

Tract-specific white matter microstructure alterations among young adult *APOE* ε4 carriers: A replication and extension study

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Abstract

The parahippocampal cingulum bundle (PHCB) connects regions known to be vulnerable to early Alzheimer's disease (AD) pathology, such as posteromedial cortex and medial temporal lobe. While AD-related pathology has been robustly associated with alterations in PHCB microstructure, specifically lower fractional anisotropy (FA) and higher mean diffusivity (MD), emerging evidence indicates that the reverse pattern is evident in younger adults at-risk of AD. In one such study, Hodgetts et al. (2019) reported that healthy young adult carriers of the apolipoprotein-E (*APOE*) $\epsilon 4$ allele – the strongest common genetic risk factor for AD – showed higher FA and lower MD in the PHCB but not the inferior longitudinal fasciculus (ILF). These results are consistent with proposals claiming that heightened neural activity and connectivity have a significant role in posteromedial cortex vulnerability to amyloid- β and tau spread beyond the medial temporal lobe. Given the implications for understanding AD risk, here we sought to replicate Hodgetts et al.'s finding in a larger sample ($N = 128$; 40 *APOE* $\epsilon 4$ carriers, 88 *APOE* $\epsilon 4$ non-carriers) of young adults (age range: 19-33). Extending this work further, we also conducted exploratory analyses using a more advanced measure of microstructure: hindrance modulated orientational anisotropy (HMOA). These analyses included an investigation of hemispheric asymmetry in PHCB and ILF HMOA. Contrary to the original study, we observed no difference in PHCB microstructure between *APOE* $\epsilon 4$ carriers and non-carriers. Bayes factors (BFs) further revealed moderate-to-strong evidence in support of these null findings. *APOE* $\epsilon 4$ -related differences in ILF HMOA asymmetry were evident, however, with carriers demonstrating lower leftward asymmetry. Our findings indicate that young adult *APOE* $\epsilon 4$ carriers do not show alterations in PHCB microstructure, as observed by Hodgetts et al., but may show altered asymmetry in ILF microstructure.

Keywords: *APOE*, Alzheimer's disease, parahippocampal cingulum bundle, inferior longitudinal fasciculus, diffusion MRI, structural connectivity

1. Introduction

Alzheimer's disease (AD) is a chronic, progressive disease and the most common cause of dementia (Scheltens et al., 2021). The hallmark pathological features of AD are the presence of extracellular amyloid- β -containing plaques and intracellular tau-containing neurofibrillary tangles (DeTure & Dickson, 2019; Trejo-Lopez et al., 2021). Although controversial (Frisoni et al., 2022; Herrup, 2015), the dominant hypothesis in the field – the amyloid cascade hypothesis – holds that the accumulation of amyloid- β peptide is the critical factor in AD pathogenesis (Selkoe & Hardy, 2016). Amyloid- β accumulation follows a relatively distinct spatiotemporal pattern in the ageing brain, beginning preferentially in posteromedial regions, including retrosplenial/posterior cingulate cortices and precuneus (Mattsson et al., 2019; Palmqvist et al., 2017; Villeneuve et al., 2015). Collectively, these regions are sometimes referred to as posteromedial cortex (Parvizi et al., 2006). The vulnerability of posteromedial cortex to AD pathology has been linked to its hub-like properties (Jagust, 2018), in particular its high-levels of baseline metabolic/neural activity and high intrinsic/extrinsic connectivity (Bero et al., 2012; Buckner et al., 2009; de Haan et al., 2012). Notably, posteromedial cortex is densely connected with several medial temporal lobe structures, such as parahippocampal cortex and hippocampus, forming a “posterior medial” or “extended navigation” network (Murray et al., 2017; Ranganath & Ritchey, 2012). This broader network is implicated in several inter-related cognitive functions that are impaired early in AD, such as episodic memory (Rajah et al., 2017), perceptual scene discrimination (Lee et al., 2006), and spatial navigation (Coughlan et al., 2018). Given this, there is a pressing need to identify biomarkers that capture the functional and/or structural integrity of this AD-vulnerable brain network. In this context, the parahippocampal cingulum bundle (PHCB) – a prominent white matter tract linking posteromedial cortex with the medial temporal lobe (Bubb et al., 2018; Heilbronner & Haber, 2014; Jitsuishi & Yamaguchi, 2021) – represents a strong candidate for understanding and characterising connectivity alterations associated with AD.

Increasing evidence indicates that PHCB connectivity is altered in AD. Using diffusion magnetic resonance imaging (dMRI), a non-invasive method that examines the random, microscopic movement of water molecules, it is possible to delineate the major white matter tracts of the brain and evaluate their microstructural properties in vivo (Assaf et al., 2019). In most AD-relevant dMRI studies, white matter microstructure is assessed via measures derived from the diffusion tensor, notably fractional anisotropy (FA) and mean diffusivity (MD; Harrison et al., 2020). Low FA and high MD are widely interpreted as representing poorer microstructural integrity and thus lower connectivity (Yeh et al., 2021), although multiple biological factors – including neuroinflammation (Kor et al., 2022) – can influence these measures (Jones, Knösche, & Turner, 2013). Studies comparing AD patients to cognitively normal older adults using dMRI have reliably observed both lower FA and higher MD in the cingulum bundle and the PHCB in particular (Acosta-Cabronero et al., 2010; Bozzali et al., 2012; Choo et al., 2010; Kantarci et al., 2017). In addition, longitudinal changes in PHCB microstructure – reduced FA, increased MD – have been reported among AD patients but not cognitively normal older adults (Mayo et al., 2017). Indeed, it has recently been suggested that PHCB FA constitutes a highly effective biomarker for differentiating between typical ageing and AD (Dalboni da Rocha et al., 2020).

Studies of amnesic mild cognitive impairment (aMCI), a transitional stage between typical ageing and AD (Albert et al., 2011), further highlight that PHCB alterations precede the onset of AD dementia. In one region-of-interest (ROI) meta-analysis, for example, Yu et al. (2017) identified robust alterations in PHCB microstructure (lower FA, higher MD) among individuals with aMCI. This is congruent with the notion that cingulum bundle alterations predict cognitive decline in aMCI and may even predict conversion to AD (Gozdas et al., 2020). Studies combining positron emission tomography and dMRI have also allowed PHCB changes to be linked directly to AD pathology. For example, amyloid- β burden has been associated with longitudinal changes in white matter microstructure that are consistent with patterns observed in aMCI and AD (Rieckmann et al., 2016; Song et al., 2018; Vipin et al.,

2019). In particular, high levels of cortical amyloid- β burden at baseline have been associated with accelerated decline in PHCB FA and a trend-level increase in PHCB MD (Rieckmann et al., 2016). In keeping with this tract-specific finding, one recent cross-sectional study reported that lower FA and higher MD in the PHCB was associated with greater cortical amyloid- β and entorhinal tau burden, especially in those with high levels of pre-existing pathology (Pichet Binnette et al., 2021). It thus appears that PHCB microstructure is detrimentally impacted over the course of AD, including stages prior to the onset of dementia symptoms.

