

NIH Public Access

Author Manuscript

Psychol Bull. Author manuscript; available in PMC 2012 September 1

Published in final edited form as:

Psychol Bull. 2011 September; 137(5): 753–784. doi:10.1037/a0023262.

Neuroanatomical substrates of age-related cognitive decline

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Abstract

There are many reports of relations between age and cognitive variables and of relations between age and variables representing different aspects of brain structure, and a few reports of relations between brain structure variables and cognitive variables. These findings have sometimes led to inferences that the age-related brain changes cause the age-related cognitive changes. Although this conclusion may well be true, it is widely recognized that simple correlations are not sufficient to warrant causal conclusions, and other types of correlational information, such as mediation and correlations between longitudinal brain changes and longitudinal cognitive changes, also have limitations with respect to causal inferences. These issues are discussed, and the existing results on relations of regional volume, white matter hyperintensities, and DTI measures of white matter integrity to age and to measures of cognitive functioning are reviewed. It is concluded that at the current time the evidence that these aspects of brain structure are neuroanatomical substrates of age-related cognitive decline is weak. The final section contains several suggestions concerned with measurement and methodology that may lead to stronger conclusions in the future.

The primary question addressed in this article is the extent to which relations of age with measures of cognitive functioning are attributable to relations of age with measures of brain structure. The article is organized in five sections. The first section briefly describes the primary phenomenon to be explained, namely, age-related differences and age-related changes in cognitive functioning. The second section consists of a discussion of analytical methods that can be used to examine hypothesized causal relationships among sets of variables. The third section contains a brief rationale for a focus on aspects of brain structure rather than functional activation, and the fourth section is a review of empirical research on the interrelations of age, brain structure variables, and cognitive variables based on the framework outlined in the second section. The final section summarizes the major conclusions, discusses limitations of current research, and offers suggestions for future research.

Age-related differences and changes in cognitive functioning

Age-cognition relations are well-established in cross-sectional comparisons, and are becoming better established in longitudinal comparisons. Two broad trends are typically found; an increase until about 60 years of age followed by a decrease for measures reflecting acquired knowledge, and a nearly linear decline from early adulthood for measures of the efficiency or effectiveness of processing at the time of assessment (for reviews see Craik & Salthouse, 2008; Hedden & Gabrieli, 2004; Salthouse, 2010d).

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Although longitudinal studies involving adults with mean ages of 60 or older are not uncommon (e.g., see Hofer & Piccinin, 2007, for a recent review), only a limited number of longitudinal studies have compared adults across the entire range of adulthood on measures of cognitive functioning. Nevertheless, the available studies are consistent in finding that increased age is associated with more negative cognitive change (e.g., Giambra et al., 1995; Huppert & Whittington, 1993; Schaie, 2005; Zelinski & Burnight, 1997; also see Figure 2. 2 in Salthouse, 2010d).

The patterns with variables reflecting processing efficiency and effectiveness can be illustrated with results from two projects in which both cross-sectional and longitudinal data are available from the same individuals across a wide range of ages. In order to facilitate comparisons across variables and projects, all scores are expressed in T-score units which have a mean of 50 and a standard deviation of 10. Except for the Block Design measure, each variable is represented as a composite score based on three or more separate test scores.

The left panels in Figures 1 and 2 contain data from a project by Salthouse (e.g., 2010a, b, c), that involved over 1,500 individuals, and an average retest interval of 2.5 years. The right panels contain data from a project by Ronnlund and colleagues (2005 2006), that involved an initial sample of 1,000 adults and a retest interval of 5 years. The top panels in each figure contain means, and the bottom panels contain between-person standard deviations for the same data used to compute the corresponding means.

Figure 1 portrays the cross-sectional results. Notice that there are large age-related differences in the mean level of performance for all six variables. (Age correlations for the abilities in the left panels were –.46 for reasoning, –.45 for space, –.44 for memory, and –. 63 for speed.) Although age-related increases in between-person variability are sometimes found (e.g., Christensen et al., 1999;Nelson & Dannefer, 1992;Rabbitt, 1993), the bottom panels in Figure 1 reveal that this is not the case in these data. Moreover, this finding is not specific to these two projects because nearly constant between-person variability has been found in cognitive variables from different standardized test batteries (see Figures 1.12 to 1.15 in Salthouse, 2010d). Age-related increases in variance may be more likely when the measures are in units of time in which there is typically a strong positive relation between mean and variance, or when the sample contains individuals with various pathological conditions. However, the results in Figure 1 and elsewhere clearly indicate that age-related differences in mean performance can occur without concomitant increases in between-person variability.

Longitudinal results for the same variables and individuals illustrated in Figure 1 are portrayed in Figure 2. The results in the top panels indicate that the longitudinal changes (across a span of up to 5 years) are small relative to the cross-sectional differences (across a span of about 60 years), as the changes have a range of about 8 T-score units compared to a range of about 20 T-score units for the differences. However, the cross-sectional and longitudinal data are similar in other respects. For example, the longitudinal changes are systematically related to age, but with positive values at young ages, and negative values at older ages. (Age correlations for the longitudinal changes in the left panel were -.11 for reasoning, -.21 for space, -.25 for memory, and -.17 for speed.) At least some of the positive values for adults under about 50 years of age are likely attributable to retest effects associated with prior performance of the tests (cf. Salthouse, 2010c). Additional analyses revealed that the quadratic age relations in the Salthouse data were not significantly different from zero, and because there is no evidence of a discrete step between a period of stability and a period of negative change, it does not appear to be the case that change begins only at middle age or later. Furthermore, as with the cross-sectional differences, there was little relation between age and between-person variability as the individual differences in change

were as large at age 30 as at age 70. That is, despite differences in the direction and magnitude of longitudinal change across different periods of adulthood, the magnitude of individual differences in change in these studies was nearly the same at all ages. There was also little indication of a relation of age to the individual differences in longitudinal change in other age-heterogeneous studies containing information about the variability of change (e.g., Alder et al., 1990;Giambra et al., 1995;Hertzog & Schaie, 1986;1988;Huppert & Whittington, 1993; also see Finkel et al., 1996;1998).

To summarize, nearly linear age-related declines in both cross-sectional and longitudinal comparisons have been reported in several major cognitive abilities. Although prior research has primarily focused on mean values, between-person variability is actually more important when examining relations with other variables because it sets limits on the magnitude of possible correlations the variable can have with other variables. The results summarized above are therefore noteworthy in indicating that not only is the magnitude of variability considerable in both the levels (cross-sectional) and the changes (longitudinal) in cognitive performance, but that variability does not inevitably increase with advancing age. Of particular relevance for the current review is that these results imply that there is no statistical reason why correlations involving cognitive variables would necessarily be weaker among young adults than among middle-aged or old adults.

A key question to be addressed in the remaining sections of this article is the role of neurobiological factors in these age-cognition relations. Although different mechanisms may be operating at different ages, it is important to recognize that the phenomenon to be explained is not merely relations of age in one ability within a narrow age range, but rather the nearly continuous relations of age to a wide variety of cognitive variables across nearly all of adulthood.

Analytical methods

Although most cognitive neuroscientists probably assume that individual differences in cognitive functioning have a neural basis, there is still considerable uncertainty about the role of specific brain structure characteristics on the age differences and age changes in cognitive functioning. Inferences that age-related cognitive declines are attributable to age-related changes in brain structure are sometimes based on correlations between a brain variable and a cognitive variable, on correlations of age with both the brain variable and the cognitive variable, and occasionally on correlations between longitudinal changes in a brain variable and longitudinal changes in a cognitive variable. As will be described below, all of these correlations are limited with respect to the information they provide about causal relations.

It is well-accepted that the ideal procedure for investigating causality is an experimental study in which individuals are randomly assigned either to a control group or to an experimental group, and differences between groups are examined in a relevant outcome variable. Furthermore, when the primary outcome of interest concerns effects on aging, long-term monitoring is needed to examine rates of aging in the target measures of cognitive functioning. However, even if it were ethical, it is difficult to manipulate specific aspects of brain structure, and it is seldom feasible to follow individuals long enough to observe effects on rates of aging. Correlational data are therefore the primary means of investigating interrelations of age, brain structure, and cognitive functioning in research on humans.

One type of correlational evidence simply consists of the relations of age with the relevant brain and cognitive variables. For example, as noted in several early studies (e.g., Bigler et al., 1995; Raz et al., 1993), a discovery of similar age trends in brain variables and cognitive

variables could lead to speculations that the age-related influences on the brain variables are causally related to the age-related influences on the cognitive variables. However, a very large number of variables are related to age, and therefore additional information is needed to determine which variables might be causally related to one another, and which ones have no causal relation.

Several analytical procedures have been proposed to investigate causal relations based on the principle that although correlation does not imply causation, causation does imply correlation. That is, although causal hypotheses cannot be directly investigated with correlational procedures, implications of causal hypotheses can be examined with correlational data. One of the first systematic outlines of correlation-based procedures involving mediation and moderation was published by Baron and Kenny (1986), and the fact that the article has become one of the most cited articles in psychology is an indication of the high level of interest in these procedures. Applications of these procedures in research concerned with the interrelations of age, brain variables, and cognitive variables have been discussed by a number of authors (e.g., Fjell & Walhovd, 2010; Madden, Bennett et al., 2009; Rabbitt et al., 2007), but their assessments of the issues, and of possible solutions, differ from those described here.

Mediation

The simplest versions of mediation procedures are based on relations among three variables. In the current context, the age variable will be designated A, brain structure variables will be designated B, and cognitive variables will be designated C. The top panel in Figure 3 indicates that correlations are frequently observed among the A, B, and C variables, usually in the direction of lower cognitive performance and less intact brain structure with increased age, and lower levels of cognitive performance with greater structural degradation. The remaining panels in the figure portray alternative models that could produce these types of correlations among the variables. Researchers interested in neural correlates of age differences in cognition are often most interested in Model 1, in which a brain variable is hypothesized to mediate the age-related influences on a cognitive variable. However, the other models in Figure 3 represent alternative conceptualizations of the relations among the three variables. Additional models could also be specified, such as one in which A, B, and C are all reflections of a single common factor, but the four in Figure 3 are the most frequently discussed models of the relations among these variables.

Models 1 and 2 differ in that the hypothesized causal direction is from brain structure to cognitive functioning in Model 1, whereas the causal direction between these two variables is reversed in Model 2. Model 3 is an independence model in which B and C are postulated to be related to each other only because of the common influence of A on both variables. One way to evaluate the plausibility of these models involves examining implications of specific relations under the assumption that a particular model is valid. For example, if Model 1 is valid then statistical control of the variation in B should reduce the magnitude of the A–C relation because much of the influence of A on C is postulated to be mediated through B. In a similar manner, if Model 2 is valid then statistical control of the variation in C should reduce the A–B relation. Because Model 3 specifies that B and C are actually independent, and only related to each other because of their relations to A, an implication of this model is that control of the variation in A should reduce or eliminate the B–C relation.

Model 4 postulates that some other variable, which could reflect health status, aspects of lifestyle, genetic profile, etc., is related to age, and to both the brain and cognitive variables. Unless all other relevant variables are known and represented in the model, it is difficult to rule out variants of Model 4 in which some unspecified variable X is involved in the relations among A, B, and C. Although seldom definitively rejected, interpretations based on

Model 4 can gradually be rendered less plausible as different candidates for X are examined across a series of studies, and are found to have minimal effects on the relations among A, B, and C.

A considerable number of researchers have conducted mediation analyses by postulating a particular model, such as Model 1, and concluding that the mediation interpretation is confirmed if the A–C relation is reduced after controlling the variation in B. Although seemingly straightforward, two important issues need to be considered in the application and interpretation of this strategy.

One issue is how the difference in the relevant relations before and after statistical control is evaluated. Unless no other factors are assumed to be operating, it may be unrealistic to expect complete elimination of a relation after variability in a third variable is controlled. The question therefore arises as to how much reduction of a relation should be considered meaningful, and interpreted as supporting the prediction. One possibility is to rely on statistical significance of the residual (partialled) relation, or of the indirect (mediated) path, but this has the disadvantage that significance depends on sample size. Because correlations are measures of effect size, a potentially more desirable alternative is to indicate the magnitude of the alteration in the relation by comparison of correlations. For example, expanding on Cohen's (1992) convention, correlation differences less than .1 might be considered small, those between .1 and .3 might be considered of medium size, and those greater than .5 might be considered large, and most consistent with the prediction.

A second issue to be considered in mediation analyses is that the outcomes are asymmetric because although a failure to support the prediction could be interpreted as falsifying the model, support for the prediction would merely be consistent with that model, as well as with other possible models (e.g., Edwards & Lambert, 2007; Kraemer et al., 2001; Kraemer et al., 2008; Lindenberger & Potter, 1998; Penke & Deary, 2010; Stone-Romero & Rosopa, 2008). A more informative approach than focusing exclusively on a single model might therefore involve examining the implications of multiple models, such as Models 1, 2, and 3, with the same data. That is, in addition to determining whether the data are consistent with Model 1, the data could also be examined to determine whether they are inconsistent with alternative models, in which case confidence in the target model would be enhanced relative to the alternative models.

If one assumes that increased age is associated with a decrease in some measure of brain structure, which in turn contributes to a decrease in cognitive functioning, then it is reasonable to expect that the relation between age and cognitive functioning would be smaller if there were no variation among the research participants in the brain structure measure. Mediation analyses are useful in revealing the extent to which this is the case. Nevertheless, it is important to emphasize that the results of these types of statistical control analyses do not provide a direct test of causal relations, and instead merely provide an opportunity for implications of particular hypotheses of causal relations to be disconfirmed. However, because alternative models can often be postulated to account for relations among the variables, confidence in the plausibility of the hypothesized model can be increased if the data are found to be inconsistent with alternative models of the relations.

Correlated Changes

A major limitation of analyses with cross-sectional data is that all of the observations are collected at the same point in time, and thus models with different directions of causal relations are not easily distinguished (e.g., Muller et al., 2005; Penke & Deary, 2010; Selig & Preacher, 2009; Shrout & Bolger, 2002). For example, in terms of the models in Figure 3, the absence of definitive temporal ordering means that Model 1, with B as the mediator, may

not be distinguished from Model 2, with C as the mediator, and neither may be distinguished from Model 3 in which there is no mediation.

It is sometimes assumed that the availability of longitudinal data solves the problem of temporal precedence among variables, and therefore can help distinguish among Models 1, 2, and 3. Moreover, correlations between longitudinal changes in two variables are occasionally interpreted as implying the existence of a causal relation between the variables (e.g., Van Den Heuvel et al., 2007). Although it is true that longitudinal comparisons involve observations at different points in time, longitudinal data also have limitations when used to investigate causal relations (e.g., MacKinnon et al., 2007; Selig & Preacher, 2009). One of the most serious limitations is that it may not be reasonable to assume an instantaneous causal influence, with a zero lag between changes in the relevant variables (i.e., Gollub & Reichardt, 1987). That is, in most cases a temporal lag probably exists between the changes in two sets of variables, and analyses of lagged changes are only meaningful if the longitudinal interval between observations matches the interval between early change in the presumed causal variable and later change in the presumed effect variable (e.g., Collins, 2006; Hofer & Piccinin, 2007). For example, the relation between the two changes would likely be missed with a single longitudinal interval of one year if effects do not occur until at least three years after changes begin in the causal variable.

Unfortunately, relatively little is currently known about the dynamics of changes in either brain or cognitive variables, or about the lags between changes in the two types of variables, and therefore two-occasion longitudinal information (and information from related cross-lagged panel analyses) may be of limited value for distinguishing temporal order among variables (cf. Raz et al., 2010). Increasing the number of longitudinal assessments can help address this concern because a researcher could then determine whether early changes in the brain variable are associated with later changes in the cognitive variable. However, even these types of lead-lag analyses may not eliminate the problem if the total observation interval, or the spacing of observations within the intervals, does not match the timing of critical events.

Although they have limitations, there is at least one respect in which longitudinal data are more informative than cross-sectional data. This is that with correlations among changes one can be confident that the relevant individual differences are manifested during the period between measurements. That is, correlations based on cross-sectional data (i.e., B-C) reflect relations between influences that could have occurred any time prior to the measurements, whereas correlations of longitudinal changes (i.e., $\Delta B - \Delta C$) reflect coupling of changes that occur during the interval between measurements. For example, a cross-sectional correlation between memory performance and volume of the medial temporal lobe (MTL) region could be due to factors operating at any time from conception until the period of measurement. In contrast, a discovery that people with the greatest declines in memory over a given interval also have the greatest reductions in MTL volume indicates that the relation between the two variables is evident during that period of time. We may not know why the changes in either variable occurred, or when they first started, but a significant correlation indicates that people who change the most in one of the variables over the specified interval also have a great deal of change in the other variable. Stated somewhat differently, although the influences on one or both variables could have originated at any time in the past, which limits inferences about causal order, a discovery that longitudinal changes are correlated indicates that there is a relation between the manifestations of these influences during the interval between successive observations.

