Age-related changes in brain activation during a delayed item recognition task

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Abstract

To test competing models of age-related changes in brain functioning (capacity limitation, neural efficiency, compensatory reorganization, and dedifferentiation), young (n = 40; mean age = 25.1 years) and elderly (n = 18; mean age = 74.4 years) subjects performed a delayed item recognition task for visually presented letters with three set sizes (1, 3, or 6 letters) while being scanned with BOLD fMRI. Spatial patterns of brain activity corresponding to either the slope or y-intercept of fMRI signal with respect to set size during memory set encoding, retention delay, or probe stimulus presentation trial phases were compared between elder and young populations. Age effects on fMRI slope during encoding and on fMRI y-intercept during retention delay were consistent with neural inefficiency; age effects on fMRI slope during retention delay were consistent with dedifferentiation. None of the other fMRI signal components showed any detectable age effects. These results suggest that, even within the same task, the nature of brain activation changes with aging can vary based on cognitive process engaged.

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1. Introduction

Four extant hypotheses concerning changes in brain function with aging are compensatory reorganization, dedifferentiation, computational capacity limitation, and neural inefficiency. The purpose of the current paper is to test the ability of these hypotheses to predict age-related changes in brain function associated with various aspects of cognitive processing, including verbal working memory (WM) maintenance, engaged during performance of a delayed item recognition (DIR) task for letters.

1.1. Brain areas implicated in working memory maintenance in the young

WM is a psychological construct used to describe the maintenance and manipulation of information on a time scale of seconds [7]. WM seems to be divided into verbal, spatial, and object sub-systems [8,37,82,94]. Verbal WM is thought to be critical for language comprehension and reasoning [5]. Based on neuropsychological dissociations [95,98] and word length, phonemic similarity, irrelevant speech, and articulatory suppression effects [6,11,22,51], the maintenance of information in verbal WM has been modeled as an articulatory loop in which sub-vocal rehearsal refreshes a phonological store. Experimental variation of the amount of information to be stored in verbal WM (WM load) has yielded findings of increases in fMRI signal in premotor, parietal, inferior frontal, and middle frontal areas [55,72,73,76,96]. At least some aspects of articulatory loop neural processing vary in intensity with WM load [45,96,106].

1.2. Age-related changes in the brain mechanisms of verbal WM

Even in the absence of Alzheimer’s disease (AD) and other recognized brain diseases, aging is associated with
impairment in several different memory variables [78], including WM [9,17,23,49]. In particular, load-dependent deficits in WM processing have been observed with normal aging [2,23,26,62,64]. Age-related deficits in cognition are assumed to stem from age-related brain pathology [88]. Normal aging is associated with a decrease in neuropil and neuronal number in cortex [12,21,25] and in the subiculum region of the hippocampus [84,99], an increase in the number of infarcts in cortex, basal ganglia, and white matter [65], an increase in MRI white matter lesions [80], an increase in density of neurofibrillary tangles in the CA1 region of the hippocampus [81], and a global decrease in gray matter volume [31].

There is the broad question of whether the functional neural circuitry of the brain remains static in the face of this neuropathology. Though not exhaustive, four extant hypotheses concerning changes in brain function with normal aging are compensatory reorganization, dedifferentiation, computational capacity limitation, and neural inefficiency. The purpose of the current study is to test the ability of these hypotheses to predict age-related changes in brain function associated with load-dependent and load-independent aspects of encoding, storage/rehearsal, and recognition/response components of a DIR task for letters [90], which is thought to tap verbal WM maintenance. These four hypotheses will now be briefly described, in turn.

1.3. Models which predict changes in patterns of brain activation with aging

Some have put forward a hypothesis that the brain is constructed such that it can in some sense compensate for neuropathology (such as that associated with normal aging) via macro-reorganization of neural circuits [4,10,14,33,87,100]. The teleological argument is that the effect of this reorganization would be to reduce or potentially even eliminate any behavioral consequences of the neuropathology that would otherwise occur. Compensatory reorganization, occurring to varying degrees across individuals, could potentially explain how age-associated neuropathology exists even in certain proportions of the non-demented elderly [34,81,83,85], and why variability in cognitive functioning increases with age [18]. Consistent with a special version of the compensatory reorganization hypothesis referred to as HAROLD (hemispheric asymmetry reduction in older adults [15]), a more bilateral PFC fMRI activation pattern in older adults than younger adults has been reported in word encoding [60,71], source memory [14], retrieval [53], working memory [15,69], and visual attention task contexts [15].

The types of compensatory reorganization models that we consider here (subsuming, but not limited to, the HAROLD model) posit that higher performing elders are higher performing because of a change in brain reorganization relative to both young subjects and lower performing elders. Therefore, under this type of compensatory reorganization there would be a cross-sectional correlation within elders between the degree of brain reorganization and performance, such that the brain activation patterns of higher performing elders would be more dissimilar than those of lower performing elders to young subject activation patterns [14]. We refer to all such models as cross-sectional compensatory reorganization models, to distinguish them from other types of compensatory models which do not require such cross-sectional correlations [89]. The current work can only weakly test the latter type of models, so we focus on testing cross-sectional compensatory reorganization models.