Emerging research indicates, however, that asymptomatic individuals exhibit alterations in white matter microstructure that run counter to the characteristic AD pattern. Illustrating this point, several studies have observed higher FA and lower MD in early-stage amyloid- β pathology, a pattern that is reversed as pathology further accrues (Collij et al., 2021; Dong et al., 2020; Wolf et al., 2015). These findings point to a biphasic pattern of microstructure over the disease course, with a period of high FA/low MD preceding the pattern commonly observed in patients with aMCI and AD. While increased FA in the context of early AD pathology could reflect neuroinflammation (Benitez et al., 2021; Dong et al., 2020), there is evidence that heightened activity and connectivity – including structural connectivity – may actually precede AD pathology, predisposing individuals to later amyloid- β deposition (Bero et al., 2012; Buckner et al., 2009; de Haan et al., 2012). Support for this proposal can be found in studies of young adults carriers of the apolipoprotein-E (*APOE*) ϵ 4 allele. The *APOE* ϵ 4 allele is the strongest common genetic risk factor for AD (Belloy et al., 2019), and is also associated with a younger age of onset and faster rate of posteromedial amyloid- β accumulation (Burnham et al., 2020; Mishra et al., 2018). In line with the notion that this amyloid- β accumulation is related to earlier connectivity changes, a study applying graph theoretical analysis to dMRI data observed that age was negatively associated with local interconnectivity in posteromedial regions, but only among *APOE* ϵ 4 carriers (Brown et al., 2011). Higher levels of local interconnectivity in younger adults drove this finding, such that

there was a putative *APOE* $\epsilon 4$ -related increase in connectivity early in life that was subsequently followed by a sharper decline later in the lifespan (Brown et al., 2011; see also Ma et al., 2017). Relatedly, Felsky and Voineskos (2013) further reported higher cingulum bundle FA in younger *APOE* $\epsilon 4$ carriers compared to younger non-carriers, but lower cingulum bundle FA in older *APOE* $\epsilon 4$ carriers compared to older non-carriers. Given that young adults are unlikely to possess significant amyloid- β burden (Jansen et al., 2015), these findings suggest that early-life structural alterations may precede pathology.

Consistent with this, Hodgetts et al. (2019) observed higher FA and lower MD among *APOE* $\epsilon 4$ carriers relative to non-carriers in the PHCB but not the inferior longitudinal fasciculus (ILF), a tract that connects the occipital lobe to the ventro-anterior temporal lobe (Herbet et al., 2018). Hodgetts et al. also found that PHCB microstructure was correlated with posteromedial cortex activity during perceptual scene discrimination, a task that has previously been shown to elicit heightened activity in young *APOE* $\epsilon 4$ carriers (Shine et al., 2015) and is sensitive to AD (Lee et al., 2006). Based on the proposal that heightened neural activity and connectivity can have a significant role in hub-like vulnerability to amyloid- β (Bero et al., 2012; Buckner et al., 2009; de Haan et al., 2012), it is plausible that such early-life PHCB alterations may explain why *APOE* $\epsilon 4$ is associated with earlier and faster posteromedial amyloid- β accumulation (Burnham et al., 2020; Mishra et al., 2018). Moreover, as the spread of tau has been linked to heightened functional connectivity between posteromedial cortex and the medial temporal lobe (Ziontz et al., 2021) – presumably mediated by the PHCB (Jacobs et al., 2018) – it is possible that early-life increases in structural connectivity are also related to elevated tau in *APOE* $\epsilon 4$ carriers (Therriault et al., 2020).

In view of the potential implications for understanding the role of *APOE* $\epsilon 4$ in AD risk, we sought to replicate Hodgetts et al.'s (2019) finding that healthy young adult *APOE* $\epsilon 4$ carriers demonstrate higher FA and lower MD than non-carriers in the PHCB but not the ILF. We

analysed data from an independent data set of young adults, with a total sample over four times larger than the original study. This replication attempt thus constitutes an important test of the notion that increased PHCB connectivity, as indexed by higher FA and lower MD, is evident in young adult *APOE* $\epsilon 4$ carriers, potentially increasing vulnerability to both amyloid- β accumulation and tau spread.

We also report additional exploratory analyses that seek to extend this work by incorporating a more advanced measure of microstructure: hindrance modulated orientational anisotropy (HMOA; Dell'Acqua et al., 2013). HMOA is regarded as a tract-specific measure of microstructure and is argued to be more sensitive to alterations in anisotropy than either FA or MD (Dell'Acqua et al., 2013). As such, we investigated whether *APOE* $\epsilon 4$ is associated with differences in PHCB and ILF HMOA, complementing the primary (replication) analyses. In addition, we also assessed whether *APOE* $\epsilon 4$ is associated with asymmetry in PHCB and ILF HMOA. Recent evidence suggests that AD is characterised by a loss of typical or “healthy” leftward structural and functional asymmetry in the brain (Banks et al., 2018; Roe et al., 2021; Tyrer et al., 2020), perhaps as a result of hemispheric differences in susceptibility to AD pathology (Lubben et al., 2021; Weise et al., 2018). Given the proposal that early-life *APOE* $\epsilon 4$ -related alterations in neural activity and connectivity increase vulnerability to AD pathology, notably amyloid- β accumulation but perhaps also tau spread, it is plausible that this allele may be associated with changes in the asymmetry of key white matter tracts. To our knowledge, no study to date has yet investigated this possibility, especially in healthy young adults.

2. Method

2.1. Participants

Participant data were acquired from a repository at the Cardiff University Brain Research Imaging Centre. Portions of this data have been published elsewhere (Foley et al., 2017; Koelewijn et al., 2019). Participants were healthy adults, who were screened via interview or

questionnaire for the presence of neuropsychiatric disorders. All were right-handed, had normal or corrected-to-normal vision, and provided informed consent for their data to be used in imaging genetics analyses. All procedures were originally reviewed and approved by the Cardiff University School of Psychology Research Ethics Committee. For the current study, participants were only included if they completed the requisite MRI scans, had *APOE* genotype information available, and were aged 35 years or under ($N = 148$). After additional exclusions were applied – described below (see also Supplementary Figure 1) – the final sample comprised 128 participants (86 females, 42 males) aged between 19 and 33 years ($M = 23.8$, $SD = 3.6$).

