cannot be estimated from a single slice, necessitating the use of less accurate proxy markers of the brain size, such as midsagittal cerebral area (Elster et al. 1990; Allen et al. 1991; Rauch and Jinkins 1994; Constant and Ruther 1996), midsagittal skull surface (Johnson et al. 1994), or the cube power of the basion–vertex distance (Constant and Ruther 1996).

The high-resolution 3D MRI volumes available in OASIS allowed us to correct for slice position by using a fully automated MSP detection program (Ardekani et al. 1997), following which the 3D MRI volume was resliced to correct for head tilt (i.e. to have nearly zero yaw and roll angles). The OASIS database also provides automatically determined estimates of the total intracranial volume (eTIVs; Buckner et al. 2004) that can be readily used for controlling brain size.

A further source of variability in the data is that in almost all previous work, the CC was delineated manually on the MSP, which introduces considerable subjectivity into the process. In this study, we used a fully automatic multiatlas-based method for CC segmentation (Aljabar et al. 2009; Cabezas et al. 2011; Ardekani et al. 2012) for this purpose.

Finally, a novelty of the study presented here is that we utilize publicly available MRI data from the OASIS database.

OASIS explicitly authorizes redistribution of data derived from their database. Therefore, we have made our CC segmentations and area measurements available publicly (<http://www.nitrc.org/projects/art>) which facilitates independent re-analyses and examination of the data. It is hoped that the availability of this completely transparent and easily accessible dataset would help in resolving this long-standing and often contentious debate in the scientific community.

Materials and Methods

MRI Data

We utilized the OASIS cross-sectional dataset (Marcus et al. 2007) which includes MRI brain scans from 416 right-handed subjects. Of these, we excluded the 100 subjects with dementia to avoid the confounding effects of Alzheimer's disease which is known to influence CC size (Teipel et al. 2002; Wang et al. 2006; Di Paola et al. 2010; Frederiksen et al. 2011). We used the 3D magnetization prepared rapid gradient-echo (MP-RAGE) structural MRI scans of the remaining 316 subjects (119 males and 197 females; ages: 18–94) to investigate the putative sex differences in CC size. Thus, data from “all” normal subjects in the OASIS cross-sectional dataset were employed without exception. Details of subject demographics, inclusion/exclusion criteria, MRI acquisition protocol, and preprocessing steps are given elsewhere (Marcus et al. 2007). The age and sex distribution is reproduced in Table 1.

The OASIS database also provides an automatically determined (Buckner et al. 2004) eTIV which we used as an independent variable to statistically control for brain size when comparing CCA differences between gender groups.

Corpus Callosum Segmentation

An automated multiatlas-based algorithm was used for segmentation of the CC cross-sectional area from 3D structural MRI scans (Aljabar et al. 2009; Cabezas et al. 2011; Ardekani et al. 2012). Our atlas set that consisted of 38 scans was developed as follows:

1. The MSP was detected using the automated algorithm developed by Ardekani et al. (1997).
2. The anterior and posterior commissures (AC–PC) were detected on the MSP using an automated approach (Ardekani and Bachman 2009).
3. Using the MSP and AC–PC information, a midsagittal slice in a standardized coordinates system was reconstructed using trilinear interpolation so that the center of the slice field of view was exactly the middle point between the AC and PC; the *x*-axis was parallel to the AC–PC line pointing from the AC to the PC; and the *y*-axis was perpendicular to the AC–PC line on the MSP pointing from the superior to the inferior direction.
4. The CC was manually traced on the midsagittal slice using the ITK-SNAP software (Yushkevich et al. 2006).

Figure 1 shows an example of the midsagittal slice in the standard orientation with manually traced CC. The above procedure was

Table 1
Age characteristics of the data set as given by Marcus et al. (2007)

Age group	<i>n</i>	Mean	Male	Female
<20	19	18.53	10	9
20s	119	22.82	51	68
30s	16	33.38	11	5
40s	31	45.58	10	21
50s	33	54.36	11	22
60s	25	64.88	7	18
70s	35	73.37	10	25
80s	30	84.07	8	22
≥90	8	91.00	1	7
Total	316	45.08	119	197