Dedifferentiation is another hypothesis that predicts non-identical brain activity patterns between young and elder populations. But, unlike compensatory reorganization, this change is not beneficial for the behavior in question, and is thought to represent a general deterioration in the integrity of brain circuitry [15]. Dedifferentiation and compensatory reorganization can be distinguished as the two make opposite predictions concerning the cross-sectional relationship of age-related differences in activation patterns and performance.

A critical notion concerning both of these theories is that the spatial pattern of neuronal activity in a brain that has been reorganized or de-differentiated is not identical to within a scaling factor to the corresponding canonical pattern of brain activation (in our case, that of the healthy, young population). From here on, the phrase “identical patterns” implicitly means identical to within a scaling factor. In Section 1.5, we discuss the method used to test whether elder and young activation patterns are identical.

1.4. Models which predict no change in patterns of brain activation with aging

Another general hypothesis regarding the effect of neuropathology on brain function is a reduction in the capacity of information representation or throughput in a brain circuit. This might perhaps be caused by a limitation on the amount or quality of information entering a brain circuit due to impairment in sensory systems [35,47,50]. A simple reduction in computational capacity would predict, in the context of identical task stimuli and instructions, a decrease in both performance and neurophysiologic activity (i.e., less total ionic flux across neuronal membranes due to synaptic transmission, therefore less ATP utilization through ionic pumps, and presumably less cerebral blood flow), and so would arguably be associated with identical brain activity patterns in young and elders. Reductions in activation with aging have been reported in anterior frontal cortex [32,57,79], dorsolateral PFC [41,43,58,74,75], hippocampus [57,59], anterior cingulate [58], temporal [40], parietal [40,58] and occipital cortices [15,33,40,52,57]. Some of these reduced activations have been associated with age-related impairments in certain cognitive functions, such as resolution of competing response impetuses [43], memory scanning speed [74], and feature binding [59].
Often, age-related decreases in activation in certain areas have been found concomitantly with age-related increases in other areas [15,33,57,58,75], which is inconsistent with a simple capacity limitation hypothesis. Increases in brain activation in a behaviorally impaired group have been sometimes conceived as a reduction in neural efficiency [16,76]. Here, we define neural efficiency as the amount of performance-relevant computational work (operationalized here as measures of behavioral performance) performed per unit of synaptic activity (operationalized here as BOLD fMRI signal change). We also define neural inefficiency (which might be a more stable measure than neural efficiency) as simply the reciprocal of neural efficiency. We consider neural (in)efficiency as being not the property of individual regions, but as a property of a brain system/circuit. For example, if elders show lowered performance on average compared to young subjects and engage the same brain system during task performance but to a greater degree, then we would say that the elders’ brain system is less efficient than that of young subjects. Neural efficiency, like capacity limitation, would be associated with identical patterns of brain activation in young and elders.

1.5. Comparing patterns of brain activation

We have established a dichotomy between models of age-related brain activation change with respect to changes in brain activation patterns. On the one hand, compensatory reorganization and de-differentiation both predict (on average) different activation patterns in young and elders. On the other, neural inefficiency and capacity limitation both predict identical activation patterns in young and elders. Obviously, then, discrimination between these two pairs of hypotheses requires some sort of test as to whether young and elder brain activation patterns are identical.

Certain approaches that have been used previously to compare patterns of brain activation have caveats. Direct comparison of voxel-wise signal intensities between groups via statistical parametric mapping (SPM; [15]) is an ambiguous test of identical spatial patterns, as even a pure scaling could lead to the existence of true voxel-wise intensity differences (Fig. 1). Also, visual comparison of group [14,32] or condition-specific [73] thresholded statistical maps suffers from the same problem [77] (Fig. 1). An additional difficulty of this latter method is that substantial variability between realizations of thresholded maps would be expected due to intermediate levels of statistical power [20,70]. Finally, region of interest laterality indices [13], though providing a valid test of identical patterns of region of interest effects in noiseless data, can be exceedingly unstable in practice due to their involving ratios of estimated activation; moreover, they do not compare entire brain activation patterns.

To test for non-identity of young and elder fMRI activation patterns, we used sequential latent root testing in the context of a canonical variates analysis (CVA) for imaging data with spatially correlated errors [103]. Unlike partial least-squares approximations that are valid for correlated observations, as conceived as a reduction in neural efficiency [16,76], this CVA method is invariant to linear transformations of the predictor variables and provides parametric distributional approximations that are valid for correlated observations, as is the case in our repeated measures design [103].