In some cases longitudinal information is available for only one of the variables, such that the correlation is between either the initial or final value of one variable (e.g., B) and the

To summarize, even correlations between longitudinal changes do not necessarily imply causality, and interpretations of the relations between changes can be complicated when the time courses of critical changes are not known. Nevertheless, correlations among changes can be informative because they indicate that people who change the most in one of the variables also have large changes in the other variable, and that the relevant changes are occurring in the period between the successive observations rather than at some unknown time in the past.

Moderation

Although there are many reports of correlations between brain variables and cognitive variables, several meditational analyses, and a few reports of correlations between brain changes and cognitive changes, only a small number of studies have investigated moderation by determining whether the relations between brain variables and cognitive variables vary as a function of age. This neglect is unfortunate because in some respects results of moderation analyses can be considered to provide the most valuable type of information about the role of age on the relations between brain structure differences and changes and cognitive differences and changes.

Salthouse (2010d) used a metaphor of an avalanche to describe the difficulty of inferring the causal, or triggering, variable (i.e., the first rock to move in an avalanche) on the basis of observations late in the sequence (i.e., the positions of rocks near the bottom of the mountain). However, information about temporal, and potentially causal, relations might be available by comparing relations between positions, and changes in positions, of rocks at several different locations on the mountain. An important implication of this perspective is that even longitudinal data may be of limited value if they are only available at some unknown point after the critical relations originated.

The avalanche metaphor of cognitive aging is obviously limited in many respects, but to the extent that cognitive aging is conceptualized as a long-term, dynamic phenomenon, examination of brain-cognition relations at different ages could provide valuable information about sequential relations among the variables. Unfortunately, many studies concerned with age and brain-cognition relations have focused only on the period of old age, which may be analogous to observing rock positions and motions at the bottom of the mountain, when many of the important precipitating factors occurred much earlier. Of course, if the phenomenon of interest occurs during a relatively brief period late in life, then restricting observations to the period of old age would be justified. However, because the results in Figures 1 and 2 indicate that cross-sectional differences and longitudinal changes in cognitive functioning occur continuously across adulthood, conclusions based on observations from a very restricted age range could be misleading with respect to the origin, and causal nature, of relations among variables.

Age moderation of brain-cognition relations might be expected in at least three different types of situations. First, moderation might occur if there are different determinants of C at different ages. For example, retest-related influences on cognitive variables might be stronger at younger ages than at older ages (Salthouse, 2010c), or adults of different ages might employ qualitatively different strategies to perform the tasks. Neuroanatomical substrate involvement could also vary with age, as suggested by the discovery of age

differences in the regional distribution of activation in functional neuroimaging studies (e.g., Dennis & Cabeza, 2008; Grady, 2009).

Second, moderation might be expected if there are different determinants of B at different ages. For example, weaker relations might be expected at younger ages if the B–C relations are at least partially attributable to preclinical pathology, and the pathology is less likely at younger ages.

And third, moderation could occur if there is a shift in either total variance, or the proportion of variance that is reliable, as a function of age. That is, because the key factor affecting the magnitude of relations with other variables is not the level of the variable, but rather the magnitude of the individual differences in the variable, the B–C or Δ B– Δ C relations might shift with age because of shifts in variance. For example, differences in variance might be expected when there are non-linear age trajectories for either the B or C variables, such that age-related variance is pronounced only after a particular age.

The question of the similarity of B–C relations at different ages can be investigated with versions of moderation analysis in which the focus is on interactions, or conditional relations. That is, moderation implies that the A–C relation varies as a function of the value of B (e.g., little relation of age to cognition at high values of brain volume, but more negative relations at lower values of brain volume), or equivalently, that the B–C relation varies as a function of age (e.g., weak relation of brain volume to cognition at young ages, but a stronger relation at older ages). Similar analyses could be carried out with correlations of changes to determine if the Δ B– Δ C correlations remain constant with age, or are stronger at either young or old ages.

Assuming adequate statistical power, lack of evidence of moderation would suggest that the B–C (or the Δ B– Δ C) relation is quantitatively similar at all ages, and would imply that a given value (or change) in B would be associated with a similar value (or change) in C at all ages. In contrast, evidence of moderation would imply that the B–C (or the Δ B– Δ C) relation differs as a function of age, possibly because of an age-related shift in the neural substrates of cognitive functioning. It is also important to note that discovery of a shift in the B–C (or Δ B– Δ C) relation with age would complicate interpretation of mediation analyses because it implies that control of the B variable would have different meaning at different ages (cf. MacKinnon et al., 2007).

In summary, simple correlations are not sufficient to infer causal relations between brain aging and cognitive aging, but correlation-based procedures can be informative. In particular, if the causal relation is as hypothesized, the critical relation would be expected to be reduced when variability in the relevant third variable is controlled, whereas implications of alternative models would be expected to receive less support. In addition, confidence in the hypothesized brain-cognition linkage would be strengthened if there were evidence that the relations occur across time within the same individuals in longitudinal data, and not just across individuals of different ages at a single point in time, as in cross-sectional data. Finally, moderation analyses are informative to determine whether the brain-cognition relations are invariant across all of adulthood, or whether there are shifts in magnitude, or direction, of the relations as a function of increasing age.

Brain structure versus functional activation

A considerable amount of functional neuroimaging research has been conducted in which activity in different brain regions is examined during the performance of a cognitive task (e.g., for reviews of age-comparative research see Dennis & Cabeza, 2008; Grady, 2008; and Park & Reuter-Lorenz, 2009). Because it allows brain activity to be linked to cognitive

One important issue is that increased age is often associated with lower performance in a wide variety of cognitive tasks, and consequently there could be many different patterns of age differences in functional activation. Furthermore, when only a single task is administered it is difficult to distinguish task-specific contributions to the activation from contributions associated with the construct of primary interest. One possible approach to this concern is to focus on the activation common across multiple tasks to identify aspects of shared activation (as in Collette et al., 2007; Grady et al., 2006; Nyberg et al., 2003; and Ranganath et al., 2003), but relatively few studies have collected functional neuroimaging data from multiple tasks in adults of different ages.

A second issue is that increased age is often associated with lower levels of performance, and functional activation can vary according to the level of performance in the cognitive task. The conditions of the task could be manipulated to equate the average level of performance when studying age differences in activation, but this introduces a confounding between age and task condition which can make it difficult to interpret differences in activation patterns.

A third complicating issue is that activation in functional neuroimaging may be dependent on structural characteristics, and yet direct investigations of structure-function relations are rare (cf. Dennis & Cabeza, 2008; Raz, 2000). A few studies have reported that the strength of activation is related to structural characteristics, such as brain volume (e.g., Andrews-Hanna et al., 2008; Brodtmann et al., 2003; 2009; Nordahl et al., 2006; Rosen et al., 2005; Thomsen et al., 2004; Tyler et al., 2010; Venkatraman et al., 2010), but no relations have been found in other studies (e.g., Berlingeri et al., 2010; Hazlett et al., 1998; Langennecker et al., 2007). Furthermore, relatively little is currently known about age-related influences on vascular reactivity and properties of blood oxygenation that are relevant to fMRI activation based on the BOLD signal. The inconsistent pattern regarding structure-function linkages led Grady (2008, p. 140) to conclude that: "...the influence of structural changes is far from clear and will need further work to determine whether there is an influence, which structural measure (e.g., white matter or grey matter) is most closely related to activation, and what form the influence will take (i.e., leading to an increase or a decrease of activity)."

For the preceding reasons, the current review will be limited to structural factors, and to only a subset of three of these, namely, regional volume, white matter pathology, and white matter integrity. These three characteristics have the largest amount of relevant research with concurrent brain and cognitive measurements on the same individuals across a range of ages, but it is acknowledged that they are not necessarily the most fundamental, or informative, structural characteristics. Although other structural (and physiological) characteristics such as cortical thickness, cerebral blood flow, concentration of brain metabolites, quantity of neurotransmitters or receptor sites, number of neurons, density of synapses or spines, etc. will not be considered, analytical methods similar to those described here could be applied with these measures when appropriate data become available.

Review of research

Relevant research studies were identified by searching databases (e.g., PsychInfo, Web of Science) with "brain aging," "brain cognition," "structural brain changes," "brain cognition correlations," and related terms, and manual inspection of relevant journals. In addition, potentially relevant citations from the reference lists in the retrieved articles were also examined whenever possible.

Two important limitations of the literature review need to be acknowledged at the outset. The first is that there may be some publication bias in the literature. That is, positive relations may be over-represented in the review because negative findings may not have been published, or conversely, they may be under-represented because they might have been considered mere replications of well-established results. It is difficult to estimate the magnitude of either type of publication bias, but it should be noted that many of the analyses to be reported lead to questions about the positive findings that have been published. In other words, regardless whether the proportion of positive and negative findings in the published literature is representative of the true state of affairs, concerns are raised about the interpretation of the findings that have been assumed to provide support for the hypotheses.

A second limitation of the literature review is that many published articles contain overlapping samples, often with little information about the actual degree of overlap. Although most serious when different reports focus on the same combination of brain and cognitive variables, this is also a concern across reports involving different variables from the same individuals unless there is evidence that the variables are independent of one another. As with the first limitation, the unknown degree of overlap of samples makes it difficult to derive quantitative estimates of effect sizes, but it is unlikely to have a major impact on the overall conclusions.

Relations among A (age), B (brain), and C (cognition) will be examined in cross-sectional data with the three models represented in Figure 4. The models are expressed in the form of path analyses, with the dotted line in each model representing the critical relation in the path analysis which can be compared with the simple correlation between the relevant two variables. The models are not exhaustive, in part because variants of Model 4 are not considered, and they are not mutually exclusive, because mixtures of different types of effects could occur simultaneously. Nevertheless, the three models in Figure 4 represent some of the simplest models of the relations among the A, B, and C variables, and they reflect quite different causal hypotheses.

Model 1, in which the brain variable is assumed to mediate the relations between age and cognition, is probably the most commonly assumed model. The relationship between brain and cognition is reversed in Model 2 as cognitive functioning is postulated to affect aspects of brain structure. This causal direction has become more plausible in recent years because of studies reporting changes in brain variables as a function of cognitively-demanding activities such as intensive study (e.g., Ceccarelli et al., 2009; Draganski et al., 2006), memory training (Envig et al., 2010), skill acquisition (Boyke et al., 2008; Ilg et al., 2008), and second-language learning (e.g., Mechelli et al., 2009). Model 3 portrays the possibility that the brain and cognition variables are only related to one another because they are each related to age. This is an important model to consider in age-heterogeneous samples because many brain and cognitive variables are related to age, and thus it is possible that at least some of their relations with one another are spurious, and attributable to their common relation with age.

Because confirmation of a prediction provides relatively weak evidence, whereas a pattern in which the results are not only consistent with implications of the target model but also inconsistent with implications of alternative models is more convincing, results with all three models will be considered when evaluating the plausibility of the models. However, it is important to note that simultaneous consideration of multiple models is not atheoretical or exploratory, but instead can be considered to provide a stronger test of the plausibility of a particular model. That is, because results consistent with a particular model merely indicate that the model survived an opportunity for falsification, the plausibility of the model relative to alternative models would be increased if the results were found to be inconsistent with the

implications of the other models. For example, Model 1 would be considered most plausible if the reduction implied by Model 1 (i.e., A-C > A-C.B, which means that the correlation between A and C is greater than the correlation between A and C after partialling or controlling B) is greater than the reduction implied by Model 2 (i.e., A-B > A-B.C), and the reduction implied by Model 3 (i.e., B-C > B-C.A).

The analyses require that correlations among the three variables, or R^2 values that can be used to derive the relevant correlations, were reported in the research article. Additional criteria to ensure meaningfulness of the results were that the sample size was at least 25, that the age range in a continuous sample was at least 25 years, and that cognition was assessed by sensitive performance tests rather than ratings, or with a global dementia screening test such as the Mini-Mental Status Exam (MMSE; Folstein et al., 1975). In some cases the results reported in the original article are slightly different than those reported here because the current estimates are derived from path analyses based on correlations with only three variables, whereas some studies included many brain and cognitive variables in the same analyses, some merely reported that paths from A to B and from B to C were both significantly different from zero, and some reported regression analyses and focused on ratios or proportions of variance before and after statistical control, rather than differences between correlations.

Results will be reported with all available combinations of brain variables and cognitive variables, rather than only those based on hypotheses derived from the past literature. One reason for this inclusive approach is that some hypotheses are based on the literature on focal lesions, and they may not be applicable to the gradual, and possibly diffuse, patterns associated with aging (cf. Raz & Kennedy, 2009). Another reason is that consideration of multiple combinations of brain and cognitive variables helps establish the baselines for differences between relations when there is little reason to expect a particular model to be correct.

Five questions relevant to the causal role of B on A–C relations will be examined on the basis of the available data on interrelations among age (A), measures of brain structure (B) based on volume, white-matter hyperintensities and diffusion tensor imaging, and various cognitive (C) variables.

- are the results of mediation analyses consistent with Model 1, in which B is a mediator of the A–C relation (is the correlation between age and cognition [A–C] reduced when the variability in the brain variable is held constant [A–C.B]; i.e., A– C > A–C.B?)?
- **2.** are the results of reverse mediation analyses consistent with Model 2, in which C is a mediator of the A–B relation (i.e., A–B > A–B.C?)?
- **3.** are the relations between brain and cognitive variables consistent with Model 3, such that the B–C relation is attenuated after control of age (i.e., is B–C > B–C.A?)?
- **4.** are the longitudinal changes in brain structure over a given interval correlated with the longitudinal changes in cognition (i.e., $[\Delta B \Delta C] > 0$), or are the values of B at one point in time correlated with longitudinal changes in C (i.e., $[B \Delta C] > 0$?)?
- 5. does the strength of the B–C relation, or of the Δ B– Δ C relation, vary as a function of age?

None of the information is definitive by itself, but inferences about causal relations among A, B, and C become more plausible as answers are obtained to each question, and supportive evidence accumulates. For example, the plausibility of Model 1 would be increased if the

results are consistent with positive answers to questions 1 and 4, and negative answers to questions 2 and 3. As noted earlier, question 5 is frequently neglected, and yet it is valuable in indicating whether there is a shift with age in either the B–C or Δ B– Δ C relation. A discovery that the relations are moderated by age would be consistent with the assumption that there are different neuroanatomical substrates of cognitive functioning at different ages. In contrast, lack of evidence of moderation would suggest that there may be nothing special about the period of older adulthood with respect to the qualitative nature of the B–C, or Δ B– Δ C, relations.

Regional Volume

Although shrinkage of brain volume with increased age is well-documented, reasons for age-related decreases in brain volume are not fully understood. Contrary to early views, loss of neurons does not appear to be the major factor contributing to the volume reduction (e.g., Fjell & Walhovd, 2010; Freeman et al., 2008; Morrison & Hof, 1997), but there could be shrinkage of the dendritic arbor and of cell bodies, decrease in synaptic density, loss of glial cells, reduction of myelination, and possibly decreases in vascularization. In part because of the uncertainties about the nature and causes of brain volume reduction, Van Petten (2004, p. 1395) characterized brain volume as "the crudest of neurobiological metrics". Raz and Kennedy (2009) also pointed out that the relationship between volume and cognitive performance can be difficult to interpret because "... if smaller volume means atrophy and loss of valuable neural elements, then it should predict poorer performance on cognitive tests. However, if increased regional volume reflects pathological processes, such as gliosis or failure to dispose of the unnecessary elements of the neural networks, then decreased volume is expected to go with better cognitive status (p. 51)." Despite the uncertainty about its nature, relations between brain volume measures and either age or measures of cognitive functioning have been reported in many studies.

Age Relations (AB)—Negative cross-sectional age relations with whole brain volume have been reported in a large number of studies (e.g., Abe et al., 2008; Adamson et al., 2010; Blatter et al., 1995; Carlesimo et al., 2010; Carne et al., 2006; Chee et al., 2009; Courchesne et al., 2000; DeCarli et al., 1999; 2005; Fotenos et al., 2005; 2008; Good et al., 2001; Hasan et al., 2007; Hutton et al., 2009; LeMaitre et al., in press; Marcus et al., 2007; Narr et al., 2007; Pieperhoff et al., 2008; Resnick et al., 2000; Schretlen et al., 2000; Seshadri et al., 2004; Taki et al., 2004; Van Petten et al., 2004; Wahlvold et al., 2005; Zimmerman et al., 2006; see Fjell & Walhovd, 2010, for a recent review). Global estimates of age differences across the entire period of adulthood range from about 0.2% to 0.5% per year, although there are clear regional variations. For example, volume reduction seems to be largest in the frontal and parietal lobes and least in the occipital lobe (e.g., Carne et al., 2006; Chee et al., 2009; 2010; DeCarli et al., 2005; Fjell et al., 2010; Fjell & Walhovd, 2010; Gonoi et al., 2010; Jernigan et al., 2001; Kalpouzos et al., 2009; Kennedy, Erickson et al., 2009; Lemaitre et al., in press; Raz et al., 2005; Sowell et al., 2003; Tisserand et al., 2002; Walhovd et al., 2005), and may be small in the medial temporal region and hippocampus until middle or late adulthood (e.g., Allen et al., 2005; Carlesimo et al., 2010; Fjell et al., 2010; Good, 2001; Grieve et al., 2005; Raz, et al., 2004; Sakamoto et al., 2007). Age relations may also be relatively small in the corpus callosum (e.g., Driesen & Raz, 1995; Fillipini et al., 2009; Head et al., 2005; Muller-Oehring et al., 2007; Sullivan et al., 2002).

