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## On the Impact of Interhemispheric White Matter:

## 7 Age, Executive Functioning, and Dedifferentiation in the Frontal Lobes

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## Abstract

2   **Introduction:** In middle age, declines in executive functioning (EF) are associated with  
3   decrements in the quality and/or quantity of white and grey matter. Recruitment of homologous  
4   regions has been identified as a compensatory mechanism for cognitive decline in later middle  
5   age, however research into neural substrates of EF has yet to be guided by dedifferentiation  
6   models. We hypothesized that frontal-parietal grey matter volume, interhemispheric white matter  
7   and intrahemispheric white matter fractional anisotropy (FA) will be predictive of EF. Further,  
8   we hypothesized that the comparative association between interhemispheric white matter and EF  
9   will increase with age, because of compensatory recruitment. **Methods:** Neurocognitive test data,  
10   DTI, and T1 MPRAGE scans ( $n = 444$ ) were obtained from the NKI-Rockland Sample.  
11   Structural equation modeling was used to examine the relationship between age, EF,  
12   interhemispheric white matter (forceps minor; FM), intrahemispheric white matter (superior  
13   longitudinal fasciculus; SLF), and a frontal-parietal grey matter network. EF and grey matter  
14   were modelled as latent variables, with EF examined as the criterion. Additionally, a subsample  
15   of participants aged 55-85 ( $n = 168$ ) was analyzed to examine the influence of age related  
16   compensatory mechanisms. **Results:** There was a significant relationship between FM, grey  
17   matter, and EF, which was fully mediated by age. There was a significant relationship between  
18   SLF and EF, which was not mediated by age. For older adults, only the age-mediated pathway  
19   from FM to EF was significant. **Discussion:** Using structural imaging data, support was found  
20   for age-related interhemispheric mechanisms of compensation, but not intrahemispheric  
21   mechanisms.  
22   Keywords: executive functions, dedifferentiation, compensatory recruitment, CNS white matter

1 Key points: (1) Neural substrates of executive functioning are not static across the lifespan. (2) In  
2 older adults, white matter becomes more salient as a structural correlate of executive functioning,  
3 as recruitment needs increase. (3) While the importance of interhemispheric white matter is  
4 mediated by age, intrahemispheric recruitment remains consistent across the lifespan, and is not  
5 the primary mechanism of age-based compensation in community dwelling older adults.

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## Introduction

2       Age-related decline in executive functioning (EF) starts earlier and at increased rates  
3       compared to other domains of cognitive functioning, like verbal ability (Deary et al., 2009;  
4       Strauss, Sherman, & Spreen, 2006) and memory (Grober, et al., 2008). These changes emerge  
5       around age 40 and are particularly important, because there are systematic relationships between  
6       EF, practical skills in daily living, and disability in elders (Cahn-Weiner, Boyle, & Malloy, 2002;  
7       Gansler, Suvak, Arean, & Alexopoulos, 2015; Johnson, Lui, & Yaffe, 2007; Mirelman et al.,  
8       2012; Yogeve-Seligmann, Hausdorff, & Giladi, 2008). As the proportion of US adults aged 65  
9       and older is expected to increase from 14.5% in 2014 to 21.7% by 2040 (U.S. Administration on  
10      Aging, 2016), determining behavioral and biological mechanisms and predictors for EF decline  
11      is an important tool for early detection and intervention for general cognitive and adaptive  
12      functional decline. More detailed knowledge of brain systems underlying age-related cognitive  
13      decline may inform our understanding of disability and pathological processes in aging.

14       In this study, we examined the underlying common cognitive factor of EF. EF is a broad  
15      term that refers to a group of cognitive processes needed for coordination and regulation of  
16      behavior and cognitive functions (Diamond, 2013). Models of EF generally fall into one of two  
17      categories: those based on an underlying common factor or those based on an array of specific  
18      factors. Often there are common task demands among any given neuropsychological tests of EF  
19      (i.e. goal maintenance, working memory), where that task is saturated by a common factor  
20      (Engle, Tuholski, Laughlin, & Conway, 1999; McCabe, Roediger, McDaniel, Balota, &  
21      Hambrick, 2010) with neural substrates in parietal and lateral prefrontal cortices (Collette et al.,  
22      2005). While specific aspects of dysfunction in EF processes account for differentiated but  
23      related observed behavioral effects in daily living, there are limits to predictive power with a

1 divisional approach (Burgess, Alderman, Evans, Emslie, & Wilson, 1998). While there is  
2 evidence to support the argument that specific components of EF account for a diversity of  
3 behavioral effects in adaptive functioning, general EF factors may have the most predictive  
4 power (Barbey et al., 2012) and were therefore the focus of this investigation.