2. Methods

2.1. Study population

Forty healthy, young subjects (31 M and 9 F; mean (S.D.) age = 25.1 (3.9); mean years of education = 15.7 (1.4); all right handed), recruited through flyers posted at the Columbia University campus and advertisements placed in local newspapers, and 18 healthy, elderly subjects recruited from senior centers in the New York City area (7 M and 11 F; mean (S.D.) age = 74.4 (6.9); mean (S.D.) years of education = 15.3 (2.4); all right handed) participated. Global cognitive functioning was assessed with the mMMSTotal [86], and all subjects were classified as non-demented and without serious cognitive impairment (elder mean mMMSTotal = 53.3 (2.5); young mean mMMSTotal = 55.1 (1.5)). IQ was estimated with the North American Reading Test (NART), to test if there was a systematic difference between groups in IQ [88]; the elder NART scores were significantly lower than the young (elder mean NARTIQ = 116.8 (6.4); young mean NARTIQ = 120.3 (6.1); t(56) = 2.02, two-tailed p = 0.049). All subjects supplied informed consent. Volunteers were screened for psychiatric and neurologic illness via a questionnaire. The current
2.2. Behavioral task

The behavioral task used during fMRI scanning was a DIR task for letters [90]. The task is schematized in Fig. 2. Task and training details are provided in [106]. The critical factor was set size, which is the number of letters (either 1, 3, or 6) to be remembered on each trial. Set size was varied pseudo-randomly across trials. Each of three experimental blocks contained 10 trials at each of the three set sizes, with five true negative and five true positive probes per set size, yielding a total of 30 trials per set size per subject.

2.3. fMRI data acquisition

During the performance of each block of the DIR task, 207 BOLD images [48,63], were acquired with an Intera 1.5 T Phillips MR scanner equipped with a standard quadrature head coil, using a gradient echo echo-planar (GE-EPI) pulse sequence (TE/TR = 50 ms/3000 ms; flip angle = 90°; 64 × 64 matrix, in-plane voxel size = 3.124 mm × 3.124 mm; slice thickness = 8 mm (no gap); 17 trans-axial slices per volume). Four additional GE-EPI excitations were performed before the task began, at the beginning of each run, to allow transverse magnetization immediately after radio-frequency excitation to approach its steady-state value; the images corresponding to these excitations were discarded. A T2-weighted, fast spin echo image was also acquired from each subject for spatial normalization purposes (TE/TR = 100 ms/2000 ms; flip angle = 90°, 256 × 256 matrix; in-plane voxel size = 0.781 mm × 0.781 mm; slice thickness = 8 mm (no gap); 17 trans-axial slices per volume).

Task stimuli were back-projected onto a screen located at the foot of the MRI bed using an LCD projector. Subjects viewed the screen via a mirror system located in the head coil. Responses were made on a LUMItouch response system (Photon Control Company). Task onset was electronically synchronized with the MRI acquisition computer. Task administration and collection of RT and accuracy data were controlled using PsyScope [19].

2.4. fMRI data pre-processing

All image pre-processing was implemented using the SPM99 program (Wellcome Department of Cognitive Neurology). The following steps were taken in turn for each subject’s GE-EPI dataset: data were temporally interpolated and shifted to correct for the order of slice acquisition, using the first slice acquired in the TR as the reference. All GE-EPI images were realigned to the first volume of the first session. The T2-weighted (structural) image was co-registered to the first EPI volume using the mutual information co-registration algorithm implemented in SPM99. This co-registered high-resolution image was then used to determine parameters (7 × 8 × 7 non-linear basis functions) for transformation into a Talairach standard space [92] defined by the Montreal Neurologic Institute (MNI) template brain supplied with SPM99. This transformation was then applied to the GE-EPI data, which were re-sliced using sinc-interpolation to 2 mm × 2 mm × 2 mm.
2.5. fMRI time-series modeling

All statistical analysis was implemented using the SPM99 program and other code written in MATLAB 5.3 (Mathworks, Natick MA). The fMRI data analysis comprised two levels of voxel-wise GLMs [29]. In the (subject-separable) first-level GLM, the GE-EPI time-series were modeled with regressors representing the expected BOLD fMRI response (implicitly, relative to the blank intervals) to the three DIR trial components of memory set presentation, retention delay, and probe presentation/response, separately for each crossing of the set size and true positive/true negative factors (trials on which there was no motor response from the subject during the probe period were modeled separately, and were not included for analysis at the second-level GLM). The regressors were constructed by convolutions of an indicator sequence (i.e., a train of discrete-time delta functions) representing DIR trial component onsets, an assumed BOLD impulse response function (as represented by default in SPM99), and a rectangular function of duration dictated by the duration of the assumed neural response [105]. Based in part on prior knowledge as well as regression diagnostics, two rectangular functions (and hence, two regressors) were used for the trial components of memory set presentation and probe presentation: one modeling a relatively brief (400 ms) neural response at the beginning of that trial component, and another modeling a neural response lasting throughout that entire component (3000 ms); a single rectangular function of 7000 ms duration was used for the retention delay.