Consistent with Hodgetts et al. (2019), the final sample was split into carrier and non-carrier groups based on the presence/absence of the *APOE* $\epsilon 4$ allele (Table 1). Participants carrying both risk-enhancing ($\epsilon 4$) and risk-reducing ($\epsilon 2$) *APOE* alleles were included as part of the carrier group, as the $\epsilon 2\epsilon 4$ genotype is associated with higher levels of AD pathology and risk (Goldberg et al., 2020; Jansen et al., 2015; Reiman et al., 2020). Although *APOE* is often directly genotyped, as in Hodgetts et al.'s study, here it was inferred from imputed (1000G phase 1, version 3) genome-wide genetic data (for more detail, see Foley et al., 2017). Previous research has demonstrated that it is possible to accurately infer *APOE* genotypes using this method (Lupton et al., 2018; Oldmeadow et al., 2014; Radmanesh et al., 2014). Overall, the current sample included 40 *APOE* $\epsilon 4$ carriers (4 $\epsilon 2/\epsilon 4$, 33 $\epsilon 3/\epsilon 4$, 3 $\epsilon 4/\epsilon 4$) and 88 *APOE* $\epsilon 4$ non-carriers (4 $\epsilon 2/\epsilon 2$, 14 $\epsilon 2/\epsilon 3$, 70 $\epsilon 3/\epsilon 3$). An effect size sensitivity analysis calculated using the *pwr* package (version 1.2-2; Champely, 2018) in R (version 3.6.0; R Core Team, 2019) using RStudio (version 1.3.1093; RStudio Team, 2020) revealed that the smallest effect size detectable at 80% power was Cohen's $d_s = 0.575$ ($1-\beta = .80$, Bonferroni-corrected $\alpha = .016$, directional hypothesis). By comparison, even without correcting the α level for multiple comparisons, the smallest effect size detectable at 80% power in Hodgetts et al.'s study was Cohen's $d_s = 0.931$ ($1-\beta = .80$, $\alpha = .05$, directional

hypothesis). Basic sample characteristics in this study and in Hodgetts et al.'s study are compared in Supplementary Table 1.

Table 1

Basic Sample Characteristics Separated by APOE ε4 Carrier Status.

	APOE ε4+	APOE ε4-	Statistics
	(n = 40)	(n = 88)	
Age (years; $M \pm SD$)	23.9 \pm 3.3	23.7 \pm 3.7	$t(84.84) = 0.226, p = .822$, Cohen's $d_s = 0.042$, $BF_{10} = 0.206$
Sex (Males/Females; n) ^a	12/28	30/58	$X^2(1, N = 128) = 0.209, p = .648, \phi = 0.04$, $BF_{10} = 0.241$

Note. Frequentist null hypothesis significance tests (two-sided Welch's t -test for age, chi-square test for sex) revealed no significant difference between APOE ε4 carriers and non-carriers in terms of age or sex. Effect sizes were also small, while complementary BF analyses provided moderate evidence in support of the null hypothesis of no difference. Abbreviations: APOE ε4+ = APOE ε4 carrier, APOE ε4- = APOE ε4 non-carrier, M = mean, n = number of participants, SD = standard deviation.

^aAlthough sex was self-reported, it was checked against chromosomal sex as part of genetic quality control procedures (Foley et al., 2017).

2.2. MRI scan parameters

Scanning was conducted on a GE SIGNA HDx 3T MRI system (General Electric Healthcare, Milwaukee, WI) with an eight-channel receive-only head coil. Whole-brain high angular resolution diffusion imaging data (Tuch et al., 2002) were acquired using a diffusion-weighted single-shot echo-planar imaging sequence (TE = 89ms; voxel dimensions = 2.4 x 2.4 x 2.4mm; FOV = 230mm x 230mm; acquisition matrix = 96 x 96; 60 slices aligned AC/PC with 2.4mm thickness and no gap). Gradients were applied along 30 isotropic directions (Jones et al., 1999) with $b = 1200 \text{ s/mm}^2$. Three non-diffusion-weighted images were acquired with $b = 0 \text{ s/mm}^2$. Acquisitions were cardiac-gated using a peripheral pulse oximeter. T1-weighted anatomical images were acquired using a three-dimensional fast spoiled gradient-echo sequence (TR/TE = 7.8/3s; voxel dimensions = 1mm isotropic; FOV

ranging from 256 x 256 x 168mm to 256 x 256 x 180mm; acquisition matrix ranging from 256 x 256 x 168 to 256 x 256 x 180; flip angle = 20°). These sequences were similar to those used by Hodgetts et al. (2019), with only subtle differences between the two studies (outlined in Supplementary Table 2).

2.3. dMRI

2.3.1. Pre-processing

The dMRI data were corrected for motion- and eddy current-induced distortions in ExploreDTI (version 4.8.6; Leemans et al., 2009), with an appropriate reorientation of the b-matrix (Leemans & Jones, 2009). Images were registered to down-sampled T1-weighted images (1.5mm isotropic resolution) to correct for susceptibility deformations (Irfanoglu et al., 2012). Data were visually checked as part of quality assurance procedures, leading to the removal of two participants from the analysis due to poor quality data. Consistent with Hodgetts et al. (2019), the two-compartment free-water elimination procedure was implemented using in-house MATLAB code (version R2015a; MathWorks, Inc., 2015) to correct for voxel-wise partial volume artefacts (Pasternak et al., 2009). This procedure has been shown to improve tract delineation, as well as the sensitivity and specificity of measures traditionally derived from the diffusion tensor (Pasternak et al., 2009). Free-water corrected FA and MD maps were then used in further analyses. FA represents the degree to which diffusion is constrained in a particular direction, ranging from 0 (isotropic diffusion) to 1 (anisotropic diffusion). By contrast, MD ($10^{-3}\text{mm}^2\text{s}^{-1}$) represents the average diffusivity rate.

