Appendix: Technical details, supplementary tables, and supplementary figures [posted as supplied by author]

MRI analysis

Tissue segmentation

T1-weighted images were processed using FSL tools (FMRIB Software Library, www.fmrib.ox.ac.uk/fsl) and 'fsl_anat (Beta version)' (http://fsl.fmrib.ox.ac.uk/fsl/fslwiki/fsl_anat). This reorients images to standard (MNI) space, corrects for bias field, registers the images to standard space (using linear $FLIRT^{(1, 2)}$ and nonlinear FNIRT⁽³⁾ registration), and extracts whole brain volumes ('BET').⁽⁴⁾ FMRIB's Automated Segmentation Tool (FAST) allowed extraction of measures of total grey matter (GM), white matter (WM) and cerebrospinal fluid (CSF). GM and WM volumes were adjusted for total intracranial volume.

Voxel-based morphometry (VBM)

Relationships between alcohol use and grey matter were examined initially using voxelbased morphometry, an objective method to compare grey matter density between individuals in each voxel (smallest distinguishable image volume) of the structural image. Structural data were analysed with FSL-VBM.⁽⁵⁾

http://fsl.fmrib.ox.ac.uk/fsl/fslwiki/FSLVBM), an optimised VBM protocol⁽⁶⁾ carried out with FSL tools.⁽⁷⁾ First, structural images were brain-extracted and grey matter-segmented before being registered to the MNI 152 standard space using non-linear registration.⁽³⁾ The resulting images were averaged and flipped along the x-axis to create a left-right symmetric, studyspecific grey matter template. Second, all native grey matter images were non-linearly registered to this study-specific template and "modulated" to correct for local expansion (or contraction) due to the non-linear component of the spatial transformation. The modulated grey matter images were then smoothed with an isotropic Gaussian kernel with a sigma of 3 mm. Finally, voxelwise, a Generalised Linear Model (GLM) was applied using permutationbased non-parametric testing, correcting for multiple comparisons across space (thresholdfree cluster enhancement, tfce).

Visual rating of hippocampal atrophy

Structural T1 scans were assessed independently, by three researchers for hippocampal atrophy (HA) using the Scheltens scale according to the width of the choroid fissure, width of the temporal horn, and height of the hippocampus $(0-4)$.⁽⁸⁾ Raters remained blind to all other participant data. Intra- (on a random 10% of 208 scans) and inter-rater reliability (n=208) for visual rating scores was high (intra-class correlation coefficients: 0.8 to 0.9 and 0.7 to 0.9, respectively). Left and right hippocampal atrophy was defined independently according to visual rating (Scheltens score(9) >0) by three clinicians, who reached a consensus.

Hippocampal volumes

Hippocampal volumes were calculated using FIRST (http://fsl.fmrib.ox.ac. $uk/fsl/fslwiki/FIRST)$ ⁽¹⁰⁾ an automated model-based segmentation/registration tool in a twostage process – first using all subcortical masks, and second a hippocampal mask only. Both were visually inspected to check optimal segmentation. Extracted hippocampal volumes were corrected for intracranial volume and averaged across left and right sides.

Diffusion tensor imaging

Diffusion tensor imaging (DTI) measures the directional preference of water diffusion in neural tissue and allows inferences about the structural integrity of white matter tracts. In healthy myelinated fibres diffusion is restricted perpendicular to the longitudinal axis of the fibre, i.e. it is anisotropic. Voxelwise statistical analysis of fractional anisotropy (FA), axial diffusivity (AD), radial diffusivity (RD) and mean diffusivity (MD) data was carried out using Tract-Based Spatial Statistics (TBSS).⁽¹¹⁾ This involves non-linear registration followed