For both the memory set presentation and probe presentation trial components, a linear combination (i.e., a contrast) of the two parameter estimates was computed which estimated the area under the curve of the neural response for that trial component. For the retention delay trial component, the analogous contrast was simply the coefficient of its single basis function. These contrast estimate images were intensity normalized via voxel-wise division by the time-series mean, masked with an image that represented the intersection of useable data from all subjects and had a gray matter prior probability > 0.25 in standard space using the prior probability images supplied with the SPM99 program (this eliminated the ventricles and surrounding white matter from the search volume), and spatially smoothed with an isotropic Gaussian kernel (full-width-at-half-maximum = 8 mm). The purpose of masking prior to smoothing is to avoid differences between subjects in partial volume contributions to data within the mask after smoothing. The resulting images were used as the dependent data in a second-level, voxel-wise GLM [38] that modeled 18 repeated measures per subject per voxel, with a design matrix representing two between-subjects factors (age and gender) and three repeated measure factors (trial component, set size, and probe type). Certain sets of effects of interest (see Section 2.6) from this second-level GLM were subjected to sequential latent root testing. The repeated measures covariance matrix of the second-level data was estimated at each voxel and spatially averaged. This matrix was used to approximate the known observation error covariance matrix (\(\Sigma\) in [103]). Results concerning gender are not reported here, but its inclusion as a factor meant that estimates of age effects controlled for gender.

2.6. Sequential latent root testing

In this version of CVA [103], a singular value decomposition (SVD) is performed on the whitened effects of interest, followed by sequential latent root testing (with \(\alpha\) controlled at a desired level) to assess the number of latent spatial patterns. The resulting statistical tests use parametric approximations (F-statistics), which have been validated using computer simulations [103]. The term “effects of interest” refers to a \(k\)-dimensional subspace of the design matrix of relevance for a particular hypothesis. Based on our interest in neural responses that increased with WM load as well as those that were independent of WM load, effects of interest comprised all six combinations of slopes and \(y\)-intercepts of fMRI signal with respect to set size with the three trial components and with all separately representing elder and young (thus \(k = 2\) for all of six sets of effects of interest). Effects of interest were averaged over the probe type factor (i.e., true positive and true negative trial types), thus making the effective number of trials per subject per set size equal to 30 (or close to 30 when accounting for no-response trials).

Significant latent spatial patterns are presented for descriptive purposes scaled by their singular values (analogous to SPM\(\{t\}\) images [103]), thresholded for descriptive purposes at a \(t\) value corresponding to \(p < 0.001\) uncorrected for multiple comparisons and a cluster size of 50 voxels. This threshold does not control map-wise statistical significance at \(\alpha = 0.05\) [28,101], and so is only meant to provide a somewhat condensed description of the significant latent spatial patterns.

The signs of the voxel values in a latent spatial pattern and its corresponding expression across subjects (or groups) are only meaningful in their product (i.e., the signs of each in isolation may be thought of as completely arbitrary). One multiplies a particular latent spatial pattern by its predicted expression to yield the predicted contribution from that latent pattern to the effects of interest [103].

2.7. SPM voxel-wise intensity tests

Using the same second-level GLM as in the CVA analysis, SPM was used to test for voxel-wise intensity differences between young and elder groups in the fMRI slopes and fMRI \(y\)-intercepts in all three trial components. The statistical threshold was selected to control \(\alpha\) at 0.1 per two-tailed effect, corrected for multiple voxel-wise comparisons using Gaussian random field theory [101].

2.8. Neural inefficiency

The neural inefficiency of a given latent pattern expression in each subject was computed by dividing the observed
expression in each subject by a performance value (where a higher performance value means better performance). The performance values used to compute neural inefficiency were a measure of accuracy independent of response bias ($d'$) at set size 6 (age effects on discriminability were expected to be largest at the largest set size), RT slope$^{-1}$, and RT intercept$^{-1}$.

3. Results

3.1. Reaction time

As expected [90], reaction time was affected by set size in both young ($F(1.8, 68.9) = 106.2$, $p < 0.001$) and elder ($F(1.7, 28.0) = 77.8$, $p < 0.001$) subjects (Fig. 3). The interaction of age group and set size was significant ($F(1.7, 96.1) = 6.1$, $p = 0.005$). Relatedly, a direct comparison of RT slopes revealed a significantly larger slope in elders (59.0 (30.8) ms/letter in young, 85.7 (35.7) ms/letter in elders; $t(56) = -2.91$, two-tailed $p = 0.005$). The variability of RT slope in the elder group was not significantly greater than that in the young group ($F(17, 39) = 1.35$, $p = 0.22$).

The $y$-intercepts were not significantly different between young and elders (806.6 (180.0) ms in young, 875.0 (176.1) ms in elders; $t(56) = 1.35$, two-tailed $p = 0.183$). The variability of RT intercept in the elder group was not significantly greater than that in the young group ($F(17, 39) = 0.96$, $p = 0.52$).

3.2. Accuracy

While the accuracy of young subjects was not detectably affected by set size ($F(2.0, 78) = 1.2$, $p = 0.32$), the accuracy of elder subjects decreased with set size ($F(1.6, 27.0) = 3.9$, $p = 0.04$), with the interaction of group and set size significant ($F(1.9, 103.8) = 3.5$, $p = 0.04$). This suggests that normal aging affects WM capacity (Fig. 4), even though accuracy was quite high in an absolute sense at set size 6 for elder subjects (95% confidence interval for $d' = [2.09, 3.20]$).