2.3.2. Tractography

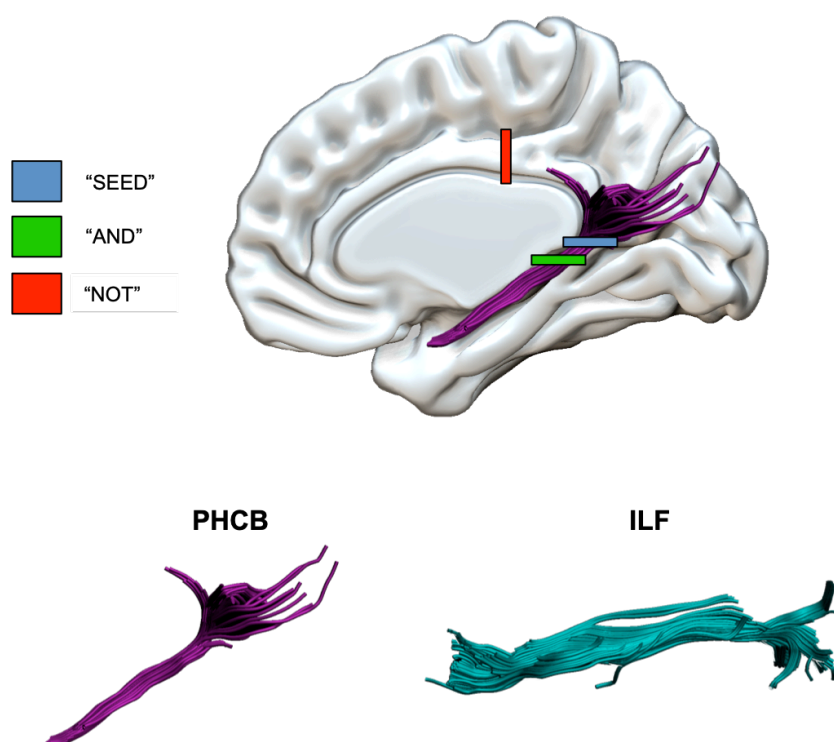
The RESDORE algorithm was used to identify outliers in the diffusion data (Parker, 2014), and then tractography was conducted in ExploreDTI using the modified damped Richardson Lucy spherical deconvolution algorithm (Dell'Acqua et al., 2010). Spherical deconvolution approaches enable multiple peaks to be extracted in the white matter fibre orientation density function (fODF) within a given voxel. This allows complex fibre arrangements, such

as crossing/kissing fibres, to be modelled more accurately (Dell'Acqua & Tournier, 2019). The current study and the original study by Hodgetts et al. (2019) both used spherical deconvolution approaches, although the latter used the constrained spherical deconvolution algorithm (Jeurissen et al., 2011). While this might lead to subtle differences between the two studies, the modified damped Richardson Lucy deconvolution algorithm was selected here because it is considered less sensitive to miscalibration (Parker et al., 2013). To minimise any further discrepancies between the studies, tracts were reconstructed using the same parameters used by Hodgetts et al. (fODF amplitude threshold = 0.1; step size = 0.5mm; angle threshold = 60°).

In-house semi-automated tractography software (Parker et al., 2012) was used to generate three-dimensional reconstructions of the PHCB and ILF in both hemispheres. The software was trained on manual reconstructions generated by author R.L. using a waypoint ROI approach in ExploreDTI, where “SEED”, “AND”, and “NOT” ROIs were used to isolate tract-specific streamlines (Figure 1). ROIs were placed in the same regions as described by Hodgetts et al. (2019). Placement was therefore guided by established protocols for the PHCB (Jones, Christiansen et al., 2013) and the ILF (Wakana et al., 2007), respectively. All reconstructions generated by the semi-automated software were visually inspected by authors R.L. and C.J.H. and, where required, manually edited post hoc to remove erroneous, anatomically implausible fibres. Participants for whom the PHCB and ILF could not be reconstructed in both hemispheres were removed from analysis ($n = 18$). Thereafter, measures of microstructure were obtained and averaged across tracts. Although the semi-automated approach used here differs to that used by Hodgetts et al., larger studies have shown this to be useful (Foley et al., 2017; Metzler-Baddeley et al., 2019). Furthermore, during visual inspection, author C.J.H. confirmed that tract reconstruction produced qualitatively similar outputs to those obtained in the original, to-be-replicated study.

Figure 1

Manual Reconstructions of the PHCB and ILF



Note. “SEED”, “AND”, and “NOT” ROIs used to manually reconstruct the PHCB are highlighted (*upper panel*). Example tract reconstructions are shown for both the PHCB and ILF (*lower panel*). The resulting tracts were used to train the semi-automated tractography software (Parker et al., 2012) and produce tracts for the entire sample. Abbreviations: ILF = inferior longitudinal fasciculus, PHCB = parahippocampal cingulum, ROI = region of interest.

2.3.3. Tract-based spatial statistics (TBSS)

Complementary voxel-wise statistical analysis of the FA and MD data was conducted using TBSS (Smith et al., 2006). Each participant’s free-water corrected FA and MD maps were first aligned in standard MNI space using nonlinear registration (Andersson et al., 2007a, 2007b). Next, the mean FA images were created and subsequently thinned (threshold = 0.2) to generate the mean FA skeleton, which represents the centre of all tracts common to the group. Each participant’s aligned FA and MD data were then projected onto the skeleton and the resulting data carried forward for voxel-wise cross-subject analysis. These analyses were performed using *randomise* (Winkler et al., 2014), a permutation-based inference tool.

For both FA and MD, a general linear model contrasting *APOE* $\epsilon 4$ carriers and non-carriers (FA: carrier > non-carrier; MD: carrier < non-carrier) was applied (n permutations = 1000). Mirroring Hodgetts et al.'s (2019) example, analyses were first restricted to the PHCB using an ROI mask [labelled "cingulum (hippo-campus)"] from the John Hopkins University ICBM-DTI-81 white-matter tractography atlas. An exploratory whole-brain analysis was then conducted. Statistically significant clusters were extracted from both analyses using threshold-free cluster enhancement with a corrected α level of 0.05 (Smith and Nichols, 2009).

2.4. Statistical analyses

Except for TBSS, all statistical analyses were conducted using R in RStudio. In addition to common frequentist null hypothesis significance tests, Bayes factors (BFs) were calculated. BFs quantify the degree to which the observed data favours predictions made by two models, in this case the null hypothesis and the alternative hypothesis. Consequently, BF analyses can provide evidence in support of the null (Dienes, 2014). In accordance with the evidence categories outlined by Lee and Wagenmakers (2013), a BF_{+0} (BF_{10} for two-sided tests) greater than 3 was considered to represent at least moderate evidence for the alternative hypothesis, whereas a BF_{+0} less than .33 was considered to represent at least moderate evidence for the null hypothesis.

