

Word Count: 3487

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Estimated regional white matter hyperintensity burden, resting state functional connectivity, and cognitive functions in older adults

Abhishek Jaywant PhD^{a,b}, Katharine Dunlop PhD^{a,c}, Lindsay W. Victoria PhD^{a,d}, Lauren Oberlin PhD^{a,d}, Charles J. Lynch PhD^{a,c}, Matteo Respingo MD^{a,d}, Amy Kuceyeski PhD^e, Matthew Scult PhD^a, Matthew Hoptman PhD^{f,g}, Conor Liston MD PhD^{a,c}, Michael W. O’Dell MD^b, George S. Alexopoulos MD^{a,d}, Roy H. Perlis MD^h, and Faith M. Gunning PhD^{a,d}

^aDepartment of Psychiatry, Weill Cornell Medicine
^bDepartment of Rehabilitation Medicine, Weill Cornell Medicine
^cFeil Family Brain and Mind Research Institute, Weill Cornell Medicine
^dWeill Cornell Institute of Geriatric Psychiatry
^eDepartment of Radiology, Weill Cornell Medicine
^fNathan Kline Institute for Psychiatric Research
^gDepartment of Psychiatry, New York University School of Medicine
^hHarvard Medical School/Massachusetts General Hospital

Corresponding author: Faith M. Gunning, Ph.D; fgd2002@med.cornell.edu
525 E 68th St, New York, NY, 10065

Keywords: Cognitive Aging; Cerebrovascular Disease; Neuroimaging

23 **Abstract**

24 Objective: White matter hyperintensities (WMH) are linked to deficits in cognitive functioning,
25 including cognitive control and memory; however, the structural and functional mechanisms are
26 largely unknown. We investigated the relationship between estimated regional disruptions to
27 white matter fiber tracts from WMH, resting state functional connectivity (RSFC), and cognitive
28 functions in older adults. Design: Cross-sectional study. Setting: Community. Participants: Fifty-
29 eight cognitively-healthy older adults. Measurements: Tasks of cognitive control and memory,
30 structural MRI, and resting state fMRI. We estimated the disruption to white matter fiber tracts
31 from WMH and its impact on gray matter regions in the cortical and subcortical frontoparietal
32 network, default mode network, and ventral attention network by overlaying each subject's
33 WMH mask on a normative tractogram dataset. We calculated RSFC between nodes in those
34 same networks. Results: The interaction of estimated regional WMH burden and RSFC in
35 cortico-striatal regions of the default mode network and frontoparietal network was associated
36 with memory retrieval. Models predicting working memory, cognitive inhibition, and set-shifting
37 were not significant. Conclusions: Findings highlight the role of circuit-level alterations at the
38 structural and functional levels in resting state networks that are related to WMH and impact
39 memory retrieval in older adults.

40

41 **MeSH Keywords:**

42 Cognitive Aging; Neuropsychology; Neuroimaging; Magnetic Resonance Imaging;

43 Cerebrovascular disease

44 **1. Objective**

45 Cerebrovascular disease is associated with mood disorders such as late-life depression¹, and
46 characterized by neurobiological abnormalities like white matter hyperintensities (WMH)².
47 WMH are linked to cognitive decline in older adults; a potential mechanism is their impact on
48 structural connectivity³. WMH affect the strength and efficiency of white matter connections,
49 which in turn are associated with slower processing speed⁴ and declines in working memory,
50 episodic memory, and cognitive flexibility^{5,6}.

51
52 Cognitive control—the ability to maintain task-relevant information in mind, flexibly shift set,
53 and inhibit irrelevant information—is susceptible to cerebrovascular disease and WMH and their
54 impact on structural connectivity. In older adults, cognitive control is linked to structural
55 disconnection in frontal-subcortical circuits arising cerebrovascular disease⁷. These processes
56 have important ramifications for older adults as they may increase risk for late-life mood
57 disorders and treatment non-response⁸ as well as loss of functional independence⁹. Memory
58 dysfunction also occurs in the context of cerebrovascular disease and is associated with WMH-
59 linked disruption in fiber tracts in frontal, temporal, and subcortical regions such as the anterior
60 thalamic tract, the uncinate fasciculus, and the forceps minor^{10,11}. Worse memory performance is
61 also associated with depression in older adults¹².

62
63 Cognitive dysfunction in the setting of cerebrovascular disease may also depend on resting state
64 functional connectivity (RSFC). In older adults with small vessel ischemic disease, worse
65 cognitive control is associated with diminished RSFC in the frontoparietal network (FPN),
66 default mode network (DMN), and in subcortical structures^{13,14}. White matter lesions are also

67 associated with altered RSFC in the DMN¹⁵ in mild cognitive impairment. These findings
68 suggest that there are complex and dynamic interrelationships between structural and functional
69 connectivity.