3.3. DIR task brain activity patterns in young and elders

At least one latent spatial pattern (the possible outcomes were 0, 1, or 2 patterns) was detected in each of the six sets of effects of interest (Table 1). Two latent spatial patterns were detected only for retention delay activation slope. Thus, the results indicate non-identical WM load-dependent, retention delay activation patterns in young and elder populations and fail to reject identity of activation patterns across age groups in the remaining five sets of effects of interest.

<table>
<thead>
<tr>
<th>2-Dimensional effects of interest</th>
<th>Test for at least one component</th>
<th>Test for two components</th>
<th>Inferred number of spatial patterns at $\alpha = 0.05$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Load-dependent processing during memory set presentation</td>
<td>3.51</td>
<td>1.07</td>
<td>0.22</td>
</tr>
<tr>
<td>Load-dependent processing during retention delay</td>
<td>1.79</td>
<td>1.41</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td>Load-dependent processing during probe presentation</td>
<td>1.12</td>
<td>0.88</td>
<td>0.92</td>
</tr>
<tr>
<td>Load-independent processing during memory set presentation</td>
<td>2.83</td>
<td>0.82</td>
<td>0.99</td>
</tr>
<tr>
<td>Load-independent processing during retention delay</td>
<td>1.22</td>
<td>0.88</td>
<td>0.92</td>
</tr>
<tr>
<td>Load-independent processing during probe presentation</td>
<td>8.96</td>
<td>1.03</td>
<td>0.37</td>
</tr>
</tbody>
</table>
Fig. 5. Scaled first latent spatial patterns (red: positive voxel weights, green: negative voxel weights) and observed (●) and predicted (——) subject-wise expressions of the corresponding pattern. (a) slope of fMRI response amplitude vs. set size associated with memory set presentation, (b) slope associated with retention delay, (c) slope associated with probe presentation, (d) y-intercept of the relationship of fMRI response amplitude vs. set size associated with memory set presentation, (e) y-intercept associated with retention delay, and (f) y-intercept associated with probe presentation. See Section 2.6 for thresholding.
The first latent spatial patterns and corresponding predicted and observed expressions for all six sets of effects of interest are shown in Fig. 5. We forego a tabular description of the latent spatial patterns in this manuscript because we have used a threshold for descriptive purposes that is more lenient than what should be used to control \( \alpha \) at a map-wise level [101], and more fundamentally because such a tabulation would be of limited direct relevance to the hypotheses we wished to test.

### 3.4. Compensatory reorganization versus dedifferentiation

As mentioned above, activation slope patterns corresponding to the retention delay were different in young and elders. The corresponding first latent spatial pattern had the same sign of expression in both age groups (Fig. 5b), and so tends to reflect where both young and elder activate in the same direction, but with a greater magnitude in elders (global maximum of negative spatial weights in left prefrontal cortex, MNI coordinate = \([-54, -642]\)). In contrast, the second latent spatial pattern had opposite signs of expression (elders: negative; young: positive; data not shown) in the two age groups, with elders having a much larger magnitude of expression than young. The only location in the second latent pattern to exceed our display threshold was an area with negative spatial weight in right parahippocampal gyrus/lingual gyrus (\([20, -50, 0]\); Fig. 6, inset) in which elders had positive (negative spatial weight multiplied by a negative expression) load-dependent retention delay effects and young subjects had essentially nil effects (data not shown). A homologous
area in the left hemisphere with similar effects was evident at the same height threshold but a more lenient cluster threshold.

Parahippocampal cortex is thought to be important for long-term memory [1,107], particularly for recollection [67]. To test if set-size dependent engagement of parahippocampal cortex (and associated areas of the second latent spatial pattern) during the retention delay confers a performance advantage (as would be predicted by cross-sectional compensatory reorganization models), we computed correlations within elders between performance ($d'$ at set size 6, RT slope, and RT intercept) and expression of this second spatial pattern. The only significant correlation was with RT slope and RT intercept. Thus, greater activation pattern similarity of young to higher performing elder brain activation patterns, is not consistent with cross-sectional compensatory reorganization.

It is also possible that the right parahippocampal gyrus itself is important for compensatory reorganization in some special way that would not be evident in the relationship of latent pattern 2 expression to behavior. All $R^2$ estimates between retention delay fMRI slope at the right parahippocampal locus and the behavioral measures in elders were <0.01, which is inconsistent with a cross-sectional compensatory role for this brain area in DIR task performance.

3.5. Neural inefficiency versus capacity limitation

For none of the effects of interest which had identical activation patterns in young and elders was pattern expression significantly lower in magnitude (i.e., closer to zero) in elder than young subjects (all $p > 0.25$). This is inconsistent with the capacity limitation hypothesis. For these same effects of interest, neural inefficiency was greater in elders than young subjects for activation slope during encoding with respect to $d'$ at set size 6 and RT slope and for activation $y$-intercept during the retention delay with respect to RT slope and RT intercept (Table 2). Also, though not sufficiently explained by the neural inefficiency model (as the young and elder activation patterns were non-identical), the first latent spatial pattern of the activation slope during the retention delay was expressed more inefficiently by elders (RT slope and RT intercept).