There have also been reports of volume declines in longitudinal comparisons (e.g., Driscoll et al., 2009; Du et al., 2006; Enzinger et al., 2005; Firbank et al., 2007; Fjell, Walhovd et al., 2009; Walhovd et al., 2010; Fotenos et al., 2005; Jack et al., 2005; 2008; Lawrie et al., 2002; Lieberman et al., 2001; Liu et al., 2003; Pfefferbaum et al., 1998; Raz et al., 2004; 2005;

2008; 2010; Resnick et al., 2003; Rettmann et al., 2006; Rodrigue & Raz, 2004; Scahill et al., 2003; Schmidt et al., 2005; Sullivan et al., 2002). Moreover, a few studies have reported greater shrinkage in longitudinal comparisons than in cross-sectional comparisons (e.g., Du et al., 2006; Raz et al., 2005; Taki et al., 2009).

Although the negative age relations are nearly linear from early adulthood in gray matter volume (e.g., Abe et al., 2008; Allen et al., 2005; Bartzokis et al., 2001; Courchesne et al., 2000; Fotenos et al., 2005; Good et al., 2001; Grieve et al., 2005; Hasan et al., 2007; Kalpouzos et al., 2009; Michielse et al., 2010; Pieperhoff et al., 2005; Sowell et al., 2003; Taki et al., 2004; Terribilli et al., 2011), there may be little difference, or possibly even an increase, in white matter volume until the 40s or 50s, albeit with some across-region variation (e.g., Abe et al., 2008; Allen et al., 2005; Courchesne et al., 2000; Fotenos et al., 2007; Michielse et al., 2005; Goud et al., 2001; Grieve et al., 2005; Guttmann et al., 1998; Hasan et al., 2007; Michielse et al., 2010; Salat, Greve et al., 2009; Salat, Lee et al., 2009; Sowell et al., 2003; Sullivan et al., 2000; Taki et al., 2004; Walhvold et al., 2005; Westlye et al., 2010). A recent integrative analysis by Walhovd et al. (2009) on data combined across multiple samples involving a large number of adults across a wide age range (i.e., N > 880, ages 18 to 90+) revealed both linear and non-linear age trends, usually with accelerated shrinkage at older ages.

Cognition Relations (B–C)—Some reviews of the literature on volume-cognition relations have stressed the inconsistencies, and have suggested that it is not yet possible to reach definitive conclusions about the relations between brain volume and cognitive performance (e.g., Raz, 2000; Van Petten, 2004). Nevertheless, many studies have reported positive correlations (i.e., better cognitive performance associated with larger volumes) between performance in various cognitive tests and regional volume. For example, two recent meta-analyses have reported correlations of .33 (McDaniel, 2005) and .40 (Rushton & Ankney, 2009) between overall brain size and measures of general cognitive ability. Some studies have relied on crude measures based on CSF volume (e.g., Cook et al., 2002; Ikram et al., 2008; Rabbitt et al., 2006; Seshardi et al., 2004; van der Werf et al., 2001; Wickett et al., 2000; Willerman et al., 1991), or other indices of whole brain volumes (e.g., Aggarwal et al., 2010; Muller et al., 2009; Posthuma et al., 2002). Other studies have reported relations between regional volume measures and either general or specific measures of cognition (e.g., Adamson et al., 2010; Andreasen et al., 1993; Colom et al., 2006a; 2006b; 2009; Eckert et al., 2010; Flashman et al., 1997; Frangou et al., 2004; Haier et al., 2004; 2009; Hulshoff Pol et al., 2006; Ikram et al., 2010; Johnson et al., 2008; Jung & Haier, 2007; Kennedy, Rodrigue et al., 2009; Luders et al., 2009; MacLullich et al., 2002; Moffat et al., 2007; Oosterman et al., 2008; Paul et al., 2005; 2009; Schwartz et al., 2007; Soderlund et al., 2004; 2006; Staff et al., 2006; Thompson et al., 2001; Ullen et al., 2008; Wickett et al., 2000). There are also quite a few reports of relations between frontal lobe volume of either gray or white matter and measures of fluid intelligence or executive functioning (e.g., Chee et al., 2009; Colcombe et al., 2005; Duarte et al., 2006; Fine et al., 2009; Gong et al., 2005; Kramer et al., 2007; Nestor et al., 2010; Newman et al., 2007; Raz et al., 1998; 2008; Schretlen et al., 2000; Ziegler et al., 2008; Zimmerman et al., 2006). Volume-cognition relations have also been reported with other combinations of variables, including memory and medial temporal lobe volume (e.g., Andreasen et al., 1993; Brickman et al., 2006; Cardenas et al., 2009; Carey et al., 2008; Chen et al., 2010; Cohen et al., 2006; Hackert et al., 2002; Kalpouzos et al., 2009; Lye et al., 2006; Mungas et al., 2005; O'Brien et al., 1997; Oosterman et al., 2008; Yonelinas et al., 2007; Ystad, et. al, 2009), general information with frontal and temporal volume (Flashman et al., 1997), vocabulary knowledge with temporal lobe volume (Colom et al., 2009) and inferior parietal volume (Lee et al., 2007), and digit symbol speed with parietal volume (Flashman et al., 1997).

Nearly all of these relations have been in the direction of bigger volumes associated with better performance. However, there have been some exceptions, as negative volume-cognition relations have also been reported (e.g., Salat et al., 2002; van Petten, 2004; van Petten et al., 2004). Unfortunately, the reasons for these inconsistencies are not yet understood.

The most detailed specification of relations between regional volumes and cognitive performance at the current time is the Parieto-Frontal Integration Theory (P-FIT) model by Jung and Haier (2007). This model postulates a distributed pattern of regional influence, with frontal, parietal, and temporal volumes related to fluid intelligence. It has also been reported that the regional involvement is more extensive (i.e., more gray matter voxels) when the tests have higher loadings on a general factor (e.g., Colom et al., 2006a, 2006b).

Mediation (A–C.B)—A relatively large number of studies have reported mediation analyses with global or regional brain volume as the hypothesized mediator. Summary results for the three models in Figure 4 for the studies with relevant information are reported in Table 1.

Perhaps the most noteworthy characteristic of the results in Table 1 is that many of the results are more consistent with alternative models than with the meditation model that was the focus in the majority of the studies. For example, Raz et al. (1998) stated that "shrinkage of the prefrontal cortex mediates age-related increases in perseveration (p. 95)". However, the pattern in Table 1 indicates that the A–C reduction implied by Model 1 was .13, whereas the B–C reduction implied by Model 3 was .17, which is more consistent with a model postulating that the brain and cognitive variables are independently related to age. A similar pattern was evident in Head et al. (2002) as the A–C reduction of .15 was smaller than the B–C reduction of .22.

Head et al. (2002) also concluded that volume of the lateral pre-frontal cortex (PFC) mediates the relation between age and verbal working memory. Table 1 indicates that the reductions in the relevant relations were .20 for A–C, .06 for A–B, and .09 for B–C, which is consistent with the B-mediation interpretation. However, these results are apparently not very robust because the pattern was not replicated in similar comparisons reported in Raz et al. (1998) and Raz et al. (2000).

Brickman et al. (2006) claimed that age-related decline in neuropsychological functioning is in part mediated by an age-related reduction in relative frontal white matter. However, the entries in Table 1 indicate that the combination of left frontal white matter (B) and the switch 2 cognitive variable (C) had a larger, rather than smaller, age relation after control of B, which is inconsistent with mediation. Furthermore, when the list learning variable served as C and right frontal white matter as B, the A–C reduction of .09 was smaller than both the A–B (.17) and B–C (.11) reductions, and thus the results are more consistent with either reverse mediation or independence than with B mediation.

Schretlen et al. (2000) also suggested that their results were consistent with mediation of the relation of age on fluid intelligence through frontal lobe volume, but the results are actually more consistent with reverse mediation as the A–C reduction was .07 compared to .16 for the AB reduction, and .10 for the B–C reduction. Finally, Rabbitt et al. (2006) claimed that brain shrinkage accounted for large proportions of the age-related variance in a variety of cognitive tests, but the results in Table 1 indicate that with many variables the results were more compatible with the independent influences model. (The article did not report A–B correlations, and thus it was not possible to examine reverse mediation in this study.) Another study by Rabbitt and colleagues (Rabbitt, Mogapi et al., 2007) reported mediation

analyses with brain volume as one indicator of a "neuro" factor, with measures of cerebral blood flow and white matter lesions as the other indicators, but correlations were not reported to allow the plausibility of alternative models to be determined.

Mediation analyses have also been reported with extreme groups of young and old adults in a number of studies. For example, Head et al. (2008, 2009) summarized their results with path analysis models in which mediation was implied. However, none of these studies considered alternative models of the relations among the variables, and therefore the results are not easily interpreted.

The overall impression from the results in Table 1 is that there does not appear to be much support for the hypothesis that age-related reductions in brain volume mediate the age differences in various cognitive variables. This interpretation is reinforced by a more detailed examination of the 12 combinations (out of the 254 entries in Table 1) in which the results were consistent with the mediation model. The results with corpus callosum volume as the B variable and measures derived from reaction time tasks as the C variable are plausible because of the importance of the corpus callosum for rapid interregional communication. However, two of the combinations involved cerebellum volume as the B variable, and although those with the pursuit rotor task might be expected because of the motor requirement in the task, those with nonverbal working memory are not easily interpreted. Several B-C combinations involved total brain volume, or CSF volume, as the B variable, but there was little coherence among the C variables as they included measures of word memory, memory scanning, and verbal ability. A few combinations consistent with the mediation model involved measures of prefrontal volume as B, and measures of working memory or perseverations as C. However, as noted above, the Head et al. (2002) pattern with perseverations was not replicated in two other studies. The other B-C combinations with prefrontal measures as B were also not replicated as the pattern reported by Head et al. (2002) with a measure of non-verbal working memory as C was not found in studies by Kennedy, Rodrigue et al. (2009) or Raz et al. (2000), and the pattern reported by Gunning-Dixon and Raz (2003) with verbal working memory as C was not replicated in Brickman et al. (2006), Head et al. (2002), Kennedy, Rodrigue et al. (2009), or Raz et al. (2000).

Correlated Change (\Delta B-\Delta C)—Relatively few studies have been reported with longitudinal B and C data, and the correlated change results have been inconsistent. Significant correlations between the two types of change have been reported in at least four studies. For example, Kramer et al. (2007) found significant relations between changes in hippocampal volume and changes in episodic memory, and between changes in cortical gray matter and changes in executive functioning. Sullivan et al. (2002) also reported a correlation between decline in corpus callosum size and change in Stroop word reading time, and Schmidt et al. (2005) found that longitudinal change in normalized brain volume was correlated with longitudinal changes in memory and attention/speed. Furthermore, Murphy et al. (2010) recently found that shrinkage of the MTL regions over a 6-month period was associated with decline in various measures of memory over a 2-year period.

However, other studies did not find significant correlations, either between hippocampal volume and memory or other cognitive variables (e.g., Cohen et al., 2001; Du et al., 2006; Ylikoski et al., 2000), or between whole brain volume and working memory (Charlton et al., 2010). Although longitudinal data were available with both brain volume and cognitive functioning variables in a project by Raz et al. (2008), no correlations between brain changes and cognitive changes were reported because the individual differences in the change in the cognitive measures were not significantly different from zero.

Partial correlated change (B–\DeltaC)—Several studies have reported that smaller baseline hippocampus volume was associated with greater subsequent (e.g., Cardenas et al., 2009; Golomb et al., 1996; Tupler et al., 2007; Woodard et al., 2010), or prior, memory decline (Borghesani et al., in press; Persson et al., 2006). In addition, Prins et al. (2005) found that greater generalized brain shrinkage at baseline was associated with steeper subsequent decline in a composite measure of cognitive functioning, and Swan et al. (1998) and Tisserand et al. (2004) found relations between prior cognitive decline and brain volume or gray matter density at the second occasion. Finally, at least three studies found Δ B–C correlations between the magnitude of longitudinal change in the hippocampus or entorhinal cortex regions and memory or fluid cognition (e.g., Cohen et al., 2006; Raz et al., 2008; Rodrigue & Raz, 2004).

Moderation—It is often claimed that relations between regional volumes and cognitive functioning are stronger at older ages (e.g., Gunning-Dixon et al., 2009; Hedden & Gabrieli, 2004; Raz & Rodrigue, 2006; Van Petten, 2004). Indeed, there does appear to be a shift with age in the nature of the relation between medial temporal lobe volume or hippocampal volume and memory functioning. That is, at least two studies with young adults have found a negative relation, with larger volumes associated with poorer memory (i.e., Chantome et al., 1999; Foster et al., 1999; but see Rajah et al., 2010, for a contradictory finding), a large study with middle-aged adults found little relation (i.e., Cherbuin et al., 2009), but a positive correlation between memory and hippocampus volume has been reported in several studies with older adults (e.g., Adamson et al., 2010; Cardenas et al., 2009; Carey et al., 2008; Dickerson et al., 2004; Golomb et al., 1996; Hackert et al., 2002; Lye et al., 2006; Mungas et al., 2001, 2005; O'Brien et al., 1997; Yonelinas et al., 2007; Ystad et al., 2009). Raz et al. (1998) also noted that there was no correlation between memory and hippocampal volume among their total sample of adults from 18 to 77, but that the correlation was positive for the subset of adults between 60 and 77 years of age. The age moderation in this case may reflect qualitative differences in the B-C relations, as some of the positive relations at older ages may be a reflection of prodomal dementia status, whereas negative relations at young ages may be related to inefficient pruning of redundant neurons during development.

Moderation results with other B–C combinations have been less consistent. Zimmerman et al. (2006) found a stronger positive relation between lateral frontal volume and executive function performance at older ages than at younger ages. In contrast, an opposite pattern was reported by Kennedy and Raz (2009), in which several interactions were in the direction of stronger brain-cognition relations at younger ages. An earlier study by these same researchers (Kennedy & Raz, 2005) did not report direct interaction tests, but did mention that separate analyses in each age group revealed a PFC - mirror tracing relation only in older adults, and not in young and middle aged adults. Finally, Moffat et al (2007) reported an interaction of age and hippocampal volume on a spatial navigation task, with a significant relation only among the younger participants.

Only a few of the longitudinal studies have involved adults under about 60 years of age, and none of them reported analyses of the relation between age and the magnitude of the correlations between brain changes and cognitive changes.

Summary of Volume Relations—Many reports have been published of simple correlations among age, performance in cognitive and neuropsychological tests, and regional or total brain volume, and the robustness of these relations may have tempted researchers to infer the existence of causal relations among the variables. However, the analyses summarized in Table 1 revealed weak evidence for the mediation pattern, with no B–C combinations exhibiting the strongest pattern of complete reduction of the A–C relation after control of B, together with no reduction in the A–B relation after control of C, and no

reduction of the B–C relation after control of A. Some results have been interpreted as support for mediation, but in many cases the results were at least as consistent with alternative models of the relations among the variables. Furthermore, in a few cases in which the pattern was most consistent with the mediation pattern, the results were not replicated in different samples. The patterns with correlated changes were also inconsistent, although there do appear to be replicable relations between B at one point in time, and prior or subsequent change in C.

Although it has sometimes been assumed that the volume-cognition relations are stronger at older ages, the evidence for this pattern was weak, and primarily limited to relations between MTL volume and memory. At the present time the available evidence does not permit strong conclusions about the role of global or regional brain shrinkage as a cause of age-related declines in cognitive functioning.

White matter hyperintensities

Age-related white matter degradation has been of considerable interest because it might affect communication efficiency across different brain regions (e.g., Bartsokis, 2004; O'Sullivan et al., 2001). That is, because myelinated white matter pathways are critical for connections across different regions of the brain, and because myelin is associated with faster conduction velocity, the presence of white matter abnormalities could have important consequences for cognition by reducing effective connectivity.