70
71 Despite advances in neuroimaging techniques—and much research on the interrelationships
72 among neuroimaging markers of cerebrovascular disease, structural anatomy and RSFC in small
73 vessel disease and mild cognitive impairment—how these interactions contribute to changes in
74 cognitive control and memory in healthy older adults is largely unknown. These interactions may
75 inform the early mechanisms that may underlie cognitive decline in older adults, which can also
76 shed light on circuit-level alterations that may predispose older adults to behavioral and mood
77 disturbances. Understanding the relationship between structure, function, cognition, and vascular
78 disease can also inform targeted and early approaches to promote and maintain cognitive health
79 prior to the onset of late-life psychiatric disorders.

80
81 The goal of this study was to investigate how estimated regional WMH burden and RSFC are
82 related to cognitive control and memory in healthy older adults. We used the Network
83 Modification Tool¹⁶ to estimate the magnitude of disruption to white matter tracts from WMH in
84 the FPN, DMN, and ventral attention network (VAN) and its impact on cognitive control and
85 memory retrieval. We evaluated the relationship between estimated burden from WMH in these
86 networks and RSFC in the same networks. We hypothesized that WMH-associated structural
87 disruption and functional connectivity measures would interact in explaining cognitive
88 performance.

89

90 **2. Methods**

91 *2.1. Participants.* Participants were 58 cognitively-healthy, independent, and community-
92 dwelling older adults aged 60-84 (M=72.9 years, SD=6.02) who were enrolled in a larger trial
93 investigating cognitive control and emotion regulation in late-life depression (ClinicalTrials.gov
94 registration NCT01728194). All participants were English-speaking, non-depressed, and
95 cognitively unimpaired (Mini Mental Status Examination $\geq 26/30$). The absence of mild
96 cognitive impairment or dementia was ensured through additional clinical assessment when
97 indicated during the initial screen, including consensus clinical conferences led by board-
98 certified geriatric psychiatrists. Because participants constituted a control group for a larger
99 study, any participant in this sample who reported subjective memory complaints during an
100 initial phone screen was screened out. Participants had no current or past history of major
101 psychiatric illness or neurologic disorder. All participants were recruited through flyers and
102 advertisements. All provided written informed consent, and the study was approved by the
103 Institutional Review Boards of Weill Cornell Medicine and the Nathan Kline Institute. The
104 participants constituted the control group for an analysis on structural connectivity and
105 performance on the Trail Making Test in late-life depression that has been previously published
106 by our group¹⁷. We extend our previous work by performing analyses of RSFC and relating it to
107 estimated regional WMH burden and to cognitive measures in healthy older adults.

108

109 *2.2. Neuropsychological Assessment.* Trained research assistants administered the
110 neuropsychological measures. To assess auditory attention and working memory, we
111 administered the Digit Span subtest (total score of the forward and backward trials) of the
112 Wechsler Adult Intelligence Scale-Fourth edition. The Stroop Color Word Test was used to

113 assess cognitive inhibition and susceptibility to interference from conflicting information. To
114 assess attentional set-shifting we used the Trail Making Test (TMT) and calculated a ratio that
115 isolates the shifting component while accounting for psychomotor processing speed, by dividing
116 the time to completion on TMT-B by the time to completion on TMT-A¹⁸. We administered the
117 Hopkins Verbal Learning Test-Revised and used the number of words recalled on the delayed
118 recall trial to evaluate memory retrieval.

119

120 *2.3. Neuroimaging.* Structural and functional MRI scans were acquired on a 3T Siemens TiM
121 Trio (Erlangen, Germany) scanner that was equipped with a 32-channel head coil at the Center
122 for Biomedical Imaging and Neuromodulation of the Nathan Kline Institute for Psychiatric
123 Research. Structural imaging included high-resolution whole brain images acquired using a 3D
124 T1-weighted MPRAGE and T2-weighted FLAIR images. The acquisition parameters for
125 MPRAGE were: TR=2500ms, TE=3.5ms, slice thickness=1mm, TI=1200ms, 192 axial slices,
126 matrix=256 x 256 (voxel size=1mm isotropic), FOV=256mm, IPAT=2, flip angle=8 degrees.
127 The acquisition parameters for the FLAIR sequence were TR=9000ms, TE=111ms, TI=2500ms,
128 FOV=192 x 256 mm, matrix 192 x 256, slice thickness=2.5 mm, number of slices=64, IPAT=2,
129 flip angle=120 degrees. The final FLAIR resolution (rectangular FOV/matrix) was 1x1 in plane.
130 Functional neuroimaging was a turbo dual echoplanar image (EPI) sequence performed while
131 participants were at rest. Acquisition parameters were repetition time=2500ms, echo time=30ms,
132 flip angle=80 degrees, slice thickness=3mm, 38 axial slices, matrix=72 x 72, 3-mm isotropic,
133 field of view=216mm, integrated parallel acquisition techniques factor=2. Resting-state image
134 acquisition was conducted in a single run that was 6 minutes and 15 seconds in duration,

135 TR=2000ms. Participants were instructed to stay awake with eyes closed. Wakefulness was
136 verified at the end of the scanning sequence by the technician.