3.6. Age group comparisons via SPM

Even though SPM could not test the primary hypothesis of interest (namely, whether spatial patterns of activation are identical in elder and young), a reviewer advised that for completeness the results of SPM analyses be presented. Voxel-wise differences between young and elder groups in the slopes or $y$-intercepts of fMRI signal with respect to set size for memory set presentation/encoding, retention delay, and probe presentation were queried with SPM using mapwise corrected thresholds of two-tailed $\alpha = 0.1$, which corresponded to $t_{\text{crit}} = 4.49$ with greater than 200 effective degrees of freedom [101]. The null hypothesis of identical activation levels in the two age groups was rejected only for slope during memory set presentation (elder > young), with local maxima in bilateral precentral gyrus (MNI coordinates of [48, −10, 34] and [−40, −14, 38]). We note that young and elder activation patterns were deemed identical by CVA for this set of effects of interest. This provides an empirical proof of the abstract concept presented in Fig. 1.

4. Discussion

4.1. Age effects on brain activation associated with performance of a DIR task

In the current study, aging was associated with set size dependent decrements in both DIR accuracy and memory scanning speed, which is consistent with previous findings [2,23,26,62]. However, as indexed with the $y$-intercept of the RT versus set size relationship, we did not detect an expected
overall slowing with aging. Assuming an age effect size on reaction time corresponding to $R = 0.50$ [97], power for the current design was >0.9 with $\alpha$ controlled at 0.05.

Nevertheless, in the context of the observed behavioral impairments, we attempted to distinguish between four broad hypotheses concerning what happens to brain function with aging from young (approximately the third decade) to elder (approximately the eighth decade) adulthood. Processes with some linear dependence on WM load were assessed with the slope of fMRI activation amplitude with respect to set size; processes that were engaged even at a theoretical zero WM load (i.e., load-independent processes) were assessed with the $y$-intercept of fMRI activation amplitude with respect to set size; there was no constraint or assumption that such processes are not co-localized in the brain. The findings are summarized in Table 3.

Critical to disambiguating the hypotheses was the concept of identity of brain activation patterns in young and elder populations to within a scaling factor. SPM does not test this; CVA does [103]. However, there is no reason why both CVA and SPM cannot be applied to the same dataset to address different hypotheses. In the current study, very few voxel-wise intensity differences between groups were detected with SPM. This might be surprising given that age effects are commonly reported in the neuroimaging literature. However, most relevant SPM analyses do not use statistical thresholds that control $\alpha$ at a map-wise level, which makes such analyses have higher type I and lower type II error rates than the current one.

Of the six sets of effects of interest examined, five had patterns that were identical (or more precisely, could not be distinguished from identical while controlling $\alpha$ at 0.05) in young and elders. As the power of spatially omnibus tests like those employed with CVA to detect diffuse differences in spatial patterns is better than that of voxel-wise tests [102,103], this finding supports a general similarity of brain activation patterns in young and elders. A visual impression of overall similarity of elder and young activation patterns was noted previously in a study that used a verbal WM task [15].

### 4.2. Capacity limitation

Of the five spatial patterns of activation identical across age groups, none were expressed to a significantly lesser degree in elders than young, and so the idea of an age-related limitation in neurophysiological processing capacity was not supported in the context of DIR for letters. To clarify, this finding does not mean that memory capacity or processing speed do not tend to decrease with age; indeed, they do [2,23,26,62,97]. Rather, it means that the data do not support that the upper bound on brain system activation in elders is substantially lower than the corresponding bound in the young [32].

#### 4.3. Neural inefficiency

The term neural (in)efficiency has been used to explain the findings of negative relationships between certain electrophysiological and intelligence measures [61,91]. It has also been used in the context of hemodynamic neuroimaging of schizophrenia [16,54,66] and of individual differences in normal populations [76]. In the current study, of the five spatial patterns of activation identical across age groups, two were expressed less efficiently in elders. These were load-dependent processing associated with memory set presentation and load-independent processing associated with the retention delay.

The lower neural efficiency associated with load-dependent memory set presentation in elders seems at odds with the hypothesis that visual areas activate less in elders than in young [15,33,40,52,57]. Less efficient neurophysiological processing during encoding of visual stimuli could be due to pathology in the substrates of retinotopic representation [25], spatial attention, or other aspects of encoding. Due to the spatial nature of the stimulus array, spatial attentional demands presumably increase with set size during this phase of the trial. However, spatial attention and its neural correlates do not seem to be affected by aging [3,27,104].