White matter hyperintensities (WMH) are abnormal clusters of white matter manifested as increased signal intensity in T2-weighted MR images, and are assumed to reflect white matter damage (cf. Gunning-Dixon et al., 2009; Mayda et al., 2009). It has sometimes been suggested that the relations with age and with measures of cognition differ according to the location of the WMH. However, the results are not very consistent as more pronounced relations with age and cognitive measures are sometimes found in deep white matter lesions (e.g., Delano-Wood et al., 2008), and sometimes in peri-ventricular regions (DeGroot et al., 2002; Van Den Heuvel et al., 2007). Furthermore, WMH lesions in the two regions are often highly correlated (e.g., DeCarli, Fletcher et al., 2005; Vannorsdall et al., 2009).

Age Relations (A–B)—Increased age has been found to be associated with higher ratings of white matter hyperintensities, or greater white matter hyperintensity volume, in both cross-sectional (e.g., DeCarli et al., 1995; 1999; Gunning-Dixon & Raz, 2003; Hentschel et al., 2007; Kochunov et al., 2008; Rovaris et al., 2003; Schaivone et al., 2009; Vannorsdall et al., 2009; Ylikoski et al., 1993), and longitudinal comparisons (e.g., Chen et al., 2006; Firbank et al., 2007; Kramer et al., 2007; Raz et al., 2007; Sachdev et al., 2007; Schmidt et al., 2005). Although some measures of white matter hyperintensities have skewed distributions, nearly linear relations with age have been found either after eliminating data from individuals with very high values (e.g., DeCarli et al., 1995), or after transforming the data to minimize skew (Gunning-Dixon & Raz, 2003).

Cognition Relations (B–C)—Reviews of relations between white matter hyperintensities and cognition have been published by Ferro and Madureira (2002), Gunning-Dixon and Raz (2000), Gunning-Dixon et al. (2009), and Mayda et al. (2009). The reviews all note that while there are some negative findings, there are also many reports of significant relations with cognition. In particular, more white matter damage has been associated with poorer performance on speed tasks (e.g., Aggarawal et al., 2010; Bunce et al., 2010; Burns et al., 2005; De Groot et al., 2000; Dufouil et al., 2003; Eckert et al., 2010; Longstreth et al., 1996; Nebes et al., 2006; Prins et al., 2005; Rabbitt et al., 2007; Saczynski et al., 2008; Soderlund et al., 2006; Ylikoski et al., 1993), and on a variety of executive functioning tasks (e.g.,

Aggarawal et al., 2010; Au et al., 2006; Bunce et al., 2010; Carey et al., 2008; Cook et al., 2002; Delano-Wood et al., 2008; Gunning-Dixon & Raz, 2003; Kramer et al., 2007; Oosterman et al., 2008; Paul et al., 2005; Prins et al., 2005; Raz et al., 2007; Tullberg et al., 2004). Relations have also been reported between white matter lesions and measures of other types of cognition (e.g., Au et al., 2006; Burns et al., 2005; DeCarli et al., 1995; de Groot et al., 2000; 2002; Kramer et al., 2007; Nordahl et al., 2006; O'Brien et al., 2002; Petkov et al., 2004; Prins et al., 2005; Schiavone et al., 2009; Van Petten et al., 2004; Vannorsdall et al., 2009; Wright et al., 2008).

Mediation (A–C.B)—Results from studies with relevant correlations based on the models in Figure 4 are contained in Table 2. It can be seen that the overall pattern was mixed, although several B–C combinations were consistent with the mediation model (Model 1). For example, Gunning-Dixon and Raz (2003) concluded that that frontal WMH mediates the relation between age and perseverations, and the results of the analyses in Table 2 are consistent with this interpretation. That is, the reduction in the A–C relation after control of B of .11 was greater than both the A–B (.07) and B–C (.08) reductions. However, it should be noted that there were no other reports with the same combination of B and C variables, and thus the replicability of this result cannot be determined at this time.

Although Rabbitt, Scott et al. (2007) titled their article "White matter lesions account for all age-related declines in speed but not in intelligence," the results in Table 2 indicate that many different types of cognitive tests had similar relations, including speed, fluency, memory span, and vocabulary. These findings suggest that WMH influences are apparently not limited to particular types of cognitive tests.

Other B–C combinations have also been examined, but have yielded mixed results, and were often more consistent with independent age relations than with B mediation. This was also true in an earlier study with only incomplete correlation information as Ylikoski et al. (1993) found reduction of the B–C relation after control of age, with decreases from -.30 to -.07 for Block Design, .41 to .23 for Trail Making A, and .42 to .27 for Stroop Word Reading Time.

Correlated Change (\Delta B-\Delta C)—Several studies have reported that increases in WMH were correlated with decreases in cognition. However, the results are apparently not very specific to a particular type of cognition because correlations have been reported for quite different types of cognitive variables. For example, correlations with WMH changes have been reported with measures of executive functioning (e.g., Kramer et al., 2007; Vannorsdall et al., 2009), speed (Longstreth et al., 2005; Schmidt et al., 2005), memory (e.g., Schmidt et al., 2005; Vannorsdall et al., 2009), fluid intelligence (Raz et al., 2007), and crystallized (verbal) intelligence (e.g., Garde et al., 2005; Vannorsdall et al., 2009). Only one study could be found with a negative finding of no significant change correlation, in this case between change in WMH and change in working memory (Charlton et al., 2010).

Partial correlated change (B–\DeltaC)—At least five studies have reported correlations between WMH at one point in time and cognitive change. For example, Garde et al. (2000) reported a correlation with WAIS IQ, Swan et al. (1998) reported correlations with digit symbol, verbal fluency, and Benton Visual Retention, Prins et al. (2005) reported correlations with speed and executive functioning, and Van Den Heuvel et al. (2007) reported a correlation with measures of speed. In a sample of adults between 55 and 90 years of age with cognitive complaints at baseline, Jacobs et al. (in press) found that greater WMH at baseline was associated with more negative change in Stroop and Trail Making performance. However, very few significant correlations between change in white matter hyperintensities and cognition at a second assessment were found by Cook et al. (2004).

Moderation—A meta-analysis of WMH-cognition relations by Gunning-Dixon and Raz (2000) found little or no differences in the relations as a function of age. However, two studies have reported stronger WMH-cognition relations over age 60 (Vannorsdall et al., 2009), or age 65 (Au et al., 2006). No reports could be found in which relations of age were examined on $\Delta B-\Delta C$ correlations, or $B-\Delta C$ correlations.

Summary of WMH Relations—As with measures of brain volume as the B variable, many simple correlations between A–B and B–C have been reported in which WMH was the B variable. Only a limited number of mediation analyses have been reported, and few of them have yielded results most consistent with the mediation pattern. Several studies have reported correlated changes in B and C, or correlations of B at one point in time with change in C. Because the cognitive variables involved in these relations have been quite diverse, relations with WMH may be relatively general. Finally, very few moderation studies have been reported, and the results have been inconsistent. The overall pattern of results from the different types of evidence could be viewed as consistent with the possibility that age-related increases in white matter lesions are contributing to at least some of the age differences and changes in cognitive functioning, but this conclusion should be considered tentative at the current time.

White matter integrity with DTI

Measures derived from diffusion tensor imaging (DTI), such as fractional anisotropy (FA) and mean diffusivity (MD), are potentially more sensitive indicators of white matter integrity than measures based on white matter lesions because they may reflect microstructural differences that occur prior to actual lesion formation. Several DTI measures are available, and all reflect aspects of water diffusion. Water molecules normally spread (diffuse) in a direction parallel to the axon and myelin sheath, and thus the flow is generally anisotropic. However, with degradation in the fibers, and in the boundaries that constrain diffusion of water molecules, the diffusion becomes more heterogeneous (mean diffusivity increases), and the degree of anisotropy decreases (becomes more isotropic). DTI measures are affected by the organization of the fiber tracts, but within a tract higher values of mean diffusivity (MD) have been interpreted as indicating lower white matter integrity, whereas higher values of fractional anisotropy (FA), which is the fraction of total diffusion that is anisotropic, have been interpreted as reflecting higher white matter integrity. Axial diffusivity is postulated to be more related to axonal damage or changes in intracellular space, whereas radial diffusivity is postulated to be more related to myelin or glial cell damage, and it is noteworthy that several studies found age effects, and strong relations with cognition, on radial diffusivity (e.g., Burzynska et al., 2010; Davis et al., 2009; Georgio et al., 2010; Hasan et al., 2009; Madden, Spaniol et al., 2009; Michielse et al., 2010; Sullivan et al., 2006; Sullivan & Pfefferbaum, 2006; Westlye et al., 2010).

Age Relations (A–B)—Several studies have reported nearly linear negative crosssectional age relations in white matter integrity assessed with DTI (e.g., Abe et al., 2008; Ardekani et al., 2007; Bendlin et al., 2010; Carlesimo et al., 2010; Charlton et al., 2006; 2008; Georgio et al., 2010; Grieve et al., 2007; Hasan et al., 2007; 2009; Hsu et al., 2008; Hugenschmidt et al., 2008; Kennedy & Raz, 2009; Kochunov et al., 2008; Lebel et al., 2010; McLaughlin et al., 2007; Michielse et al., 2010; Perry et al., 2009; Pfefferbaum et al., 2005; Pfefferbaum & Sullivan, 2003; Rovaris et al., 2003; Salat, Tuch, Greve et al., 2005; Schiavone et al., 2009; Stadlbauer et al., 2008; Sullivan et al., 2010; Voineskos et al., in press; Westlye et al., 2010; Yoon et al., 2008; Zahr et al., 2008; see reviews in Madden, Bennett et al., 2009, and Sullivan & Pfefferbaum, 2006). A number of studies have reported that the age effects are primarily, or largest, in frontal regions (e.g., Bennett et al., 2009; Burzynska et al., 2010; Damoiseaux et al., 2009; Davis et al., 2009; Gold et al., 2010; Head

et al., 2004; Lebel et al., 2010; Lehmbeck et al., 2006; Madden et al., 2004; Madden, Spaniol et al., 2009; Michielse et al., 2010; O'Sullivan et al., 2001; Pfefferbaum et al., 2005; Salat, Tuch, Greve et al., 2005; Salat, Tuch, Hevelone et al., 2005; Schiavone et al., 2009; Sullivan et al., 2006; Vernooij et al., 2008; Voineskos et al., in press), although Kennedy and Raz (2009) reported nearly identical correlations between age and FA in the frontal lobe (-.60) and in the occipital lobe (-.59), Westlye et al. (2010) also found similar age correlations in different brain regions, and Salat et al. (2005) reported negative age relations in posterior peri-ventricular regions. Longitudinal decline over two years in white matter integrity has been reported with the MD measure (Charlton et al., 2010), and the FA measure (Barrick et al., 2010).

Cognition Relations (B–C)—As with other structural variables, some negative results regarding white matter integrity and cognition have been reported, but there are also many positive results (for recent reviews see Kennedy & Raz, 2009; and Madden, Bennett et al., 2009). For example, significant correlations of DTI variables have been reported with measures of general intelligence, executive functioning, working memory, speed, and various types of memory (e.g., Begre et al., 2009; Bendlin et al., 2010; Bohr et al., 2007; Bucor et al., 2008; Charlton et al., 2006, 2008, 2010; Chiang et al., 2009; Davis et al., 2009; Deary et al., 2006; Gold et al., 2007; 2010; Goldstein et al., 2009; Grieve et al., 2007; Haut et al., 2007; Madden et al., 2004; O'Sullivan et al., 2001; Sasson et al., 2010; Schiavone et al., 2009; Sullivan et al., 2006; 2010; Tuch et al., 2005; Turken et al., 2008; Vernooij et al., 2008; Yu et al., 2008; Zahr et al., 2008; Ziegler et al., 2010). Kennedy and Raz (2009) also noted some region-specific associations, as FA in anterior regions was more closely related to speed and working memory, FA in posterior regions with executive functions, and FA in the medial-temporal lobes with memory. However, somewhat different regional associations were reported in Davis et al. (2009), who found anterior FA related to executive functioning measures and posterior FA related to memory measures.

Mediation (A–C.B)—Table 3 contains results of comparisons of the models in Figure 4 from studies with all relevant correlations. It can be seen that only a few B–C combinations were available, and in none of them was the pattern consistent with that expected from Model 1. Charlton et al. (2008) claimed that MD was a mediator of age differences in working memory. However, the results in Table 3 indicate that although there was a reduction in A–C after control of B (i.e., .17), the reduction in B–C after control of A (i.e., . 21) was actually larger, and therefore the results are at least as consistent with a model postulating independent relations of age with white matter integrity and cognition as with a model in which white matter integrity mediates age-cognition relations.

A few studies have reported reduction of A–C after control of B. For example, Sullivan et al. (2001) found that age was no longer a significant predictor of finger tapping speed when FA from posterior regions was included in the analysis as a simultaneous predictor. In samples of only young and old adults, Gold et al. (2010), Madden, Spaniol et al. (2009) and Zahr et al. (2009) all found a reduction of the age-related variance in several cognitive measures when FA measures were controlled. However, the reported information in these studies was insufficient to determine whether the results were also consistent with alternative models of the relations among the variables. Voineskos et al. (in press) described a structural equation model implying FA mediation of age-related differences in several cognitive measures, but again there was insufficient information to determine whether the results were also consistent with alternative models.

Correlated Change (\Delta B - \Delta C)—Only one study was found in which correlations of longitudinal changes were reported in DTI measures. Charlton et al. (2010) found that change in working memory was correlated –.33 with change in mean diffusivity even after

controlling the variation associated with age. However, no correlations were found with the FA measure often used in other DTI studies.

Partial correlated change (B–\DeltaC)—The one study that could be located reporting partial correlated change, Persson et al. (2006), found lower DTI-derived measures at a second time point for participants who had declined the most in memory.

Moderation—In a study involving young adults and older adults, Madden et al. (2004) reported a significant interaction in which young and old adults had different relations of FA to reaction time (RT). However, a similar study by Gold et al. (2010) did not find an interaction of age on the FA-RT relations.

Several studies did not report direct tests of the interactions, but did mention that the DTI-cognition relations were stronger at older ages (e.g., Bucor et al., 2008; Carlesimo et al., 2010; Davis et al., 2009; Vannorsdall et al., 2009). Direct interaction tests were conducted by Kennedy and Raz (2009), but surprisingly, most of the significant interactions were in the direction of stronger relations at younger ages (e.g., higher internal capsule FA – WM correlation at younger ages, higher temporal FA – delayed recall at young and middle ages).

Summary of DTI Relations—As with the other brain structure variables, when B is represented with DTI measures assumed to reflect white matter integrity, many simple correlations exist between A–B and B–C, but the results from mediation analyses were not very consistent. Only a single study was located reporting correlated change, and although there was a significant correlation with the MD measure, it was not significant with the more common FA measure. Results of moderation analyses have been inconsistent, with some reports of stronger relations at older ages, but other reports of stronger relations at younger ages, and very few direct tests of interactions involving age. Based on these results, the evidence for a causal role of DTI on age-cognition relations must be considered relatively weak at the current time.

Conclusions

The literature reviewed above reveals considerable evidence of A–C, A–B and B–C relations with B represented by global or regional volume, white matter hyperintensities, or DTI measures postulated to reflect white matter integrity, and with C represented by a variety of different cognitive variables. Statistical mediation analyses were reported in some studies, and a number of them found a reduction in the A-C relation after control of the variation in B. However, alternative models were seldom examined to rule out other interpretations, and the systematic analyses of the three models in Figure 4 revealed very few combinations in which the B-as-mediator model was more consistent with the results than alternative models. Correlated change has been examined in a very limited number of studies, and because most involved only older adults for whom both types of changes may have been occurring for many years, they may not be very informative about causal relations. Results of moderation analyses in which B-C relations are examined as a function of age have been mixed. Although there are several reasons to expect stronger B-C relations at older ages, this pattern has not always been found, and the results have occasionally been in the opposite direction. There have apparently been no studies in which the $\Delta B - \Delta C$ relations were examined as a function of age.

In an extensive review of the relevant literature, Raz and Kennedy (2009) recently concluded that: "The search for the neuroanatomical basis of cognitive aging has so far yielded limited and somewhat contradictory results (p. 59)." A similar conclusion is implied by the results reviewed above as some relevant comparisons have very little data, and the

available results have not been very consistent. In light of the inconclusive results, it is worth considering how stronger inferences might eventually be possible. The following paragraphs therefore describe suggestions for future research based on limitations of current studies with respect to samples, variables, and analytical methods.

Several suggestions concern relatively basic aspects of methodology that, if neglected, can weaken the contribution of the research. For example, the samples in many studies have been relatively small, which means that confidence intervals around the correlations are large, and the power to detect differences in simple correlations and partial correlations is weak. Power depends on the values of all relevant correlations, but if A-C = .5 and B-C = .5, then detection of an A–C.B correlation of .3, corresponding to difference of .2 correlation units from the A–C correlation, will have power of only .17 with a sample of 20, .36 with a sample of 50, .62 with a sample of 100, and reaches .80 only with a sample of 150 (Calculations based on G-Power 3.1, Faul, Erdfelder,Buchner & Lang, 2009). Because neuroimaging is expensive, it is understandable that many studies have had small samples, but the cost of low power also needs to be recognized (see Fjell & Walhovd, 2010, for a similar observation).