5 Declines in EF in the aging brain have been linked to age-related degradations in  
6 neuroanatomical structures. White matter (WM) and grey matter (GM) tissue deterioration  
7 related to aging is primarily seen in the prefrontal, temporal, and parietal lobes (Raz & Rodriguez,  
8 2006). These areas are a critical basis of the neural representations of the EF domains of task  
9 coordination, planning, goal maintenance, working memory, and task switching (Daniels, Toth,  
10 & Jacoby, 2006). Specifically, degradation of anterior cingulate cortex (Botvinick, Cohen, &  
11 Carter, 2004; Van Veen, Cohen, Botvinick, Stenger, & Carter, 2001), the dorsolateral prefrontal  
12 cortex (Kane & Engle, 2002; MacPherson, Phillips, & Sala, 2002), and WM of the genu of the  
13 corpus callosum (Kochunov, et al., 2012) have all been predictive of age-related changes in EF.  
14 However, there is limited research on the comparative contribution of white and GM in  
15 executive networks in large samples of community dwelling older adults.

16 Some studies also suggest that the root of these age-related changes in EF is a function of  
17 either, under recruitment, or, nonselective recruitment of related brain regions (Logan, Sanders,  
18 Snyder, Morris, & Buckner, 2002). These findings suggest that because EF often involves the  
19 coordination of several brain regions, individual differences in WM may increase in influence as  
20 age increases due to the need for broader recruitment. One of the most well studied neural  
21 compensatory mechanisms observed in older adults involves increased homologous recruitment  
22 and dedifferentiation of brain regions, which seem to mitigate the effect of age-related  
23 deterioration in cortical areas and its effects on EF. In this context, dedifferentiation refers to the

1 shift from focal to more diffuse recruitment of brain networks. Previous research has established  
2 that this process emerges around age 55 (Rypma & D'Esposito, 2000), while evidence suggests  
3 EF decline begins earlier. For younger adults (age 18-25), activation is found primarily in the left  
4 hemisphere for verbal working memory tasks, and in the right hemisphere for spatial working  
5 memory. However, older adults (age 65-75) show patterns of anterior bilateral activation for both  
6 verbal and spatial working memory, recruiting homologous brain regions in the opposite  
7 hemisphere to do the same tasks (Reuter-Lorenz et al., 2000).

8 These patterns of dedifferentiation are often found in areas associated with working  
9 memory in the dorsolateral prefrontal cortex, an important component of EF (Reuter-Lorenz et  
10 al., 2000). Younger adults only showed high activation in dorsolateral areas when switching  
11 between tasks, whereas older adults showed activation in these regions during both isolated tasks  
12 and when switching (DiGirolamo et al., 2001). Expanded distributions of activity may be the  
13 result of greater demands for executive control as EF becomes less automated with age. For high  
14 performing, community dwelling older adults, there may be a hierarchy of compensatory  
15 activation, where homologous regions are recruited before heterologous regions (Cabeza et al,  
16 2002).

17 Although the associations between prefrontal cortex volume and EF do remain consistent  
18 throughout the lifespan (Yuan & Raz, 2014), the ability to increase recruitment in older adults is  
19 largely dependent on the integrity of WM pathways (Just, Cherkassky, Keller, & Minshew,  
20 2004; Nordahl, et al., 2006). For older adults, the ability to activate areas bilaterally results in  
21 better performance on letter matching tasks in visual laterality studies (Reuter-Lorenz, Stanczak,  
22 & Miller, 1999). For older adults, there are associations between individual differences in  
23 integrity of these interhemispheric pathways and the degree of observed compensatory activation

1 (Perrson et al., 2006). Because of this mechanism, interhemispheric WM (e.g., forceps major)  
2 connectivity may play a more important role in older adults compared to both GM and  
3 intrahemispheric WM (e.g., superior longitudinal fasciculus).

4 To examine the interaction between age and the neuroanatomical predictors of EF, we  
5 hypothesized that:

6 *Hypothesis 1.* Higher WM integrity (FA) in the forceps minor (FM) and superior longitudinal  
7 fasciculus (SLF) would predict better EF performance.

8 *Hypothesis 2.* Larger GM volumes in a frontal-parietal system would predict better EF  
9 performance.

10 *Hypothesis 3.* The influence of the FM would be more robust as participant age increases, while  
11 the predictive value of GM and SLF would be constant across ages.

## 12 Methods

13 **Participants.** More detailed methodology is available in Supplementary Materials. A total of  
14 444 adult participants (Table 1) with complete neurocognitive and scan data were included in  
15 these analyses. The data were selected from de-identified phenotypic and neuroimaging data for  
16 645 participants, which was made available via the enhanced Nathan Kline Institute – Rockland  
17 Sample (NKI-RS), an open-access, cross sectional, community sample (Nooner, et al., 2012).  
18 Rockland County's economic and ethnic demographics are representative of the United States  
19 census (U.S. Census Bureau, 2009), making the NKI-RS generalizable to the U.S. population.  
20 Data use agreement was accepted by NKI-RS and data handling procedures were approved by  
21 the Institutional Review Board at Suffolk University.