Functionally, the $y$-intercept of fMRI signal associated with the retention delay would represent all aspects of WM that do not depend on set size. The lower neural efficiency of expression of this pattern in elders is difficult to speculate upon, as the predicted expression in the most heavily (positively) weighted voxels was negative, which thus corresponds to negative predicted $y$-intercepts in both age groups with respect to baseline levels of neural activity. Thus, more inefficient expression in the elders means that the elder activation $y$-intercepts (when normalized by performance) tended to be

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**Table 3**

<table>
<thead>
<tr>
<th>2-Dimensional effects of interest</th>
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– means inconsistent with the hypothesis.
more negative than those in the young. We can say that if a reduction of set size zero neural activity from baseline is integral to retention delay processing, then elders brains manifest a more inefficient form of this mechanism than young subjects. But we nevertheless recognize the more contentious nature of comparing the intensities of deactivations, since metabolic deactivations appear to be different from inhibitory synaptic activity, which ceteris paribus should lead to an increase in metabolic activity [44].

The first latent spatial pattern associated with load-dependent retention delay processing was also less efficiently expressed in elders than young. But since the spatial activation patterns for this effect in young and elders were different, the simple neural efficiency model does not explain these data. However, it is conceivable that age-related decreases in neural efficiency of canonical brain networks and engagement of non-canonical networks are related. Consistent with this speculation, elder subjects who expressed the first latent spatial pattern the most tended to have greater expression of the second latent spatial pattern ($R^2 = 0.66$) while the opposite was true of young subjects, with those young subjects evidencing the greatest expression of the first latent spatial pattern having expression of the second latent spatial pattern closest to zero ($R^2 = 0.19$).

The nature of the mechanisms of reduced neural efficiency in elders could presumably be informed by cross-sectional correlations of inefficiency with age-related neuropathology, modeling of neural circuit bioenergetics, and in vivo and in vitro pharmacological studies. We feel that more needs to be done to see if the definition of neural efficiency adopted here is a useful descriptor for the effects of brain pathology in aging. For example, one obvious issue is the construct’s sensitivity to the units of measurement.

### 4.4. Compensatory reorganization and dedifferentiation

Non-identity of young and elder brain activation patterns was detected for load-dependent processing associated with the retention delay. This effect is presumed to relate to load-dependent verbal WM maintenance. Thus, our findings support the existence of some age-related change in the brain networks that mediate this process that cannot be simply explained as a scaling. However, this does not mean that these networks in the young and elder populations are non-overlapping, as the first spatial pattern was expressed with the same sign and similar magnitudes in both age groups. In contrast, the second latent spatial pattern was expressed with different signs in the two age groups, and primarily in elders. This second pattern was most heavily weighted in right parahippocampal gyrus/lingual gyrus, where elders showed positive and young subjects showed near nil load-dependent activation during the retention delay. It is important to stress that not all important elements of latent spatial pattern 2 might be above visualization thresholds, and that a more powerful study would be required to localize them with reasonable control over type II error rate.

When the second latent pattern has null expression in both age groups, then both age groups have identical spatial patterns (equal to the first latent pattern). Expression of the second latent pattern was near zero for both age groups at faster memory scanning speeds. This means that young activation patterns were more similar to those of higher performing elders than those of lower performing elders. This is not consistent with cross-sectional models of compensatory reorganization such as HAROLD [14]. In contrast, this finding is consistent with, but does not imply, that the age-related change in activation pattern is a cause of impaired performance.

This result is also consistent with the idea of neural compensation that we have discussed previously in the context of cognitive reserve [89]. According to this idea, age-related changes in activation patterns might arise when age-related brain pathology has impaired the brain networks normally used by younger individuals: The expression of these new activation patterns might be compensatory in that they could have a beneficial effect on performance relative to their theoretical absence, even if they are not associated with better task performance in a cross-sectional sense. Also, although this compensatory reorganization would arise in response to age-related brain pathology, individuals with same degree of underlying pathology might vary in their ability to compensate [89]. In order to test this interaction between brain reorganization, behavioral performance, and cognitive reserve one would need an index of brain pathology, which the current study lacks.

### 4.5. Consideration of population under study

High densities of plaques and tangles, the main pathological markers of AD, are not a necessary consequence of aging, as the majority of elders who are cognitively intact at the time of death have minimal amounts of AD-like cortical neuropathology [34,46,81]. However, it is possible for a cognitively intact person to have a high degree of AD-type neuropathology [85]. As assessed by MMSETotal [86], the population of interest in the current study comprises elders absent dementia or mild cognitive impairment [36], but we do not know positively whether these behaviorally normal elders are absent any AD pathology. The elders were also well matched with the young subjects on years of education, but tended to have somewhat lower IQ values. Future work will investigate the relationship between IQ and other hypothesized aspects of cognitive reserve on neural correlates of WM. Another caveat is that the sampling for the study was not random, with certain subpopulations of elderly (e.g., those associated with senior centers) and young (e.g., Columbia University students) being over-represented.

### 4.6. Relationship to previous findings

Greater levels of activation for elders than young subjects have been reported in some brain areas under various
conditions [15,33,57,58,71,75]. One influential study reported greater DIR task activation (ostensibly related to WM) in elders compared to younger subjects in premotor cortex, parietal cortex, and ventral and DLPFC [15]. Unfortunately there were problems in the fMRI time-series modeling of that report that makes its neuroimaging results difficult to interpret. In contrast, the strength of the current study is that it modeled several features (across trial components and set size) of the functional responses associated with DIR trials in a rational and interpretable manner.

