



Whole Brain Magnetic Resonance Image Atlases: A Systematic Review of Existing Atlases and Caveats for Use in Population Imaging

David Alexander Dickie^{1,2*†}, Susan D. Shenkin^{1,3,4†}, Devasuda Anblagan^{1,2,5}, Juyoung Lee⁶, Manuel Blesa Cabeza⁵, David Rodriguez^{1,2}, James P. Boardman⁵, Adam Waldman¹, Dominic E. Job^{1,2} and Joanna M. Wardlaw^{1,2,4*}

¹ Brain Research Imaging Centre, Neuroimaging Sciences, Centre for Clinical Brain Sciences, Royal Infirmary of Edinburgh, The University of Edinburgh, Edinburgh, UK, ² Scottish Imaging Network, A Platform for Scientific Excellence (SINAPSE) Collaboration, Glasgow, UK, ³ Geriatric Medicine Unit, Royal Infirmary of Edinburgh, The University of Edinburgh, Edinburgh, UK, ⁴ Department of Psychology, Centre for Cognitive Ageing and Cognitive Epidemiology, The University of Edinburgh, Edinburgh, UK, ⁵ MRC Centre for Reproductive Health, Queen's Medical Research Institute, Edinburgh, UK, ⁶ Graduate Training Centre of Neuroscience, International Max Planck Research School, University of Tübingen, Tübingen, Germany

OPEN ACCESS

Edited by:

David N. Kennedy,
University of Massachusetts Medical
School, USA

Reviewed by:

Xi-Nian Zuo,
Chinese Academy of Sciences, China
Frithjof Kruggel,
University of California, Irvine, USA

*Correspondence:

David Alexander Dickie
david.dickie@ed.ac.uk
Joanna M. Wardlaw
joanna.wardlaw@ed.ac.uk

†These authors have contributed
equally to this work.

Received: 25 August 2016

Accepted: 04 January 2017

Published: 19 January 2017

Citation:

Dickie DA, Shenkin SD, Anblagan D, Lee J, Blesa Cabeza M, Rodriguez D, Boardman JP, Waldman A, Job DE and Wardlaw JM (2017) Whole Brain Magnetic Resonance Image Atlases: A Systematic Review of Existing Atlases and Caveats for Use in Population Imaging. *Front. Neuroinform.* 11:1. doi: 10.3389/fninf.2017.00001

Brain MRI atlases may be used to characterize brain structural changes across the life course. Atlases have important applications in research, e.g., as registration and segmentation targets to underpin image analysis in population imaging studies, and potentially in future in clinical practice, e.g., as templates for identifying brain structural changes out with normal limits, and increasingly for use in surgical planning. However, there are several caveats and limitations which must be considered before successfully applying brain MRI atlases to research and clinical problems. For example, the influential Talairach and Tournoux atlas was derived from a single fixed cadaveric brain from an elderly female with limited clinical information, yet is the basis of many modern atlases and is often used to report locations of functional activation. We systematically review currently available whole brain structural MRI atlases with particular reference to the implications for population imaging through to emerging clinical practice. We found 66 whole brain structural MRI atlases world-wide. The vast majority were based on T1, T2, and/or proton density (PD) structural sequences, had been derived using parametric statistics (inappropriate for brain volume distributions), had limited supporting clinical or cognitive data, and included few younger (>5 and <18 years) or older (>60 years) subjects. To successfully characterize brain structural features and their changes across different stages of life, we conclude that whole brain structural MRI atlases should include: more subjects at the upper and lower extremes of age; additional structural sequences, including fluid attenuation inversion recovery (FLAIR) and T2* sequences; a range of appropriate statistics, e.g., rank-based or non-parametric; and detailed cognitive and clinical profiles of the included subjects in order to increase the relevance and utility of these atlases.

Keywords: brain mapping, MRI imaging, atlases as topic, brain, systematic review, aging, neurodevelopment, neurodegeneration

INTRODUCTION

Structural magnetic resonance imaging (MRI) brain atlases, frequently also referred to in the literature as templates, are important tools for research and, increasingly, clinical practice. Individual brain scans from several individuals can be combined to form a brain image bank, which can in turn be used to form a brain atlas—an anatomical representation of the brain showing group-wise or study population global or regional brain features.

The terms “brain atlas” and “brain template” have both been used commonly in the literature to date; while they may have different meanings in some situations, many papers do not make this clear but rather appear to use the terms interchangeably. Therefore, for the interests of this paper, we focus on using the term “atlas” but use both terms interchangeably. Atlases are derived by statistically summarizing, e.g., averaging, voxel-wise, regional, or global brain MRI measures from several individuals and they may be used in research as registration targets for functional activation, segmentation, and statistical mapping, for example in analysis of population imaging datasets (Good et al., 2001; Buckner et al., 2004; Avants et al., 2008). In the future, atlases may also be used in clinical practice as reference images to support diagnoses of age-related neurodegenerative disorders (Farrell et al., 2009); therefore their reliability and relevance to the clinical population on which they are being used is paramount.

Brain structure in old age and early life is different to brain structure in younger and middle-aged adults (Gur et al., 1991; Courchesne et al., 2000; Good et al., 2001; Sowell et al., 2003). For example, the developing brain presents specific challenges to atlas construction because of marked variations in head size and shape in early life, maturational processes leading to changes in signal intensity profiles (for example, reducing brain water content and increasing cell density over the perinatal period), relatively lower spatial resolution (cortical patterning at term birth is broadly similar to adult patterns but is approximately one third of the volume at adulthood), and lower contrast between tissue classes (Matsuzawa et al., 2001). In children >5 years, the brain is still developing at an accelerated rate. These issues invalidate the application of adult atlases to data acquired during development, because of misclassification of tissues and structures (Muzik et al., 2000; Yoon et al., 2009), and have led to the development of age-specific atlases for early life studies.

In older age the ventricles, particularly the lateral ventricles, and sulci spaces are generally larger, the gray matter and white matter atrophy in varying proportions, and white matter hyperintensities (WMH) are often present (Lemaitre et al., 2005; Dickie et al., 2015b, 2016b). These and the other many features of brain aging, e.g., lacunes, microbleeds and enlarged perivascular spaces, require specific T2-based sequences, such as fluid attenuated inversion recovery (FLAIR) and T2*, to be captured effectively (Wardlaw et al., 2013). Because of these differences in brain structure, the use of an atlas based on only younger subjects and a limited range of sequences can create a bias in life course population studies, e.g., systematic overexpansion (Buckner et al., 2004) or regional distortion of older brains. Even within restricted age bands brain structure is highly variable due to various factors such as ethnicity, medical

history, e.g., hypertension, smoking and cognition (Farrell et al., 2009; Wardlaw et al., 2011). Therefore, population brain atlases must include information on age, sex, ethnicity, relevant medical history, and cognitive testing to have broad uses and relevance. Further, brain atlases should be derived using statistical methods that effectively characterize the wide and irregular variance in brain structure across the life course (Dickie et al., 2013). Attempts to understand this variation and create brain atlases have increased exponentially with the advent of MR and other non-invasive imaging techniques but the origins of this pursuit extend back many thousands of years.

The gyral and sulcal pattern of the human brain is thought to have been first described in 3000 B.C. by Imhotep, an Egyptian “god” of medicine (Adelman and Smith, 1987). Although study of the structure of the brain continued for more than 4500 years, it was not until 1664 when Thomas Willis published *Cerebri Anatome* (“Anatomy of the Brain”) that robust methods for measuring brain structure started to be developed (O’connor, 2003). Willis directed novel autopsies of the brain in which it was first removed from the skull, in contrast to the traditional *in situ* dissections of the time, and then sliced from the base upwards. The slices were then viewed with a microscope and drawn by Christopher Wren (O’connor, 2003). These 350 year old drawings arguably represent the first attempt to create a brain atlas but more detailed atlases of the brains’ cyto- and myelo-architecture did not emerge until the late nineteenth/early twentieth century (Betz, 1874; Brodmann, 1909, 1994; Von Economo and Koskinas, 1925). Such atlases are useful to understand the distribution of tissue types and fibers, but they have little use in modern clinical practice. One of the first clinically relevant atlases was published by Talairach et al. (1967), who developed a 3D coordinate system to assist deep-brain surgery.

The subsequent Talairach and Tournoux atlas (Talairach and Tournoux, 1988) has become one of the most influential atlases in brain imaging (Evans et al., 2012). This atlas provides a standardized set of coordinates to determine specific sites within the brain. It has been used to describe the site of a biopsy, or to compare data from structural MRI, functional MRI (fMRI), SPECT, and PET studies. However, the Talairach and Tournoux atlas has been described as “woefully inadequate” (Toga and Thompson, 2007). The reasons for this, including that it was derived from a single fixed cadaveric brain from an elderly female with limited clinical information, have been listed by many and well-known since the atlases’ inception (Evans et al., 1993, 2012; Devlin and Poldrack, 2007). Indeed, they were noted in the original author’s foreword, “this method is valid with precision only for the brain under consideration” (Talairach and Tournoux, 1988), but this may not be commonly known amongst users of this and derived atlases, e.g., Montreal Neurological Institute (MNI)152 (Brett et al., 2001). Population brain atlases, many of which were descended from Talairach (Evans et al., 2012), may therefore be lacking in age-appropriate, clinically, and cognitively described subjects that were synthesized via appropriate image analysis and statistical methods. It is for this reason that we undertook the following systematic review to identify, collate, and describe existing structural MRI brain atlases.

In this review, we aim to summarize the currently available structural MRI brain atlases across the life span—published in journals and/or on the internet—for researchers in population based imaging. Following our review we discuss the practical, technical, and statistical considerations that should be borne in mind when using brain image atlases.