Another concern is that the samples in many studies had a narrow age range, which precludes tests of age-related moderation of the B–C relations. Furthermore, when the samples consist of only older adults interpretations of the B–C relations may also be limited by the fact that both sets of variables could have been declining for decades prior to the period of observation, and may no longer reflect causal relations. In other words, relations between two variables after an extended period of change could be quite different from those apparent when one or both variables are just beginning to change. A related point is that some studies only compared extreme groups of young and old adults instead of adults throughout the entire range of adulthood. The use of extreme groups is more efficient for the detection of age differences than a continuous sample, but this practice inflates estimates of the age relations because variance associated with middle-aged adults is omitted, and it precludes analyses of the age at which possible shifts in B–C relations occur.

Four other methodological limitations are not as widely recognized as those just mentioned. First, in many studies cognitive functioning was assessed with a single variable from one task, and therefore the variable can be assumed to reflect a mixture of task-specific influences, influences of the construct of primary interest, and influences of measurement error. Because individual variables seldom exclusively and exhaustively reflect a single theoretical construct, it can be difficult to determine how much of a relation is attributable to influences. Furthermore, the presence of measurement error in individual variables means that they often have low reliability, which may attenuate relations of the variable with other variables.

Some researchers have attempted to deal with these problems by analyzing composite scores postulated to represent broader cognitive constructs. Aggregation of variables typically results in an increase in reliability, but composites can be formed from any combination of variables, and the constituent variables do not necessarily represent a single meaningful construct. An alternative approach to the problem is to investigate both convergent and discriminant validity with a measurement model based on factor analysis, and then use factor scores defined by the reliable shared variance among the variables in subsequent analyses. An advantage of a procedure such as this is that one can be confident that the variables represent a common construct that is distinct from other constructs. Unfortunately, this approach requires moderately large samples of individuals and a relatively large number of variables, which are not always feasible. Nevertheless, whenever variables are combined,

it is important to be sensitive to the possibility that they may not all represent the same construct, and to a similar extent at every age.

A second limitation of prior studies is that different types of structural variables have been considered separately even though there is evidence that they are not independent. Anatomical regions are often considered separately for a given type of brain variable because of the well-documented regional specialization. However, moderate correlations have often been reported between the same type of measure in different regions (e.g., Alexander et al., 2006; Bergfield et al., 2010; Brickman et al., 2007; 2008; DeCarli et al., 2005; Gunning-Dixon & Raz, 2003), which raises the possibility that it may be informative to consider both general and specific relations of brain structure, as in Ecker et al. (2009) with cortical thickness measures, and Penke et al. (2010) with FA measures of white matter integrity.

Figure 5 portrays the contrast between the typical approach (top panel), involving a relation between a brain structure characteristic in a single anatomical region and a single cognitive variable, and an alternative multivariate approach (bottom panel), involving relations between higher-order constructs which represent many variables, as well as relations between specific brain regions and cognitive variables. Data from a project by Kennedy, Rodrigue et al. (2009) can be used to illustrate the two approaches. The simple correlation between lateral prefrontal volume and fluid intelligence in this study was .38, which is similar to values reported in other studies (e.g., Raz et al., 1993;Schretlen et al., 2000). However, volumes of six other regions were also analyzed in this study, and each was correlated with the other volume measures and with various cognitive measures. Furthermore, a re-analysis of the data based on the reported correlations revealed that a latent brain volume construct had significant loadings from each regional volume measure, and that a latent cognitive construct had significant loadings from the fluid intelligence, verbal working memory, and nonverbal working memory variables. Of greatest interest in the current context was that the correlation between the two latent constructs was .44, and that none of the relations between specific regions and cognitive variables (including the relation between lateral prefrontal volume and fluid intelligence) was significant after taking the relation between the two constructs into consideration. At least in the Kennedy, Rodrigue et al. (2009) project, therefore, it appears that all of the brain-cognition relations were operating at a relatively general level, with no evidence of specific relations after taking the general relation into consideration. Penke and Deary (2010) recently made a similar point regarding the desirability of considering higher-order factors based on reanalyses of data from a study by Charlton et al. (2008).

Because multiple-level analyses allow both specific and general relations to be examined, analytical methods based on models such as that portrayed in Figure 5 are likely to be more informative than the bivariate analyses that have dominated earlier studies. Some researchers adopt a narrow focus because they are primarily interested in a particular brain region, or in measures from a cognitive task designed to isolate critical cognitive processes. However, unless the brain-cognition relations are examined in a broader context it is impossible to determine if those relations are truly specific, or are merely another manifestation of a broader phenomenon.

Third, many of the measures of longitudinal changes are not very reliable, which limits the magnitude of correlations with other variables, and can lead to weak correlations of changes for statistical, rather than substantive, reasons. Longitudinal change is sometimes assessed with a simple difference or residual score, which tend to have low reliability because reliability of difference scores is inversely related to the (frequently high) correlation between the scores across the two occasions. A more desirable alternative approach is to use

methods based on latent difference score or latent growth curve models which minimize measurement error when assessing change (cf. Ferrer & McArdle, 2010; Raz et al., 2005, 2008, 2010; Salthouse, 2010b).

And finally, as discussed earlier in this article, inferences about causal relations are often based on correlations, but simple correlations are very limited with respect to causal information. Moreover, when mediation results have been reported the importance of examining alternative models, and establishing that the results are replicable, has often been neglected. In addition, very few studies have involved longitudinal data to examine correlated changes, or have examined whether the brain-cognition relations vary as a function of age, and yet both types of information are relevant to the causal role of brain changes in cognitive changes. Because every methodological approach has limitations, inferences will generally be stronger when they are based on converging results from several different analytical methods rather than those from a single procedure.

Acknowledgments

This research was supported by NIA Grant R37AG024270. I would like to thank David Madden and Jason Steffener for their constructive comments on earlier drafts of this manuscript.

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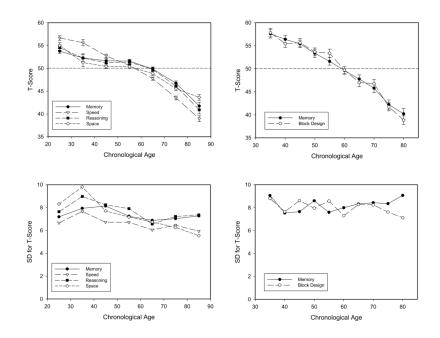


Figure 1.

Means (and standard errors) in different cognitive variables (top panel), and standard deviations corresponding to the means (bottom panel) as a function of age in cross-sectional comparisons in two projects. The panels on the left portray data from projects by Salthouse (2010a, 2010b, 2010c), and those on the right portray data from Ronnlund et al. (2005, 2006).

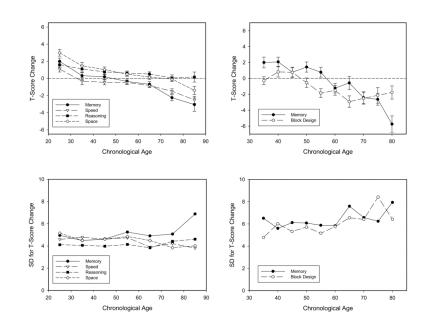


Figure 2.

Means (and standard errors) of longitudinal changes in different cognitive variables (top panel), and standard deviations corresponding to the means (bottom panel) as a function of age in two projects. The panels on the left portray data from projects by Salthouse (2010a, 2010b, 2010c), and those on the right portray data from Ronnlund et al. (2005, 2006).

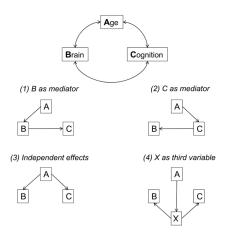


Figure 3.

Schematic illustration of correlations among age, brain, and cognitive variables (top), and four models of the relations among the variables which could produce the correlations.



Figure 4.

Schematic illustration of three models of the relations among the A, B, and C variables, and the relation (dotted line) hypothesized to be small or non-existent according to the model.

Figure 5.

Illustration of univariate (top panel) and multivariate (bottom panel) analyses of the relations among A, B, and C variables. Note that in the bottom panel both specific (dotted lines) and general (involving the circled B and C variables) relations are examined.

Table 1

Analyses with Volume as the Brain Structure Variable

С	В	A-C	A-C.B	A-B	A–B.C	$\mathbf{B}^{-\mathbf{C}}$	B-C.A
	Raz et al. (1993), N = 29, Ages 18 to 78	, Ages 1	8 to 78				
Cattell Culture Fair IQ	Total intracranial volume	72	65	40	19	.43	.22
	Dorso-lateral prefrontal	72	61	43	13	.51	.32
	Primary Somato-Sensory	72	74	09	47	19	37
	Inferior Parietal Lobule	72	72	22	23	.15	01
	Prefrontal White Matter	72	70	36	27	.32	60.
	Hippocampus	72	72	03	.03	.06	90.
Vocabulary	Total intracranial volume	.26	.36	40	46	.10	.23
	Dorso-lateral prefrontal	.26	.34	43	47	.03	.16
	Primary Somato-Sensory	.26	.24	09	04	20	18
	Inferior Parietal Lobule	.26	.32	22	30	.21	.28
	Prefrontal White Matter	.26	.30	36	38	01	60.
	Hippocampus	.26	.26	03	.04	.03	.04
	Raz et al. (1998), N = 95,	= 95, Ages 18 to 77	8 to 77				
WCST (Perseverations)	Prefrontal cortex	44.	.31	51	40	42	25
	Limbic cortex	44.	.41	18	10	23	17
	Visual cortices	.44	.39	35	28	29	16
	Inferior parietal lobule	<u>44</u> .	.45	15	19	.01	60.
Verbal Memory	Prefrontal cortex	39	41	51	52	.17	04
	Limbic cortex	39	38	18	15	.13	.07
	Visual cortices	39	41	35	38	.08	07
	Inferior parietal lobule	39	40	15	18	01	08
Nonverbal Memory	Prefrontal cortex	38	38	51	51	.20	.01
	Limbic cortex	38	37	18	15	.14	.08
	Visual cortices	38	35	35	32	.21	60.
	Inferior parietal lobule	38	38	15	16	.04	02
	Raz et al. (1998), $N = 95$, Ages 18	, Ages 1	8 to 77				
Priming	Prefrontal cortex	10	16	51	52	04	11
	Limbic cortex	10	10	18	18	.01	01