137

138 [Insert Figure 1 about here]

139

140 Procedures to segment and estimate WMH burden and to estimate the impact of WMH on white
141 matter fiber tracts have been previously described in detail¹⁷. Figure 1 illustrates the flow of
142 neuroimaging analysis and statistical procedures. In brief, two raters performed a visual rating
143 using operational criteria of the Age-Related White Matter Change scale¹⁹. Inter-rater reliability
144 was strong (Intraclass Correlation Coefficient=.95). FSL²⁰ and the BIANCA program were
145 subsequently used to segment WMH lesions and create WMH masks. WMH lesions smaller than
146 3 voxels were removed. A visual check was performed on each individual WMH mask and
147 minimal manual adjustments were made. Each final WMH mask was nonlinearly registered to
148 MNI space and binarized.

149

150 Regional WMH burden was estimated using the Network Modification Tool, which has
151 previously been used in aging populations²¹. Each WMH mask was overlaid on a normative
152 sample of 420 healthy participants' tractograms from the Human Connectome Project. The
153 program estimates the disruption to white matter fiber tracts and the effects on gray matter
154 regions, incorporating the potential distal effects of white matter lesions on gray matter, by
155 removing those tracts passing through a subject's WMH mask. This procedure was conducted
156 using a combined atlas of the Yeo et al²² 7-network cortical parcellation and the Choi et al²³ 7-
157 network corresponding striatal parcellation. We were primarily interested in the FPN, DMN, and

158 VAN, the latter because it includes nodes of the salience network such as the dorsal anterior
159 cingulate that are implicated in cognitive control. For our analysis, we evaluated the
160 proportional loss (due to WMH mask overlap) in tracts cortically in the three networks, and
161 between the cortex and striatum because of the known association between cortico-striatal
162 connectivity and cognition. The proportional loss of tracts due to the WMH mask was calculated
163 for each network (cortical and cortico-striatal separately) by dividing the number of streamlines
164 in that subject's modified connectome by the mean number of streamlines in that region pair
165 from the 420 healthy control tractogram set. Values range from 0 to 1, with higher values
166 indicating greater estimated regional WMH burden.

167
168 To extract corresponding RSFC values, each subject's EPI images were motion corrected using
169 FSL's MCFLIRT, linearly registered to the anatomical T1 scan, and normalized to the MNI 152
170 template. Additional preprocessing including slice-timing correction and spatial smoothing
171 (6mm FWHM) was performed in AFNI²⁴. Further removal of motion artifacts was performed
172 using ICA-AROMA²⁵. Motion was regressed out (demeaned and first derivative), artifacts were
173 further removed using AFNI's *anaticor*, and the time-series was bandpass filtered. Time-series
174 data were extracted using a previously published parcellation of 277 functional nodes²⁶ that
175 includes the 264 nodes from the Power atlas²⁷ combined with 13 additional nodes in the caudate,
176 amygdala, hippocampus, nucleus accumbens, subgenual anterior cingulate, locus coeruleus,
177 ventral tegmental area, and the raphe nucleus. Each subject's 277x277 matrix was Fisher r-to-Z
178 transformed. We subsequently extracted the modified Power atlas ROIs overlapping with the
179 combined Yeo-Choi parcellation by calculating the Dice similarity coefficient for each network
180 (FPN, VAN, DMN cortical and striatal). The 10 functional nodes in each network that had the

181 largest Dice coefficient > 0 (i.e. had > 1 voxel overlapping the structural ROI) were retained. We
182 averaged the RSFC for the nodes within each network to extract a per-network RSFC score. Note
183 that we elected not to use the same 7-network parcellation for RSFC because current
184 recommendations suggest using functionally-based parcellations with much more fine-grained
185 nodes for fMRI data²⁸.

186
187 *2.4. Statistical Analysis.* Statistical analyses were conducted using IBM SPSS Statistics version
188 25 (IBM, Armonk, NY), Matlab, and RStudio. Analyses focused on the FPN, DMN, and VAN
189 from the combined Yeo-Choi parcellation as described above. We computed Spearman rank-
190 order correlations (because of the possibility of a non-linear relationship) to evaluate the
191 relationship between estimated regional WMH burden and RSFC in each network, cortical and
192 cortico-striatal separately. We used a conservative alpha of .01 given 36 correlations in the
193 matrix, rather than a Bonferroni correction, to best balance the risk of type I and type II error and
194 to ensure that we could detect correlations of modest strength. To further evaluate the robustness
195 of these relationships, we compared model fit using the likelihood ratio test in predicting RSFC
196 from regional WMH burden beyond the effect of age, education, and gender (see Supplemental
197 Digital Content 1).

198
199 We conducted a multivariate general linear model with $\alpha=.05$ to determine how estimated
200 regional WMH burden and functional connectivity were associated with cognitive performance.
201 Outcome variables were Digit Span, Stroop Interference, TMT B/A ratio, and HVLT-R Delayed
202 Recall. As predictors, we entered as main effects: (1) estimated regional WMH burden within the
203 cortical FPN, DMN, and VAN; (2) cortico-striatal estimated regional WMH burden within the

204 FPN, DMN, and VAN; (3) cortico-cortical RSFC within the FPN, DMN, and VAN; and (4)
205 cortico-striatal RSFC within the FPN, DMN, and VAN. We entered as interaction effects—
206 which was the primary test of our hypothesis of structure-function interactions—the estimated
207 regional WMH burden x RSFC for each network (i.e., cortico-cortical estimated regional WMH
208 burden x RSFC in the FPN, cortico-striatal estimated regional WMH burden x RSFC in the FPN,
209 cortico-cortical estimated regional WMH burden x RSFC in the DMN, cortico-striatal estimated
210 regional WMH burden x RSFC in the DMN, cortico-cortical estimated regional WMH burden x
211 RSFC in the VAN, cortico-striatal estimated regional WMH burden x RSFC in the VAN). All
212 predictors were Z-normalized. Age, education, and gender were entered as covariates.