2.4.1. Primary (replication) analyses

To test whether *APOE* $\epsilon 4$ carriers showed higher FA and lower MD in the PHCB but not the ILF, one-sided Welch's t -tests were conducted. As in Hodgetts et al. (2019), all tests were repeated, once with male participants removed and once with $\epsilon 2$ carriers removed. These additional tests – performed independently of each other – were originally conducted based on evidence that *APOE* $\epsilon 4$ may have a stronger effect on AD biomarkers in females than males (Riedel et al., 2016), whereas *APOE* $\epsilon 2$ may have a protective effect on AD biomarkers (Suri et al., 2013). To ensure that the probability of falsely rejecting the null – the

Type I error rate – was not inflated, a Bonferroni correction was applied to the α level ($.05 / 3 = .016$). Two BFs were also calculated: a default JZS BF and a replication BF. The default JZS BF, which uses a default prior distribution and was computed using the *BayesFactor* package (version 0.9.12-4.2; Morey & Rouder, 2018), examines whether an effect is present or absent in the data collected in the replication study regardless of the original effect. Here, one-sided (directional) default JZS BFs were calculated. The replication BF, by contrast, uses the posterior distribution of the original study as the prior distribution in the replication study, examining whether the original effect is present or absent in the data collected in the replication study. This BF was computed using previously published R code (Verhagen & Wagenmakers, 2014).

2.4.2. Secondary (extension) analyses

2.4.2.1. HMOA index

It remains to be seen whether *APOE* $\epsilon 4$ -related differences in PHCB microstructure are better captured by measures other than FA and MD, which are sensitive to various aspects of white matter microstructure without being specific to any one (Jones, Knösche, & Turner, 2013). One such measure is HMOA, which is defined as the absolute amplitude of each fODF lobe (Dell'Acqua et al., 2013). This is normalised using a reference amplitude in order to create an index bound between zero and one. A value of zero reflects the absence of a fibre, whereas a value of one reflects the highest fODF signal that can realistically be detected in biological tissue (Dell'Acqua et al., 2013).

Given the lack of a directional hypothesis relating to HMOA, two-sided Welch's *t*-tests and two-sided default JZS BFs were used to identify any differences between *APOE* $\epsilon 4$ carriers and non-carriers. In keeping with the primary (replication) analyses described above, these tests were repeated with males removed and then with $\epsilon 2$ carriers removed. These analytical steps were performed independently. A Bonferroni correction was applied to the nominal α level ($.05 / 3 = .016$).

2.4.2.2. Hemispheric asymmetry

Despite reports linking AD with a loss of leftward structural and functional asymmetry (Banks et al., 2018; Roe et al., 2021; Tyrer et al., 2020), which may be related to differences in hemispheric susceptibility to pathology (Lubben et al., 2021; Weise et al., 2018), no study to our knowledge has yet investigated whether the *APOE* ϵ 4 allele is associated with asymmetry in PHCB or ILF microstructure. Moreover, considering the proposed interaction between *APOE* ϵ 4 and sex in the context of AD risk (Riedel et al., 2016), there is also an interesting question as to whether sex moderates any potential *APOE* ϵ 4-related association with hemispheric asymmetry. We therefore examined whether HMOA – a more tract-specific measure – was lateralised to the left or right hemisphere, and whether this was impacted by *APOE* ϵ 4, sex, or their interaction.

As with the analyses described previously, the ILF was included as a comparison tract. Lateralisation indices (LIs) were calculated for HMOA in both the PHCB and ILF [$LI = (right - left) / (right + left)$]. For any given participant, a negative LI score indicates that HMOA was higher in the left hemisphere, whereas a positive LI score indicates that HMOA was higher in the right hemisphere (Zhao et al., 2016). These LI_{HMOA} scores were subsequently analysed using robust multiple linear regression, which was carried out via the *lmrob* function from the *robustbase* package (version 0.93-7; Maechler et al., 2021). The fitted models were as follows:

$$LI_{HMOA} \sim APOE \epsilon 4 \text{ carrier status} \times \text{sex} + \text{age} \quad (1)$$

LIs were entered as dependent variables. *APOE* ϵ 4 carrier status and sex were treated as categorical variables and coded using deviation coding. Age – included as a covariate of “no interest” – was centred and scaled. The interaction between *APOE* ϵ 4 carrier status and sex was included in the model. Results were deemed statistically significant if the observed *p* value was smaller than the nominal α level of 0.05.

2.5. Data and code availability

R code used to analyse and visualise data in the current study is made publicly available via the Open Science Framework (<https://osf.io/f6jp3/>). Due to the sensitive nature of the data, the original ethics do not allow for the public archiving of study data (for more information, see Koelewijn et al., 2019). Access to pseudo-anonymised data may be granted, however, after the signing and approval of suitable data-transfer agreements. Readers seeking access through this mechanism should contact Professor Krish D. Singh at the Cardiff University Brain Research Imaging Centre (singhkd@cardiff.ac.uk).

3. Results

3.1. Primary (replication) analyses

3.1.1. Effect of *APOE* $\epsilon 4$ on PHCB FA and MD

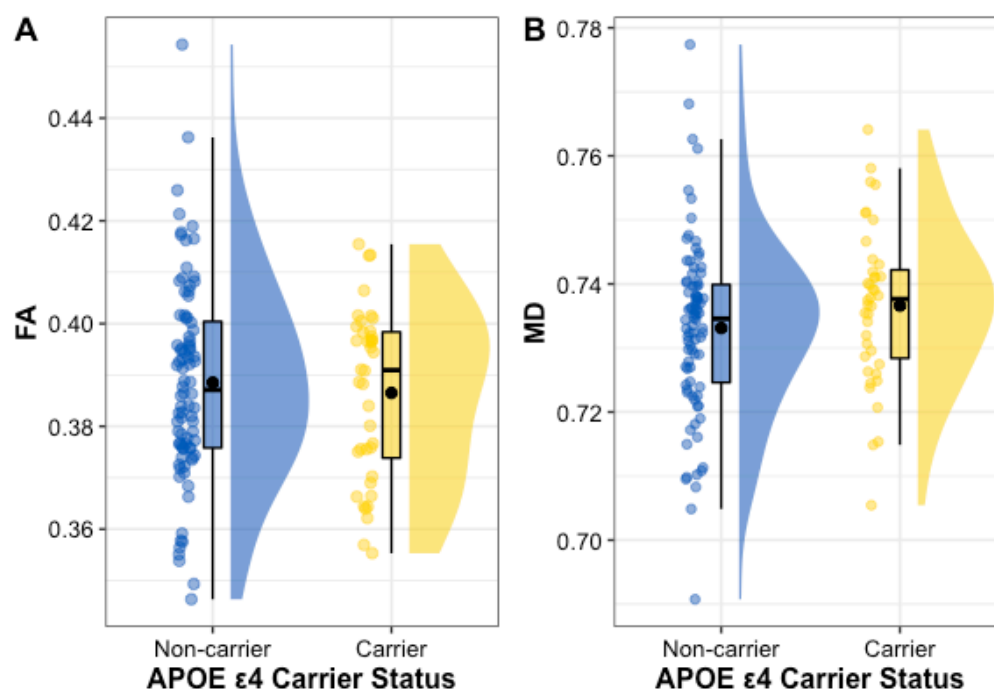
FA values for the PHCB – separated by *APOE* $\epsilon 4$ carrier status – are shown in Figure 2A. Contrary to our initial hypothesis, PHCB FA was not significantly higher for *APOE* $\epsilon 4$ carriers than non-carriers ($t(87.559) = -0.606$, $p = .727$, Cohen's $d_s = -0.112$). Supporting this, BF analysis produced moderate evidence in favour of the null (default JZS $BF_{+0} = 0.138$, replication $BF_{10} = 0.141$). Removing males from the analysis did not alter the results in any meaningful way ($t(57.685) = 0.045$, $p = .482$, Cohen's $d_s = 0.01$, default JZS $BF_{+0} = 0.246$, replication $BF_{10} = 0.168$), nor did removing $\epsilon 2$ carriers ($t(84.459) = -0.923$, $p = .821$, Cohen's $d_s = -0.183$, default JZS $BF_{+0} = 0.125$, replication $BF_{10} = 0.271$).