22 [Table 1]

1 Participants were excluded for major psychiatric or neurological conditions, or problems  
2 with MRI scans (see Supplementary Material 1. Participant Exclusion). The NKI-RS was  
3 selected for this study due to its unique properties. With minimal exclusion criteria, the NKI-RS  
4 represented the distribution of psychiatric disorders among the U.S. population. Consistent with  
5 epidemiological research showing lifetime prevalence of 46.4% (Kessler, Chiu, Demler, &  
6 Walters, 2005), 45.7% of this sample had at least one *relatively less severe* DSM-IV Diagnosis  
7 (current or past), as determined by a semi-structured clinical interview. In addition to imaging  
8 data and demographic data, the NKI-RS included a battery of neurocognitive tasks that was used  
9 to assess brain-behavior relationships.

10 **Measures.**

11 **EF.** In order to examine the unity aspect of EF, it was conceptualized as a latent variable  
12 created from five indicators. As an assessment of broad EF, participants completed the Delis-  
13 Kaplan Executive Function System Test (D-KEFS; Delis, Kaplan, & Kramer, 2001) and the  
14 Penn Computerized Neurocognitive Battery (PennCNB; Gur et al., 2001). Because age was a  
15 predictor variable of interest, raw scores were used rather than age-scaled scores. All measures  
16 were standardized and timed tasks were reverse coded, so that all positive scores reflected better  
17 performance. See Supplementary Table 1 and 2 for descriptive statistics of EF measures, and  
18 Supplementary Material for detailed descriptions of neuropsychological assessment.

19 **D-KEFS.** Measures of EF were extracted from the following subtests: Trails (TMT),  
20 Design Fluency (DF), Color Word Interference (CWIT), Tower, and Verbal Fluency (VF). For  
21 each measure the raw scores, selected from the array of subtest scores, were selected based on  
22 the literature and included in analyses (Gunning-Dixon & Raz, 2000; Kaplan, et al., 2001).

1        **PennCNB N-Back Task (N-Back).** This measure assesses working memory (Gur et al.,  
2        2001) and has been validated with functional neuroimaging meta-analysis showing activation in  
3        front-parietal networks (Owen, McMillan, Laird, & Bullmore, 2005).

4        **Imaging.** Diffusion weighted images (voxel size = 2×2×2 mm, 137 directions) and  
5        structural T1 MPRAGE (voxel size = 1x1x1 mm) were acquired using a 3.0 T Siemens Trio  
6        scanner (see Supplementary Material 3. Imaging). Unless otherwise stated, data were obtained  
7        from NKI-RS in their raw form and processing was completed as part of this investigation.

8        **GM Image Processing.** T1 MPRAGE images were automatically reconstructed in  
9        Freesurfer 5.3 (<http://surfer.nmr.mgh.harvard.edu>; see Supplementary Material 3.1 GM Image  
10      Processing). Cortical volumes from the Desikan-Killiany atlas (Desikan et al., 2006) were  
11      calculated as a ratio of estimated total intracranial volume (Buckner et al., 2004), and averaged  
12      across hemispheres. Volumes from frontal and parietal regions were chosen to reflect a unitary  
13      biological construct associated with EF (Collette et al., 2005; Neindam et al., 2012; Spreng,  
14      Sepulcre, Turner, Stevens, & Schacter, 2013), and constructed using measurement models for  
15      latent variables.

16        **WM Image Processing.** DTI was processed using TRACULA, included in Freesurfer 5.3  
17      (Yendiki et al., 2011; see Supplementary Material 3.2 WM Image Processing). Fractional  
18      anisotropy (FA) from two WM pathways relevant to the EF and the GM latent variable was  
19      extracted (Schmahmann, & Pandya, 2009): the FM and parietal branch of the SLF. The FM,  
20      which courses through the anterior corpus callosum, is the largest anterior inter-hemispheric  
21      tract. The SLF, a major intrahemispheric deep WM tract connects frontal and parietal regions  
22      indicated in many EF tasks. The SLF FA values were averaged across hemispheres.

23        **Statistical Analysis Plan.**

1        This study took a structural equation modeling approach (Bentler & Weeks, 1980) to  
2        assess both *direct* and age-mediated *indirect* relationships between brain structures (Figure 1,  
3        Figure 2, and Figure S1), and EF (the criterion). This approach was chosen to (1) examine the  
4        underlying common factor of EF and (2) assess the relative GM and WM contributions. EF and  
5        GM were assessed as latent variables given the available array of appropriate variable  
6        constituents. FM, SLF, and age were assessed as indicators.

7        In order to examine the specific effect of compensatory recruitment in older adults, the  
8        sample was split at age 55, based on the previously reported age-marker of compensatory  
9        dedifferentiation (Rypma & D'Esposito, 2000). We assessed model fit on two nested samples,  
10      older adults (age 55-85;  $n = 168$ ) and younger adults (age 20-54;  $n = 276$ ), and compared them to  
11      the full sample.