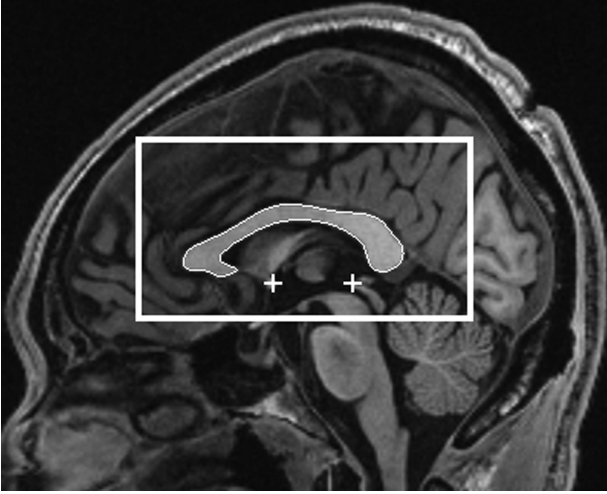


Figure 1. Automatically detected MSP in a standard orientation. The automatically detected AC–PC locations are shown by the plus signs. The rectangular bounding box is the search region for the CC determined based on a priori information obtained from the manually defined atlas set.

repeated for each of the 38 scans in our atlas set. A pair of 2D images was stored for each atlas: the gray-scale midsagittal slice and a binary image of the CC segmentation.

To automatically find the CC on a test image:

1. The midsagittal slice in the standard orientation was found using the exact procedure as above steps (1–3).
2. A rectangular bounding box was defined on the midsagittal slice that contained the CC. The rectangular region was specified in terms of its “superior-anterior” and “inferior-posterior” corners. The coordinates of these points were determined based on information obtained from the atlas set. The “superior-anterior” corner coordinates were $(x, y) = (-57.75, -47.75)$ mm and the “inferior-posterior” corner coordinates were $(x, y) = (52.25, 12.25)$ mm. The bounding box is shown in Figure 1.
3. All atlases were nonlinearly registered to the test image using the Automatic Registration Toolbox intersubject registration module (Ardekani et al. 2005) which has been found to be one of the most accurate intersubject registration algorithms available (Klein et al. 2009). The nonlinear transformation was only found within the rectangular bounding box, greatly accelerating the registration process.
4. The nonlinear transformations were applied to the binary CC images of the atlas set. Each transformation independently predicted the CC location on the test image. The final consensus segmentation was obtained using the vote rule (Rohlfing et al. 2004).

We segmented all 316 scans in our cohort using this method. The computation time per scan was less than 1 min on a Linux workstation with 2.4 GHz clock speed. The segmentations as well as the midsagittal slices of all images are available as Supplementary Material. In approximately 20% of the cases, minor manual editing of the CC was required. This was also performed using the ITK-SNAP software by one of the authors (K.F.) who was blind to the subjects’ gender and age.

Statistical Analysis

The following multiple regression model was considered for statistical testing:

$$y = \mu_f I_f + \mu_m I_m + \beta_1 (x_1 - m_1) + \beta_2 (x_2 - m_2) + e \quad (1)$$

The precise definitions of all variables in this model are given in Table 2. Briefly, y represents the CCA, I_f and I_m are female and male indicator functions, respectively; x_1 represents age; m_1 is the sample mean of x_1 ; x_2 is eTIV raised to the power two-third as suggested by

Table 2
Definitions of variables in the statistical model (1)

Notation	Definition
y	Corpus CCA (mm ²)
x_1	Age (years)
$m_1 = \frac{1}{n} \sum_{i=1}^n x_{1i}$	Sample mean of x_1
x_2	Two-third power of the eTIV (cm ²)
$m_2 = \frac{1}{n} \sum_{i=1}^n x_{2i}$	Sample mean of x_2
β_1	Model parameter (mm ² /year)
β_2	Model parameter (mm ² /cm ²)
$I_f = \begin{cases} 1 & \text{if subject is female} \\ 0 & \text{if subject is male} \end{cases}$	Female indicator function
$I_m = \begin{cases} 0 & \text{if subject is female} \\ 1 & \text{if subject is male} \end{cases}$	Male indicator function
μ_f	Model parameter (mm ²)—marginal mean for females
μ_m	Model parameter (mm ²)—marginal mean for males
e	Zero-mean normally distributed error

Smith (2005); m_2 is the sample mean of x_2 ; μ_f , μ_m , β_1 , and β_2 are unknown model parameters, and e is a random error term assumed to have zero mean and be normally distributed. A priori, we expect β_1 to be negative, that is, the CCA to decrease with age. We expect β_2 to be positive, that is, the CCA to increase with cranial capacity. Parameters μ_f and μ_m are female and male marginal means, respectively.