MD values for the PHCB – separated by *APOE* $\epsilon 4$ carrier status – are shown in Figure 2B. Again, contrary to prior expectations, PHCB MD was not significantly lower for *APOE* $\epsilon 4$ carriers than non-carriers ($t(83.625) = 1.429$, $p = .922$, Cohen's $d_s = 0.267$). Here, BF analysis revealed strong evidence in favour of the null (default JZS $BF_{+0} = 0.092$, replication $BF_{10} = 0.057$). As with FA, the results for MD did not change substantively after removing males ($t(59.729) = 1.515$, $p = .933$, Cohen's $d_s = 0.341$, default JZS $BF_{+0} = 0.106$, replication

BF₁₀ = 0.054) or after removing ε2 carriers ($t(79.581) = 1.328, p = .906$, Cohen's $d_s = 0.267$, default JZS BF₊₀ = 0.103, replication BF₁₀ = 0.1).

Figure 2

Differences in PHCB FA and MD Between APOE ε4 Carriers and Non-Carriers



Note. Differences in (A) PHCB FA and (B) MD ($10^{-3}\text{mm}^2\text{s}^{-1}$) between APOE ε4 carriers and non-carriers are shown. Individual data points, each representing a single participant, are shown alongside boxplots and density plots (“raincloud plots”; Allen et al., 2021). A small amount of jitter has been added to each data point for clarity. To facilitate interpretation, the mean value (black circle) and median value (a black line) for each group are both shown. Abbreviations: FA = fractional anisotropy, MD = mean diffusivity.

3.1.2. Effect of APOE ε4 on ILF FA and MD

The same analysis was conducted on ILF FA and MD. Analysis revealed that ILF FA was not significantly higher for APOE ε4 carriers than non-carriers ($t(86.143) = -0.864, p = .805$, Cohen's $d_s = -0.16$). BF analysis provided moderate-to-strong evidence favouring the absence of an effect (default JZS BF₊₀ = 0.12), as well as anecdotal-to-moderate evidence favouring the absence of the effect reported by Hodgetts et al. (replication BF₁₀ = 0.309).

This slight discrepancy between BFs is likely because the original to-be-replicated effect was also small and did not reach the threshold for statistical significance, meaning that the informed prior used was already more “sceptical” than the default prior. Results remained largely unchanged when males were removed ($t(49.129) = -0.069$, $p = .527$, Cohen’s $d_s = -0.016$, default JZS $BF_{+0} = 0.226$, replication $BF_{10} = 0.308$) and when $\epsilon 2$ carriers were removed ($t(79.5) = -0.893$, $p = .813$, Cohen’s $d_s = -0.179$, default JZS $BF_{+0} = 0.126$).

ILF MD was not significantly lower for *APOE* $\epsilon 4$ carriers than non-carriers ($t(81.941) = 0.54$, $p = .705$, Cohen’s $d_s = 0.101$). BFs again provided evidence in support of the null (default JZS $BF_{+0} = 0.142$, replication $BF_{10} = 0.446$). Removing males had no notable impact on the results ($t(55.856) = 0.818$, $p = .792$, Cohen’s $d_s = 0.187$, default JZS $BF_{+0} = 0.144$, replication $BF_{10} = 0.613$) nor did removing *APOE* $\epsilon 2$ carriers ($t(75.242) = 0.713$, $p = .761$, Cohen’s $d_s = 0.145$, default JZS $BF_{+0} = 0.137$).

3.1.2. TBSS

Consistent with the tractography analysis, PHCB-restricted TBSS analysis revealed no significant differences between *APOE* $\epsilon 4$ carriers and non-carriers. This was true of both FA (contrast: carriers > non-carriers) and MD (contrast: carriers < non-carriers). Adopting an uncorrected α level of $p = .005$, as has been done previously (Hodgetts et al., 2019; Postans et al., 2014), did not alter this outcome. Exploratory whole-brain TBSS analysis provided complementary evidence, with no differences evident between *APOE* $\epsilon 4$ carriers and non-carriers.

3.2. Secondary (extension) analyses

3.2.1. Effect of *APOE* $\epsilon 4$ on PHCB and ILF HMOA

Analysis revealed no significant difference between *APOE* $\epsilon 4$ carriers and non-carriers in terms of PHCB HMOA ($t(90.357) = -0.399$, $p = .691$, Cohen’s $d_s = -0.073$). BF analysis also provided moderate evidence in favour of the null (default JZS $BF_{10} = 0.215$). These results

were largely unaffected by the removal of males ($t(58.33) = 0.445$, $p = .658$, Cohen's $d_s = 0.10$, default JZS $BF_{10} = 0.258$) or the removal of $\epsilon 2$ carriers ($t(85.926) = -0.844$, $p = .401$, Cohen's $d_s = -0.167$, default JZS $BF_{10} = 0.283$).