12      **Structural Equation Modeling.** Development of the latent variables and the  
13      measurement model occurred in two phases: exploratory and confirmatory (see Supplementary  
14      Material 4. Structural Equation Modeling). The EF and GM latent variables were created, and a  
15      structural model evaluated the influences of the age, GM, FM, and SLF on EF, as indicated by  
16      the standardized regression weights of corresponding pathways at or below statistical threshold  
17      ( $p \leq 0.05$ ). Specifically, the mediating effect of age between different neural substrates and EF  
18      was compared. In addition to the model fit indices described for the measurement model,  
19      changes in the Akaike information criterion ( $\Delta AIC > 2$ ) was used for model comparison. The use  
20      of  $\Delta AIC$  was appropriate as all variables were included in all models, and could therefore be  
21      considered nested.

22      Bootstrapping was completed to create 95% confidence intervals and significance values  
23      for both direct and indirect paths, with 5000 bootstrapped samples extracted (Preacher & Hayes,

1 2008). The standardized regression weights for total, direct, and indirect effects were used to  
2 compare associations at or below statistical threshold ( $p \leq 0.05$ )

3 **Results**

4 **Pre-analysis.** Correlations between all indicator variables can be found in Table 2. The final  
5 factor fit for the latent constructs of EF ( $\chi^2 = 22.65, p = 0.007$ ) and GM ( $\chi^2 = 239.53, p < 0.001$ )  
6 was good (see Supplementary 5. Pre-analysis). The final EF and GM latent variables were  
7 derived from the indicators found in Table 3.

8 [Table 2]

9 **Measurement Model.** To understand the association between the constituents, both latent  
10 variables and stand-alone indicators, a measurement model was created in which EF and GM  
11 were set as related to one another (Supplementary Figure 1). To set the metric of latent variables,  
12 the first factor loading of each latent variable was set to 1. Model fit was good ( $\chi^2 = 353.89, df$   
13 =228, Cmin/df = 1.55, CFI = 0.98, RMSEA = 0.03). All indicators were significantly associated  
14 to their latent variable ( $ps < 0.01$ ; factor loadings of above 0.40). There was no significant change  
15 in model fit after setting the bidirectional path between the two latent variables. However, the EF  
16 and GM latent variables were significantly associated ( $r = 0.11, p = 0.016$ ).

17 [Table 3]

18 **Structural Model.** Age, FM, and SLF were added into the model as indicators. Model fit  
19 remained adequate ( $\chi^2 = 878.95, df = 366, Cmin/df = 2.40, CFI = 0.93, RMSEA = 0.04$ ).  
20 Directional paths from age, GM, FM, and SLF to EF were specified to assess the influence of  
21 each variable on EF, while controlling for the level of the other variables. Additionally, the effect  
22 of age as a mediator was considered by specifying directional paths from each structure to age.  
23 Variables were entered into the model together, so the effect of each variable was examined

1 while holding the other variables constant. Four different models were compared in the full  
2 sample (Figure 1). Provided that all other model fit indices were comparable, the best model was  
3 chosen based on  $\Delta$ AIC. The fit of the best model was then examined in the split age sample to  
4 assess the direct and indirect pathways indicated by the dedifferentiation hypothesis.

5 [Figure 1]

6 **Full Sample.** For the full sample (age 20-85,  $n = 444$ ), the best model by  $\Delta$ AIC had good  
7 fit ( $\chi^2 = 643.50$ ,  $df = 345$ ,  $p < 0.001$ ,  $C_{min}/df = 1.87$ ,  $CFI = 0.96$ ,  $RMSEA = 0.03$ ). The effect of  
8 GM ( $\beta = 0.13$ ,  $p = 0.003$ ) and FM ( $\beta = 0.17$ ,  $p = 0.008$ ) on EF was mediated by age (Figure 2).  
9 The slopes from the model indicated a  $690 \text{ mm}^3$  increase in GM volume was associated with a  
10  $0.09 SD$  increase in EF performance, and a  $0.06$  increase in FM FA, was associated with a  $0.08$   
11  $SD$  increase in EF performance. For FM and GM, there was full mediation (Baron & Kenny,  
12 1986), with the indirect, age-mediated pathway accounting for 67.3% and 74.8% respectively of  
13 the brain structure-EF relationship. CIs and p-values generated from bootstrapped samples  
14 (Table 4) for the standardized regression weights of the indirect effect supported this  
15 interpretation, for both GM ( $\beta = 0.10$ ,  $p < 0.001$ ) and FM ( $\beta = 0.11$ ,  $p < 0.001$ ).

16 The direct path from SLF ( $\beta = 0.14$ ,  $p = 0.023$ ) to EF was significant, while the age-  
17 mediated indirect path accounted for only 21.2% of the brain structure-EF relationship (Figure 2)  
18 and was not significant ( $\beta = -.02$ ,  $p > 0.05$ ). The model slope indicated a  $0.05$  increase in SLF FA  
19 was associated with a  $0.07 SD$  increase in EF performance. This suggests that age played an  
20 insubstantial role in mediating the relationship between EF and SLF.