The model parameters are estimated using the standard procedure:

$$[\hat{\mu}_f \quad \hat{\mu}_m \quad \hat{\beta}_1 \quad \hat{\beta}_2]^T = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y} \quad (2)$$

where \mathbf{y} is an $(n \times 1)$ matrix with elements y_i and \mathbf{X} is an $(n \times 4)$ matrix with rows $[I_{fi} \quad I_{mi} \quad x_{1i} - m_1 \quad x_{2i} - m_2]$, where the subscript i indicates the value of the variable for subject i .

Based on the above model, we tested three null hypotheses: (1) $\beta_1 = 0$, that is, there is no linear association between age and the CCA; (2) $\beta_2 = 0$, that is, there is no linear association between CCA and the cranial capacity variable x_2 ; and (3) $\mu_f = \mu_m$, that is, all things being equal (age and eTIV), there is no difference between the male and female CCA. The three hypotheses were tested using the following three statistics:

$$\frac{\hat{\beta}_1}{\hat{\sigma} \sqrt{(\mathbf{X}^T \mathbf{X})_{33}^{-1}}} \quad (3)$$

$$\frac{\hat{\beta}_2}{\hat{\sigma} \sqrt{(\mathbf{X}^T \mathbf{X})_{44}^{-1}}} \quad (4)$$

$$\frac{(\hat{\mu}_f - \hat{\mu}_m)}{\hat{\sigma} \sqrt{(\mathbf{X}^T \mathbf{X})_{11}^{-1} + (\mathbf{X}^T \mathbf{X})_{22}^{-1} - 2(\mathbf{X}^T \mathbf{X})_{12}^{-1}}} \quad (5)$$

where $\hat{\sigma}^2$ estimates the noise variance and is given by

$$\hat{\sigma}^2 = \frac{1}{(n-4)} \mathbf{y}^T [\mathbf{I} - \mathbf{X}(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T] \mathbf{y} \quad (6)$$

Under the null hypothesis, all three statistics in (3–5) have central t distributions with $(n-4)$ degrees of freedom.

Pairing for Intracranial Volume

As mentioned previously, a major persistent criticism of the studies that find greater either relative or corrected CCA in females is that the observed differences are not gender specific but due to different average brain sizes (Peters 1988; Going and Dixson 1990; Jäncke et al. 1997; Bermudez and Zatorre 2001; Leonard et al. 2008; Tepest et al. 2010; Bruner et al. 2012). To address this issue, we analyzed data from a subset of subjects that were closely matched for eTIV and age. In this section, we explain the automated and objective method

used for pairing subjects with similar eTIV, each pair consisting of one female and one male subject. To avoid the confounding age effects we only considered the subjects between 18 and 29 years old for pairing because the CCA does not change appreciably with age in normal adults in this age range (Johnson et al. 1994). There were $n_m=61$ males and $n_f=77$ females in this age group (Table 1). To match for eTIV, we only considered pairing females and males whose cranial capacity measure x_2 (Table 2) differed by less than 1 cm^2 . However, this criterion alone is not sufficient for objective pairings of subjects because, for example, a given male subject could be matched to more than 1 female and vice versa. To objectively and uniquely match the subjects, we considered every possible pairings $n_m \times n_f$ and computed all the corresponding absolute differences in the cranial capacity measure $|x_{2i} - x_{2j}|$, where $i=1,2,\dots,n_m$ indexes the male subjects, and $j=1,2,\dots,n_f$ indexes the female subjects. We then paired the 1 male and the 1 female with the minimum distance $|x_{2i} - x_{2j}|$ and removed them from the cohort. Following this step, the number of possible pairings was reduced to $(n_m - 1) \times (n_f - 1)$. Among these, again we chose the pair with the minimum distance $|x_{2i} - x_{2j}|$. The procedure was repeated until all possible pairings had been achieved, that is, there were no remaining pairs with $|x_{2i} - x_{2j}| < 1.0\text{ cm}^2$. The above procedure resulted in 37 pairs of males and females. We then tested for CCA differences in this group using the statistical model described above. A spreadsheet of the pairings that resulted from this procedure is provided as Supplementary Material.