MATERIALS AND METHODS

We followed “Preferred reporting items for systematic reviews and meta-analyses (PRISMA)” reporting guidelines (Moher et al., 2009) in preparation of this manuscript. From October 2010 to April 2015, we systematically searched for “normal” brain structural MRI atlases. From April 2015 to August 2016, we supplemented this search with: hand searching of reference sections in previous review articles and records we included here (e.g., Mazziotta et al., 2001; Toga et al., 2006; Evans et al., 2012); periodical searching of Google with a subset of these terms; review of content alerts distributed by relevant journal articles, e.g., *NeuroImage* (<http://www.journals.elsevier.com/neuroimage/>), *Human Brain Mapping* [[http://onlinelibrary.wiley.com/journal/10.1002/\(ISSN\)1097-0193](http://onlinelibrary.wiley.com/journal/10.1002/(ISSN)1097-0193)], and *Frontiers in Neuroscience* (<http://journal.frontiersin.org/journal/neuroscience>); and, finally, hand searching of neuroimaging data sharing initiatives *NeuroVault* (<http://neurovault.org/>) and *NITRC* (<http://www.nitrc.org/>). Two authors (DAD and JYL) independently and systematically searched PubMed (including MEDLINE; <http://www.ncbi.nlm.nih.gov/pubmed/>), and the internet using Google (<http://www.google.co.uk/>) and Google Scholar (<http://scholar.google.co.uk/>) with the terms: “Magnetic Resonance Imaging” or “Magnetic Resonance Image” or “Magnetic Resonance Images” or “MRI” or “MR” and “brain” and “template” or “atlas” or “stereotactic” or “stereotaxic” and “human.”

October 2010–August 2016 was the time during which we conducted our search, there were no publication date restrictions on eligibility for inclusion and we included all normal MRI atlases of whole brain structures from across the lifespan. We included atlases with “anatomical” or “structural” sequences and probability maps, e.g., T1-, T2-, T2*-weighted images, and gray matter (GM), white matter (WM), and cerebrospinal fluid (CSF) probability maps. We did not include atlases solely of segmented regional structures (ROI), such as subcortical GM or individual cortical areas (e.g., Westbury et al., 1999; Ahsan et al., 2007), or histological sections (e.g., Eickhoff et al., 2005), but did include atlases that had whole brain and regional structures. We excluded: (1) non-human brain atlases, e.g., macaque; (2) diffusion or functional MRI connectively atlases without anatomical/structural components, e.g., JHU ICBM-DTI-81 and NTU-90 (Yeh and Tseng, 2011); (3) functional MRI brain atlases only, e.g., <http://www.brainmap.org/>; (4) records that described atlas methods only (e.g., Maldjian et al., 2003; Wilke et al., 2008; Van Leemput, 2009; Chen et al., 2012); and (5) atlases that included patients with known neurological or central nervous system disease, e.g., Alzheimer’s disease (Desikan et al., 2006; Loni, 2011).

We provide information reported in each structural MRI brain atlas on the number, age, and sex of participants; sequences collected; statistical derivation method; and clinical/cognitive data found.

RESULTS

We identified 543 potentially eligible records (**Figure 1**) of which 66 met inclusion criteria. Descriptions of each atlas are provided in **Table 1**.

We found 66 structural brain MRI atlases with a total of 10,354 subjects (median = 43, mean = 157, range = 1–2762), including European, North American, Chinese, Japanese, Korean, Indian, and Malay participants.

We identified 19 fetal, neonate and infant (0–5 years); six childhood (5–18 years); 23 young or middle aged adult (18–60 years); seven older adult (aged >60 years); and six life-course atlases including several age groups. Five atlases did not report the age of included subjects.

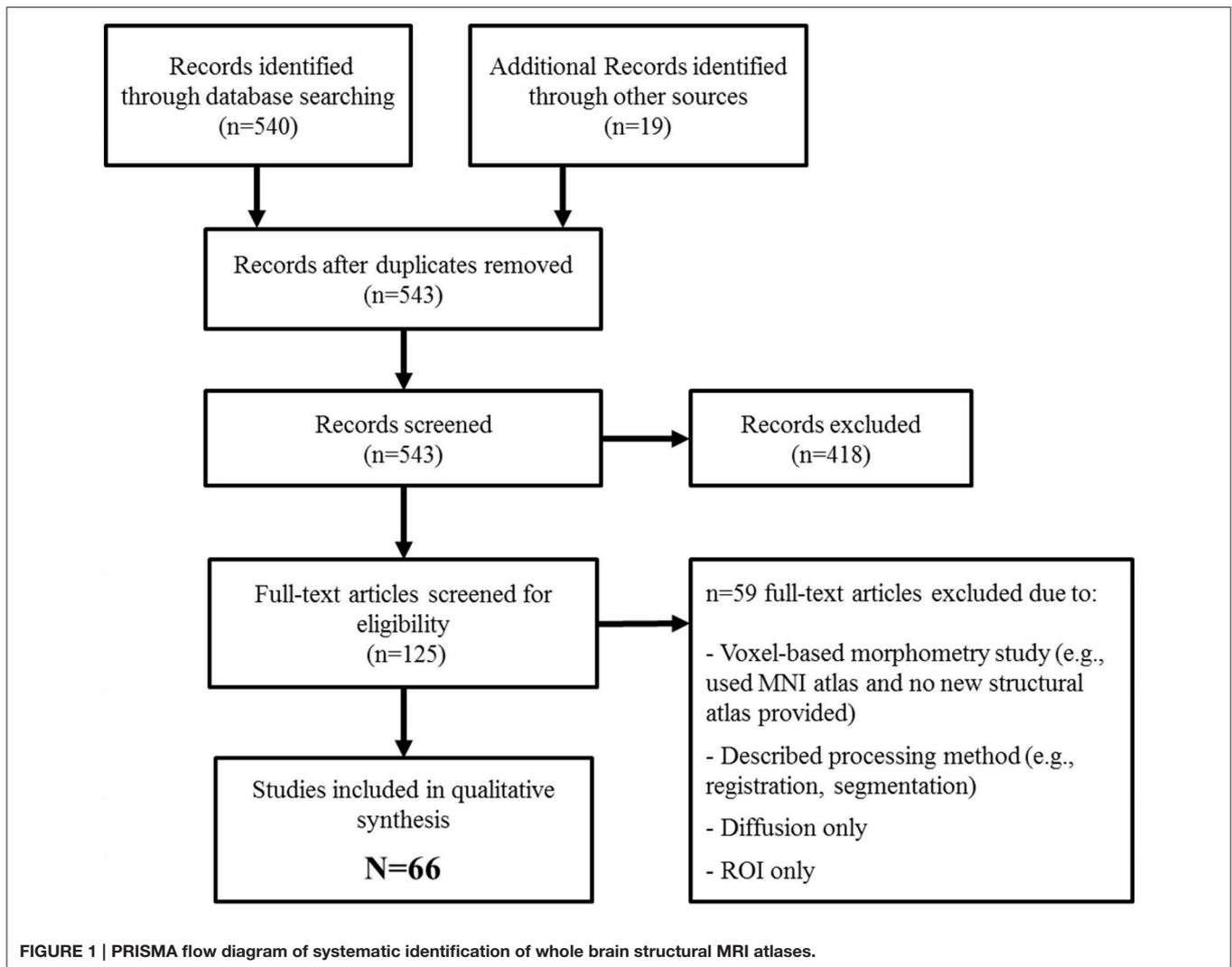
Twenty-seven atlases (41%) reported cognitive/clinical data but this was generally in summary form, e.g., “subjects had no history of neurological, psychiatric or other significant medical illnesses” (Lee et al., 2005) rather than summarized measures from individual subjects. One atlas of the elderly brain reported data on age, handedness, MMSE, education level, and proportion of hypertensive subjects (Lemaitre et al., 2005), but we found no atlas that reported a comprehensive battery of cognitive, medical, and demographic data that are increasingly found in large cohort studies (Wardlaw et al., 2011; Deary et al., 2012).

All atlases were based on T1, T2, and/or PD structural sequences. No atlas included FLAIR or T2* sequences. Almost all multiple subject atlases (except Farrell et al., 2009; Dickie et al., 2015a); were derived using parametric mean-based methods rather than non-parametric percentile ranks or ranges.

Some atlases used the same publicly available databases, e.g., Open Access Series of Imaging Studies (OASIS) data were used in at least two atlases (Dickie et al., 2015a; Richards et al., 2016). We were not able to quantify the subject overlap between atlases as subject identifiers were generally not provided. Ten atlases were based on a single subject. We identified 13 atlases (19.7%) that were developed by or descended from Talairach and Tournoux (labeled “T&T” in **Table 1**).

DISCUSSION

Brain atlases are an important resource for neuroanatomical definition and are often the basis for automated image analyses, which are likely to become increasingly used for population imaging studies. It is important that users are aware of the origins and assumptions underlying these atlases. We identified 66 whole brain structural MRI atlases with a total of 10,354 “normal” subjects from 15 weeks gestational age to 92 years. The number of subjects in each atlas was generally rather small (median = 43; mean = 157; range = 1–2762; $n \geq 100 = 18$; $n \geq 1000 = 3$) given that several hundreds or even thousands of subjects are required to represent population brain structure adequately



(Mazziotta et al., 2001; Toga, 2002; Toga et al., 2006; Evans et al., 2012). Only 622 subjects (6%) had measures of medical, cognitive, and demographic data to support their classification as normal (Lemaitre et al., 2005). Thirteen atlases (~20%) were descended from the Talairach and Tournoux atlas (Talairach and Tournoux, 1988), e.g., MNI, ICBM, and “Brain atlas for healthy elderly.”