21 [Figure 2]

22 **Older Subsample.** Group statistics for indicator variables can be found in Supplementary  
23 Table 3. Model fit in the older adults ( $n = 168$ ) was comparable to that of the full sample ( $\chi^2 =$

1 159.66,  $df = 115$ ,  $p = 0.004$ ,  $C_{min}/df = 1.39$ ,  $CFI = 0.96$ ,  $RMSEA = 0.05$ ). However, the  
2 RMSEA CI was wider in the older group (range = 0.04) compared to the full sample (range =  
3 0.01), which suggested lower precision and less certainty of model fit (Kenny, Kaniskan, &  
4 McCoach, 2015). However, the CI was still sufficiently narrow to infer accurate estimation of  
5 relationships (Byrne, 2013). This was also reflected in the CIs and p-values for the standardized  
6 total, direct, and indirect relationships between EF GM and SLF (Table 4).

7 The age-mediated pathway from FM to age to EF was significant for older adults only ( $\beta$   
8 = 0.06,  $p = 0.016$ ) with the indirect path accounting for 47.4% of the relationship (Table 4).  
9 Comparatively, the age-mediated pathway for GM accounted for only 6.3% of the relationship.  
10 Although the direct pathway from SLF to EF appeared strongest ( $\beta = 0.19$ ,  $p = 0.055$ ), the  
11 corresponding CI included zero (-.01. 0.36) and it was not significant. This could be an artifact of  
12 the lowered precision of the model (and thus wider confidence intervals). The direct and indirect  
13 paths from GM and SLF to EF were not significant in the older adults.

14 **Younger Subsample.** For younger adults, the model fit indices ranged from good to  
15 acceptable ( $\chi^2 = 216.77$ ,  $df = 115$ ,  $p < 0.001$ ,  $C_{min}/df = 1.89$ ,  $CFI = 0.95$ ,  $RMSEA = 0.06$ ). The  
16 relationships between EF and neuroanatomical structures were not significant (Table 4). Possibly  
17 due to restricted variance in EF among younger adults (0.57) compared to older adults (1.01),  
18 age was no longer associated with EF (Beta = -.09,  $p = 0.253$ ) and all age-mediated pathways  
19 became non-significant.

20 [Table 4]

## 21 Discussion

22 The principal findings of our study were that: (1) while higher FA in both the FM and the  
23 SLF was associated with better EF performance, only FM was fully mediated through age (2) the

1 effect of GM on EF was similarly mediated through age, and (3) the influence of FM is more  
2 salient for older adults compared to SLF and frontal-parietal GM as evidenced by significance of  
3 that pathway only among older adults.

4 Compared to frontal GM and the intrahemispheric WM pathway (SLF), the  
5 interhemispheric WM pathway (FM) explained the most variance in EF for older adults,  
6 consistent with the dedifferentiation hypothesis of compensatory homologous recruitment  
7 (Cabeza et al, 2002). The consistency of this relationship after the age-split suggests that the FM  
8 is a particularly relevant substrate for high performing, community dwelling older adults.

9 The relationship between intrahemispheric WM and EF was not mediated by age, but  
10 instead had a direct effect on EF in the full age range. This suggests that intrahemispheric  
11 structures necessary for recruitment in the parietal region are relevant across the lifespan, not as  
12 an age-related compensatory mechanism. Although many neuropsychological tests are designed  
13 as “frontal batteries”, tasks of EF often require recruitment in multiple regions across a frontal-  
14 parietal network (Collette et al., 2005). The direct effect of intrahemispheric WM on EF suggests  
15 that the recruitment of intrahemispheric regions may stay relatively constant across the lifespan,  
16 reflecting the centrality of frontal-parietal recruitment is central to these tasks. The increased  
17 variability between intrahemispheric WM and EF in the older adults also supports the conclusion  
18 that intrahemispheric recruitment is not the primary mechanism of age-based compensation in  
19 community dwelling older adults.

20 The influence of GM on EF was mediated by age, which is not consistent with literature  
21 showing the static association between single EF measures and brain regions (Yuan & Raz,  
22 2014). Our findings may be a consequence of reduced variance in younger adults (age 20-54)  
23 compared to older adults (age 55-85). The reduced variance seems to be the result of ceiling

1 effects in the timed tasks used in the latent variable. Notably though, the relationship between  
2 GM and EF was also significantly reduced in older adults, as compared to the full sample. This  
3 may be because the network is of paramount importance, not its constituent parts. It is also  
4 possible that influence of decreasing volume in GM regions may be more salient in middle aged  
5 adults, which were not captured as an age group due to limits in sample size.

6 Future research should investigate specific thresholds for age-mediated pathways between  
7 brain structures and EF, specifically the changing effects of GM in the transition period of  
8 middle age. We would also expect that these relationships would change in the presence of  
9 neuropathology. While the NKI-RS offers unique advantages as a highly representative sample,  
10 these results may not generalize to certain clinical populations characterized by pathological and  
11 age-related structural changes. In clinical samples there may be additional recruitment needs  
12 both inter- and intrahemispherically, for which identification is critical in dementia progression  
13 prevention.

