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The Reference Ability Neural Network Study: Life-time stability of reference-ability neural networks derived from task maps of 2

young adults 3

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ABSTRACT

Analyses of large test batteries administered to individuals ranging from young to old have consistently yielded a 15 set of latent variables representing reference abilities (RAs) that capture the majority of the variance in agerelated cognitive change: Episodic Memory, Fluid Reasoning, Perceptual Processing Speed, and Vocabulary. In a 17 previous paper (Stern et al., 2014), we introduced the Reference Ability Neural Network Study, which adminis- 18 ters 12 cognitive neuroimaging tasks (3 for each RA) to healthy adults age 20-80 in order to derive unique neural 19 networks underlying these 4 RAs and investigate how these networks may be affected by aging. 20We used a multivariate approach, linear indicator regression, to derive a unique covariance pattern or Reference 21 Ability Neural Network (RANN) for each of the 4 RAs. The RANNs were derived from the neural task data of 64 22 younger adults of age 30 and below. We then prospectively applied the RANNs to fMRI data from the remaining 23 sample of 227 adults of age 31 and above in order to classify each subject-task map into one of the 4 possible ref- 24 erence domains. Overall classification accuracy across subjects in the sample age 31 and above was 0.80 ± 0.18 . 25 Classification accuracy by RA domain was also good, but variable; memory: 0.72 ± 0.32 ; reasoning: 0.75 ± 0.35 ; 26 speed: 0.79 ± 0.31 ; vocabulary: 0.94 ± 0.16 . Classification accuracy was not associated with cross-sectional age, 27 suggesting that these networks, and their specificity to the respective reference domain, might remain intact 28 throughout the age range. Higher mean brain volume was correlated with increased overall classification 29 accuracy; better overall performance on the tasks in the scanner was also associated with classification accuracy. 30 For the RANN network scores, we observed for each RANN a higher score was associated with a higher corre- 31 sponding classification accuracy for that reference ability. Despite the absence of behavioral performance infor- 32 mation in the derivation of these networks, we also observed some brain-behavioral correlations, notably for 33 the fluid-reasoning network whose network score correlated with performance on the memory and fluid- 34 reasoning tasks. While age did not influence the expression of this RANN, the slope of the association between 35 network score and fluid-reasoning performance was negatively associated with higher ages. These results pro- 36 vide support for the hypothesis that a set of specific, age-invariant neural networks underlies these four RAs, 37 and that these networks maintain their cognitive specificity and level of intensity across age. 38 Activation common to all 12 tasks was identified as another activation pattern resulting from a mean-contrast 39 Partial-Least-Squares technique. This common pattern did show associations with age and some subject demo- 40 graphics for some of the reference domains, lending support to the overall conclusion that aspects of neural pro- 41 cessing that are specific to any cognitive reference ability stay constant across age, while aspects that are common 42 to all reference abilities differ across age. 43

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49Introduction

Analyses of large test batteries administered to individuals ranging 50from young to old, have consistently yielded latent variables, or refer-5152 ence abilities (RAs) that capture the majority of the variance in age-

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related cognitive change. Salthouse et al. have identified four domains: 53 episodic memory, fluid reasoning, perceptual speed, and vocabulary 54 (Salthouse, 2005, 2009; Salthouse et al., 2008). Based on these findings, 55 Salthouse et al. have argued that a productive and efficient approach to 56 cognitive aging research is to try to understand how aging impacts 57 performance of this small set of RAs, rather than on specific tasks 58 (Salthouse and Ferrer-Caja, 2003). Similarly, for cognitive neuroimaging 59 research in aging the emphasis on age-related differences in a set of 60

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C. Habeck et al. / NeuroImage xxx (2015) xxx-xxx

broad neural networks underlying the reference abilities for the four 61 62 cognitive domains would be more productive than piecemeal approach 63 focusing on separate individual tasks without consideration of com-64 monalities between these tasks. This would allow us to more reliably explore the neural basis of aging's influence on key cognitive abilities. 65 The Reference Ability Neural Network (RANN) Study is designed to 66 identify networks of brain activity uniquely associated with perfor-67 68 mance across adulthood of each of the four reference abilities described 69 above. In the RANN study, 12 tasks, three from each domain, that have 70reliably been associated with the corresponding RA, are administered 71to subjects in the scanner. Using analytic approaches that parallel those used to derive latent variables from cognitive psychometric 72data, we aim to determine whether four spatial fMRI networks can be 73 74 derived that serve as the neural substrate for the latent cognitive structure of the reference abilities. 75

76 In a previous report (Stern et al., 2014) we introduced the RANN study and presented details of its acquisition and analysis procedures. 77 We described an analysis intended to provide an initial representation 78 of actual RANNs for each ability. We used a general linear model ap-79 proach to summarize each subject's activation for each task into a single 80 contrast. We then used a multivariate technique, linear indicator regres-81 sion analysis, to derive four unique linear combinations of Principal 82 83 Components (PC) of imaging data, one for each RA. We then investigated the ability of these constructed patterns to predict the reference do-84 main using the activation of individual subjects for each task in held-out 85 data. Median accuracy rates for associating component task activation 86 with its corresponding reference ability were quite good: memory: 87 88 76%; reasoning: 82%; speed: 79%; vocabulary: 71%. We took this as an indication that it will be possible to identify unique networks associated 89 90 with each reference ability.