Another study reported no effect of age on ventral PFC set size-dependent activation and greater DLPFC activation in younger than elder subjects [75]. This is quite different from our findings in that DLPFC and anterior insula (which is proximal to and could be potentially reported as ventral PFC) had substantial spatial weights in latent spatial pattern 1 of load-dependent processing during the retention delay, which was expressed to a greater degree in elders. But that study collapsed across all trial phases [75], and had other methodological differences from the current study, making them difficult to meaningfully compare.

In a different study by the same group, greater DLPFC activation was found in younger than older adults only during the probe phase of a DIR task [74], with probe phase DLPFC activation correlating negatively with RT in elders. As that study collapsed fMRI signal over set sizes, it is not easily comparable with the results from the current study. Nevertheless, there was no correlation detected between expressions of latent spatial pattern 1 of load-independent or load-dependent probe period processing with either RT slope or RT intercept in the elders (all $R^2$ estimates $<0.01$). However, there were significant correlations between RT slope and expression of latent spatial patterns 1 ($R^2 = 0.33$) and 2 ($R^2 = 0.37$) of load-dependent retention delay processing in the elders. But, the direction of the correlation they reported in elders collapsed over set sizes during the probe phase (slower RT being associated with smaller DLPFC activation) is opposite in sign from the one detected here (taking into account the relative signs of the spatial weights in right DLPFC and expression of latent spatial pattern 1) which had slower RT being associated with greater pattern expression. Finally, our findings of greater overall activation during the retention delay in elders go against their conclusion that age-related differences in the neural correlates of DIR task performance are limited to the probe phase.

In contrast to increases in activation in elders, there have also been many reports of reduced activations in elders [15,32,40,41,43,57–59,74,75,79]. We did not detect any such effects at the spatial pattern level (which would have been manifest as a lower, i.e., closer to zero, expression of a pattern in elders than young) or at the voxel-wise level using SPM. Thus this study also yielded no results consistent with the unsubtle idea that some parts of elder brains might simply stop working altogether [32].

The broader significance of the current study is that it is first time the idea has been formally tested that normal aging might simply vary the gain of neural activity in an age-independent brain network. Vis a vis Fig. 1, greater activation detected via SPM in elders over young has tended to be interpreted as compensatory reorganization [4,10,14,33,87,100], without consideration of the alternative that there is a general increase in expression of an age-independent brain activation pattern, and that what is tending to be detected are simply the largest ensuing age-dependent intensity differences. Critical to appreciating this alternative is the fact that a simple scaling (i.e., gain) difference in brain pattern expression between two groups can lead to spatially varying activation intensity differences between groups.

4.7. Alternative explanations of current findings

For comparisons of BOLD signal changes across groups to be neurophysiologically meaningful, the coupling of neural activity to BOLD signal must be essentially the same in the two groups. On average, the gain of coupling of neural activity to the BOLD signal has been reported to be essentially unchanged with normal aging, though there is greater within-group variability in elder subjects [24,39]. This evidence that neural-hemodynamic coupling is not changed with normal aging is critical as it allows other differences in fMRI response to be interpreted, at least tentatively, as reflective of neurophysiological differences.

The brain tends to atrophy with aging [31], and it is possible that gray matter loss is closely related to any age effects on brain activation. This is because analyses of imaging data performed in standard brain space do not typically involve normalization of the activation measures by gray matter concentration per voxel, and so a standard space voxel with low gray matter concentration would be expected ceteris paribus to have a low activation signal. Such effects have been observed [30,68,93], but not always [42]. Also, in Alzheimer’s disease patients an inverse relationship between gray matter concentration in hippocampus and memory task-related activation in inferior frontal gyrus and cerebellum has been observed [30]. In the current experiment, total (i.e., averaged over the entire brain in standard space) elder gray matter concentration was not correlated with the expression of any of the latent spatial patterns (all two-tailed $p > 0.05$). This does not rule out that local gray matter concentrations have relationships with activation levels. However, that elders did not tend to have lower activation levels than young subjects seems to speak against the possibility that atrophy is responsible in a direct way for any of our fMRI findings. Indeed, if fMRI signal had been normalized by GM concentrations in this study, it seems quite likely that the resulting elder activations would have been that much greater than those in young.

A reviewer raised the issue of whether differences in the quality of coregistration of structural to functional data between the two age groups could account for any of the findings. If there were increased coregistration noise in the elderly, this would be expected ceteris paribus to reduce
observed amplitudes of fMRI activation in the elderly in the second-level GLM, which was the opposite of what tended to be observed.

5. Conclusions

None of the models under consideration were able to explain all of the age-related differences in brain activity during performance of a DIR task. There was evidence for neural inefficiency (load-dependent aspects of encoding and load-independent aspects of verbal WM) and dedifferentiation (or possibly compensation [89]; load-dependent aspects of verbal WM maintenance). Neither capacity limitation nor load-independent aspects of encoding or response planning/execution, no effect of aging was detected at all. These results suggest that, even within the same task, the existence and nature of brain activation changes with aging can vary based on cognitive process engaged.

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References