Specific populations should be analyzed using an atlas derived from other subjects in that population, or a closely relevant population, otherwise systematic errors may be introduced, e.g., the overexpansion of atrophied brains registered to younger subject atlases (Buckner et al., 2004). Relevant to this, we suggest that the most appropriate atlas for a given study (should there be multiple atlases available with similar demographic, clinical, and cognitive profiles) is the one which requires the least amount of global or regional warping from native subject space to atlas space (and vice-versa). The consequences of various degrees of processing and warping individual subjects to an atlas space have previously been analyzed and discussed (Dickie et al., 2015a). The presence of cognitive deficits and medical conditions, e.g.,

vascular risk factors, also affect brain structure (Ritchie et al., 2015; Dickie et al., 2016b) and therefore it is essential for this information to be measured and tabulated in brain atlases. Although we appreciate that such depths of data may be difficult and expensive to acquire their strong influence on brain structure makes them imperative for understanding the appearance and structure of brain atlases. Medical, cognitive, and demographic data that may be useful in understanding the structure of atlases at different stages of life have been described previously (Job et al., 2016). Given the wide variation and features of brain structure across the life course (Good et al., 2001; Sowell et al., 2003; Allen et al., 2005; Raz et al., 2010), reliable studies, particularly at the extremes of life, require atlases with many more subjects including clinical and cognitive data and additional structural MRI sequences, e.g., T2-based sequences for measuring burden of small vessel disease (Wardlaw et al., 2013).

Such “big-data” approaches including a wide number of imaging sequences and supporting textual information have been successfully applied in studies with limited age ranges such as the “Human Connectome Project” which aims to map structural and

TABLE 1 | Whole brain structural MRI atlases (alphabetical order by name).

Name	Age ^a	N (Sex)	Sequences/contents	Derivation method	Clinical/cognitive data
10–20 sensor placement system structural atlas (Kabdebon et al., 2014)	7.1 weeks	1 (M = 0; F = 1)	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Single subject 	Not reported
4D dynamic probabilistic atlas of developing brains (Kuklisova-Murgasova et al., 2011)	36.6 ± 4.9 weeks GA	142 (M = 70; F = 72)	<ul style="list-style-type: none"> • T2 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise weighted intensity averaging 	Not reported
83 ROI 2-year old atlas (Gousias et al., 2008)	21.4–34.4 (24.8 ± 2.4) months	33 (M = 17; F = 16)	<ul style="list-style-type: none"> • T1 • T2 • ROI 	<ul style="list-style-type: none"> • Single subjects 	Not reported
A database of age-appropriate average MRI templates (Fillmore et al., 2015; Richards et al., 2016)	2 weeks–89 years*	2762	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
A multi-channel 4D probabilistic atlas of the developing fetal brain (Serag et al., 2012b)	29.6 ± 4.6 weeks GA	80	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise weighted intensity averaging 	Not reported
A multi-modal map of human cerebral cortex (Glasser et al., 2016)	22–35 years	210 (M = 80; F = 130)	<ul style="list-style-type: none"> • T1 • T2 • fMRI • rfMRI 	<ul style="list-style-type: none"> • Group average parcellation 	Not reported
A neonatal atlas template (Kazemi et al., 2007)	39–42 weeks GA	7 (M = 4; F = 3)	<ul style="list-style-type: none"> • T1 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
A spatiotemporal atlas of MR intensity, tissue probability and shape of the fetal brain (Habas et al., 2010)	20.57–24.71 weeks GA	20	<ul style="list-style-type: none"> • SSFSE T2 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging • Single subjects 	Not reported
Adult brain maximum probability map: "Hammers adult atlases" (Hammers et al., 2003)	31.6 ± 9.9 years	30 (M = 15; F = 15)	<ul style="list-style-type: none"> • T1 • ROI 	<ul style="list-style-type: none"> • Voxel-wise probabilities 	Reported
Age-specific MRI templates for pediatric neuroimaging (Sanchez et al., 2012a)	4.5–24 years*	1289 (M = 636; F = 653)	<ul style="list-style-type: none"> • T1 • T2/PD 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Allen Human Brain Atlas (Allen Institute for Brain Science, 2009)	24–57 years* (post-mortem)	8 (M = 6; F = 2)	<ul style="list-style-type: none"> • T1 • T2 	<ul style="list-style-type: none"> • Single subjects 	Not reported
Automatic analysis of cerebral atrophy (Subsol et al., 1997)	37 years (mean)	10 (M = 10; F = 0)	<ul style="list-style-type: none"> • T1 • Ventricle map 	<ul style="list-style-type: none"> • Average and SD feature positions 	Reported

(Continued)

TABLE 1 | Continued

Name	Age ^a	N (Sex)	Sequences/contents	Derivation method	Clinical/cognitive data
Bayesian inference atlases (Van Leemput, 2009)		18	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Bayesian inference averaging 	Not reported
Brain atlas for healthy elderly ^{T&T} (Lemaitre et al., 2005)	63–75 years	662 (M = 331; F = 331)	<ul style="list-style-type: none"> • T1 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Brain Characterization Using Normalized Quantitative Magnetic Resonance Imaging (Warrnities et al., 2013)	26–67 (45 ± 11) years	31 (M = 14; F = 17)	<ul style="list-style-type: none"> • R₁ • R₂ • PD 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
Brain Imaging of Normal Subjects (BRAINS) age-specific MRI atlases from young adults to the very elderly (Dickie et al., 2016a)	25–92 years*	225	<ul style="list-style-type: none"> • T1 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Brain template for children from 2 weeks to 4 years age (Sanchez et al., 2012b)	8 days–4.4 years*	154 (M = 83; F = 71)	<ul style="list-style-type: none"> • T1 • T2/PD • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Brainetome atlas (Fan et al., 2016)	22–35 years	49 (M = 17; F = 32)	<ul style="list-style-type: none"> • T1 • T2 • Diffusion • fMRI • ROI 	<ul style="list-style-type: none"> • Voxel-wise probabilities 	Not reported
Cerefy brain atlas ^{T&T} (Nowinski, 2005)	60 years	1 (M = 0; F = 1)	<ul style="list-style-type: none"> • Digitised Talairach plates • ROI 	<ul style="list-style-type: none"> • Single subject 	Not reported
Chinese probabilistic atlas (Xing et al., 2013)	18–70 years*	1000	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Chinese ₅₆ ^{T&T} (Tang et al., 2010)	24.46 ± 1.81 years	56 (M = 56; F = 0)	<ul style="list-style-type: none"> • T1 • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Clinical toolbox ^{T&T} (Forden et al., 2007)	72.9 ± 7.63years	50 (M = 18; F = 32)	<ul style="list-style-type: none"> • T1 • Tissue maps • CT 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
Consistent high-definition spatio-temporal atlas of the developing brain (Serag et al., 2012a)	28–44 (37.3 ± 4.8) weeks PMA	204	<ul style="list-style-type: none"> • T1 • T2 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
Construction of multi-region-multi-reference atlases (Shi et al., 2010)	1.3 ± 0.7 months	68 (M = 38; F = 30)	<ul style="list-style-type: none"> • T2 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported

(Continued)

TABLE 1 | Continued

Name	Age ^a	N (Sex)	Sequences/contents	Derivation method	Clinical/cognitive data
Contributions to 3D Diffeomorphic Atlas Estimation: Application to Brain Images (Bossa et al., 2007)		19	• T1	• Voxel-wise averaging and SD	Not reported
Cortical gray matter of young adults (Luders et al., 2005)	25 ± 4 years	60 (M = 30; F = 30)	• T1 • Tissue maps • ROI	• Average and SD gyral locations	Not reported
Deformable Spatiotemporal MRI Atlas of the Fetal Brain (Gholipour et al., 2014)	26.14–35.86 (30.50 ± 3.05) weeks GA	40	• SSFSE	• Voxel-wise averaging	Not reported
Digital Pediatric Brain Structure Atlas (Shan et al., 2006)	9 years	1 (M = 0; F = 1)	• T1 • ROI	• Single subject	Reported
EvePM (Lim et al., 2013)	33 years	1 (M = 0; F = 1)	• T1 • Diffusion • ROI • susceptibility	• Single subject	Not reported
FreeSurfer “Destrieux” cortical atlas (Destrieux et al., 2010)	18–33 years	12 (M = 6; F = 6)	• T1 • ROI	• Vertex-wise probabilities	Not reported
Group-specific brain tissue probability map (Yoon et al., 2005)	26.07 ± 5.32 years	59 (M = 36; F = 23)	• T1 • Tissue maps • ROI	• Voxel-wise averaging	Reported
Harvard brain atlas (Shenton et al., 1995)	25 years	1 (M = 1; F = 0)	• T1 • ROI	• Single subject	Reported
Harvard-Oxford cortical and subcortical structural (Fmrib, 2008)	18–50 years	37 (M = 21; F = 16)	• T1 • ROI	• Voxel-wise probabilities	Not reported
Human cortical development map (Gogtay et al., 2004)	13.0 ± 4.8 years*	13 (M = 6; F = 7)	• T1 • GM map • ROI	• Average gyral locations	Reported
ICBM452 T&T (Lancaster et al., 2007)	20–40 years (27.8 ± 5.1) years	452	• T1 • T2 • Tissue maps • ROI	• Voxel-wise averaging	Not reported
Infant brain atlas (Altaie et al., 2009)	9–15 months	76 (M = 31; F = 45)	• T1 • Tissue maps	• Voxel-wise averaging	Not reported
Japanese pediatric standard brain (Uchiyama et al., 2013)	6–9 years	45 (M = 22; F = 23)	• T1	• Voxel-wise averaging	Reported

(Continued)