Here we report an extension of this analysis in a larger group of par-91 92ticipants. In our original report, we attempted to identify networks 93unique to each ability using data from subjects of all ages. Since the RANN study is intended to understand the sources of age-related cogni-94tive change, it would be important to identify RANNs in younger people, 95 and then investigate how these networks change as a function of aging. 96 97 In the current study, we again used linear indicator regression analysis 98 to derive a unique spatial covariance pattern (from a set of Principal Components) for each reference ability, but this analysis focused only 99 on 64 individuals age 30 and below. We then investigated whether ex-100 pression of these covariance patterns could successfully predict the ref-101 102 erence domain associated with the activation of individual subjects and tasks in participants age 31 to 80. To the extent that these patterns are 103 consistently expressed across age, this association should remain stable. 104 105 However, a worsening in the ability to categorize abilities for older participants might indicate some age-related change. To the extent that we 106 107 observed differences in classification accuracy, we planned to investigate the basis of these differences taking several approaches. Here we 108 assessed whether classification accuracy 1) was lower for higher age 109for specific reference abilities or specific individuals, 2) was associated 110 with the degree to which these patterns were expressed, and 3) was as-111 112 sociated with observed age differences in mean cortical volume, cortical 113 thickness and white-matter hyper-intensity burden. In addition to the activation particular to each reference domain, we also identified a 114common activation pattern in the derivation sample of participants 115aged 20-30. Brain-behavioral correlations and correlation with demo-116 117 graphics was also assessed in the validation sample of participants aged 31 and above. 118

119 Material and Methods

120 Subjects

121 291 healthy adults were included in these analyses. All subjects na-122 tive English speakers, strongly right-handed, and have at least a fourth 123 grade reading level. Subjects were screened for MRI contraindications and hearing or visual impairment that would impede testing. Subjects 124 were free of medical or psychiatric conditions that could affect cogni-125 tion. Careful screening ensured that the elder subjects did not meet 126 criteria for dementia or Mild Cognitive Impairment (MCI). A score greater than 130 was required on the Mattis Dementia Rating Scale (Mattis, 128 1988). Further, performance was required to be within age-adjusted 129 normal limits on a list-learning test, and participants were required to 130 have no or minimal complaints on a functional impairment questionnaire (Blessed et al., 1968).

Procedure

All subjects completed screening for dementia or MCI prior to partic- 134 ipating in the remainder of the study. They participated in two 2-hour 135 scanning sessions. Six tasks were administered in each session in the 136 context of fMRI studies. One session presented three Vocabulary tasks 137 and three Perceptual Speed tasks interspersed in a fixed order: Syno- 138 nyms, Digit-Symbol, Antonyms, Letter Comparison, Picture Naming, 139 and Pattern Comparison; and the other session presented three Episodic 140 Memory tasks and three Fluid Reasoning tasks, also interspersed in a 141 fixed order: Logical Memory, Paper Folding, Word Order Recognition, 142 Matrix Reasoning, Paired Associates, Letter Sets. The order of tasks with- 143 in session was not varied, but the order of the two sessions was 144 counterbalanced across subjects, with equal numbers having each 145 order. The activation tasks were supplemented with other imaging pro-146 cedures described below. At a separate session subjects completed a 147 battery of neuropsychological tests as well as a set of questionnaires. 148 These will not be discussed in the current report. 149

Stimulus presentation

Task stimuli were back-projected onto a screen located at the foot of 151 the MRI bed using an LCD projector. Participants viewed the screen via a 152 mirror system located in the head coil and, if needed, had vision 153 corrected to normal using MR compatible glasses (manufactured by 154 SafeVision, LLC. Webster Groves, MO). Responses were made on a 155 LUMItouch response system (Photon Control Company). Task administration and collection of reaction time (RT) and accuracy data were controlled by EPrime (v2.08) running on a PC computer. Task onset was electronically synchronized with the MRI acquisition computer. 159

Reference Ability tasks

In the scanner, participants performed a battery of twelve computerized tasks based on the cognitive tasks that have been used to derive the RAs that are addressed in this report. Prior to the scan session, computerized training was administered for the six tasks included in that session. At the completion of training for each task, participants had the option of repeating the training. The tasks are described in detail in (Stern et al., 2014). For all tasks, except picture naming, responses made on the computer keyboard and during scans they were made on the LUMItouch response system.

In the remainder of the manuscript, we will use the following short-171 hand notation for the reference abilities: episodic memory – MEM, fluid reasoning – FLUID, perceptual processing speed – SPEED, and vocabu-173 lary – VOCAB. 174

Vocabulary Tests. The primary dependent variable for all VOCAB tasks is 175 the proportion of correct items. 176

Synonyms (Salthouse, 1993). Subjects have to match a given word to 177 its synonym, or to the word most similar in meaning. The probe word is 178 presented in all capital letters at the top of the screen, and four numbered choices are presented below. 180

Antonyms (Salthouse, 1993). Participants match a given word to its 181 antonym, or to the word most different in meaning. 182

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