For completeness, the same analysis was conducted for ILF HMOA. Results revealed that *APOE* $\epsilon 4$ carriers and non-carriers did not differ significantly in terms of ILF HMOA ($t(94.682) = -0.762$, $p = .448$, Cohen's $d_s = -0.139$). BF analysis provided complementary evidence, largely favouring the null (default JZS $BF_{10} = 0.251$). This remained the case when males were removed ($t(48.941) = 0.394$, $p = .696$, Cohen's $d_s = 0.092$, default JZS $BF_{10} = 0.256$) and when individuals possessing the $\epsilon 2$ allele were removed ($t(84.914) = -0.819$, $p = .415$, Cohen's $d_s = -0.162$, default JZS $BF_{10} = 0.279$).

3.2.2. Hemispheric asymmetry in PHCB and ILF HMOA

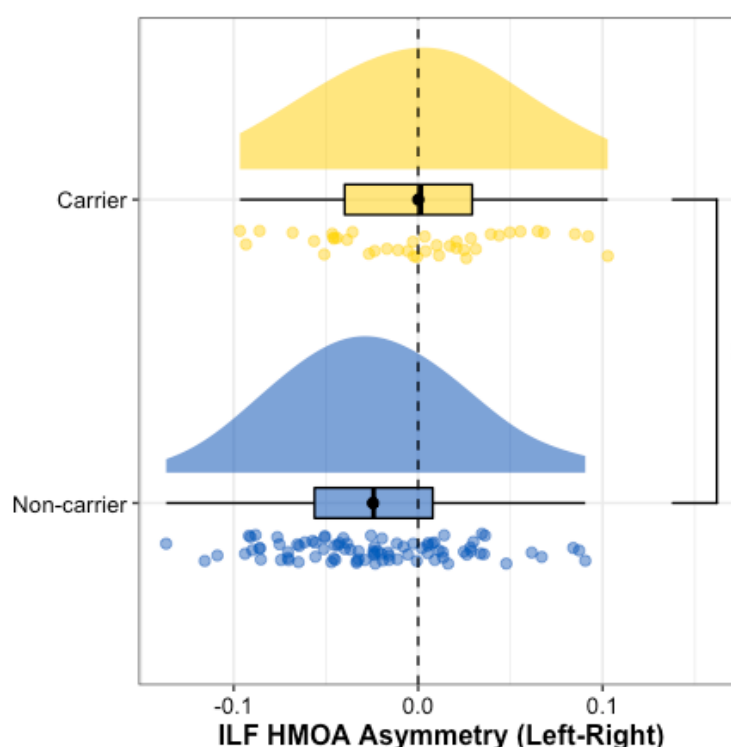
In terms of hemispheric asymmetry, analysis revealed that HMOA was higher in the right ($M = .234$, $SD = .015$) than the left ($M = .224$, $SD = .018$) PHCB ($t(127) = -6.631$, $p < .001$, Cohen's $d_z = -0.586$, default JZS $BF_{10} > 100$). Nevertheless, for PHCB LI_{HMOA} , there was no significant association with *APOE* $\epsilon 4$ ($b < -.001$, $p = .911$), sex ($b = -.002$, $p = .743$), or their interaction ($b = -.008$, $p = .558$). Consequently, we observed no evidence indicating that *APOE* $\epsilon 4$, sex, or their interactions influenced hemispheric asymmetry in PHCB microstructure.

The same analysis was conducted on ILF microstructure. HMOA was higher in the left ($M = .302$, $SD = .027$) than the right ($M = .293$, $SD = .029$) hemisphere ($t(127) = 3.778$, $p < .001$, Cohen's $d_z = 0.334$, default JZS $BF_{10} = 74.09$). Examining whether this hemispheric asymmetry was influenced by *APOE* $\epsilon 4$, sex, or their interaction, LIs were again calculated and analysed. In the case of ILF LI_{HMOA} , there was a significant association with *APOE* $\epsilon 4$ ($b = 0.027$, $p = .005$) but not with sex ($b = 0.014$, $p = .156$) or their interaction ($b = 0.007$, $p =$

.674). Figure 3 highlights the group-level differences in ILF LI_{HMOA} . As shown, this was driven by reduced leftward asymmetry in this tract among *APOE* $\epsilon 4$ carriers than non-carriers.

Figure 3

*Difference in ILF LI_{HMOA} Between *APOE* $\epsilon 4$ Carriers and Non-Carriers*



Note. Differences in ILF LI_{HMOA} between *APOE* $\epsilon 4$ carriers and non-carriers are shown. Negative values indicate that HMOA was higher in the left hemisphere, whereas positive values indicate that HMOA was higher in the right hemisphere. Zero – highlighted by a dashed line – indicates no asymmetry. Individual data points, each representing a single participant, are shown alongside boxplots and density plots (“raincloud plots”; Allen et al., 2021). A small amount of jitter has been added to each data point for clarity. To facilitate interpretation, the mean value (black circle) and median value (a black line) for each group are both shown. Abbreviations: HMOA = hindrance modulate orientation anisotropy, LI = lateralisation index.

4. Discussion

In this study, we aimed to replicate Hodgetts et al.’s (2019) findings that healthy young *APOE* $\epsilon 4$ carriers show higher FA and lower MD than non-carriers in the PHCB but not the

ILF. Such a pattern would be in line with suggestions that individuals with pre-existing “hyper-connectivity” between posteromedial cortex and the medial temporal lobe may be more vulnerable to amyloid- β accumulation (Buckner et al., 2009; Bero et al., 2012; de Haan et al., 2012) and/or tau spread (Jacobs et al., 2018; Ziontz et al., 2021). Extending this work, we also conducted analyses on HMOA, a measure that is proposed to be more sensitive to alterations in tract microstructure than FA or MD (Dell’Acqua et al., 2013). This included an investigation into hemispheric asymmetry in PHCB and ILF HMOA, as prior reports indicate that AD impacts brain asymmetry (Banks et al., 2018; Roe et al., 2021; Tyrer et al., 2020).

In contrast to the original study, we did not observe higher FA or lower MD in the PHCB of young *APOE* $\epsilon 4$ carriers compared to non-carriers. Rather, we found: no statistically significant effects in the expected direction (all $ps \geq .482$); relatively small effect sizes (Cohen’s d_s range from -0.183 to 0.341); and BFs providing evidence in favour of the null (default JZS BF_{+0} range from .092 to .246, replication BF_{10} range from .054 to .273). Crucially, these BFs represent moderate-to-strong evidence in support of the null hypothesis (Lee & Wagenmakers, 2013). As such, we not only failed to replicate the effect reported by Hodgetts et al. (2019), but also found evidence against the presence of such an effect. There are several plausible explanations for this, although they are not necessarily mutually exclusive.