213

214 [Insert Table 1 and Figure 2 about here]

215 **3. Results**

216 *3.1. Demographic and clinical variables.* Table 1 displays the mean, SD, and ranges for age,
217 education, and performance on the dementia screening assessments (Mini Mental Status
218 Examination) and neuropsychological outcome measures for our sample. The ratio of
219 male:female participants is also provided. Figure 2 shows the WMH distribution in the sample.
220

221 *3.2. Correlations between estimated regional WMH burden and RSFC.* Table 2 shows the
222 correlations between estimated regional WMH burden and RSFC for the FPN, VAN, and DMN,
223 both cortically and between the cortex and striatum. Six participants were excluded from the
224 correlation analyses and the multivariate general linear model (below) due to missing data,
225 resulting in a final analyzed sample of N=52. There were significant negative correlations
226 ($p < .01$) between RSFC in the cortical DMN and estimated regional WMH burden in the cortico-

227 cortical FPN and cortico-striatal FPN. There were significant negative correlations between
228 RSFC in the cortico-cortical VAN and estimated regional WMH burden in the cortico-cortical
229 VAN. Of these correlations, only the relationship between estimated regional WMH burden in
230 the cortico-striatal FPN and RSFC in the cortico-cortical DMN improved model fit beyond using
231 age, education, and gender to predict RSFC (Supplemental Digital Content 1).

232

233 [Insert Table 2 about here]

234

235 *3.3. Associations between estimated regional WMH burden, RSFC, and cognition.* Results of the
236 omnibus test of significance for the multivariate general linear model indicated a significant
237 effect on HVLT-R Delayed Recall, $F(24)=3.43$, $p=.001$, adjusted $R^2=.50$. The omnibus test was
238 not significant for Stroop Interference, $F(24)=.90$, $p=.59$, adjusted $R^2=-.04$; Digit Span,
239 $F(24)=1.37$, $p=.21$, adjusted $R^2=.13$; or TMT B/A, $F(24)=.22$, $p=1.0$, adjusted $R^2=-.47$). We did
240 not evaluate the significance of individual predictors for these non-significant outcomes.

241

242 For the outcome variable of HVLT-R Delayed Recall, our primary analysis was the interaction
243 between regional WMH burden and RSFC. There were significant interaction effects of
244 estimated regional WMH burden x RSFC in the cortico-striatal FPN ($\beta=-.89$, $SE=.31$, $t=-2.89$,
245 $p=.007$), cortico-striatal DMN ($\beta=.68$, $SE=.31$, $t=2.18$, $p=.04$), and cortico-cortical DMN
246 ($\beta=1.67$, $SE=.79$, $t=2.11$, $p=.04$). For the interaction between estimated regional WMH burden
247 and RSFC in the cortical DMN, there was a negative relationship between RSFC and memory
248 retrieval, although the slope of that relationship was weaker in individuals with low estimated
249 regional WMH burden (Figure 3a; Figure 4). For the interaction between estimated regional

250 WMH burden and RSFC in the cortico-striatal DMN, higher RSFC in the DMN was associated
251 with better memory retrieval for individuals with high estimated regional WMH burden; lower
252 RSFC was associated with better memory retrieval for those with low estimated regional WMH
253 burden; and there was minimal relationship between RSFC and memory retrieval for those with
254 moderate estimated regional WMH burden (Figure 3b; Figure 4). For those with high estimated
255 regional WMH burden in the cortico-striatal FPN, lower RSFC was associated with better
256 memory retrieval, while the opposite was true of those with low and moderate estimated regional
257 WMH burden (Figure 3c; Figure 4). Though not of primary interest in evaluating our hypotheses,
258 significant main effects included education ($\beta=.30$, $SE=.12$, $t=2.45$, $p=.02$), RSFC in the cortical
259 FPN ($\beta=-2.68$, $SE=1.06$, $t=-2.53$, $p=.02$), and RSFC in the cortico-striatal FPN ($\beta=.76$, $SE=.34$,
260 $t=2.20$, $p=.04$).