TABLE 1 | Continued

Name	Age ^a	N (Sex)	Sequences/contents	Derivation method	Clinical/cognitive data
JHU-neonatal brain atlas (Oishi et al., 2011)	0–4 days	25 (M = 15; F = 10)	<ul style="list-style-type: none"> • T1 • T2 • Diffusion 	<ul style="list-style-type: none"> • Voxel-wise averaging • Single subject 	Not reported
Korean standard brain template (Lee et al., 2005)	18–77 (44.6 ± 19.4) years*	78 (M = 49; F = 29)	<ul style="list-style-type: none"> • T1 • F-18-FDG PET 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
LPBA40 ^{T&T} (Shattuck et al., 2008)	19–39 (29 ± 6) years	40 (M = 20; F = 20)	<ul style="list-style-type: none"> • T1 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging • Voxel-wise probabilities 	Reported
Merged young- and old-adult atlas target: "Washington 711" ^{T&T} (Buckner et al., 2004)	49 years	24 (M = 9; F = 15)	<ul style="list-style-type: none"> • T1 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Mindboggle-101 (Klein and Tourville, 2012)	19–61 years	101 (M = 57; F = 44)	T1 ROI	<ul style="list-style-type: none"> • Single subjects 	Not reported
MNI/ICBM 152 ^{T&T} (Mazziotta et al., 2001)	18–44 (24 ± 7) years	152 (M = 86; F = 66)	<ul style="list-style-type: none"> • T1 • T2/PD • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
MNI 305 ^{T&T} (Evans et al., 1993)	23.4 ± 4.1 years	305 (M = 239; F = 66)	<ul style="list-style-type: none"> • T1 • Brain masks 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
MNI Pediatric atlases ^{T&T} (Fonov et al., 2011)	0–18.5 years*	324	<ul style="list-style-type: none"> • T1 • T2/PD • Tissue maps • Brain masks 	<ul style="list-style-type: none"> • Voxel-wise averaging and SD 	Not reported
MNI-Colin27 ^{T&T} (Holmes et al., 1998; Aubert-Broche et al., 2006)		1 (M = 1; F = 0)	<ul style="list-style-type: none"> • T1 • T2/PD • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging (of repeated single subject scans) 	Not reported
Neonatal brain atlas: "ALBERT" ^{T&T} (Gousias et al., 2012)	39–45 (41) weeks PMA	5 (M = 3; F = 2)	<ul style="list-style-type: none"> • T1 • T2 • ROI 	<ul style="list-style-type: none"> • Single subjects 	Reported
Neonatal brain template of 1 week newborn (Hashicka et al., 2012)	5.6 ± 17.6 days	14 (M = 11; F = 3)	<ul style="list-style-type: none"> • T2 	<ul style="list-style-type: none"> • Voxel-wise averaging • Single subjects 	Not reported

(Continued)

TABLE 1 | Continued

Name	Age ^a	N (Sex)	Sequences/contents	Derivation method	Clinical/cognitive data
Neonatal probabilistic models (Kazemi et al., 2008)	39–42 weeks	7 (M = 3; F = 4)	<ul style="list-style-type: none"> • T1 • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
Non-parametric percentile rank atlas of the aging brain (Dickie et al., 2015a)	55–90 years	98 (M = 40; F = 58)	<ul style="list-style-type: none"> • T1 • GM map 	<ul style="list-style-type: none"> • Voxel-wise non-parametric percentile ranking 	Reported
Normal Brain F-18 FDG-PET and MRI Atlas (Schiffner et al., 1993)		1	<ul style="list-style-type: none"> • T1 • T2 • FDG-PET 	<ul style="list-style-type: none"> • Co-registration of within subject images 	Not reported
Normal reference MR images for aging brain (Fairrell et al., 2009)	65–80 years*	79 (M = 61; F = 18)	<ul style="list-style-type: none"> • T1 • T2 	<ul style="list-style-type: none"> • Qualitative percentile ranking • Voxel-wise averaging 	Reported
NTU standard Chinese brain template (Jao et al., 2009)	19–42 (25.7) years	95 (M = 50; F = 45)	<ul style="list-style-type: none"> • T1 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Parcellation of the Healthy Neonatal Brain into 107 Regions (Blesa et al., 2016)	39–47 ⁺¹ (42 ⁺²) weeks	33	<ul style="list-style-type: none"> • T1 • T2 • Diffusion • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise majority voting 	Reported
Population difference in brain among Chinese, Malay and Indian neonates (Bai et al., 2012)	5–17 days	177 (M = 94; F = 83)	<ul style="list-style-type: none"> • T2 • Diffusion 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Population-Average, Landmark- and Surface-based (PALS) atlas (Van Essen, 2005)	18–24 years	12 (M = 6; F = 6)	<ul style="list-style-type: none"> • T1 • Cortical surface 	<ul style="list-style-type: none"> • Selected landmark averaging 	Not reported
Regional growth and atlas of the developing human brain (Makropoulos et al., 2016)	39 ⁺¹ (27 ⁺¹ –44 ⁺⁶) weeks PMA	338	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
Resource atlases for multi-atlas brain segmentations with multiple ontology levels based on T1-weighted MRI (Wu et al., 2016)	4–82 years*	90	<ul style="list-style-type: none"> • T1 • ROI 	<ul style="list-style-type: none"> • Hierarchical ontology 	Not reported
Spatial-temporal fetal atlas (Zhan et al., 2013)	15–22 weeks GA*	34 (M = 12; F = 22)	<ul style="list-style-type: none"> • T2 	<ul style="list-style-type: none"> • Voxel-wise averaging and SD 	Reported
SRI24 (Rohlfing et al., 2010)	19–84 (62 ± 5) years	24 (M = 12; F = 12)	<ul style="list-style-type: none"> • T1 • T2/PD • Diffusion • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported

(Continued)

TABLE 1 | Continued

Name	Age ^a	N (Sex)	Sequences/contents	Derivation method	Clinical/cognitive data
Symmetric atlas in normal older adults ^{T&T} (Grabner et al., 2006)	75 ± 6 years	153	<ul style="list-style-type: none"> • T1 • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Not reported
Talairach and Tournoux ^{T&T} (Talairach and Tournoux, 1988; Brett et al., 2001)	60 years	1 (M = 0; F = 1)	<ul style="list-style-type: none"> • Histological slices • Photographs • Hand drawings • Stereotactic coordinates 	<ul style="list-style-type: none"> • Postmortem slicing • Photography • Drawing 	Not reported
The human brain in 1700 pieces (Nowinski et al., 2012)		1 (M = 0; F = 1)	<ul style="list-style-type: none"> • T1 • 3D TOF • SWI • Diffusion • ROI 	<ul style="list-style-type: none"> • Single subject 	Not reported
The pediatric template of brain perfusion (Avants et al., 2015)	7–18 years	120 (M = 59; F = 61)	<ul style="list-style-type: none"> • T1 • BOLD • Diffusion • pCASL • ROI • Tissue maps 	<ul style="list-style-type: none"> • Voxel-wise averaging 	Reported
Three-dimensional digitized mono-subject anatomical template (Lalys et al., 2010)	45 years	1 (M = 1; F = 0)	<ul style="list-style-type: none"> • T1 • T2 	<ul style="list-style-type: none"> • Voxel-wise kappa-sigma clipping average (of repeated single subject scans) 	Not reported
UNC Infant 0–1–2 atlases (Shi et al., 2011)	0–2 years	95 (M = 56; F = 39)	<ul style="list-style-type: none"> • T1 • T2 • Tissue maps • ROI 	<ul style="list-style-type: none"> • Voxel-wise averaging • Voxel-wise majority voting (maximum probability) 	Reported

Empty or partially empty cells indicate that we could not find relevant data in original manuscripts; ^aage is reported as in the original manuscript and is shown "range (mean ± SD)" if available; MRI, magnetic resonance imaging; SD, standard deviation; ROI, region of interest; PD, proton density; SWI, susceptibility weighted imaging; fMRI, task-based functional magnetic resonance imaging; rfMRI, resting-state functional magnetic resonance imaging; PMA, post-menstrual age; GA, gestational age; pCASL, pseudo continuous arterial spin labeled; BOLD, blood oxygen level-dependent; SSFSE, single shot fast spin echo; M, male; F, female; T&T, developed by or descended from Talairach and Tournoux.

functional connections in the healthy brain between ages 22 to 35 years (Van Essen et al., 2012) and UK Biobank (Miller et al., 2016). The challenge is to collect similarly rich and relevant data, including sequences such as T2* and FLAIR and vascular risk factor measures for appropriately characterizing cerebrovascular and cognitive development/aging effects on brain structure, at the extremes of life. An international collaborative and aggregative approach may be the best way of achieving this goal as was recently agreed by a panel of experts in structural brain mapping in 2014 (Job et al., 2016) and as is evidenced in similar efforts in functional imaging (Zuo et al., 2014). Although there are challenges to aggregating brain MRI from multiple centers/scanners, particularly in functional connectomics (Zuo and Xing, 2014), these issues have received great attention (e.g., Gountouna et al., 2010; Gradin et al., 2010) and the variability between scanners has often shown to be nominal compared to the great variability in brain structure among even people of the same age, gender, and cognitive status (Dickie et al., 2013; Ritchie et al., 2015; Miller et al., 2016).

High resolution structural MRI is increasingly used in population imaging to study brain development in fetal (pre-birth), neonatal (birth to 4 weeks corrected gestational age) and pediatric (1 month to 18 years) populations because of its utility to: provide quantitative measures of typical brain growth; map atypical growth following complications such as preterm birth, perinatal asphyxia and stroke; evaluate tissue effects of neuroprotective treatment strategies; identify the neural substrates of long-term neurodevelopmental impairments; and because it has potential to uncover early life origins of adult neurological and psychiatric disease. All of these applications benefit from the anatomic context provided by atlases.

There are challenges in analyzing structural images in early and late life. These begin during image acquisition and extend into image analysis. For example, infant participants are asleep during scanning while adults are usually awake; motion artifacts are generally low in mid-life but increase at the extremes of life; and heart and respiratory rates also vary greatly through life (Zuo et al., 2017). Brain structural patterns also vary greatly through life: in early life growth is rapid and head shape and size varies, with a changes in tissue composition and relatively low spatial resolution (Matsuzawa et al., 2001). In older people there is accelerated brain tissue loss, reduced cortical contrast, white matter disease, enlarged perivascular spaces, stroke infarcts, and microbleeds, among other features (Raz et al., 2010; Wardlaw et al., 2013; Dickie et al., 2016b). There have been several ($N = 19$) fetal, neonate, or infant (<age 5) atlases published, but our review found relatively limited age-specific childhood ($N = 6$: >5 and <18 years) and older adult atlases ($N = 7$: >60 years) compared to young/middle-aged adult atlases ($N = 23$). Despite their current under-representation in the literature, age-specific atlases in childhood, and old age may have important uses in research and clinical practice, such as providing targets for aiding classification and diagnoses of developmental and neurodegenerative diseases (Farrell et al., 2009; Dickie et al., 2013, 2014), particularly since better understanding of normal development, aging, and

dementia prevention are major focuses of many large population studies.