Results

For the entire cohort of 316 subjects, there was a significant linear association between age and CCA (the null hypothesis $\beta_1=0$ was rejected [$t(312) = -5.39$; $P < 10^{-6}$]). CCA decreased with increasing age. There was also a significant linear association between the intracranial capacity and CCA (the null hypothesis $\beta_2=0$ was also rejected [$t(312) = 7.56$; $P < 10^{-12}$]). CCA increased with cranial capacity. Most importantly, a gender difference in CCA was observed with females having a larger area than males (the null hypothesis of equal sex effects $\mu_f = \mu_m$ was rejected [$t(312) = 2.24$; $P < 0.03$]). Estimates of the marginal means for females and males were $\hat{\mu}_f = 634$ and $\hat{\mu}_m = 611\text{ mm}^2$, respectively.

In the cohort of 74 young adults matched for intracranial capacity, age was not associated with CCA [$t(70) = 1.52$; $P = 0.133$]. However, intracranial capacity was associated with CCA [$t(70) = 2.32$; $P = 0.023$]. Most significantly, when males and females were matched for intracranial capacity, female CCA was still significantly greater than male CCA [$t(70) = 3.69$; $P < 0.0005$].

Discussion

Using a public database consisting of 316 normal right-handed subjects, with ages that encompass a broad range of the adult human life span, we were able to confirm that there is a sexual dimorphism in the CCA, being larger in females in comparison with males by a few percent. As predicted, there was also a significant linear tendency for the CCA to decrease with age. Not surprisingly, larger brains tended to have larger CCA.

When reviewing the literature on sexual dimorphism of CC, it is helpful to consider 3 distinct random variables. The first random variable is some measure of "brain size," for example, the brain weight (Clarke et al. 1989; Witelson 1985) or its two-third power (Going and Dixson 1990; Holloway et al. 1993) as used in some postmortem studies, brain volume (Tepest et al. 2010), forebrain volume (Jäncke et al. 1997;

Bermudez and Zatorre 2001) or total intracranial volume (Johnson et al. 1994; Sullivan et al. 2001; Mitchell et al. 2003) or its two-third power (Smith 2005). Regardless, let x_2 denote this random variable. It is well known that the average x_2 in males is larger than in females. The second random variable is the CCA, which we denote by y . Finally, the third random variable, denoted by r , is the ratio of y to x_2 , that is, $r = y/x_2$. Thus, r is not directly measured but is obtained by dividing the first 2 random variables. This is the random variable considered in the majority of studies on the present topic. There is a fairly broad consensus in the literature (Holloway et al. 1993; Smith 2005) that r in females is statistically significantly larger than in males. The main criticism of using r is that the observed differences are spuriously created because the random variable r is formed by using a denominator (x_2) that is known to be sexually dimorphic (Bishop and Wahlsten 1997). However, others point out that this statistical pattern of sexual dimorphism is specific to CC, in that, in other neural structures, such as cerebellum, hippocampus, and thalamus, while the absolute sizes are larger in males, the relative sizes are not significantly different between genders (Holloway et al. 1993).

The question studied in the present paper is not whether r is different between genders, but whether the second random variable y representing CCA is different between females and males after statistically controlling for the confounding effect of the first random x_2 representing brain size. There is no dispute between researchers that to study this particular question an appropriate statistical method is multiple regression analysis, analysis of covariance, or partial correlation (Holloway et al. 1993; Constant and Ruther 1996; Bishop and Wahlsten 1997; Smith 2005). In this paper, we used multiple linear regression to show that the CCA in females is on average larger than in males after controlling brain size and age.

Regardless of whether the ratio random variable r or the random variable y representing CCA is compared between female and male groups, it has been contended by many authors including in recent papers (Peters 1988; Going and Dixson 1990; Jäncke et al. 1997; Bermudez and Zatorre 2001; Leonard et al. 2008; Tepest et al. 2010; Bruner et al. 2012) that the observed differences are not gender specific but brain-size specific. The idea is that as brain size increases, the CC does not keep pace, such that subjects with larger brain size have a relatively smaller CC regardless of their gender. Peters notes (Peters 1988) that: "...there is no reason to assume that whatever can or cannot be said about sex differences in corpus callosum parameters cannot also be said more generally about large and small brains." Since the 2 groups are not only different in gender, but also different in average brain size, this contention has been difficult to refute. To address this issue, we used a completely automated method that matched 37 pairs of female–male subjects very closely for intracranial capacity. The mean (\pm SD) x_2 was $133.43 (\pm 5.64)\text{ cm}^2$ in females and $133.47 (\pm 5.63)\text{ cm}^2$ in males. Thus, any difference in CC size could not be attributed to different brain sizes. The gender differences detected in CCA in this sub-cohort were even more reliable than the entire cohort. Not surprisingly, there was no significant linear association between age and CCA in this gender/brain size matched sub-cohort because, in the narrow age range considered, the brain was not expected to undergo major

structural changes. However, as expected the linear association between the CCA and the intracranial capacity still remained in this sub-cohort. As far as we know, this is the first study in which the female and male groups have been matched for brain size. Therefore, the differences detected cannot be attributed to groups having different brain sizes.

