# Supplementary Information

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## 1. Age prediction models

SI Figure 1: Age predictions for global white matter (WM) and grey matter (GM). The left plots show true age versus predicted age, with the red lines indicating the model fits (r = 0.72[0.71, 0.73] for the WM model and 0.57[0.55, 0.58] for the GM model). The plots on the right show the predictions after correcting for age-bias [1, 2], which was done by first fitting  $Y = \alpha \times \Omega + \beta$ , where Y is the modelled predicted age as a function of chronological age ( $\Omega$ ), and  $\alpha$  and  $\beta$  represent the slope and intercept. The derived values of  $\alpha$  and  $\beta$  was then used to correct predicted age with *Corrected Predicted Age = Predicted Age* +[ $\Omega - (\alpha \times \Omega + \beta)$ ]. This correction gives equivalent results to using age as a covariate in the linear regressions against number of previous childbirths [3].



# 2. Correlations between brain age data, previous childbirths, demographics and covariates

SI Figure 2: Correlations (Pearson's r) between global white matter (WM) brain-age gaps (BAG), number (N) of previous childbirths, demographics and covariates. IQ = intelligence quotient; BMI = body mass index; OC = oral contraceptive; HRT = hormone replacement therapy; pregnancy loss = experience with stillbirth, miscarriage or pregnancy termination.

## 3. Associations with conventional MRI summary measures

SI Table 1: Associations between number of previous childbirths and the MRI summary measures total grey matter (GM) volume, mean fractional anosotropy (FA) and mean diffusivity (MD), as well as FA and MD in Corpus Callosum (CC) - the tract showing the strongest effects in the brain age analyses. Chronological age was included in the analyses for covariate purposes and *p*-values are reported before and after correction for multiple comparisons. All variables were standardised before entered into the analysis. SE = standard error.

Measure	eta	SE	t	p	$p_{corr}$
Total GM volume	0.0204	0.010	2.009	0.045	0.064
Mean FA	0.0179	0.010	1.854	0.064	0.064
Mean MD	-0.0262	0.010	-2.675	0.007	0.022

#### 4. Feature importance ranking for brain age prediction in the corpus callosum

SI Table 2: Feature importance ranking for the Corpus callosum-based age prediction, with gain for each diffusion metric. The Gain indicates the relative contribution of the corresponding feature to the prediction model, calculated based on each feature's contribution for each tree in the model. A higher value implies that the feature was more important for generating the prediction. DTI = diffusion tensor imaging; DKI = diffusion kurtosis imaging; WMTI = white matter tract integrity; SMT = spherical mean technique; FA = fractional anisotropy; MD = mean diffusivity; AD = axial diffusivity; RD = radial diffusivity; MK = mean kurtosis, AK = axial kurtosis; RK = radial kurtosis; AWF = axonal water fraction; axEAD = extra-axonal axial diffusivity; radEAD = extra-axonal radial diffusivity; INVF = intra-neurite volume fraction; exMD = extra-neurite mean diffusivity; exRD = extra-neurite radial diffusivity.

Feature	Gain		
WMTI-radEAD	2068.467539		
SMT-exMD	1343.625053		
DTI-MD	446.814543		
DTI-RD	230.211839		
DTI-AD	226.769257		
DTI-FA	220.537127		
SMT-INVF	213.289021		
DKI-AK	210.312535		
SMT-exRD	198.772824		
DKI-MK	194.508129		
DKI-RK	194.375870		
WMTI-AWF	162.461406		
WMTI-axEAD	157.201242		

### References

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- [2] S. M. Smith, D. Vidaurre, F. Alfaro-Almagro, T. E. Nichols, K. L. Miller, Estimation of brain age delta from brain imaging, NeuroImage (2019).
- [3] A.-M. G. de Lange, J. H. Cole, Commentary: Correction procedures in brain-age prediction, NeuroImage: Clinical 26 (2020).