Most atlases we found were based on mean/parametric statistics and designed to provide a standard space for voxel-wise analyses or support tissue/ROI volume segmentation. In contrast, the “Normal reference MR images for the brain” atlas was based on qualitatively determined percentile ranks of brain volumes during normal aging and designed to support clinical diagnoses of whole brain volume loss in aging (65–70 and 75–80 year old) patients (Farrell et al., 2009). These clinical atlases are designed to “calibrate” differences in perception between neuroradiologists and have been of growing interest and in increased use since their inception in 2009 (Farrell et al., 2009; Hoggard, 2009; Job et al., 2016). Additionally, increased interest in use of computational automated image processing in clinical practice, e.g., to assess brain, hippocampus, or white matter lesion volumes, relies on availability of relevant and reliable age-relevant atlases. Atlases based on parametric statistics, e.g., mean and standard deviation, are not suitable to define the irregular brain volume distributions in old age (Dickie et al., 2013, 2015a). Therefore, non-parametric statistics were recently applied quantitatively to derive voxel-based percentile ranks and limits of normal aging GM, but this atlas was limited by the use of only T1 sequences and a wide age range (Dickie et al., 2015a). Further, work in developing non-parametric distributional representations of the brain, including a broad range of sequences in well-described (cognitively and medically) age-specific groups, may lead to clinically useful atlases for supporting diagnoses of developmental and neurodegenerative disease (Farrell et al., 2009; Wardlaw et al., 2013; Dickie et al., 2014).

The strengths of our review include the use of structured methods, that were reported following the PRISMA Guidelines (Moher et al., 2009), over ~6 years. We also conducted an exhaustive manual search of printed and online materials, and provided a structured evaluation of brain atlases according to pre-specified criteria. This allowed us to produce a holistic review of structural MRI brain atlases from across the life course in detail that we have not found previously. But despite these strengths, our review also has some limitations. The atlases we found were openly published, and identified through a formal search thus we may not have identified all relevant atlases, e.g., those described as part of larger studies (and therefore potentially not visible through traditional search methods) or those not published/openly accessible. We report data as described in the paper or website, and it is possible that additional data, e.g., on subjects’ age, sex, clinical information, was collected and may have been published elsewhere. We did not contact authors for additional information. Further, we did not investigate potential uses for atlases beyond those described in the original manuscripts/sources. It could be that any one of these atlases may be modified to serve additional purposes. Related to this, we described the methods and uses of each atlas according to our interpretation of the source manuscripts/reference manuals, which may differ from the meaning intended by the original authors.

Notwithstanding these limitations, we have reviewed and described structural MRI brain atlases from across the life course and found that they were mostly of modest size with limited supporting subject information, developed with restricted image sequences for specific processing purposes, and that childhood and elderly populations were under-represented. We conclude that there is a continuing need for multi-sequence structural MRI, and the associated clinical, medical, and demographic data, collected in population imaging studies to be made widely available (with appropriate legal and ethical approvals) to create non-parametric brain atlases that adequately reflect the variability and features of brain changes throughout the life course. Brain image databanks, such as Brain Imaging in Normal Subjects (BRAINS; <https://www.brainsimagebank.ac.uk/>; Job et al., 2016), should work together to maximize sample sizes, generalizability and optimize data use to benefit analyses in population imaging studies and in future clinical practice.

AUTHOR CONTRIBUTIONS

DAD and JL conducted systematic searches of the literature and internet. DAD, SS, JL, DA, MBC, and JB, conducted hand searching and reviewing of the literature and internet. DAD and SS wrote the manuscript. DAD, SS, JL, DA, MBC, JB, AW, DR, DJ, and JW edited the manuscript. DAD, SS, DR, DJ, and JW conceptualized and designed the study.

REFERENCES

- Adelman, G., and Smith, B. H. (1987). *Encyclopedia of Neuroscience*. Boston, MA: Birkhäuser.
- Ahsan, R. L., Allom, R., Gousias, I. S., Habib, H., Turkheimer, F. E., Free, S., et al. (2007). Volumes, spatial extents and a probabilistic atlas of the human basal ganglia and thalamus. *Neuroimage* 38, 261–270. doi: 10.1016/j.neuroimage.2007.06.004
- Allen Institute for Brain Science (2009). *Allen Human Brain Atlas*. Available online at: http://human.brain-map.org/mri_viewers/data (Accessed August 26, 2013).
- Allen, J. S., Bruss, J., Brown, C. K., and Damasio, H. (2005). Normal neuroanatomical variation due to age: the major lobes and a parcellation of the temporal region. *Neurobiol. Aging* 26, 1245–1260. doi: 10.1016/j.neurobiolaging.2005.05.023
- Altay, M., Holland, S. K., Wilke, M., and Gaser, C. (2008). Infant brain probability templates for MRI segmentation and normalization. *Neuroimage* 43, 721–730. doi: 10.1016/j.neuroimage.2008.07.060
- Aubert-Broche, B., Evans, A. C., and Collins, L. (2006). A new improved version of the realistic digital brain phantom. *Neuroimage* 32, 138–145. doi: 10.1016/j.neuroimage.2006.03.052
- Avants, B. B., Duda, J. T., Kilroy, E., Krasileva, K., Jann, K., Kandel, B. T., et al. (2015). The pediatric template of brain perfusion. *Sci. Data* 2, 150003. doi: 10.1038/sdata.2015.3
- Avants, B. B., Epstein, C. L., Grossman, M., and Gee, J. C. (2008). Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain. *Med. Image Anal.* 12, 26–41. doi: 10.1016/j.media.2007.06.004
- Bai, J., Abdul-Rahman, M. F., Rifkin-Graboi, A., Chong, Y.-S., Kwek, K., Saw, S.-M., et al. (2012). Population differences in brain morphology and microstructure among Chinese, Malay, and Indian Neonates. *PLoS ONE* 7:e47816. doi: 10.1371/journal.pone.0047816
- Betz, W. (1874). Anatomischer nachweis zweier gehirncentra. *Zentralbl Med Wiss* 12, 578–595.

ACKNOWLEDGMENTS

The authors would like to gratefully acknowledge and thank the following centres and funders. This work was carried out in The University of Edinburgh Brain Research Imaging Centre (BRIC; <http://www.bric.ed.ac.uk/>) within the Department of Neuroimaging Sciences. BRIC is part of the Scottish Imaging Network, A Platform for Scientific Excellence (SINAPSE) collaboration (<http://www.sinapse.ac.uk/>), funded by the Scottish Funding Council, Scottish Executive Chief Scientist Office, and the six collaborator Universities. Professor JW was funded by the Scottish Funding Council and Scottish Executive Chief Scientist Office through the SINAPSE collaboration. DAD was funded by a SINAPSE industrial collaboration (SPIRIT) Ph.D. scholarship, a Medical Research Council (MRC) scholarship, and the Tony Watson Scholarship bequest to The University of Edinburgh; and is currently funded by Innovate UK. Dr. DJ was funded by Wellcome Trust Grant 007393/Z/05/Z. Funding from Edinburgh and Lothians Health Foundation 53/311 and BBSRC Sparking Impact SI 2013-0210 is gratefully acknowledged. The University of Edinburgh Centre for Cognitive Aging and Cognitive Epidemiology (SS) is part of the cross council Lifelong Health and Wellbeing Initiative (G0700704/84698). Funding from the Biotechnology and Biological Sciences Research Council, Engineering and Physical Sciences Research Council, Economic and Social Research Council, Medical Research Council is also gratefully acknowledged.

- Blesa, M., Serag, A., Wilkinson, A. G., Anblagan, D., Telford, E. J., Pataky, R., et al. (2016). Parcellation of the healthy neonatal brain into 107 regions using atlas propagation through intermediate time points in childhood. *Front. Neurosci.* 10:220. doi: 10.3389/fnins.2016.00220
- Bossa, M., Hernandez, M., and Olmos, S. (2007). “Contributions to 3D diffeomorphic atlas estimation: application to brain images,” in *Proceedings of the 10th International Conference on Medical Image Computing and Computer-Assisted Intervention - Volume Part I* (Brisbane, QLD: Springer-Verlag).
- Brett, M., Christoff, K., Cusack, R., and Lancaster, J. (2001). Using the Talairach atlas with the MNI template. *Neuroimage* 13, S85. doi: 10.1016/S1053-8119(01)91428-4
- Brodman, K. (1909). *Vergleichende Lokalisationslehre der Großhirnrinde in ihren Prinzipien dargestellt auf Grund des Zellenbaues*. Leipzig: Barth.
- Brodman, K. (1994). *The Principles of Comparative Localisation in the Cerebral Cortex Based on Cytoarchitectonics*. London: Smith-Gordon.
- Buckner, R. L., Head, D., Parker, J., Fotenos, A. F., Marcus, D., Morris, J. C., et al. (2004). A unified approach for morphometric and functional data analysis in young, old, and demented adults using automated atlas-based head size normalization: reliability and validation against manual measurement of total intracranial volume. *Neuroimage* 23, 724–738. doi: 10.1016/j.neuroimage.2004.06.018
- Chen, T., Rangarajan, A., Eisenschenk, S. J., and Vemuri, B. C. (2012). Construction of a neuroanatomical shape complex atlas from 3D MRI brain structures. *Neuroimage* 60, 1778–1787. doi: 10.1016/j.neuroimage.2012.01.095
- Courchesne, E., Chisum, H. J., Townsend, J., Cowles, A., Covington, J., Egaas, B., et al. (2000). Normal brain development and aging: quantitative analysis at *in vivo* MR imaging in healthy volunteers. *Radiology* 216, 672–682. doi: 10.1148/radiology.216.3.r00au37672
- Deary, I. J., Gow, A. J., Pattie, A., and Starr, J. M. (2012). Cohort profile: the Lothian birth cohorts of 1921 and 1936. *Int. J. Epidemiol.* 41, 1576–1584. doi: 10.1093/ije/dyr197
- Desikan, R. S., Ségonne, F., Fischl, B., Quinn, B. T., Dickerson, B. C., Blacker, D., et al. (2006). An automated labeling system for subdividing the human