Visual cortices -10 -15 -36	С	в	A-C	A-C.B	A-B	A–B. C	$\mathbf{B}^{-\mathbf{C}}$	B-C.A
Inferior parietal lobule -10 -11 -15 -15 -16 -202 Prefrontal cortex -39 -34 -51 -46 27 Limbic cortex -39 -36 -11 -12 21 Visual cortices -39 -36 -16 -73 -12 21 Inferior parietal lobule -38 -27 -36 -11 29 211 Visual cortex -38 -27 -36 -36 -11 29 211 Visual cortex -38 -27 -36 -13 211 29 Visual cortex -38 -27 -36 -112 21 21 Visual cortices -36 -37 -161 -161 211 29 Inferior parietal lobule -38 -37 -161 -12 221 Visual cortices -36 -17 -56 -161 210 2		Visual cortices	10	15	35	36	09	13
Prefrontal cortex 39 34 51 46 27 Limble cortex 39 36 16 12 21 Visual cortices 39 36 16 12 21 Visual cortices 39 37 16 12 21 halt Prefrontal cortex 38 32 27 36 11 97 Visual cortices 38 32 37 16 17 29 Visual cortices 35 27 36 13 11 27 Visual cortices 35 27 36 13 11 13 11 13 $$		Inferior parietal lobule	10	11	15	15	02	04
Limble cortex 39 36 12 2.1 Visual cortices 39 37 16 36 111 Inferior parietal lobule 39 37 15 09 20 bul) Prefrontal cortex 38 27 61 91 2.0 Limbic cortex 35 37 16 11 2.0 Visual cortices 47 47 46 11 2.0 Visual cortices 47 41 20 17 2.1 Visual cortex 47 41 20 17 <	Working Memory (Verbal)	Prefrontal cortex	39	34	51	48	.27	60.
Visual cortices 39 40 35 36 $.11$ Inferior parietal lobule 39 37 15 36 $.11$ Prefrontal cortex 35 27 51 47 20 Limbic cortex 35 27 57 27 23 Visual cortices 35 32 16 27 23 Visual cortices 35 37 16 12 23 Visual cortices 35 37 16 17 23 Visual cortices 36 17 57 13 11 Raz et al. (2000), N = 68, Ages 21 to 80 41 56 13 13 Dorso-lateral prefrontal 47 46 31 29 32 Undate 47 41 26 17 29 18 Dorso-lateral prefrontal 47 46 06 17 29 Dorso-lateral prefrontal 47 46 09 19 29 Dorso-lateral prefrontal 47 41 29 29 29 Dorso-lateral prefrontal 47 46 06 17 29 Dorso-lateral prefrontal 47 41 29 29 29 Dorso-lateral prefrontal 47 41 29 29 29 Dorso-lateral prefrontal 28 20 29 29 29		Limbic cortex	39	36	18	12	.21	.15
Inferior partetal lobule 39 37 15 09 20 bal) Prefrontal cortex 35 27 51 61 2.9 Limbic cortex 35 27 51 12 2.9 Visual cortices 35 27 35 27 2.9 Visual cortices 35 34 15 13 2.1 Visual cortices 35 37 36 37 37 2.3 Visual cortices 35 34 55 64 17 2.9 Visual cortices 47 55 64 17 2.3 Dorso-lateral prefrontal 47 46 21 2.4 Pulamen 47 46 21 2.4 2.4 Pulamen 28 28 21 2.6 2.6 Pulamen 28 21 26 26 2.6 Pulamen 26 20 26		Visual cortices	39	40	35	36	.11	03
bal)Prefrontal cortex 35 27 51 47 29 Limbic cortex 35 32 18 12 21 Visual cortices 35 34 13 12 21 Visual cortices 35 34 13 13 21 Visual cortices 35 34 13 13 21 Pripocampus 47 55 64 68 23 Dorso-lateral prefrontal 47 46 09 23 Hippocampus 47 46 01 29 23 Dorso-lateral prefrontal 47 46 09 29 23 Dorso-lateral prefrontal 47 46 09 04 24 Dorso-lateral prefrontal 47 40 29 22 22 Dorso-lateral prefrontal 28 20 29 22 23 Dorso-lateral prefrontal 28 20 29 22 Hippocampus 28 20 29 22 Dorso-lateral prefrontal 28 20 29 23 Dorso-lateral prefrontal 28 20 29 22 Dorso-lateral prefrontal 28 29 29 23 Dorso-lateral prefrontal 28 29 29 23 Dorso-lateral prefrontal 28 29 20 23 Dorso-lateral prefrontal $-$		Inferior parietal lobule	39	37	15	09	.20	.16
Limbic cortex 35 32 16 12 2.7 3.3 Visual cortices 35 27 35 13 1.11 Raz et al. (2000), N = 68, Ages 22 to 80 47 55 64 68 2.3 Hippocampus 47 46 69 29 1.8 Dorso-lateral prefrontal 47 46 09 2.3 Hippocampus 47 46 09 2.9 Cerebellum 47 46 09 2.9 1.1 Putamen 47 46 09 17 2.4 Hippocampus 28 23 21 2.2 2.9 Hubu 28 29 29 2.2 2.4 Hippocampus 28 29 29 2.2 2.2 Hubu 28 29 29 2.2 2.4 2.4 Putamen 28 29 29 29 2.2 <td< td=""><td>Working Memory (NonVerbal)</td><td>Prefrontal cortex</td><td>35</td><td>27</td><td>51</td><td>47</td><td>.29</td><td>.14</td></td<>	Working Memory (NonVerbal)	Prefrontal cortex	35	27	51	47	.29	.14
Visual cortices 35 27 35 27 3.3 Inferior parietal lobule 35 34 15 13 11 Raz et al. (2000), N = 68, Ages 22 to 80 Dorso-lateral prefrontal 47 55 64 68 23 Hippocampus 47 56 64 68 23 Dorso-lateral prefrontal 47 46 31 29 11 Val 47 46 31 29 11 Dorso-lateral prefrontal 47 46 31 29 11 Utatamen 47 46 31 29 22 Hippocampus 28 20 31 29 23 Utatamen 28 20 29 22 23 Utatamen 28 29 29 22 23 Utatamen 28 29 29 22 23 Utatamen 26 20 29		Limbic cortex	35	32	18	12	.21	.16
Inferior parietal lobule 35 34 15 13 $.11$ Raz et al. (2000), N = 68, Ages 22 to 80 Dorso-lateral prefrontal 47 55 64 66 $.23$ Hippocampus 47 56 64 66 $.23$ Hippocampus 47 41 29 $.18$ Cerebellum 47 46 04 $.11$ Putamen 47 46 06 $.20$ $.18$ Mathoocampus 47 41 29 $.18$ $.21$ $.24$ Putamen 47 46 66 06 $.20$ $.24$ Hippocampus 28 16 26 $.22$ $.22$ $.22$ Utamen 28 20 29 $.22$ $.23$ $.24$ Utamen 26 10 26 $.22$ $.23$ $.24$ Dorso-lateral prefrontal 26 20 $.29$ $.24$ $.24$ $.24$ $.24$		Visual cortices	35	27	35	27	.33	.24
Raz et al. (2000), N = 68, Ages 22 to 80 Dorso-lateral prefrontal 47 55 64 68 23 Hippocampus 47 46 31 29 $.18$ Cerebellum 47 41 29 $.18$ Caudate 47 41 29 $.18$ Caudate 47 41 29 $.18$ Dorso-lateral prefrontal 47 46 06 $.16$ $.21$ Putamen 47 40 32 21 $.34$ Dorso-lateral prefrontal 28 16 06 $.23$ Hippocampus 28 22 22 $.32$ Dorso-lateral prefrontal 26 21 $.26$ $.24$ Hippocampus 28 23 22 $.32$ Dorso-lateral prefrontal 26 29 $.22$ $.32$ Hippocampus 28 20 $.22$ $.32$ Dorso-lateral prefrontal 26 20		Inferior parietal lobule	35	34	15	13	.11	90.
Dorso-lateral prefrontal -47 55 64 68 23 Hippocampus 47 47 46 29 $.18$ Hippocampus 47 41 29 $.17$ $.34$ Cerebellum 47 46 08 04 $.11$ Putamen 47 46 08 01 $.34$ Putamen 47 40 32 21 $.34$ Hippocampus 47 40 32 21 $.29$ Hippocampus 28 16 26 $.24$ $.26$ Cerebelum 28 20 22 $.22$ $.22$ $.23$ Putamen 26 21 26 $.26$ $.26$ $.26$ $.26$ $.26$ Hippocampus 26 21 29 $.22$ $.23$ $.21$ Value 26 21 26 $.26$ $.26$ $.26$ $.26$ $.26$ $.26$		Raz. et al. (2000), N = 68	8, Ages 22	2 to 80				
Hippocampus 47 46 31 -29 $.18$ Cerebellum 47 41 29 17 $.34$ Cerebellum 47 46 08 04 11 Putamen 47 46 08 04 11 Putamen 47 46 29 17 34 Putamen 47 46 28 21 29 Hippocampus 28 23 21 29 29 Uadate 28 20 29 22 29 Cerebelum 28 20 29 22 23 Dorso-lateral prefrontal 28 20 29 22 25 Putamen 26 19 29 22 29 29 Dorso-lateral prefrontal 26 19 29 22 29 29 Putamen 26 19 29 29 29 29 29 Dorso-lateral prefrontal 26 19 29 29 29 29 29 Hippocampus 26 19 29 29 29 29 29 29 Dorso-lateral prefrontal 26 19 29 29 29 29 29 Dorso-lateral prefrontal 26 19 29 29 29 29 29 Putamen 26 $-$	Working Memory (Verbal)	Dorso-lateral prefrontal	47	55	64	68	.23	10
Cerebelum 47 41 29 17 $.34$ Caudate 47 46 08 04 $.11$ Putamen 47 40 32 21 $.34$ Putamen 47 40 32 21 $.34$ Putamen 47 40 32 21 $.34$ Hippocampus 28 20 29 26 $.24$ Utamen 28 20 29 26 $.24$ $.26$ $.24$ Putamen 28 20 29 26 $.25$ $.25$ $.25$ Putamen 26 19 26 $.24$ $.26$ $.26$ $.26$ $.26$ $.26$ $.27$ Hippocampus 26 19 21 26 $.26$ $.23$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$ $.26$		Hippocampus	47	46	31	29	.18	.04
Caudate 47 46 08 04 $.11$ Putamen 47 46 08 04 $.11$ Putamen 47 40 32 21 $.34$ Hippocampus 28 16 61 $.29$ Hippocampus 28 21 26 $.24$ Cerebelum 28 20 22 $.22$ $.24$ Putamen 28 20 29 22 $.24$ Dorso-lateral prefrontal 28 20 22 $.23$ $.24$ Hippocampus 28 20 29 22 $.23$ Dorso-lateral prefrontal 26 19 26 $.25$ $.25$ Hippocampus 26 19 20 $.23$ $.26$ $.26$ Dorso-lateral prefrontal 26 19 22 $.27$ $.26$ $.25$ Hippocampus 26 10 22 $.27$ $.26$ $.26$		Cerebellum	47	41	29	17	.34	.24
Putamen 47 40 32 -2.1 $.34$ bal)Dorso-lateral prefrontal 28 16 61 $.29$ Hippocampus 28 23 31 26 $.24$ Cerebelum 28 20 29 $.22$ $.32$ Caudate 28 20 29 22 $.32$ Putamen 28 20 29 22 $.32$ Dorso-lateral prefrontal 26 19 66 $.39$ Utamen 26 19 29 20 $.31$ Dorso-lateral prefrontal 26 19 62 $.32$ Dorso-lateral prefrontal 26 19 29 $.20$ Dorso-lateral prefrontal 26 19 29 $.20$ Dorso-lateral prefrontal 26 19 29 $.23$ Hippocampus 26 19 29 $.29$ $.39$ Utamen 26 117 29 $.20$ $.39$ Putamen 26 17 32 25 $.35$ Head et al. (2002), N = 68, Ages 22 to 80 64 67 $.28$ Lateral Prefrontal 50 49 17 23 Lateral Prefrontal 50 46 13 $.15$ Putamen 50 46 13 23 Putamen 50 46 13 $.27$ Putamen 50 47 $-$		Caudate	47	46	08	04	H.	.08
bal)Dorso-lateral prefrontal 28 16 61 29 Hippocampus 28 23 31 61 29 Hippocampus 28 20 29 26 24 Cerebelum 28 20 29 26 $.21$ Putamen 28 20 29 26 $.21$ Putamen 28 21 08 02 $.23$ Putamen 28 20 19 66 $.23$ Dorso-lateral prefrontal 26 19 66 $.23$ Hippocampus 26 19 26 $.20$ $.39$ Dorso-lateral prefrontal 26 16 26 $.23$ Huppocampus 26 16 29 $.23$ Dorso-lateral prefrontal 26 16 26 $.23$ Huppocampus 26 16 29 $.23$ Duramen 26 17 29 $.26$ $.23$ Head et al. (2002), N = 68, Ages 22 to 80 06 17 13 $.27$ Lateral Prefrontal 50 46 17 23 $.28$ Lateral Prefrontal 50 46 23 $.23$ Putamen 50 46 17 23 $.23$ Putamen 50 47 29 23 $.27$ Putamen 50 46 31 23 $.27$ Putamen <th< td=""><td></td><td>Putamen</td><td>47</td><td>40</td><td>32</td><td>21</td><td>.34</td><td>.23</td></th<>		Putamen	47	40	32	21	.34	.23
Hippocampus 28 23 31 26 $.24$ Cerebellum 28 20 29 22 $.32$ Caudate 28 27 08 02 $.31$ Putamen 28 23 25 $.32$ Dorso-lateral prefrontal 26 19 25 $.23$ Dorso-lateral prefrontal 26 19 26 $.23$ Dorso-lateral prefrontal 26 10 29 $.23$ Dorso-lateral prefrontal 26 117 29 $.23$ Duranen 26 17 29 29 39 Lateral Prefrontal 50 917 23 27 Lateral Prefrontal 50 49 17 23 27 Putamen 50 46 17 23 27 Putamen 50 46 17 23 27 Derebellum 50 46 13 23 23 Duranen 50	Working Memory (NonVerbal)	Dorso-lateral prefrontal	28	16	64	61	.29	.15
Cerebelium 28 20 22 22 22 22 22 Caudate 28 21 08 02 21 Putamen 28 23 22 25 35 Dorso-lateral prefrontal 26 19 64 62 23 Hippocampus 26 19 64 62 29 Hippocampus 26 10 31 26 29 Cerebelium 26 17 29 20 39 Putamen 26 17 23 25 35 Head et al. (2002), N = 68, Ages 22 to 80 64 67 28 Lateral Prefrontal 50 49 17 13 15 Putamen 50 49 17 23 16 Caudate 50 49 17 13 15 Putamen 50 49 17 23 28 Crebellum 50 46 17 23 27		Hippocampus	28	23	31	26	.24	.17
Caudate 28 27 08 02 $.21$ Putamen 28 23 32 $.2.5$ $.35$ Putamen 26 19 64 62 $.23$ Hippocampus 26 19 64 62 $.23$ Hippocampus 26 16 29 $.20$ $.39$ Cerebelum 26 16 29 20 $.39$ Caudate 26 17 29 20 $.39$ Putamen 26 17 23 $.27$ Head et al. (2002), N = 68, Ages 22 to 80 67 $.28$ $.36$ Lateral Prefrontal 50 49 17 23 $.27$ Putamen 50 46 67 $.28$ $.27$ Cerebelum 50 46 23 $.27$		Cerebellum	28	20	29	22	.32	.26
Putamen 28 23 25 25 $.35$ Dorso-lateral prefrontal 26 19 64 62 $.23$ Hippocampus 26 20 31 26 $.25$ Cerebellum 26 16 29 20 $.39$ Caudate 26 17 29 20 $.39$ Putamen 26 17 29 26 $.37$ Head et al. (2002), N = 68, Ages 22 to 80 49 67 $.28$ Lateral Prefrontal 50 49 13 $.15$ Putamen 50 49 13 $.15$ Putamen 50 46 67 $.28$ Carbalte 50 46 13 $.12$		Caudate	28	27	08	02	.21	.20
Dorso-lateral prefrontal 26 19 64 62 $.23$ Hippocampus 26 26 31 62 $.23$ Cerebellum 26 16 $.29$ 20 $.39$ Caudate 26 17 29 20 $.39$ Caudate 26 17 22 06 $.12$ Putamen 26 17 32 25 $.35$ Head et al. (2002), N = 68, Ages 22 to 80 64 67 $.28$ Lateral Prefrontal 50 94 67 $.28$ Caudate 50 49 17 23 23 Putamen 50 46 67 $.28$ Putamen 50 46 13 15 Putamen 50 47 23 $.27$		Putamen	28	23	32	25	.35	.14
Hippocampus 26 20 31 26 $.25$ Cerebellum 26 16 $.29$ $.20$ $.39$ Caudate 26 16 $.29$ 20 $.39$ Caudate 26 17 29 20 $.39$ Putamen 26 17 32 25 $.12$ Head et al. (2002), N = 68, Ages 22 to 80 32 25 $.35$ Lateral Prefrontal 50 94 67 $.28$ Caudate 50 49 17 23 $.27$ Putamen 50 46 13 23 $.27$ Cerebellum 50 47 29 22 $.25$	Pursuit Rotor, Day 1	Dorso-lateral prefrontal	26	19	64	62	.23	60.
Cerebellum 26 16 29 20 $.39$ Caudate 26 26 26 05 $.12$ Putamen 26 17 32 25 $.35$ Head et al. (2002), N = 68, Ages 22 to 80Lateral Prefrontal 50 54 67 $.28$ Lateral Prefrontal 50 49 17 13 $.15$ Putamen 50 49 17 23 $.27$ Putamen 50 46 23 $.23$ Cerebellum 50 47 23 $.25$		Hippocampus	26	20	31	26	.25	.19
Caudate 26 25 08 05 $.12$ Putamen 26 17 32 25 $.35$ Head et al. (2002), N = 68, Ages 22 to 80Lateral Prefrontal 50 54 64 67 $.28$ Caudate 50 49 17 13 $.15$ Putamen 50 46 23 $.27$ Creebellum 50 47 29 $.25$		Cerebellum	26	16	29	20	.39	.34
Putamen 26 17 32 25 $.35$ Head et al. (2002), N = 68, Ages 22 to 80Lateral Prefrontal 50 54 67 $.28$ Caudate 50 49 17 13 $.15$ Putamen 50 46 23 $.27$ Crebellum 50 47 29 $.25$		Caudate	26	25	08	05	.12	.10
Head et al. (2002), N = 68, Ages 22 to 80Lateral Prefrontal 50 54 67 $.28$ Caudate 50 49 17 13 $.15$ Putamen 50 46 31 23 $.27$ Cerebellum 50 47 29 22 $.25$		Putamen	26	17	32	25	.35	.29
Lateral Prefrontal 50 54 64 67 .28 Caudate 50 49 17 13 .15 Putamen 50 46 31 23 .27 Cerebellum 50 47 29 22 .25		Head et al. (2002), N = 68	8, Ages 2	2 to 80				
50491713 .15 50463123 .27 	Working Memory (Verbal)	Lateral Prefrontal	50	54	64	67	.28	06
50463123 .27 50472922 .25		Caudate	50	49	17	13	.15	.08
50472922 .25		Putamen	50	46	31	23	.27	.14
		Cerebellum	50	47	29	22	.25	.13

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0		A_C	A_C.B	A_B	A-B.C	B_C	B_C.A
	1) :		1)	
Η	Hippocampus	50	53	20	28	02	14
	Visual Cortex	50	50	.03	.07	.05	.08
Working Memory (NonVerbal)	Lateral Prefrontal	24	04	64	59	.34	.25
0	Caudate	24	20	17	10	.30	.27
Η	Putamen	24	18	31	27	.24	.18
0	Cerebellum	24	22	29	28	.12	.05
Η	Hippocampus	24	20	20	16	.22	.18
	Visual Cortex	24	24	.03	.05	.07	.08
WCST Composite (Perseverations) I	Lateral Prefrontal	.45	.30	64	56	43	21
0	Caudate	.45	.42	17	07	25	20
I	Putamen	.45	.40	31	23	29	18
0	Cerebellum	.45	.43	29	25	20	08
I	Hippocampus	.45	.43	20	16	17	09
	Visual Cortex	.45	.45	.03	60.	10	13
Tower Hanoi, Time/Move Blkl I	Lateral Prefrontal	.57	.66	64	71	28	.13
	Caudate	.57	.56	17	14	13	04
I	Putamen	.57	.56	31	27	22	06
0	Cerebellum	.57	.60	29	36	08	II.
ł	Hippocampus	.57	.54	20	06	28	21
	Visual Cortex	.57	.57	.03	.12	-00	13
Tower Hanoi, Time/Move Blk 2	Lateral Prefrontal	.51	.44	64	60	39	06
0	Caudate	.51	.49	17	10	19	12
I	Putamen	.51	.47	31	23	27	14
	Cerebellum	.51	.52	29	30	13	.02
I	Hippocampus	.51	.52	20	24	04	.07
	Visual Cortex	.51	.51	.03	.08	05	08
	Head et al. (2002), $N = 68$, Ages 22 to 80	, Ages 22	2 to 80				
Tower Hanoi, Moves/Blk, Blk 1	Lateral Prefrontal	.33	.15	64	58	38	23
0	Caudate	.33	.32	17	15	10	05
I	Putamen	.33	.33	31	31	11	01
0	Cerebellum	.33	.34	29	30	08	.02

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H Vv V V V V C C C C C C C C C C C C C C	Hippocampus Visual Cortex	.33					
v Tower Hanoi, Moves/Blk, Blk 2 L ¹ C C P P H H C C C C C C C C C C C C C N H H N V V V V V V V V V V V V V V V V	isual Cortex		.28	20	12	29	24
Tower Hanoi, Moves/Blk, Blk 2 La C: C: Pt H: H: Vv Vv Vv Vv Vv Vv Vv Vv Vv Vv Vv Vv Vv		.33	.33	.03	.01	90.	.05
(lac (lac	Lateral Prefrontal	.11	01	64	63	18	14
(lac (lac	Caudate	H.	11.	17	17	03	01
(lac	Putamen	.11	.08	31	30	12	09
aal) aal)	Cerebellum	.11	.12	29	29	01	.02
(lac)	Hippocampus	.11	.15	20	22	.15	.18
(lac	Visual Cortex	.11	.11	.03	.04	09	09
aal)	Gunning-Dixon & Raz (2003), $N = 139$, Ages 50 to 81	V = 139, <i>i</i>	Ages 50 to	81			
(lac	Prefrontal	27	19	29	22	.33	.27
oal)	Fusiform gyrus	27	21	20	12	.33	.29
	Prefrontal	.27	.20	29	23	30	24
	Fusiform gyrus	.27	.22	20	13	30	26
	Kennedy, Rodrigue et al. (2009), $N = 169$, Ages 18 to 80	N = 169,	Ages 181	o 80			
	Lateral Prefrontal cortex	37	34	58	57	.25	.05
Ō	Orbital frontal	37	33	35	31	.22	.10
Pr	Prefrontal white matter	37	32	33	28	.25	.15
H	Hippocampus	37	35	30	27	.18	.08
C	Caudate	37	36	29	27	.15	.05
Fu	Fusiform gyrus	37	36	41	40	.18	.03
V.	Visual cortex	37	36	12	08	.15	.11
Kennedy	Kennedy, Rodrigue et al. (2009), $N = 169$, Ages 18 to 80	N = 169,	Ages 181	io 80			
Working Memory (NonVerbal) L6	Lateral Prefrontal cortex	34	26	58	54	.29	.12
Ō	Orbital frontal	34	29	35	30	.25	.15
Pr	Prefrontal white matter	34	28	33	27	.28	.19
H	Hippocampus	34	29	30	24	.25	.17
C	Caudate	34	32	29	27	.15	.06
Fu	Fusiform gyrus	34	30	41	38	.23	II.
N.	Visual cortex	34	33	12	07	.16	.13
Fluid Intelligence L ₂	Lateral Prefrontal cortex	54	48	58	53	.38	.10
Ō	Orbital frontal	54	49	35	26	.31	.15