14 These results suggest that the neural substrates of EF are not static across the lifespan, but  
15 change in later life, as paralleled in early lifespan work (Horowitz-Kraus, Holland, & Freund,  
16 2016). The impact of these findings, in terms of variance accounted for, was moderately small,  
17 but does have theoretical and clinical ramifications. As detection and prevention programs for  
18 cognitive decline identify neural targets of interest, WM substrates may be more relevant for  
19 geriatric populations. Consistent with previous work in dedifferentiation, the efficiency of  
20 activation mediated through WM may be more important than any one GM region (i.e. salience  
21 of network over node).

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## Tables

2 Table 1. Demographics for full sample (n = 444)

	<i>M(SD)</i>
Age	48.45(17.33)
Education	15.67(2.24)
%	
Sex	
Male	34.9
Female	65.1
Race	
Native	
American	0.7
Asian	4.3
Black	16.4
White	75.7
Other	3
Ethnicity	
Hispanic	10.4
Non-Hispanic	89.6

3

4

1 *Table 2.* Correlations between all indicator variables.

	Age	SLF	FM	POP	PT	POR	MOFC	SF	CMF	RMF	RACC	PCUN	IP	TMT	Tower	CWIT	Tower	CWIT	N-Back
Age	---																		
SLF	-.06	---																	
FM	-.32**	.40*	---																
POP	-.22**	-.03	-.08	---															
PT	-.27**	0.01	-.03	.67**	---														
POR	-.20**	-.08	-.08	.53**	.59**	---													
MOFC	-.17**	-.10*	-.11*	.59**	.55**	.63**	---												
SF	-.29**	-.02	-.04	.69**	.68**	.66**	.70**	---											
CMF	-.26**	-.06	-.04	.57**	.52**	.52**	.55**	.67**	---										
RMF	-.26**	-.06	-.09	.57**	.62**	.68**	.69**	.79**	.65**	---									
RACC	-.14**	-.09	-.09	.48**	.51**	.53**	.56**	.62**	.54**	.63**	---								
PCUN	-.23**	-.01	-.07	.62**	.64**	.65**	.70**	.79**	.63**	.74**	.60**	---							
IP	-.26**	-.04	-.08	.62**	.62**	.62**	.64**	.73**	.56**	.70**	.59**	.74**	---						
TMT	-.27**	.15**	.16**	0.09	0.06	0.02	0.08	.11*	0.09	.10*	0.08	.10*	0.08	0.06	0.08	---			
Tower	-.12*	0.01	0.05	0.07	0.05	0.01	1.0*	0.08	0.05	.10*	0.05	.10*	0.05	0.04	.39**	---			
CWIT	-.22**	.15**	.15**	.01	0.03	-.02	0.04	0.08	0.00	0.05	0.01	0.01	0.01	.53**	.28**	---			
N-Back	-.07	.11*	0.03	0.00	-.03	0.02	0.01	0.04	0.01	0.04	-.01	0.00	-.01	.40**	.31**	.29**	---		
DF	-.17**	.10*	.12*	0.07	0.02	-.02	0.03	0.07	0.05	0.03	0.07	0.05	0.03	0.00	0.04	.40**	.22**	.34**	.19**

2

3 *Note.* \* $p < .05$  \*\* $p < .01$  \*\*\* $p < .001$ . Pearson's  $r$ . Abbreviations are as follows: rostral anterior  
4 cingulate (RACC), superior frontal (SF), rostral middle frontal (RMF), caudal middle frontal  
5 (CMF), pars opercularis (POP), pars triangularis (PT), pars orbitalis (POR), precuneus (PCUN)  
6 and inferior parietal (IP) regional volumes.

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2

*Table 3.* Confirmatory factor loadings of indicators on latent variables.

	EF	GM
	Factor Loading (95% CI)	
DF	.57(.48-.64)	---
N-Back	.49(.34-.63)	---
CWIT	.65(.54-.74)	---
TMT	.80(.71-.87)	---
Tower	.50(.39-.59)	---
IP	---	.82(.78-.85)
PCUN	---	.87(.84-.89)
RACC	---	.70(.65-.75)
RMF	---	.86(.83-.89)
CMF	---	.73(.68-.78)
SF	---	.91(.89-.93)
POR	---	.74(.66-.79)
PT	---	.76(.71-.79)
POP	---	.74(.69-.78)

3  
4

1 *Table 4.* Standardized total, direct, and indirect effects ( $\beta$ ) of brain structures on executive  
2 functioning.