First, it could be the case that Hodgetts et al.’s (2019) findings were false positives (see also Dell’Acqua et al., 2015). Hodgetts et al.’s study included just 15 participants in the *APOE* $\epsilon 4$ carrier and non-carrier groups and, as such, was likely underpowered to detect an effect of the magnitude one might expect from this common genetic variant, especially in early adulthood (Henson et al., 2020). Given that low statistical power reduces the probability that an observed effect represents a true effect (Button et al., 2013), it is possible that the effects reported by Hodgetts et al. were false positives, although it is unclear how this relates to their observation that PHCB microstructure correlated with posteromedial cortex activity

during perceptual scene discrimination (see also Shine et al., 2015). The BF analyses conducted here provide complementary support for this assertion, demonstrating that the observed data favour the null. Taken at face value, this interpretation casts doubt on the notion that increased connectivity between posteromedial cortex and the medial temporal lobe – mediated by individual differences in PHCB microstructure – represents a pre-existing *APOE* $\epsilon 4$ -related trait enhancing vulnerability to amyloid- β accumulation and/or tau spread.

Alternatively, it could be the case that Hodgetts et al. (2019) observed a true effect, but its magnitude was exaggerated. Effect size inflation is most likely to occur in studies with small sample sizes, a phenomenon referred to as the “winner’s curse” (Button et al., 2013). If true, the analysis reported in this replication attempt might itself be underpowered to detect the effect of *APOE* $\epsilon 4$ on PHCB FA and MD, thereby constituting a Type II error or false negative. Such an explanation would help to reconcile the observed findings with prior results indicating that *APOE* $\epsilon 4$ does have an impact on posteromedial connectivity early in life (Brown et al., 2011; Felsky & Voineskos, 2013; Hodgetts et al., 2019). While this cannot currently be ruled out, it should be noted that an effect size sensitivity analysis revealed that the smallest effect size detectable at 80% power in the current study was Cohen’s $d_s = 0.57$. In addition, the BF analyses conducted here indicated that the observed data provided moderate-to-strong evidence in favour of the null, as opposed to simply providing inconclusive evidence. This shows that, even with the current sample size, our findings have relatively high evidential value (Dienes, 2014).

Another potential explanation is that the *APOE* $\epsilon 4$ carriers and non-carriers included in the two studies differed in other AD-relevant factors. It is well established that while *APOE* $\epsilon 4$ carriers are at increased risk of developing AD relative to non-carriers, not all go on to develop the disease (Liu et al., 2013). In fact, only ~50% of individuals with AD possess one or more copies of the *APOE* $\epsilon 4$ allele (Karch et al., 2014), highlighting the importance of other factors – genetic and environmental – in disease risk/protection (Jagust & Mormino,

2011; Silva et al., 2019). Following this line of reasoning, it is possible that – due to sampling variation – the *APOE* $\epsilon 4$ carrier and non-carrier groups included in the two studies differed in their overall AD risk profiles, with potential implications for white matter microstructure. This would at least partly explain why we failed to replicate the effect originally reported by Hodgetts et al. (2019). Nevertheless, it is important to recognise that this remains an open question, and large-scale dMRI studies are required to test this possibility.

Regarding the asymmetry of PHCB microstructure, we found that HMOA was higher in the right hemisphere. This is consistent with some previous reports using diffusion tensor metrics (Metzler-Baddeley et al., 2012; Powell et al., 2012), although certainly not all (Lebel et al., 2012; Thiebaut de Schotten et al., 2011). Prior research has suggested that while left-hemispheric networks exhibit increased nodal efficiency in brains areas supporting language, right-hemispheric networks exhibit increased nodal efficiency in brain areas related to episodic memory (Caeyenberghs & Leemans, 2014). This potentially highlights a functional role for the observed rightward asymmetry in PHCB microstructure. However, we did not observe an effect of *APOE* $\epsilon 4$ or sex on the degree of PHCB asymmetry.

A different pattern emerged in the analysis of ILF microstructure, with HMOA characterised by leftward asymmetry. As with the PHCB, this finding is consistent with a number of studies examining asymmetry in ILF volume and diffusion tensor-derived measures of microstructure (Banfi et al., 2019; Panesar et al., 2018; Thiebaut de Schotten et al., 2011). We also observed that the degree of asymmetry in this tract was associated with *APOE* $\epsilon 4$ carrier status, such that asymmetry was lower in carriers relative to non-carriers, mirroring to some extent the loss of leftward asymmetry in AD (Banks et al., 2018; Roe et al., 2021; Tyrer et al., 2020). The ILF connects occipital and ventro-anterior temporal lobe (Herbet et al., 2018), underpinning a network involved in representing item information, including semantic and perceptual information (Murray et al., 2017; Ranganath & Ritchey, 2012). Recent research suggests that complex item discrimination is impaired in AD risk (Fidalgo et

al., 2016; Mason et al., 2017), which has in turn been linked to the structure and function of components within this network (Berron et al., 2018; Olsen et al., 2017; Reagh et al., 2016). Indeed, complex item discrimination has been proposed as a useful measure for the detection of early AD (Gaynor et al., 2019). In addition, a recent study of young adult *APOE* $\epsilon 4$ carriers in the Human Connectome Project failed to replicate enhanced intrinsic functional connectivity between posteromedial cortex and the medial temporal lobe, as observed previously (Filippini et al., 2009), but found heightened activity in left hemisphere regions connected by the ILF during face encoding (Mentink et al., 2021), possibly suggestive of a lifelong neural inefficiency (Jagust & Mormino, 2011). Future research should seek to replicate further the effect of *APOE* $\epsilon 4$ on reduced structural (and functional) left hemispheric asymmetry, especially given potential implications for later life cognition (Jiang et al., 2021; Maass et al., 2019).

5. Summary

In this study, we failed to replicate Hodgetts et al.'s (2019) finding that, relative to non-carriers, healthy young adult *APOE* $\epsilon 4$ carriers show higher FA and lower MD in the PHCB but not the ILF. Rather, the observed data strongly supported the null hypothesis of no difference. Our findings thus suggest that young adult *APOE* $\epsilon 4$ carriers do not show alterations in PHCB microstructure that might enhance vulnerability – via excessive connectivity-dependent neuronal activity – to amyloid- β accumulation and/or tau spread. Nevertheless, marked patterns of hemispheric asymmetry were evident in PHCB and ILF microstructure, although only the latter was associated with *APOE* $\epsilon 4$ carrier status. Given the potential implications for later life cognition, our study highlights an important area for future research seeking to understand how this AD risk factor impacts neural and cognitive efficiency years prior to the onset of clinical symptoms.

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640

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