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	В	A-C	A-C.B	A-B	A-B.C	B-C	B-C.A
	Prefrontal white matter	54	48	33	21	.33	.19
	Hippocampus	54	51	30	23	.26	.12
	Caudate	54	53	29	26	.20	.05
	Fusiform gyrus	54	48	41	32	.34	.15
	Visual cortex	54	52	12	02	.20	.16
Priming Training	Lateral Prefrontal cortex	.38	.35	58	57	25	04
	Orbital frontal	.38	.39	35	36	12	.02
	Prefrontal white matter	.38	.39	33	34	11	.02
	Hippocampus	.38	.37	30	28	15	04
	Caudate	.38	.38	29	30	10	.01
	Fusiform gyrus	.38	.33	41	37	25	11
	Visual cortex	.38	.37	12	10	10	06
Priming Repeated	Lateral Prefrontal cortex	.50	.48	58	57	31	03
	Orbital frontal	.50	.52	35	39	12	.07
	Prefrontal white matter	.50	.52	33	37	11	.07
	Hippocampus	.50	.50	30	29	16	01
	Caudate	.50	.49	29	27	18	04
	Fusiform gyrus	.50	.48	41	38	25	06
	Visual cortex	.50	.48	12	03	19	15
Priming Novel	Lateral Prefrontal cortex	.47	.46	58	57	29	02
	Orbital frontal	.47	.49	35	38	12	.05
	Prefrontal white matter	.47	.48	33	34	14	.02
	Hippocampus	.47	.46	30	28	18	05
	Caudate	.47	.49	29	32	09	.06
	Fusiform gyrus	.47	44.	41	37	26	08
	Visual cortex	.47	.46	12	06	16	12
	Schretlen et al. (2000) , N = 112, Ages 20 to 92	112, Age	s 20 to 92				
Fluid Intelligence	Frontal volume	52	45	28	12	.37	.27
	Tisserand et al. (2000) , N = 61, Ages 21 to 81	: 61, Ages	: 21 to 81				
Word Recall Immediate	Total brain volume	33	07	78	73	39	.22
	Third ventricle	33	27	.61	.59	26	08

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	Hippocampus	33	37	32	36	00.	12
	Parahippocampal gyrus	33	28	42	38	.24	.12
Word Recall Delayed	Total brain volume	28	06	78	75	.33	.19
	Third ventricle	28	19	.61	.58	26	12
	Hippocampus	28	31	32	35	00.	10
	Parahippocampal gyrus	28	24	42	40	.20	.10
Stroop Control Time	Total brain volume	.41	.27	78	75	39	12
	Third ventricle	.41	.33	.61	.57	.33	.11
	Hippocampus	.41	.33	32	21	36	27
	Parahippocampal gyrus	.41	.31	42	32	37	24
Stroop Conflict Time	Total brain volume	.58	.49	78	74	50	09
	Third ventricle	.58	.49	.61	.53	45	.15
	Hippocampus	.58	.58	32	31	20	02
	Parahippocampal gyrus	.58	.58	42	42	24	.01
Memory Scanning 1 item	Total brain volume	.81	69.	78	64	69	16
	Third ventricle	.81	.75	.61	.46	.56	.14
	Hippocampus	.81	LL.	32	06	37	20
	Parahippocampal gyrus	.81	.75	42	14	46	23
Memory Scanning 3 items	Total brain volume	.53	.24	78	67	56	28
	Third ventricle	.53	.49	.61	.58	.36	90.
	Hippocampus	.53	.50	32	25	26	11
	Parahippocampal gyrus	.53	.46	42	32	36	18
	Cook et al. (2002), $N = 43$, Ages 60 to 83	3, Ages 6	0 to 83				
Trail Making B	Global atrophy	.57	.42	.47	.27	.51	.33
Shipley Abstraction	Global atrophy	57	54	.47	.43	32	07
	Walhovd et al. (2004), $N = 54$, Ages 20 to 88	54, Ages	20 to 88				
Memory - 5 minutes recall	Cortical volume	63	62	85	85	.54	.01
	White matter volume	63	62	52	51	.34	.02
	Hippocampus	63	53	50	34	.47	.23
Memory - 30 minutes recall	Cortical volume	64	65	85	85	.54	01
	White matter volume	64	64	52	51	.34	.01

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C	В	A-C	A-C.B	A-B	A-B.C	B-C	B-C.A
	Hippocampus	64	56	50	37	44.	.18
Memory – multi-week recall	Cortical volume	57	37	85	80	.55	.15
	White matter volume	57	52	52	46	.37	.11
	Hippocampus	57	38	50	26	.57	.40
	Walhovd et al. (2005), $N = 71$, Ages 21 to 88	71, Ages	21 to 88				
Performance IQ	Cortical volume	74	57	65	41	.63	.29
Verbal IQ	Cortical volume	15	90'-	65	64	.18	.11
	Walhovd & Fjell (2007), N =	= 71, Age	= 71, Ages 20 to 88				
Performance IQ	Cortical gray matter	67	52	78	69	.60	.17
	White matter	67	62	50	41	.41	.12
Verbal IQ	Cortical gray matter	14	06	78	77	.15	.07
	White matter	14	13	50	50	.08	.01
Mean Reaction Time	Cortical gray matter	.15	12	78	76	25	22
	White matter	.15	.10	50	49	15	-00
	Brickman et al. (2006), $N = 199$, Ages 21 to 79	199, Age	s 21 to 79				
Choice Reaction Time	L. frontal rel. white matter	.42	.42	18	17	09	02
	R. frontal rel. white matter	.42	.37	35	28	28	16
	L. temporal rel. white matter	.42	.43	18	21	02	.06
	R. temporal rel. white matter	.42	.43	22	23	07	.03
Digits Forward	L. frontal rel. white matter	20	20	18	18	.05	.02
	R. frontal rel. white matter	20	20	35	35	.07	00.
	L. temporal rel. white matter	20	20	18	18	.04	00.
	R. temporal rel. white matter	20	20	22	22	.06	.02
Digits Backward	L. frontal rel. white matter	33	33	18	17	.08	.02
	R. frontal rel. white matter	33	31	35	33	.16	.05

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С	В	A-C	A-C.B	A-B	A-B.C	B-C	B-C.A
	L. temporal rel. white matter	33	34	18	18	00.	06
	R. temporal rel. white matter	33	32	22	20	.12	.05
List learning	L. frontal rel. white matter	48	43	18	02	.34	.29
	R. frontal rel. white matter	48	37	35	18	.44	.33
	L. temporal rel. white matter	48	47	18	14	.15	.07
	R. temporal rel. white matter	48	46	22	21	.20	.11
Verbal interference	L. frontal rel. white matter	62	62	18	15	.14	.04
	R. frontal rel. white matter	62	59	35	28	.29	.10
	L. temporal rel. white matter	62	62	18	17	.12	.01
	R. temporal rel. white matter	62	62	22	21	.15	.02
Fluency (FAS)	L. frontal rel. white matter	10	10	18	18	.02	00.
	R. frontal rel. white matter	10	05	35	34	.17	.15
	L. temporal rel. white matter	10	10	18	18	.03	.01
	R. temporal rel. white matter	10	07	22	21	.17	.15
Fluency (Animals)	L. frontal rel. white matter	51	51	18	17	.10	.01
	R. frontal rel. white matter	51	49	35	31	.24	.08
	L. temporal rel. white matter	51	49	18	12	.18	.10
	R. temporal rel. white matter	51	48	22	13	.24	.15
Attention Switching 1	L. frontal rel. white matter	.48	.48	18	19	07	.02
	R. frontal rel. white matter	.48	.47	35	33	20	04

С	В	A-C	A-C.B	A-B	A-B.C	B-C	B-C.A
	L. temporal rel. white matter	.48	.51	18	27	.05	.16
	R. temporal rel. white matter	.48	.47	22	19	15	.05
Attention Switching 2	L. frontal rel. white matter	.57	.59	18	28	.02	.15
	R. frontal rel. white matter	.57	.55	35	31	25	07
	L. temporal rel. white matter	.57	.59	18	25	02	.10
	R. temporal rel. white matter	.57	.58	22	27	07	.07
M	Muller-Ochring et al. (2007) , N = 37, Ages 26 to 79	V = 37, A	ges 26 to 7	6/			
Precedence in Global Local RT	Corpus callosum genu	.32	.21	43	36	34	24
Interference in Global Local RT	Corpus callosum genu	.28	60.	43	30	48	42
Resp. Conflict in Global Local RT	Corpus callosum genu	.31	.15	43	33	43	35
	Rabbitt et al. (2006), N = 62, Ages 69	52, Ages	69 to 85				
Culture Fair IQ	CSF Volume	45	42	NA	NA	28	05
AH4-1 IQ	CSF Volume	46	35	NA	NA	40	22
AH4-2 IQ	CSF Volume	48	45	NA	NA	30	06
Memory for People	CSF Volume	39	30	NA	NA	33	18
Memory for Objects in a Circle	CSF Volume	61	46	NA	NA	37	12
Visual Search	CSF Volume	37	25	NA	NA	35	23
Letter-Digit Substitution	CSF Volume	42	23	NA	NA	46	35
Category fluency	CSF Volume	35	34	NA	NA	42	34
Brixton	CSF Volume	.40	.28	NA	NA	.37	.23
Rey-Osterrieth Figure	CSF Volume	52	44	NA	NA	39	16
	Charlton et al. (2010) , N = 1	104, Ages	= 104, Ages 50 to 90				
Immediate memory	Hippocampus	57	56	19	15	.16	.06
	Whole brain volume	57	65	53	63	.19	16
	Chee et al. (2009), $N = 24$	= 248, Ages 55 to 86	55 to 86				
Attention	Right white matter	21	17	41	39	.17	60.
Speed	Right white matter	41	36	41	36	.27	.12
Verbal Memory	Right white matter	22	15	41	38	.23	.16

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Visual Memory	Right white matter	22	17	41	39	.19	.11
Executive functions	Right white matter	30	22	41	36	.29	.19
Language	Right white matter	22	19	41	40	.15	.07
	Chen et al. (2010), $N = 147$, Ages 55 to 83	: 147, Ages 5	55 to 83				
Memory – RAVLT Immediate	L Hippocampus	15	15	31	31	90.	.01
	R Hippocampus	15	11	36	34	.16	.12
Memory – RAVLT Delayed	L Hippocampus	05	07	31	31	05	07
	R Hippocampus	05	05	36	36	.03	.01
Memory – Visual Repr. Immed.	L Hippocampus	32	30	31	29	.15	.06
	R Hippocampus	32	29	36	33	.19	60.
Memory – Visual Repr. Delayed	L Hippocampus	21	19	31	30	.13	.07
	R Hippocampus	21	20	36	36	.10	.03
Mazes Efficiency	L Hippocampus	26	24	31	29	.14	.07
	R Hippocampus	26	23	36	34	.17	60.
Mazes Errors	L Hippocampus	.23	.27	31	34	.03	.11
	R Hippocampus	.23	.29	36	39	.05	.15
Simple Reaction Time	L Hippocampus	.27	.29	31	32	04	.05
	R Hippocampus	.27	.31	36	39	01	.10

vas at least some overlap in the samples reported in different articles by the p new and old participants, the data were not suitable for meta-analyses. à Entries in bold represent combinations that were most consistent with the mediation model in that the reduction in A-C after control of B was larger than both the reduction of A-B after control of C and the reduction of B-C after control of A.

NA indicates that the estimates could not be computed because the relevant correlations were not available.

Table 2

Analyses with White Matter Hyperintensities (WMH) as the Brain Structure Variable