- cerebral cortex on MRI scans into gyral based regions of interest. *Neuroimage* 31, 968–980. doi: 10.1016/j.neuroimage.2006.01.021
- Destrieux, C., Fischl, B., Dale, A., and Halgren, E. (2010). Automatic parcellation of human cortical gyri and sulci using standard anatomical nomenclature. *Neuroimage* 53, 1–15. doi: 10.1016/j.neuroimage.2010.06.010
- Devlin, J. T., and Poldrack, R. A. (2007). In praise of tedious anatomy. *Neuroimage* 37, 1033–1041. doi: 10.1016/j.neuroimage.2006.09.055
- Dickie, D. A., Job, D. E., Gonzalez, D. R., Shenkin, S. D., Ahearn, T. S., Murray, A. D., et al. (2013). Variance in brain volume with advancing age: implications for defining the limits of normality. *PLoS ONE* 8:e84093. doi: 10.1371/journal.pone.0084093
- Dickie, D. A., Job, D. E., Gonzalez, D. R., Shenkin, S. D., and Wardlaw, J. M. (2015a). Use of brain MRI atlases to determine boundaries of age-related pathology: the importance of statistical method. *PLoS ONE* 10:e0127939. doi: 10.1371/journal.pone.0127939
- Dickie, D. A., Job, D. E., Rodriguez, D., Robson, A., Danso, S., Pernet, C., et al. (2016a). *Brain Imaging of Normal Subjects (BRAINS) Age-Specific MRI Atlases from Young Adults to the Very Elderly (v1.0)*, [dataset]. University of Edinburgh, Edinburgh Imaging, CCBS, BRAINS Imagebank. doi: 10.7488/ds/1369
- Dickie, D. A., Job, D. E., Sparrow, S., Piyasena, C., Wilkinson, G., Wardlaw, J. M., et al. (2014). “Preterm infant brain pathology revealed in individuals by voxel ranking against a normal term atlas” in *Proceedings of the 20th Annual Meeting of the Organization for Human Brain Mapping* (Hamburg).
- Dickie, D. A., Karama, S., Ritchie, S. J., Cox, S. R., Sakka, E., Royle, N. A., et al. (2015b). Progression of white matter disease and cortical thinning are not related in older community-dwelling subjects. *Stroke* 47, 410–416. doi: 10.1161/STROKEAHA.115.011229
- Dickie, D. A., Ritchie, S. J., Cox, S. R., Sakka, E., Royle, N. A., Aribisala, B. S., et al. (2016b). Vascular risk factors and progression of white matter hyperintensities in the Lothian Birth Cohort 1936. *Neurobiol. Aging* 42, 116–123. doi: 10.1016/j.neurobiolaging.2016.03.011
- Eickhoff, S. B., Stephan, K. E., Mohlberg, H., Grefkes, C., Fink, G. R., Amunts, K., et al. (2005). A new SPM toolbox for combining probabilistic cytoarchitectonic maps and functional imaging data. *Neuroimage* 25, 1325–1335. doi: 10.1016/j.neuroimage.2004.12.034
- Evans, A. C., Collins, D. L., Mills, S. R., Brown, E. D., Kelly, R. L., and Peters, T. M. (1993). “3D statistical neuroanatomical models from 305 MRI volumes,” in *IEEE Nuclear Science Symposium and Medical Imaging Conference* (San Francisco, CA), 1813–1817.
- Evans, A. C., Janke, A. L., Collins, D. L., and Baillet, S. (2012). Brain templates and atlases. *Neuroimage* 62, 911–922. doi: 10.1016/j.neuroimage.2012.01.024
- Fan, L., Li, H., Zhuo, J., Zhang, Y., Wang, J., Chen, L., et al. (2016). The human Brainnetome atlas: a new brain atlas based on connective architecture. *Cereb. Cortex* 26, 3508–3526. doi: 10.1093/cercor/bhw157
- Farrell, C., Chappell, F., Armitage, P. A., Keston, P., MacLulich, A., Shenkin, S., et al. (2009). Development and initial testing of normal reference MR images for the brain at ages 65–70 and 75–80 years. *Eur. Radiol.* 19, 177–183. doi: 10.1007/s00330-008-1119-2
- Fillmore, P. T., Phillips-Meek, M., and Richards, J. E. (2015). Age-specific MRI brain and head templates for healthy adults from twenty through eighty-nine years of age. *Front. Aging Neurosci.* 7:44. doi: 10.3389/fnagi.2015.00044
- Fmrib (2008). *Atlases Included with FSL*. Available online at: <http://www.fmrib.ox.ac.uk/fsl/data/atlas-descriptions.html> (Accessed March 25, 2011).
- Fonov, V., Evans, A. C., Botteron, K., Almli, C. R., McKinstry, R. C., and Collins, D. L. (2011). Unbiased average age-appropriate atlases for pediatric studies. *Neuroimage* 54, 313–327. doi: 10.1016/j.neuroimage.2010.07.033
- Gholipour, A., Limperopoulos, C., Clancy, S., Clouchoux, C., Akhondi-Asl, A., Estroff, J. A., et al. (2014). Construction of a deformable spatiotemporal mri atlas of the fetal brain: evaluation of similarity metrics and deformation models. *Med. Image Comput. Comput. Assist. Interv.* 17, 292–299. doi: 10.1007/978-3-319-10470-6_37
- Glasser, M. F., Coalson, T. S., Robinson, E. C., Hacker, C. D., Harwell, J., Yacoub, E., et al. (2016). A multi-modal parcellation of human cerebral cortex. *Nature* 536, 171–178. doi: 10.1038/nature18933
- Gogtay, N., Giedd, J. N., Lusk, L., Hayashi, K. M., Greenstein, D., Vaituzis, A. C., et al. (2004). Dynamic mapping of human cortical development during childhood through early adulthood. *Proc. Natl. Acad. Sci. U.S.A.* 101, 8174–8179. doi: 10.1073/pnas.0402680101
- Good, C. D., Johnsrude, I. S., Ashburner, J., Henson, R. N. A., Friston, K. J., and Frackowiak, R. S. J. (2001). A voxel-based morphometric study of ageing in 465 normal adult human brains. *Neuroimage* 14, 21–36. doi: 10.1006/nimg.2001.0786
- Gountouna, V., Job, D., McIntosh, A., Moorhead, T., Lymer, G., Whalley, H., et al. (2010). Functional magnetic resonance imaging (fMRI) reproducibility and variance components across visits and scanning sites with a finger tapping task. *Neuroimage* 49, 552–560. doi: 10.1016/j.neuroimage.2009.07.026
- Gousias, I. S., Edwards, A. D., Rutherford, M. A., Counsell, S. J., Hajnal, J. V., Rueckert, D., et al. (2012). Magnetic resonance imaging of the newborn brain: manual segmentation of labelled atlases in term-born and preterm infants. *Neuroimage* 62, 1499–1509. doi: 10.1016/j.neuroimage.2012.05.083
- Gousias, I. S., Rueckert, D., Heckemann, R. A., Dyet, L. E., Boardman, J. P., Edwards, A. D., et al. (2008). Automatic segmentation of brain MRIs of 2-year-olds into 83 regions of interest. *Neuroimage* 40, 672–684. doi: 10.1016/j.neuroimage.2007.11.034
- Grabner, G., Janke, A. L., Budge, M. M., Smith, D., Pruessner, J., and Collins, D. L. (2006). Symmetric atlas and model based segmentation: an application to the hippocampus in older adults. *Med. Image Comput. Comput. Assist. Interv.* 4191, 58–66. doi: 10.1007/11866763_8
- Gradin, V., Gountouna, V. E., Waiter, G., Ahearn, T. S., Brennan, D., Condon, B., et al. (2010). Between-and within-scanner variability in the CaliBrain study n-back cognitive task. *Psychiatry Res. Neuroimaging* 184, 86–95. doi: 10.1016/j.psychres.2010.08.010
- Gur, R. C., Mozley, P. D., Resnick, S. M., Gottlieb, G. L., Kohn, M., Zimmerman, R., et al. (1991). Gender differences in age effect on brain atrophy measured by magnetic resonance imaging. *Proc. Natl. Acad. Sci. U.S.A.* 88, 2845–2849. doi: 10.1073/pnas.88.7.2845
- Habas, P. A., Kim, K., Corbett-Detig, J. M., Rousseau, F., Glenn, O. A., Barkovich, A. J., et al. (2010). A spatiotemporal atlas of MR intensity, tissue probability and shape of the fetal brain with application to segmentation. *Neuroimage* 53, 460–470. doi: 10.1016/j.neuroimage.2010.06.054
- Hammers, A., Allom, R., Koepp, M., Free, S., Myers, R., Lemieux, L., et al. (2003). Three-dimensional maximum probability atlas of the human brain, with particular reference to the temporal lobe. *Hum. Brain Mapp.* 19, 224–247. doi: 10.1002/hbm.10123
- Hashioka, A., Kobashi, S., Kuramoto, K., Wakata, Y., Ando, K., Ishikura, R., et al. (2012). A neonatal brain MR image template of 1 week newborn. *Int. J. Comput. Assist. Radiol. Surg.* 7, 273–280. doi: 10.1007/s11548-011-0646-5
- Hoggard, N. (2009). Re: development and initial testing of normal reference MR images for the brain at ages 65–70 and 75–80 years. *Eur. Radiol.* 19, 1025–1025. doi: 10.1007/s00330-008-1230-4
- Holmes, C. J., Hoge, R., Collins, L., Woods, R., Toga, A. W., and Evans, A. C. (1998). Enhancement of MR images using registration for signal averaging. *J. Comput. Assist. Tomogr.* 22, 324. doi: 10.1097/00004728-199803000-00032
- Jao, T., Chang, C. Y., Li, C. W., Chen, D. Y., Wu, E., Wu, C. W., et al. (2009). “Development of NTU standard Chinese brain template: Morphologic and functional comparison with MNI template using magnetic resonance imaging,” in *2009 Annual International Conference of the IEEE Engineering in Medicine and Biology Society* (Minneapolis, MN), 4779–4782.
- Job, D. E., Dickie, D. A., Rodriguez, D., Robson, A., Danso, S., Pernet, C., et al. (2016). A brain imaging repository of normal structural MRI across the life course: Brain Images of Normal Subjects (BRAINS). *NeuroImage* 144, 299–304. doi: 10.1016/j.neuroimage.2016.01.027
- Kabdebon, C., Leroy, F., Simmonet, H., Perrot, M., Dubois, J., and Dehaene-Lambertz, G. (2014). Anatomical correlations of the international 10–20 sensor placement system in infants. *Neuroimage* 99, 342–356. doi: 10.1016/j.neuroimage.2014.05.046
- Kazemi, K., Ghadimi, S., Abrishami-Moghaddam, H., Grebe, R., Gondry-Jouet, C., and Wallois, F. (2008). “Neonatal probabilistic models for brain, CSF and skull using T1-MRI data: Preliminary results,” in *2008 30th Annual International Conference of the IEEE Engineering in Medicine and Biology Society* (Vancouver, BC), 3892–3895.
- Kazemi, K., Moghaddam, H. A., Grebe, R., Gondry-Jouet, C., and Wallois, F. (2007). A neonatal atlas template for spatial normalization of whole-brain