by projection onto an alignment-invariant tract representation (the "mean FA skeleton"). This avoids alignment problems for multiple subjects and avoids arbitrariness of spatial smoothing extent, improving the sensitivity, objectivity and interpretability of analysis of multi-subject diffusion imaging studies.^{(12)} Multiple diffusion indices were analysed to allow a richer investigation of localised connectivity related changes. AD describes diffusion parallel to, and RD perpendicular to the to the principal fibre direction. MD is the apparent diffusion coefficient averaged over all directions. FA reflects AD in relation to RD and is widely used as a marker of tract integrity.^{(13),(14)} Diffusion images were corrected for head movement and eddy currents (eddy_correct) and brain masks generated using BET. Fractional anisotropy, mean diffusivity, axial diffusivity and radial diffusivity maps were generated using DTIFit (http://fsl.fmrib.ox.ac.uk/fsl/fdt) that fits a diffusion tensor model at each voxel. Tract-based spatial statistics (TBSS) were used in a 4-stage process. Pre-processing prepared images for registration to standard space. Mean fractional anisotropy (FA), diffusivity (MD), radial diffusivity (RD), axial diffusivity (AD) and skeletonized FA, MD, RD and AD images were created, and thresholded. Lastly each FA, MD, RD and AD image was projected onto the relevant skeleton and 'randomize' used for statistical analyses. Masks of specific white matter tracts were created using ICBM-DTI-81 white-matter labels atlas^{(15)} and used to extract mean white matter integrity indices.

Supplementary results

Table A: Comparison of the imaging sample with the Whitehall II cohort at Phase 11.

Table B: Summary of alcohol consumption by gender at each study phase (n=527) including: weekly units (grams) consumed; percent drinking over safe
weekly limits, as defined by pre-2016 (>14 units/112g women, >21 units/168g

Table C: Summary of CAGE (screen for alcohol dependence) scores across study phases.

Table D: Mean (S.D.) raw and adjusted hippocampal volumes as percentage of intracranial volume (%ICV) extracted using FIRST for the sample.

* For every 10 unit increase in alcohol weekly

¹Analyses adjusted for age, sex, premorbid IQ, education, social class, marital status, Framingham Risk Score, smoking, history of Major Depressive Disorder (SCID), exercise frequency, club attendance, social visits, current psychotropic medication.

²Hippocampal volume squared.

Table E: Outlier analysis of multiple linear regression results, with squared hippocampal volume (% of ICV) as the dependent variable and average alcohol consumption as an independent variable.

Figure A: Partial regression plots showing the relationships between a number of variables and hippocampal volume (extracted from FIRST, as %ICV, squared), controlling for other variables in the model. For example, the last plot visualises the relationship between alcohol and hippocampal size, after controlling for for: age, sex, social class, education, FSIQ, marital status, club participation, social visits, psychotropic medication, education, FSIQ, marital status, club participation, social visits, psychotropic medicatio
exercise, Framingham Risk Score and history of Major Depressive Disorder. The x-axis units are residuals from a regression model omitting alcohol. The y-axis units are residuals from a regression of alcohol against the remainder of the independent variables. The slope of the red line represents the partial regression coefficients coefficients.

¹ Females vs. reference group males. ** P value <0.01

²Classes 2,3 and 4 vs. reference group class 1. *** P value <0.001

²Classes 2,3 and 4 vs. reference group class 1.

³Estimates represent odds for memory recall (binomial regression) and mean count for lexical and semantic fluency (Poisson regression).

Table F: Results from mixed effects models fitting longitudinal cognitive test scores over time according to average alcohol consumption (reference group abstainers) across the study. For lexical and semantic fluency Poisson regression is applied, and for memory recall binomial regression. Adjusted for: age, sex, education, social class, FSIQ, Framingham Risk Score, learning effects of the test and time from study baseline.

¹Adjusted age, sex, education, FSIO

²As in ¹ and additionally corpus callosum MD

*As a multiple of 100

Table G: Results from multiple linear regression analysis, with decline in lexical fluency (slopes, Phase 3 to time of scan) as the dependent variable, and average alcohol consumption (weekly units) as an independent variable.

 1 Mean (S.D.) for normally distributed continuous variables.

2 Median (IQR) for discrete variables.

 3 Hypothesis test for significant difference between models with and without alcohol included.

* Not significant after Bonferroni correction (0.05/9 = 0.006 significance level).

Table H: Summary of cross-sectional cognitive test data and its relationship with average alcohol consumption across the study. Means and standard deviations are given for normally distributed data and medians and interquartile ranges for non-normally distributed data. MoCA=Montreal Cognitive Assessment, TMT=Trail making test, Rey=complex figure task, HVLT=Hopkins Verbal Learning Test, BNT=Boston naming test (associated with IQ, hence positive association). P values result from hypothesis tests (likelihood tests) comparing regression models with cognitive test as the dependent variable, adjusted for age, sex, education and FSIQ, with and without alcohol consumption in the model.

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