3  
4

	Full (Age 20-85)			Younger (Age 20-54)			Older (Age 55-85)		
	GM	FM	SLF	GM	FM	SLF	GM	FM	SLF
Total	.13**	.17**	.12	.05	.09	.14	.11	.14	.15
Direct	.03	.05	.14*	.02	.06	.14	.10	.07	.19
Indirect	.10***	.11***	-.02	.02	.03	.00	.01	.06*	-.04

5 *Note.* \* $p < .05$  \*\* $p < .01$  \*\*\* $p < .001$

6

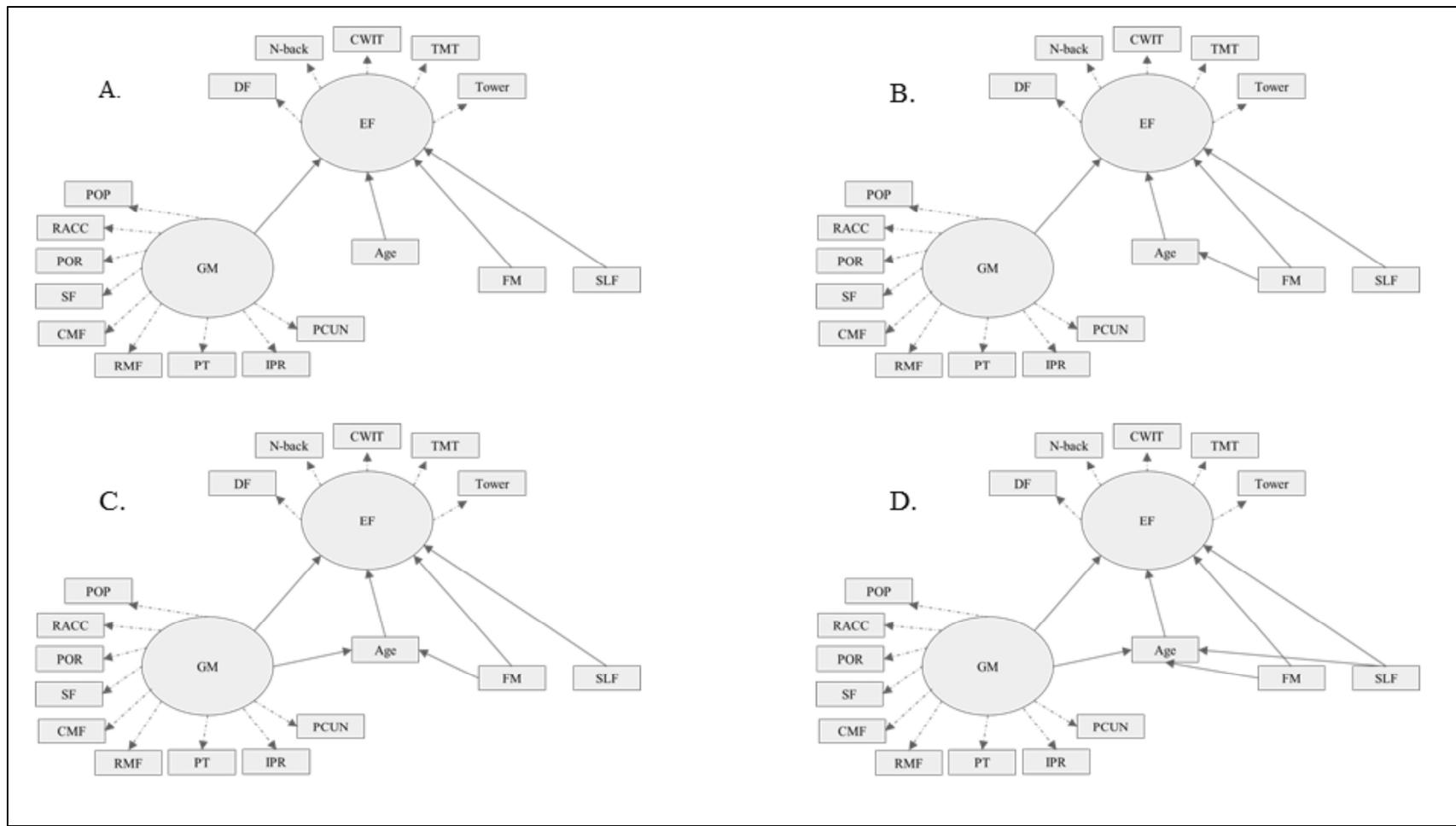
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*Figure 1.* Four structural models that were compared. Model D represents the model with the best fit. Dashed lines indicate factor loadings to latent variables. Abbreviations are as follows: rostral anterior cingulate (RACC), superior frontal (SF), rostral middle frontal (RMF), caudal middle frontal (CMF), pars opercularis (POP), pars triangularis (PT), pars orbitalis (POR), precuneus (PCUN) and inferior parietal (IP) regional volumes. Model fit indices for the four models were as follows: (A)  $\chi^2 = 797.14$ , df = 354,  $p < .001$ , Cmin/df = 2.26, CFI = 0.94, RMSEA = 0.04, AIC = 1109.14; (B)  $\chi^2 = 788.67$ , df = 351,  $p < 0.001$ , Cmin/df = 2.25, CFI = 0.94, RMSEA = 0.04, AIC = 1106.87; (C)  $\chi^2 = 732.55$ , df = 348,  $p < .001$ , Cmin/df = 2.10 CFI = 0.94, RMSEA = 0.04, AIC = 1056.25; (D)  $\chi^2 = 643.50$ , df = 345,  $p < 0.001$ , Cmin/df = 1.87, CFI = 0.96, RMSEA = 0.03, AIC = 973.50