- magnetic resonance images of newborns: preliminary results. *Neuroimage* 37, 463–473. doi: 10.1016/j.neuroimage.2007.05.004
- Klein, A., and Tourville, J. (2012). 101 labeled brain images and a consistent human cortical labeling protocol. *Front. Neurosci.* 6:171. doi: 10.3389/fnins.2012.00171
- Kuklisova-Murgasova, M., Aljabar, P., Srinivasan, L., Counsell, S. J., Doria, V., Serag, A., et al. (2011). A dynamic 4D probabilistic atlas of the developing brain. *Neuroimage* 54, 2750–2763. doi: 10.1016/j.neuroimage.2010.10.019
- Lalys, F., Haegelen, C., Ferre, J.-C., El-Ganaoui, O., and Jannin, P. (2010). Construction and assessment of a 3-T MRI brain template. *Neuroimage* 49, 345–354. doi: 10.1016/j.neuroimage.2009.08.007
- Lancaster, J. L., Tordesillas-Gutiérrez, D., Martínez, M., Salinas, F., Evans, A., Zilles, K., et al. (2007). Bias between MNI and Talairach coordinates analyzed using the ICBM-152 brain template. *Hum. Brain Mapp.* 28, 1194–1205. doi: 10.1002/hbm.20345
- Lee, J. S., Lee, D. S., Kim, J., Kim, Y. K., Kang, E., Kang, H., et al. (2005). Development of Korean standard brain templates. *J. Korean Med. Sci.* 20, 483–488. doi: 10.3346/jkms.2005.20.3.483
- Lemaitre, H., Crivello, F., Grassiot, B., Alperovitch, A., Tzourio, C., and Mazoyer, B. (2005). Age- and sex-related effects on the neuroanatomy of healthy elderly. *Neuroimage* 26, 900–911. doi: 10.1016/j.neuroimage.2005.02.042
- Lim, I. A. L., Faria, A. V., Li, X., Hsu, J. T. C., Airan, R. D., Van Zijl, P. C. M., et al. (2013). Human brain atlas for automated region of interest selection in quantitative susceptibility mapping: application to determine iron content in deep gray matter structures. *Neuroimage* 82, 449–469. doi: 10.1016/j.neuroimage.2013.05.127
- Loni (2011). *Alzheimer's Disease Template*. Available online at: <http://www.loni.usc.edu/atlas/> (Accessed).
- Luders, E., Narr, K. L., Thompson, P. M., Woods, R. P., Rex, D. E., Jancke, L., et al. (2005). Mapping cortical gray matter in the young adult brain: effects of gender. *Neuroimage* 26, 493–501. doi: 10.1016/j.neuroimage.2005.02.010
- Makropoulos, A., Aljabar, P., Wright, R., Hüning, B., Merchant, N., Arichi, T., et al. (2016). Regional growth and atlas of the developing human brain. *Neuroimage* 125, 456–478. doi: 10.1016/j.neuroimage.2015.10.047
- Maldjian, J. A., Laurienti, P. J., Kraft, R. A., and Burdette, J. H. (2003). An automated method for neuroanatomic and cytoarchitectonic atlas-based interrogation of fMRI data sets. *Neuroimage* 19, 1233–1239. doi: 10.1016/S1053-8119(03)00169-1
- Matsuzawa, J., Matsui, M., Konishi, T., Noguchi, K., Gur, R. C., Bilker, W., et al. (2001). Age-related volumetric changes of brain gray and white matter in healthy infants and children. *Cereb. Cortex* 11, 335–342. doi: 10.1093/cercor/11.4.335
- Mazziotta, J., Toga, A., Evans, A., Fox, P., Lancaster, J., Zilles, K., et al. (2001). A probabilistic atlas and reference system for the human brain: International Consortium for Brain Mapping (ICBM). *Philos. Trans. R. Soc. Lond. B Biol. Sci.* 356, 1293–1322. doi: 10.1098/rstb.2001.0915
- Miller, K. L., Alfaro-Almagro, F., Bangarter, N. K., Thomas, D. L., Yacoub, E., Xu, J., et al. (2016). Multimodal population brain imaging in the UK Biobank prospective epidemiological study. *Nat. Neurosci.* 19, 1523–1536. doi: 10.1038/nn.4393
- Moher, D., Liberati, A., Tetzlaff, J., and Altman, D. G. (2009). Preferred reporting items for systematic reviews and meta-analyses: the PRISMA statement. *PLoS Med.* 6:e1000097. doi: 10.1371/journal.pmed.1000097
- Muzik, O., Chugani, D. C., Juhász, C., Shen, C., and Chugani, H. T. (2000). Statistical parametric mapping: assessment of application in children. *Neuroimage* 12, 538–549. doi: 10.1006/nimg.2000.0651
- Nowinski, W. L. (2005). The cerefy brain atlases. *Neuroinformatics* 3, 293–300. doi: 10.1385/NI:3:4:293
- Nowinski, W. L., Chua, B. C., Qian, G. Y., and Nowinska, N. G. (2012). The human brain in 1700 pieces: design and development of a three-dimensional, interactive and reference atlas. *J. Neurosci. Methods* 204, 44–60. doi: 10.1016/j.jneumeth.2011.10.021
- O'Connor, J. P. B. (2003). Thomas Willis and the background to Cerebri Anatomie. *J. R. Soc. Med.* 96, 139–143. doi: 10.1258/jrsm.96.3.139
- Oishi, K., Mori, S., Donohue, P. K., Ernst, T., Anderson, L., Buchthal, S., et al. (2011). Multi-contrast human neonatal brain atlas: application to normal neonate development analysis. *Neuroimage* 56, 8–20. doi: 10.1016/j.neuroimage.2011.01.051
- Raz, N., Ghisletta, P., Rodrigue, K. M., Kennedy, K. M., and Lindenberger, U. (2010). Trajectories of brain aging in middle-aged and older adults: regional and individual differences. *Neuroimage* 51, 501–511. doi: 10.1016/j.neuroimage.2010.03.020
- Richards, J. E., Sanchez, C., Phillips-Meek, M., and Xie, W. (2016). A database of age-appropriate average MRI templates. *NeuroImage* 124(Pt B), 1254–1259. doi: 10.1016/j.neuroimage.2015.04.055
- Ritchie, S. J., Dickie, D. A., Cox, S. R., Valdes Hernandez, M. D. C., Corley, J., Royle, N. A., et al. (2015). Brain volumetric changes and cognitive ageing during the eighth decade of life. *Hum. Brain Mapp.* 36, 4910–4925. doi: 10.1002/hbm.22959
- Rohlfing, T., Zahr, N., Sullivan, E., and Pfefferbaum, A. (2010). The SRI24 multichannel atlas of normal adult human brain structure. *Hum. Brain Mapp.* 31, 798–819. doi: 10.1002/hbm.20906
- Rorden, C., Bonilha, L., and Nichols, T. E. (2007). Rank-order versus mean based statistics for neuroimaging. *Neuroimage* 35, 1531–1537. doi: 10.1016/j.neuroimage.2006.12.043
- Sanchez, C. E., Richards, J. E., and Almli, C. R. (2012a). Age-specific MRI templates for pediatric neuroimaging. *Dev. Neuropsychol.* 37, 379–399. doi: 10.1080/87565641.2012.688900
- Sanchez, C. E., Richards, J. E., and Almli, C. R. (2012b). Neurodevelopmental MRI brain templates for children from 2 weeks to 4 years of age. *Dev. Psychobiol.* 54, 77–91. doi: 10.1002/dev.20579
- Schifter, T., Turkington, T. G., Berlangieri, S. U., Hoffman, J. M., Macfall, J. R., Pelizzari, C. A., et al. (1993). Normal brain F-18 FDG-PET and MRI anatomy. *Clin. Nucl. Med.* 18, 578–582. doi: 10.1097/00003072-199307000-00008
- Serag, A., Aljabar, P., Ball, G., Counsell, S. J., Boardman, J. P., Rutherford, M. A., et al. (2012a). Construction of a consistent high-definition spatio-temporal atlas of the developing brain using adaptive kernel regression. *Neuroimage* 59, 2255–2265. doi: 10.1016/j.neuroimage.2011.09.062
- Serag, A., Kyriakopoulou, V., Rutherford, M., Edwards, A., Hajnal, J., Aljabar, P., et al. (2012b). A multi-channel 4D probabilistic atlas of the developing brain: application to fetuses and neonates. *Ann. BMVA*, 2012, 1–14.
- Shan, Z. Y., Parra, C., Ji, Q., Ogg, R. J., Zhang, Y., Laningham, F. H., et al. (2006). “A digital pediatric brain structure atlas from T1-weighted MR images,” in *Medical Image Computing and Computer-Assisted Intervention – MICCAI 2006: 9th International Conference, Copenhagen, Denmark, October 1–6, 2006, Proceedings, Part II*, eds R. Larsen, M. Nielsen, and J. Sparring. (Berlin; Heidelberg: Springer Berlin Heidelberg), 332–339.
- Shattuck, D. W., Mirza, M., Adisetiyo, V., Hojatkashani, C., Salamon, G., Narr, K. L., et al. (2008). Construction of a 3D probabilistic atlas of human cortical structures. *Neuroimage* 39, 1064–1080. doi: 10.1016/j.neuroimage.2007.09.031
- Shenton, M. E., Kikinis, R., McCarley, R. W., Saiviroonpom, P., Hokama, H. H., Robatino, A., et al. (1995). “Harvard brain atlas: a teaching and visualization tool,” in *Proceedings of the 1995 Biomedical Visualization* (Washington, DC), 10–17.
- Shi, F., Yap, P.-T., Fan, Y., Gilmore, J. H., Lin, W., and Shen, D. (2010). Construction of multi-region-multi-reference atlases for neonatal brain MRI segmentation. *Neuroimage* 51, 684–693. doi: 10.1016/j.neuroimage.2010.02.025
- Shi, F., Yap, P.-T., Wu, G., Jia, H., Gilmore, J. H., Lin, W., et al. (2011). Infant Brain Atlases from Neonates to 1- and 2-Year-Olds. *PLoS ONE* 6:e18746. doi: 10.1371/journal.pone.0018746
- Sowell, E. R., Peterson, B. S., Thompson, P. M., Welcome, S. E., Henkenius, A. L., and Toga, A. W. (2003). Mapping cortical change across the human life span. *Nat. Neurosci.* 6, 309–315. doi: 10.1038/nn1008
- Subsol, G., Roberts, N., Doran, M., Thirion, J.-P., and Whitehouse, G. H. (1997). Automatic analysis of cerebral atrophy. *Magn. Reson. Imaging* 15, 917–927. doi: 10.1016/S0730-725X(97)00002-7
- Talairach, J., and Tournoux, P. (1988). *Co-planar Stereotactic Atlas of the Human Brain: 3-dimensional Proportional System: An Approach to Cerebral Imaging*. Stuttgart: Georg Thieme Verlag.