261

262 [Insert Figures 3 and 4 about here]

263

264 **4. Discussion**

265 We found that in cognitively-healthy older adults, (1) there are modest associations between
266 estimated regional WMH burden and RSFC within and across networks, with WMH burden in
267 the FPN and RSFC in the DMN having a relatively robust association; and (2) the relationship
268 between estimated regional WMH burden and resting state functional connectivity predicts
269 memory retrieval, but not cognitive inhibition, attention/working memory, or set-shifting.
270 Specifically, estimated regional WMH burden interacts with RSFC in cortico-cortical and
271 cortico-striatal regions of the FPN and DMN, to predict memory retrieval. These findings extend
272 prior work relating white matter lesions and RSFC in small vessel disease and mild cognitive

273 impairment. Our findings indicate that alterations in structural and functional circuitry from
274 WMH are both associated with select aspects of cognitive functioning in healthy older adults.
275
276 We found that structure-function relationships may differ by network in cognitively-healthy
277 older adults without a diagnosed cerebrovascular disorder. These findings suggest that greater
278 estimated regional WMH burden is associated with an age-related decline in RSFC within and
279 between networks, though the magnitude of the correlation coefficients were small-moderate.
280 We found a relatively robust negative association between cortico-striatal estimated regional
281 WMH burden in the FPN and RSFC in the cortico-cortical DMN. This finding is consistent with
282 the known dense connections that exist between networks²⁹. It is also consistent with
283 observations that RSFC decreases in functional networks, including the DMN, in normal aging³⁰,
284 small vessel ischemic disease¹³, and stroke³¹. Although RSFC in the FPN was not associated
285 with structural disruption from WMH, there is suggestion that structure-function coupling within
286 this network may weaken in older adults³².
287
288 Our examination of estimated regional WMH burden replicates prior studies demonstrating a
289 relationship between disruptions in white matter pathways and cognitive decline in older
290 adults^{4,6}, and the effects of WMH on distal white matter tracts and cognition³³, and extends them
291 to identify interactions with RSFC. We found that the relationship between estimated regional
292 WMH burden and RSFC in the FPN and DMN predicted memory retrieval. This finding is
293 consistent with the known role of the DMN in episodic memory³⁴ and the importance of the role
294 of executive control processes subserved by the FPN in facilitating effortful search and retrieval
295 from long-term memory³⁵.

296
297 The interactions observed in the DMN and FPN and memory retrieval indicates that in
298 individuals with high estimated regional WMH burden in the cortico-striatal DMN, RSFC was
299 positively associated with memory retrieval, while RSFC was negatively associated with
300 memory retrieval in those with low estimated regional WMH burden. This pattern suggests that
301 functional connectivity in the DMN may support behavioral performance in a “compensatory”
302 manner in the face of increasing WMH-related structural disruption. Increases in functional
303 connectivity may be protective against the negative effects of white matter lesions³⁶. Our result is
304 also similar to the finding that greater functional connectivity between the medial prefrontal
305 cortex and inferior parietal cortex is associated with better episodic memory in older adults with
306 greater gray matter atrophy³⁷.

307
308 Consistent with the known role of functional activation in the FPN during memory retrieval
309 processes, increasing RSFC in the cortico-striatal FPN was positively associated with better
310 memory retrieval in those with low and moderate levels of estimated regional WMH burden.
311 However, at high levels of WMH burden, increasing RSFC in the FPN was associated with
312 worse memory performance. This pattern may represent an unsuccessful or maladaptive
313 compensatory response, reflected in an overreliance on the FPN. A similar dissociation between
314 functional activity in the DMN and FPN at high levels of WMH-related structural disruption has
315 been observed recently in multiple sclerosis in which cognitive deficits were linked to decreased
316 DMN activity and increased FPN activity³⁸. That cortico-striatal connectivity in the FPN and
317 DMN was implicated may be because the WMH in our sample were frequently in periventricular
318 regions.

319

320 Contrary to our initial hypothesis, we did not find significant associations with cognitive control.

321 The lack of association with cognitive control may be related to the assessment measures

322 themselves, as the neuropsychological measures we used may not have been sensitive or

323 comprehensive enough to detect subtle age-related decline in cognitive control functions. In

324 addition, subjective cognitive decline may be a marker of subtle deficits prior to objectively

325 observable cognitive impairment, thus, it is possible that estimated regional WMH burden and

326 RSFC may have predicted subjective complaints in cognitive control functions.

327

328 Limitations include the length of the resting state scan (6 minutes and 15 seconds), which is of

329 short duration compared to current standards. Given that reliability tends to increase with longer

330 durations replication of our RSFC findings with longer scan times is warranted. Our age range

331 was relatively restricted and future studies would benefit from including a broader range of ages

332 and comparing our findings to middle-aged and old-old adults. The normative tractograms used

333 by the Network Modification Tool are drawn from relatively young adults in the Human

334 Connectome Project and the discrepancy in age with our sample may have induced additional

335 error. However, this weakness may be partially mitigated by studies demonstrating that structural

336 disconnection maps were similar in younger adults versus older adults³⁹ and a strong correlation

337 ($r=.87$) between disconnection maps that use age-matched reference tractograms and young adult

338 tractograms from the Human Connectome Project⁴⁰. Our measure of estimating regional burden

339 from WMH takes an “all or none” approach (presence or absence of a streamline) which likely

340 oversimplifies the process of white matter injury and reorganization; nonetheless, this approach

341 can be useful in clinical populations in determining inferred disconnection in a cost-efficient
342 manner when diffusion MRI is not feasible.