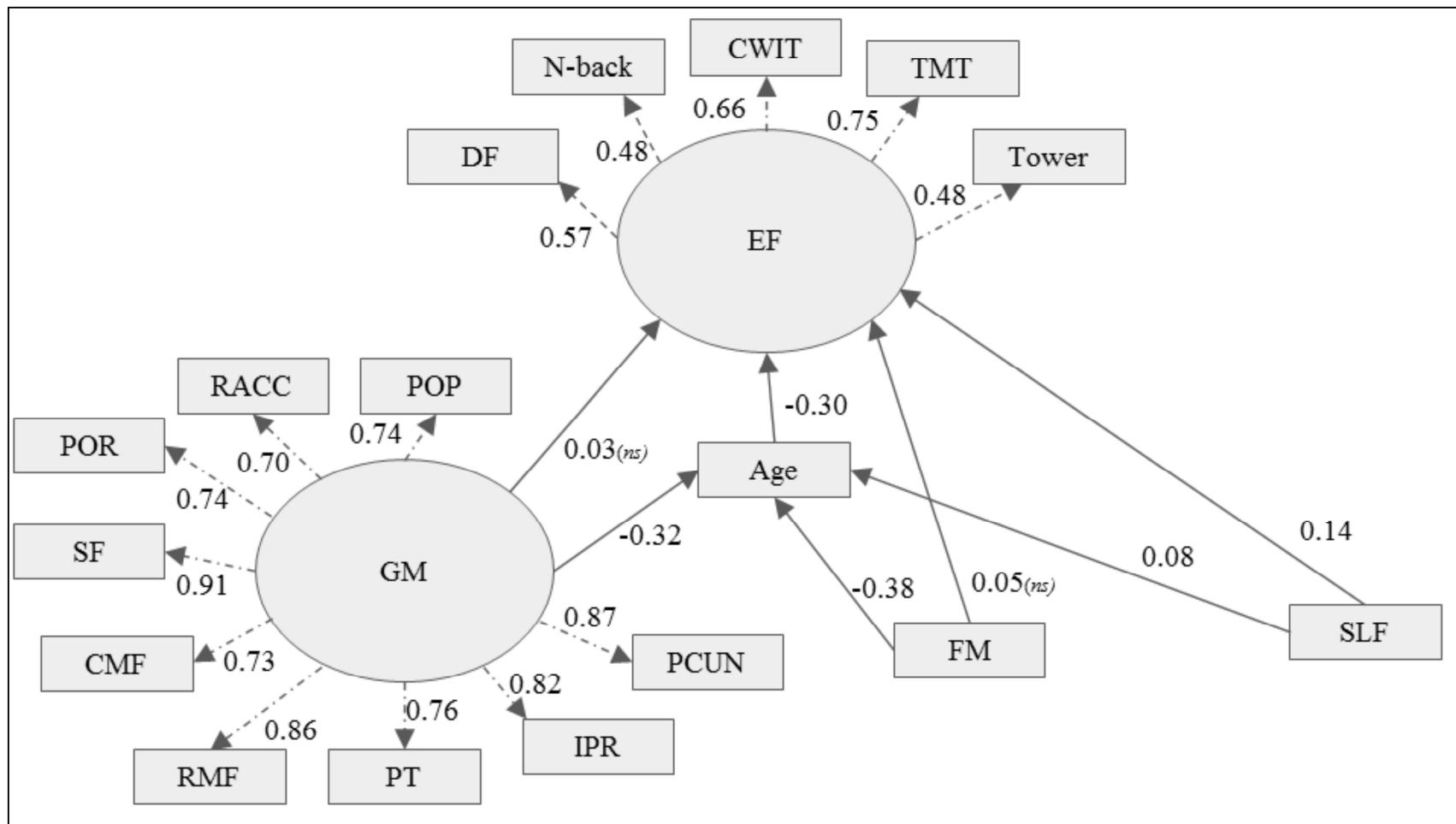


Figure 2. Final model for the full sample ( $n = 444$ ). Dashed lines indicate factor loadings to latent variables. Abbreviations are as follows: rostral anterior cingulate (RACC), superior frontal (SF), rostral middle frontal (RMF), caudal middle frontal (CMF), pars opercularis (POP), pars triangularis (PT), pars orbitalis (POR), precuneus (PCUN) and inferior parietal (IP) regional volumes. Unless otherwise indicated as not significant (ns), all pathways are significant.