- Talairach, J., Szikla, G., Tournoux, P., Prosalentis, A., Bordas-Ferrier, M., Covelto, L., et al. (1967). *Atlas D'Anatomie Stereotaxique du Telencephale*. Paris: Masson.
- Tang, Y., Hojatkishani, C., Dinov, I., Sun, B., Fan, L., Lin, X., et al. (2010). The construction of a Chinese MRI brain atlas: a morphometric comparison study between Chinese and Caucasian cohorts. *Neuroimage* 51, 33–41. doi: 10.1016/j.neuroimage.2010.01.111
- Toga, A. W. (2002). Neuroimage databases: the good, the bad and the ugly. *Nat. Rev. Neurosci.* 3, 302–309. doi: 10.1038/nrn782
- Toga, A. W., and Thompson, P. M. (2007). What is where and why it is important. *Neuroimage* 37, 1045–1068. doi: 10.1016/j.neuroimage.2007.02.018
- Toga, A. W., Thompson, P. M., Mori, S., Amunts, K., and Zilles, K. (2006). Towards multimodal atlases of the human brain. *Nat. Rev. Neurosci.* 7, 952–966. doi: 10.1038/nrn2012
- Uchiyama, H. T., Seki, A., Tanaka, D., Koeda, T., and Group, J. C. S. (2013). A study of the standard brain in Japanese children: morphological comparison with the MNI template. *Brain Dev.* 35, 228–235. doi: 10.1016/j.braindev.2012.04.005
- Van Essen, D. C. (2005). A population-average, landmark-and surface-based (PALS) atlas of human cerebral cortex. *Neuroimage* 28, 635–662. doi: 10.1016/j.neuroimage.2005.06.058
- Van Essen, D. C., Ugurbil, K., Auerbach, E., Barch, D., Behrens, T. E. J., Bucholz, R., et al. (2012). The Human Connectome Project: a data acquisition perspective. *Neuroimage* 62, 2222–2231. doi: 10.1016/j.neuroimage.2012.02.018
- Van Leemput, K. (2009). Encoding probabilistic brain atlases using Bayesian inference. *IEEE Trans. Med. Imaging* 28, 822–837. doi: 10.1109/TMI.2008.2010434
- Von Economo, C., and Koskinas, G. N. (1925). *Die Cytoarchitektur der Hirnrinde des Erwachsenen Menschen*. Berlin: Julius Springer.
- Wardlaw, J. M., Bastin, M. E., Valdés Hernández, M. C., Mu-oz Maniega, S., Royle, N. A., Morris, Z., et al. (2011). Brain aging, cognition in youth and old age and vascular disease in the Lothian Birth Cohort 1936: rationale, design and methodology of the imaging protocol. *Int. J. Stroke* 6, 547–559. doi: 10.1111/j.1747-4949.2011.00683.x
- Wardlaw, J. M., Smith, E. E., Biessels, G. J., Cordonnier, C., Fazekas, F., Frayne, R., et al. (2013). Neuroimaging standards for research into small vessel disease and its contribution to ageing and neurodegeneration. *Lancet Neurol.* 12, 822–838. doi: 10.1016/S1474-4422(13)70124-8
- Warntjes, J. B. M., Engström, M., Tisell, A., and Lundberg, P. (2013). Brain characterization using normalized quantitative magnetic resonance imaging. *PLoS ONE* 8:e70864. doi: 10.1371/journal.pone.0070864
- Westbury, C. F., Zatorre, R. J., and Evans, A. C. (1999). Quantifying variability in the planum temporale: a probability map. *Cereb. Cortex* 9, 392–405. doi: 10.1093/cercor/9.4.392
- Wilke, M., Holland, S., Altaye, M., and Gaser, C. (2008). Template-O-Matic: a toolbox for creating customized pediatric templates. *Neuroimage* 41, 903–913. doi: 10.1016/j.neuroimage.2008.02.056
- Wu, D., Ma, T., Ceritoglu, C., Li, Y., Chotiyononta, J., Hou, Z., et al. (2016). Resource atlases for multi-atlas brain segmentations with multiple ontology levels based on T1-weighted MRI. *Neuroimage* 125, 120–130. doi: 10.1016/j.neuroimage.2015.10.042
- Xing, W., Nan, C., Zhentao, Z., Rong, X., Luo, J., Zhuo, Y., et al. (2013). Probabilistic MRI brain anatomical atlases based on 1,000 Chinese subjects. *PLoS ONE* 8:e50939. doi: 10.1371/journal.pone.0050939
- Yeh, F.-C., and Tseng, W.-Y. I. (2011). NTU-90: a high angular resolution brain atlas constructed by q-space diffeomorphic reconstruction. *Neuroimage* 58, 91–99. doi: 10.1016/j.neuroimage.2011.06.021
- Yoon, U., Fonov, V. S., Perusse, D., and Evans, A. C. (2009). The effect of template choice on morphometric analysis of pediatric brain data. *Neuroimage* 45, 769–777. doi: 10.1016/j.neuroimage.2008.12.046
- Yoon, U., Lee, J.-M., Koo, B. B., Shin, Y.-W., Lee, K. J., Kim, I. Y., et al. (2005). Quantitative analysis of group-specific brain tissue probability map for schizophrenic patients. *Neuroimage* 26, 502–512. doi: 10.1016/j.neuroimage.2005.01.056
- Zhan, J., Dinov, I. D., Li, J., Zhang, Z., Hobel, S., Shi, Y., et al. (2013). Spatial-temporal atlas of human fetal brain development during the early second trimester. *Neuroimage* 82, 115–126. doi: 10.1016/j.neuroimage.2013.05.063
- Zuo, X.-N., and Xing, X.-X. (2014). Test-retest reliabilities of resting-state fMRI measurements in human brain functional connectomics: a systems neuroscience perspective. *Neurosci. Biobehav. Rev.* 45, 100–118. doi: 10.1016/j.neubiorev.2014.05.009
- Zuo, X.-N., Anderson, J. S., Bellec, P., Birn, R. M., Biswal, B. B., Blautzik, J., et al. (2014). An open science resource for establishing reliability and reproducibility in functional connectomics. *Sci. Data* 1, 140049. doi: 10.1038/sdata.2014.49
- Zuo, X.-N., He, Y., Betzel, R. F., Colcombe, S., Sporns, O., and Milham, M. P. (2017). Human connectomics across the life span. *Trends Cogn. Sci.* 21, 32–45. doi: 10.1016/j.tics.2016.10.005

Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Copyright © 2017 Dickie, Shenkin, Anblagan, Lee, Blesa Cabez, Rodriguez, Boardman, Waldman, Job and Wardlaw. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) or licensor are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.