Cook et al. (2002), N = 43, Ages 60 to 93 Trail Making B Peri-ventricular WMH 57 53 47 42 33 00 Shipley Abstraction Peri ventricular WMH -57 53 47 42 33 -09 Shipley Abstraction Peri ventricular WMH -57 -53 47 42 -33 -09 Shipley Abstraction Peri ventricular WMH -57 -56 22 11° -03 -09 Working Memory (Verbal) Frontal WMH -27 -28 47 47 -11° -28° -21° -23° 47° -21° -22° -21° -22°	С	В	A-C	A-C.B	A–B	A-B.C	$\mathbf{B}^{-\mathbf{C}}$	B-C.A
Peri-ventricular WMH $.57$ $.53$ $.47$ $.42$ $.33$ Deep WMH $.57$ $.53$ $.47$ $.42$ $.30$ Peri-ventricular WMH $.57$ $.53$ $.47$ $.42$ $.30$ Peri-ventricular WMH $.57$ $.53$ $.47$ $.42$ $.30$ Deep WMH $.57$ $.53$ $.47$ $.42$ $.30$ Jab Frontal WMH $.57$ $.53$ $.49$ $.47$ $.17$ Jab Frontal WMH $.27$ $.23$ $.47$ $.17$ $.222$ Jab Frontal WMH $.27$ $.23$ $.49$ $.47$ $.17$ Jab Frontal WMH $.27$ $.23$ $.48$ $.47$ $.17$ Jab Temporal WMH $.27$ $.23$ $.48$ $.47$ $.17$ Jab Temporal WMH $.27$ $.23$ $.49$ $.47$ $.17$ Jab Temporal WMH $.27$ $.23$ $.48$ $.46$ $.19$ Jabit Temporal WMH		Cook et al. (2002), N =	= 43, Age	s 60 to 93				
	Trail Making B	Peri-ventricular WMH	.57	.53	.47	.42	.33	60.
Peri-ventricular WMH 57 57 53 4.7 4.2 33 Deep WMH 57 55 2.2 1.5 21 Adming-Dixon & Raz (2003), N = 113, Ages 50 to 81 22 4.0 3.7 22 Auning-Dixon & Raz (2003), N = 139, Ages 50 to 81 Temporal WMH 27 23 4.8 4.6 19 Aunual WMH 27 23 4.8 4.6 19 22 Frontal WMH 27 23 4.8 4.7 22 24 Frontal WMH 27 23 4.8 4.7 22 Frontal WMH 27 23 4.8 4.7 22 Coorjes et al. (2007), N = 36, Ages 50 to 81 74 74 74 74 Deep WMH 57 57 57 6.7 6.6 74 MMH 74 57 57 6.7 6.7 22 WMH 74 74 74 74 21		Deep WMH	.57	.53	.22	.07	.30	.22
	Shipley Abstraction	Peri-ventricular WMH	57	53	.47	.42	33	09
al) Fontal WMH 27 23 40 37 22 al) Fontal WMH 27 23 48 46 19 Temporal WMH $.27$ $.23$ 48 46 19 Frontal WMH $.27$ $.23$ 48 47 19 Frontal WMH $.27$ $.25$ 48 47 $.117$ Gootjes et al. (2007), N = 36, Ages 50 to 81 $.47$ $.117$ Deep WMH $.29$ 57 $.67$ $.66$ 41 MH 50 53 $.48$ $.47$ $.117$ Deep WMH $.27$ $.23$ $.48$ $.46$ $.177$ MH 59 57 $.67$ $.66$ 41 VMH 57 $.53$ $.48$ $.46$ 27 WMH 41 46 $.33$ $.26$ $.27$ WMH 41 40 $.38$ $.37$ $.20$ WMH 41 41 $.26$		Deep WMH	57	55	.22	.15	21	11
ai) Fontial WMH 27 23 40 37 22 Temporal WMH 27 23 48 46 19 Frontial WMH $.27$ $.23$ 48 19 19 Frontial WMH $.27$ $.27$ $.23$ $.49$ 19 17 Temporal WMH $.27$ $.27$ $.25$ $.48$ $.47$ $.17$ Temporal WMH $.27$ $.25$ $.48$ $.47$ $.17$ Gootjes et al. (2007), N = 56, Ages 50 to 81 17 17 17 Deep WMH 59 57 $.67$ $.66$ 16 WMH 57 53 34 30 31 WMH 51 57 56 58 56 56 WMH 41 36 38 34 51 50 WMH 41 56 56 58 56 50 50 WMH 41	Gu	nning-Dixon & Raz (2003), N = 13	9, Ages 50	to 81			
Temporal WMH -27 -23 48 46 -19 Frontal WMH $.27$ $.16$ 40 33 $.34$ Frontal WMH $.27$ $.25$ $.48$ $.47$ $.17$ Gootjes et al. (2007), N = 36 , Ages 50 to 81 $.67$ $.66$ 69 57 $.67$ $.66$ 17 Deep WMH 59 57 $.57$ $.67$ $.66$ 17 $.17$ Deep WMH 59 57 $.67$ $.66$ 14 $.17$ Deep WMH 59 57 $.67$ $.66$ 14 $.17$ Deep WMH 59 57 $.57$ $.67$ $.66$ 14 WMH 51 57 $.53$ $.38$ $.37$ 18 WMH 41 41 60 $.38$ $.37$ 18 WMH 16 13 $.38$ $.37$ 18 WMH 16 13 $.38$ $.37$ 18	Working Memory (Verbal)	Frontal WMH	27	22	.40	.37	22	13
Frontal WMH.27.16.40.33.34Temporal WMH.27.25.48.47.17Temporal WMH.27.25.48.47.17Gootjes et al. (2007), $N = 36$, Ages 50 to 81.67.66 41 Deep WMH 59 57 .67.66 41 Charlton et al. (2010), $N = 104$, Ages 50 to 93.67.66 41 WMH 57 55 .48.46 30 WMH 57 55 .48.46 30 WMH 51 57 55 .48.46WMH 51 57 53 .48.40WMH 51 53 .38.3718WMH 41 40 .38.3718WMH 41 40 .38.3718WMH 16 12 .38.3718WMH 16 12 .38.3713WMH 21 12 .38.3721WMH 16 12 .38.3631WMH 16 12 13.3834WMH 16 12 16.3821WMH 16 12 16.3831WMH 21 12 16.3831WMH 29 20 .38.3631WMH 52 16 <td></td> <td>Temporal WMH</td> <td>27</td> <td>23</td> <td>.48</td> <td>.46</td> <td>19</td> <td>07</td>		Temporal WMH	27	23	.48	.46	19	07
Temporal WMH $.27$ $.28$ $.47$ $.17$ Gootjes et al. (2007), N = 36, Ages 50 to 81Gootjes et al. (2007), N = 36, Ages 50 to 81 $.41$ $.17$ Deep WMH 59 $.57$ $.67$ $.66$ 41 Deep WMH 57 55 $.48$ 30 Charlton et al. (2010), N = 104 , Ages 50 to 93 $.46$ 30 WMH 57 55 $.48$ $.46$ 30 WMH 57 53 $.48$ $.46$ 30 WMH 57 53 $.48$ $.46$ 30 WMH 57 56 $.48$ $.46$ 30 WMH 51 53 $.48$ 31 30 WMH 41 36 $.38$ $.37$ 20 WMH 41 36 $.38$ $.37$ 20 unWMH 41 36 $.38$ 21 unWMH 16 13 $.38$ $.37$ 20 unWMH 16 13 $.38$ $.37$ 21 unWMH 16 13 $.38$ $.37$ 21 unWMH 16 13 $.38$ $.37$ 21 unWMH 16 13 $.28$ 21 unWMH 21 13 13 21 unWMH 29 13 21 21 unWMH 21 13 21 21 <td>WCST (Perseverations)</td> <td>Frontal WMH</td> <td>.27</td> <td>.16</td> <td>.40</td> <td>.33</td> <td>.34</td> <td>.26</td>	WCST (Perseverations)	Frontal WMH	.27	.16	.40	.33	.34	.26
Gooijes et al. (2007) , $N = 36$, $Ages 50$ to 81 Deep WMH 59 $.57$ $.66$ 41 Deep WMH 59 $.57$ $.66$ 41 Charlton et al. (2010) , $N = 104$, $Ages 50$ to 93 WMH 57 $.56$ 41 WMH 57 55 $.48$ $.46$ 30 WMH 57 55 $.48$ $.46$ 30 Rabbit et al. (2007) , $N = 65$, $Ages 64$ to 85 $.48$ $.46$ 30 WMH 41 67 $.38$ $.37$ 16 WMH 41 40 $.38$ $.37$ 16 WMH 41 40 $.38$ $.37$ 21 MMH 16 13 $.38$ $.37$ 21 InvWMH 16 13 $.38$ $.37$ 21 InvWMH 16 13 $.38$ $.37$ 21 InvWMH 19 12 $.38$ $.37$ 21 InvWMH 19 13 $.38$ $.37$ 21 InvWMH 19 19 $.38$ $.37$ 21 InvWMH 19 19 19 21 21 InvWMH 21 13		Temporal WMH	.27	.25	.48	.47	.17	.05
Deep WMH 59 57 $.67$ $.66$ 41 Charlton et al. (2010), N = 104, Ages 50 to 93WMH 57 55 $.48$ $.46$ 30 WMH 57 57 48 46 30 WMH 57 57 38 37 18 WMH 41 40 38 37 20 WMH 41 40 38 37 28 MMH 41 40 38 37 20 MMH 21 12 38 37 28 MMH 21 12 38 37 28 MMH 21 12 38 37 21 unWMH 16 13 38 21 unWMH 16 19 38 21 unWMH 12 19 38 21 unWMH 12 16 38 21 unWMH 12 19 19 21 unWMH 12 12 12 14 unWMH 12 12 12 14 unWMH 12 12 12		Gootjes et al. (2007), N		es 50 to 81				
Charlton et al. (2010), $N = 104$, Ages 50 to 93WMH 57 55 48 30 WMH 57 55 48 30 Rabbit et al. (2007), $N = 65$, Ages 64 to 85 33 37 30 WMH 41 40 38 37 18 WMH 41 40 38 37 27 WMH 41 36 33 37 27 WMH 41 36 33 37 27 WMH 41 47 38 37 20 WMH 41 47 38 37 27 unyWMH 41 36 33 21 unyWMH 16 13 33 21 unWMH 16 13 33 21 unWMH 16 13 33 21 unWMH 11 09 38 $.35$ 21 unWMH 11 16 $.38$ $.36$ 21 unWMH 11 09 $.38$ $.36$ 21 sin a CircleWH 29 20 $.38$ $.36$ 21 un distanceWMH 31 19 $.38$ $.36$ 21 unWMH 21 21 $.216$ $.38$ $.24$ unWMH 31 19 $.36$ 21 unWMH 31 <td< td=""><td>Dichotic listening</td><td>Deep WMH</td><td>59</td><td>57</td><td>.67</td><td>.66</td><td>41</td><td>03</td></td<>	Dichotic listening	Deep WMH	59	57	.67	.66	41	03
WMH 57 56 $.48$ 46 30 Rabbitt et al. (2007), N = 65 , Ages 64 to 85 40 38 46 30 WMH 41 40 38 37 16 WMH 41 40 38 37 27 WMH 41 36 32 27 WMH 41 36 32 27 WMH 41 36 37 27 WMH 41 36 37 27 WMH 21 12 38 27 WMH 21 12 38 29 anyWMH 16 13 38 29 unWMH 16 19 38 21 unWMH 19 20 21 unWMH 13 19 21 sin a CircleWMH 33 19 21 utionWMH 33 19 21 utionWMH 33 19 21 utionWMH 19 19 21 utionWMH 19 19 21 utionWMH 21 21 21 utionWMH		Charlton et al. (2010), N		ges 50 to 9	3			
Rabbitt et al. (2007), $N = 65$, $Ages 64 to 85$ WMH 41 40 38 37 18 WMH 41 40 38 37 18 WMH 41 36 38 37 27 WMH 41 36 38 37 20 WMH 21 12 38 37 20 WMH 21 12 38 37 20 MMH 21 12 38 37 20 MMH 21 12 38 37 20 MMH 16 12 38 37 20 MMH 16 12 38 37 21 MM WHH 17 09 38 21 MMH 17 09 38 21 MMH 12 16 38 21 MION 12 12 21 21	Immediate memory	MMH	57	55	.48	.46	30	04
WMH 41 40 $.38$ $.37$ 18 WMH 41 36 $.38$ $.37$ 27 WMH 41 36 $.38$ 27 27 WMH 21 47 $.38$ 27 20 WMH 21 12 38 27 20 WMH 21 12 38 26 21 unWMH 16 13 38 26 21 anyWMH 16 13 38 26 21 unWMH 11 09 38 26 21 sin a CircleWMH 11 09 38 36 31 utionWMH 31 19 38 36 31 utionWMH 31 19 38 36 14 utionWMH 33 19 36 44 utionWMH 33 16 38 44		Rabbitt et al. (2007), N	= 65, Ag	es 64 to 85				
WMH 41 36 $.38$ $.32$ 27 WMH 41 47 $.38$ 2 20 WMH 21 47 $.38$ 21 20 WMH 21 12 38 21 20 WMH 21 12 38 24 28 wMH 16 13 38 21 13 aryWMH 16 13 38 21 unWMH 19 16 38 24 unWMH 19 09 38 37 21 unWMH 19 01 38 36 24 unWMH 19 09 38 36 31 win CircleWH 19 09 38 32 31 sin a CircleWH 31 19 38 34 34 utionWMH 31 19 38 31 34 win WH 31 31 31 31 31 34 utionWMH 31 31 31 31 31 win WH 31 31 31 $.$	Culture Fair IQ	HMM	41	40	.38	.37	18	03
WMH 48 47 $.38$ $.37$ 20 WMH 21 12 $.38$ $.37$ 28 WMH 21 12 $.38$ $.37$ 20 WMH 21 12 $.38$ $.37$ 28 May WMH 16 13 $.38$ $.37$ 13 May WMH 16 16 38 37 21 MM WMH 17 16 38 37 21 MM WMH 17 09 38 37 21 MM WMH 17 09 38 37 21 Min a Circle WMH 17 09 38 32 31 Min a Circle WMH 29 20 31 31 31 31 31 31 Min a Circle WMH 31 31 31 32 31 $$	AH4-1 IQ	НММ	41	36	.38	.32	27	14
WMH 21 12 $.38$ $.34$ 28 laryWMH 16 13 $.38$ $.37$ 13 laryWMH 16 13 $.38$ $.36$ 13 laryWMH 16 $.09$ $.38$ $.36$ 13 laryWMH 19 15 $.38$ 21 laryWMH 19 15 38 21 larWMH 11 09 38 24 larWMH 17 09 38 24 larWMH 29 20 38 34 sin a CircleWMH 29 20 38 34 utionWMH 31 19 38 34 utionWMH 33 19 36 44	AH4-2 IQ	НММ	48	47	.38	.37	20	02
aryWMH 16 13 $.38$ $.37$ 13 aryWMH 16 09 $.38$ $.37$ 13 aryWMH 21 15 $.38$ $.36$ 21 WMH 21 15 $.38$ $.35$ 21 unWMH 19 16 $.38$ 23 21 unWMH 19 16 38 35 21 wMH 17 09 38 35 24 sin a CircleWMH 22 46 38 32 31 utomWMH 31 19 38 32 34 utomWMH 31 19 38 36 34 utomWMH 31 19 38 36 34 utomWMH 32 16 38 35 22	WAIS Vocabulary	HIMM	21	12	.38	.34	28	22
aryWNH 16 09 $.38$ $.36$ 21 WMH 15 15 38 35 21 WMH 19 07 38 33 34 WMH 17 09 38 33 34 WMH 17 09 38 32 24 WMH 17 09 38 32 31 WMH 29 20 38 32 31 $sin a Circle$ WMH 31 19 38 36 WMH 31 19 38 36 34 $wionwMH31193836wionwMH31193836wionwMH33193836wionwMH33193836wionwMH32163836$	Mill Hill A Vocabulary	НММ	16	13	.38	.37	13	08
WMH 21 15 .38 .35 21 un WMH 19 07 .38 .33 34 WMH 17 09 .38 .35 34 WMH 17 09 .38 .35 34 WMH 17 09 .38 .35 34 WMH 29 20 .38 .35 31 sin a Circle WMH 52 46 .38 .28 34 win WH 31 19 .38 .28 34 ution WMH 31 19 .38 .26 44 WMH 33 19 .38 .26 44 ution WMH 33 19 .38 .36 WMH 33 19 .38 .36 44 WMH 33 16 .38 .36 .44	Mill Hill B Vocabulary	HIMM	16	60	.38	.36	21	16
	Forward Digit Span	HMM	21	15	.38	.35	21	14
WMH 17 09 .38 .35 24 . WMH 29 20 .38 .35 24 . WMH 29 20 .38 .32 31 . wMH 52 46 .38 .32 31 . sin a Circle WMH 52 46 .38 .28 34 wMH 31 19 .38 .29 38 . ution WMH 33 19 .38 .26 44 . wMH 22 16 .38 .35 22 .	Backward Digit Span	HIMM	19	07	.38	.33	34	30
WMH 29 20 .38 .32 31 s in a Circle WMH 52 46 .38 .28 34 34 WMH 31 19 .38 .29 38 .36 .14 WMH 31 19 .38 .29 38 .14 ution WMH 33 19 .38 .26 44 . wMH 22 16 .38 .35 22 .22 .	Word Span	HIMM	17	-00	.38	.35	24	19
WMH 52 46 .38 .28 34 WMH 31 19 .38 .29 38 WMH 33 19 .38 .26 44 WMH 22 16 .38 .35 22	Memory for People	HIMM	29	20	.38	.32	31	23
WMH 31 19 .38 .29 38 WMH 33 19 .38 .26 44 WMH 22 16 .38 .35 22	Memory for Objects in a Circle	НММ	52	46	.38	.28	34	18
WMH3319 .38 .2644 · WMH2216 .38 .3522 ·	Visual Search	HIMM	31	19	.38	.29	38	30
WMH2216 .38 .3522	Letter-Digit Substitution	HIMM	33	19	.38	.26	44	36
	Phonemic fluency	HIMM	22	16	.38	.35	22	15

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c	B	A-C	A-C.B	A-B	A-C A-C.B A-B A-B.C B-C B-C.A	B-C	B-C.A
Category fluency	HMM	35	24 .38	.38	.28	38	29
Brixton	HMW	.37	.34	.34 .38	.35	.22	60.
Rey-Osterrieth Figure	ММН	56	51	.38	.28	33	15

Note: All of the B-C combinations within a given study presumably involved the same research participants. It is likely that there was at least some overlap in the samples reported in different articles by the same research team. However, the degree of overlap was not always stated, and because results were not reported separately for the new and old participants, the data were not suitable for meta-analyses. Entries in bold represent combinations that were most consistent with the mediation model in that the reduction in A-C after control of B was larger than both the reduction of A-B after control of C and the reduction of B-C after control of A.

NA indicates that the estimates could not be computed because the relevant correlations were not available.

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Table 3

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Analyses

c	В	A-C	A-C.B	A-B	A-B.C	B_C	B-C.A
	Charlton et al. (2008) , N = 118, Ages 50 to 90	= 118, A ₅	ges 50 to 9	0			
Working Memory	Mean diffusivity	35	18	LT.	.73	36	15
Flexibility	Mean diffusivity	54	60	LT.	.79	39	.05
Speed	Mean diffusivity	55	53	LT.	LT.	43	02
Fluid Intelligence	Mean diffusivity	42	48	LL.	79.	29	.05
	Charlton et al. (2009), $N = 106$, Ages 50 to 90	= 106, A§	ges 50 to 9	0			
Theory of Mind	Mean diffusivity	30	29	.78	.78	24	01
	Charlton et al. (2010), $N = 104$, Ages 50 to 90	= 104, A _§	ges 50 to 9	0			
Episodic memory	Mean diffusivity	57	69	.78	.84	38	12
	Carlesimo et al. (2010), $N = 76$, Ages 20 to 80	= 76, A	ges 20 to 8	0			
Word memory, delayed	L. Hippocampal diffusivity	42	57	.75	79.	23	.14
	R. Hippocampal diffusivity	42	45	.70	.71	27	.04
	L. Hippocampal FA	42	73	60	77	08	46
	R. Hippocampal FA	42	74	67	80	.02	39
Rey Visual Figure Memory	L. Hippocampal diffusivity	52	91	.75	.91	16	.41
	R. Hippocampal diffusivity	52	73	.70	.81	21	.25
	L. Hippocampal FA	52	63	60	69	.19	18
	R. Hippocampal FA	52	75	67	80	.16	30

Psychol Bull. Author manuscript; available in PMC 2012 September 1.

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NA indicates that the estimates could not be computed because the relevant correlations were not available.