343

344 *4.1. Conclusion*

345 We demonstrate that in cognitively-healthy older adults, estimated regional WMH burden is
346 associated with a decline in resting state functional connectivity in the DMN, and that the
347 interaction between resting state functional connectivity and estimated regional WMH burden in
348 cortical and subcortical regions of the frontoparietal and default mode networks are both
349 associated with memory retrieval. Our findings highlight the role that alterations in the structural
350 and functional connectome both play in older adults in the setting of cerebrovascular processes.
351 Early, targeted approaches directed towards the FPN and DMN may promote and maintain
352 cognitive health in older adults, prior to the onset of mild cognitive impairment or late-life mood
353 disorders.

354 **Acknowledgements**

355 We thank Cristina Pollari for assistance with database management, Naib Chowdhury for help
356 acquiring behavioral data, Keith Jamison for assistance in implementing the Network
357 Modification Tool, and Elvisha Dhamala for help in creating the figures. We are grateful to the
358 staff of the Center for Biomedical Neuroimaging and Brain Modulation at the Nathan Kline
359 Institute for Psychiatric Research for assistance with neuroimaging data collection. This work
360 was supported by the National Institute of Mental Health (NIMH) grants R01 MH097735
361 (Gunning) and T32 MH019132 (Alexopoulos). The sponsor did not have any role in the study
362 design, the acquisition, analysis, or interpretation of the data, the writing of the report, or the
363 decision to submit the manuscript for publication.

364

365 **Author Contributions**

366 AJ: conceptualization, methodology, formal analysis, data curation, writing, and visualization.
367 KD: conceptualization, methodology, analysis validation, and writing. LWV: conceptualization,
368 methodology, data curation, writing, and visualization. LO: conceptualization, writing; CJL:
369 conceptualization, writing; MR: methodology, conceptualization, software implementation; data
370 curation, formal analysis; AK: methodology, software development, conceptualization, writing;
371 MS: writing; MJH: writing; CL: supervision; MWO: writing; GSA: writing, funding acquisition;
372 RHP: formal analysis, writing; FMG: conceptualization, methodology, writing, supervision,
373 funding acquisition.

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377 **Financial Disclosures/Conflicts of Interest**

378 GSA has participated in the advisory boards of Janssen and Otsuka, and has served on the
379 speakers bureaus of Allergan, Otsuka, Takeda, and Lundbeck. RHP receives consulting fees
380 from Genomind, RID Ventures, Outermost Therapeutics, Psy Therapeutics, Burrage Capital, and
381 Takeda. He is a paid associate editor for the JAMA Network. AJ, KD, LWV, LO, CJL, AK, MR,
382 MJH, MS, CL, MWO, and FMG report no financial disclosures.

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384

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498 with function and behaviour. *Nat Commun.* 2020;11(5094). doi:10.1038/s41467-020-
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502 **Table 1.** Demographic and clinical characteristics of the analyzed study sample (N=52).

	Mean	SD	Minimum	Maximum
Age	72.7	5.98	60	84
Education	16.8	2.27	12	20
Gender (M/F)	30/22	--	--	--
Mini Mental Status Examination	28.71	1.07	27	30
Stroop Interference	-3.54	7.96	-27.6	12.9
TMT B/A Ratio	2.36	0.91	0.40	5.41
Digit Span Total	15.0	3.34	9	25
HVLT-R Delayed Recall	8.85	2.21	0	12

Table 2. Spearman rank-order correlations between estimated regional WMH burden (proportion loss of connections) and resting state functional connectivity (RSFC; Fisher r to z transformed) in networks of interest.

Estimated Regional WMH Burden:	RSFC Cortical VAN	RSFC Cortical FPN	RSFC Cortical DMN	RSFC Cortico-Striatal VAN	RSFC Cortico-Striatal FPN	RSFC Cortico-Striatal DMN
Cortical VAN	-.36, p=.009	-.24, p=.09	-.29, p=.04	.09, p=.51	-.23, p=.10	-.18, p=.21
Cortical FPN	-.25, p=.08	-.15, p=.30	-.38, p=.005	-.11, p=.43	-.19, p=.18	-.16, p=.27
Cortical DMN	-.30, p=.03	-.15, p=.30	-.35, p=.01	-.13, p=.35	-.18, p=.20	-.09, p=.55
Cortico-Striatal VAN	-.34, p=.01	-.31, p=.03	-.33, p=.02	-.13, p=.35	-.27, p=.05	-.14, p=.33
Cortico-Striatal FPN	-.29, p=.04	-.11, p=.45	-.36, p=.008*	-.14, p=.32	-.20, p=.16	-.06, p=.66
Cortico-Striatal DMN	-.33, p=.02	-.08, p=.59	-.27, p=.06	-.04, p=.76	-.13, p=.36	-.13, p=.36

Note. VAN = Ventral Attention Network; FPN = Frontoparietal Network; DMN = default mode network. To reduce the probability of a Type I error, we interpret as significant correlations with $p < .01$. *Denotes relationship in which estimated regional WMH burden improved model fit in predicting RSFC beyond age, education, and gender (see Supplement 1).