The current study has a number of advantages over previous studies on gender differences in CCA. The MRI data are taken from an easily accessible public database, which also provides eTIVs. The entire available data of normal subjects in the OASIS cross-sectional study were used. The CC segmentations and the corresponding area measurements are made publicly available for independent analyses. Using these data, our results can be easily replicated by using the “general linear model” analysis under standard statistical software (e.g. SPSS, IBM Corp.) with CCA as the dependent variable, sex as a fixed factor, and age and intracranial capacity variable (x_2) as covariates. Our sample size of 316 is the largest sample size used to date for addressing gender differences in CC size. As mentioned above, the issue of brain size differences is addressed by an automated cranial capacity matching process which produces a unique set of pairings between male and female subjects with nearly equal brain size. Furthermore, we have made the pairings and the corresponding measurements publicly available. We used an automated technique (Ardekani et al. 1997) for detecting the MSP on the 3D high-resolution MP-RAGE MRI scans, and measured the CCA on the MSP after reslicing the MRI volume by trilinear interpolation to zero the yaw and roll angles. The importance of applying a consistent and repeatable method for defining the midsagittal slice in all subjects has been emphasized by Rauch and Jinkins (1996) and Mitchell et al. (2003). We also used an automated multiatlas-based method for CC segmentation in the midsagittal slice (Aljabar et al. 2009; Cabezas et al. 2011; Ardekani et al. 2012), which required only minimal manual intervention in a small number of cases, thus largely avoiding the subjectivity of manual segmentations that have been hitherto used when studying the CC.

We did not segment the CC into its subdivisions (splenium, isthmus, body, genu, rostrum, etc.). To this date, the majority of methods for subdividing the CC have been purely geometrically based (e.g. splenium as the posterior fifth of CC) (Witelson 1985; Weis et al. 1991; Hampel et al. 1998). However, it has been shown that significant variability can be introduced into the measurements with different methods of subdivision (Constant and Ruther 1996). We believe that using the standard methods of subdivision would make analyses prone to Type I and Type II errors. Currently, however, there is a promising technique based on fiber tracking using diffusion tensor imaging (DTI) that has the potential of producing anatomically based segmentations of the CC (Abe et al. 2004; Huang et al. 2005; Styner et al. 2005; Hofer and Frahm 2006). In the OASIS database, however, DTI was not available. In the future, we propose to investigate sex differences in CC subdivisions that have been obtained based on DTI and tractography.

The CCA can be affected by the number, density, and composition of callosal fibers with different diameters. In this paper, it has been shown that on average, for pairs of female and male subjects with equal brain sizes and similar ages, the CCA is larger in the female by a few percent. Given that postmortem studies of callosal fibers in normal subjects

have either found no difference in fiber density between sexes (Aboitiz et al. 1992) or a denser fiber packing in females (Highley et al. 1999), it can be inferred that for a given brain size, the female cerebral hemispheres are more extensively interconnected. The relevance of this finding to theories of the evolution of the brain that depend upon the mechanism of sexual selection (Holloway 1990; Geary 1998; Hirnstein et al. 2008), and theories that relate individual differences in behavior (McGee 1979) and functional and structural brain asymmetries (McGlone 1980; Voyer 1996; Sommer et al. 2004; Wallentin 2009) to CC morphology remains a topic of future investigation. In particular, future studies are needed that relate macro- and microstructural CC morphology to behavior and functional and structural brain asymmetries.

Supplementary Material

The standardized midsagittal slices, automated CC segmentations, and area measurements of all 316 subjects studied in this paper are available online at <http://www.nitrc.org/projects/art>. In addition, the list of the 74 paired young adult subjects and their corresponding CCA measurements are provided. The data can also be obtained directly from the corresponding author.

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Notes

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