Figure Legends

Figure 1. Flowchart of neuroimaging analysis procedures and statistical procedures for the primary analyses.

Figure 2: Heatmap depicting the distribution of white matter hyperintensities (WMH) in the N=52 participants analyzed in the final sample. The heatmap was created by concatenating all participants' binary lesion masks and thus values range from 0 to 52.

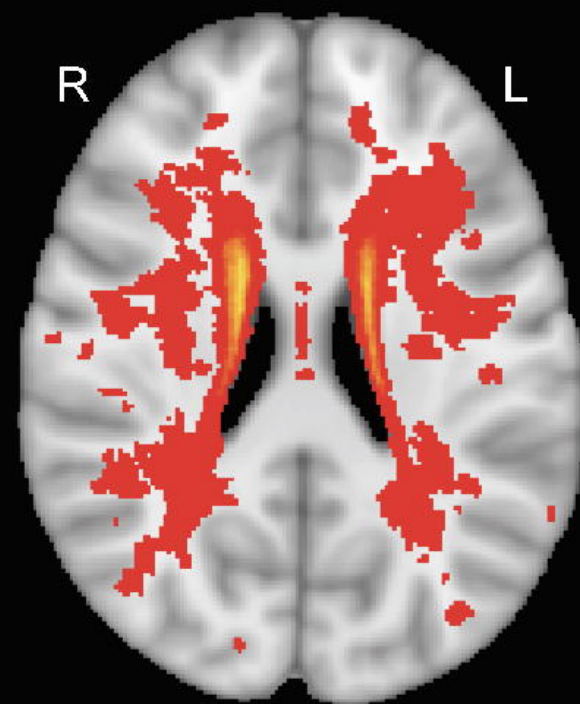
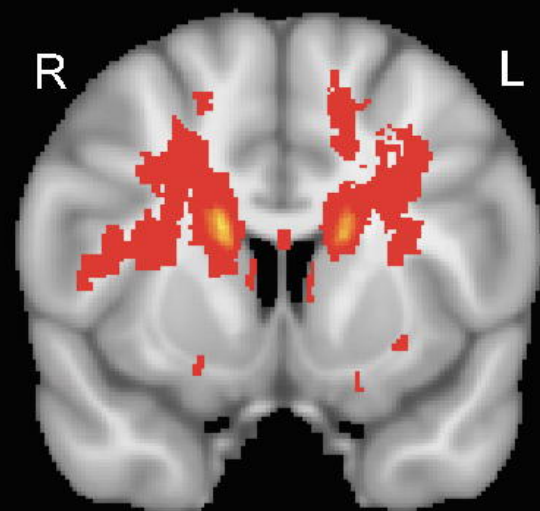
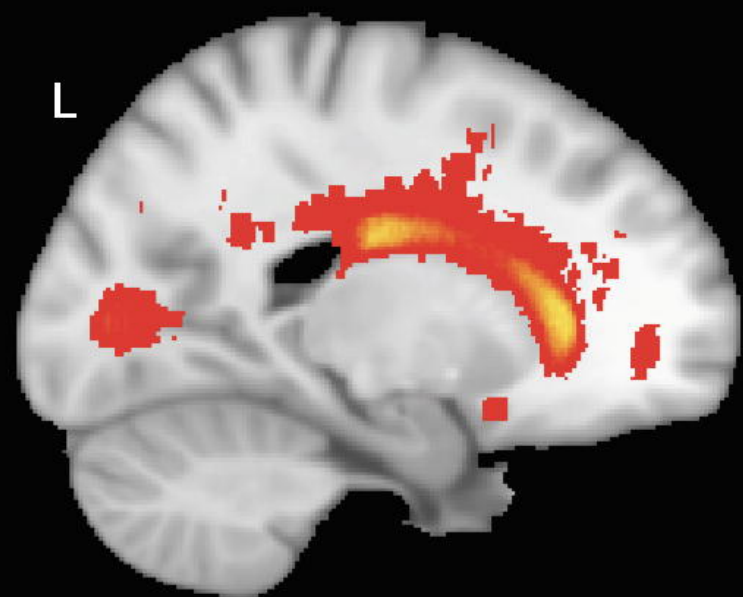
Figure 3. Scatterplots (with best fit lines) depicting the relationship between HVLTR Delayed Recall performance predicted by the general linear model, estimated regional white matter hyperintensity burden, and resting state functional connectivity (RSFC) in the (a) cortical default mode network (DMN); (b) cortico-striatal DMN; and (c) cortico-striatal frontoparietal network (FPN). For visualization, estimated regional white matter hyperintensity burden was divided into three equal groups: “low” (blue dots), “moderate” (red dots), and “high” (green dots) burden. RSFC values are normalized to z-scores as in the general linear model.

Figure 4. Glassbrains depicting the frontoparietal network (FPN; top left) and default mode network (DMN; top right) as defined by the combined Yeo-Choi atlas and used to calculate the estimated regional white matter hyperintensity burden. Glassbrains also depict functional nodes in the FPN (bottom left) and DMN (bottom right) as defined by the Power atlas and used to calculate resting state functional connectivity.

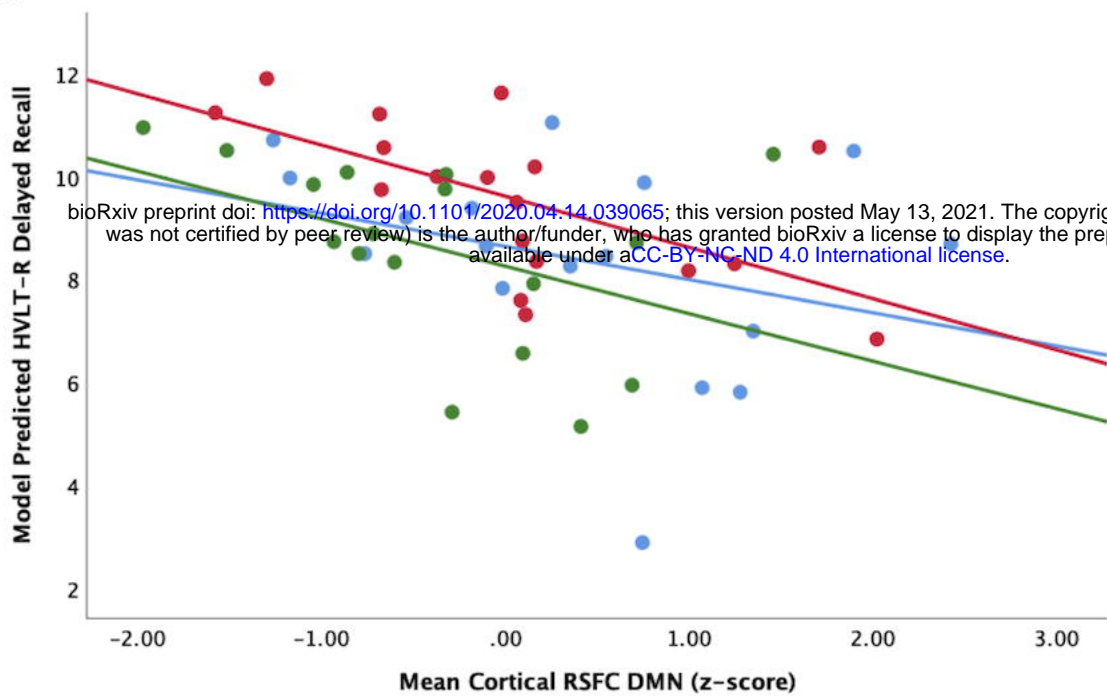
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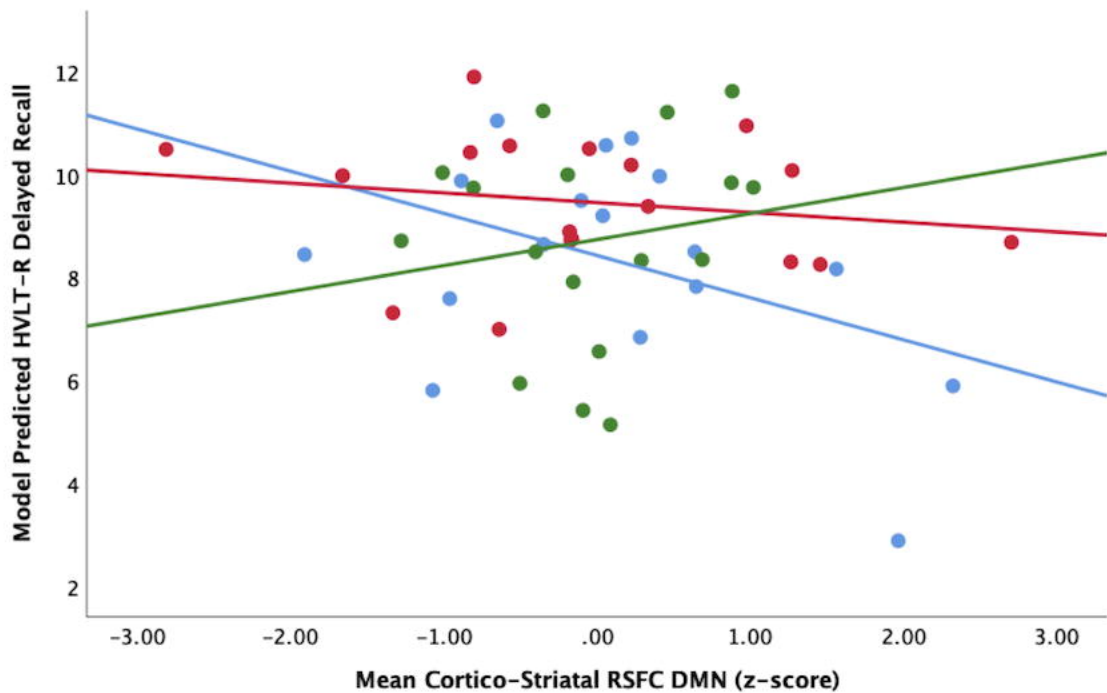




a.



b.



c.

