Distinctive Neurochemistry in Alzheimer’s Disease via 7 T In Vivo Magnetic Resonance Spectroscopy

Małgorzata Marjańska a,*, J. Riley McCarten b,c, James S. Hodges d, Laura S. Hemmy b,c and Melissa Terpstra a

aCenter for Magnetic Resonance Research and Department of Radiology, University of Minnesota, Minneapolis, MN, USA
bGeriatric Research, Education and Clinical Center, Veterans Affairs Health Care System, Minneapolis, MN, USA
cDepartment of Neurology, University of Minnesota, Minneapolis, MN, USA
dDivision of Biostatistics, School of Public Health, University of Minnesota, Minneapolis, MN, USA
eDepartment of Psychiatry, University of Minnesota, Minneapolis, MN, USA

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Abstract. This study’s objective was to increase understanding of biological mechanisms underlying clinical Alzheimer’s disease (AD) by noninvasively measuring an expanded neurochemical profile and exploring how well this advanced technology distinguishes AD from cognitively normal controls. We measured concentrations of 14 neurochemicals using ultra-high field (7 T) ultra-short echo time (8 ms) magnetic resonance spectroscopy (MRS) in 16 participants with mild to moderate clinical AD and 33 age- and gender-matched control participants. MRS was localized to the posterior cingulate cortex (PCC), a region known to be impacted by AD, and the occipital cortex (OCC), a control region. Participants with AD were recruited from dementia specialty clinics. Concentration of the antioxidant ascorbate was higher (p<0.0007) in both brain regions. Concentrations of the glial marker myo-inositol and the choline-containing compounds involved in membrane turnover were higher (p ≤ 0.0004) in PCC of participants with AD. Ascorbate and myo-inositol concentrations were strongly associated, especially in the PCC. Random forests, using the 14 neurochemicals in the two regions, distinguished participants with AD from controls: same-sample sensitivity and specificity were 88% and 97%, respectively, though out-of-sample-values would be lower. Ultra-high field ultra-short echo time MRS identified the co-occurrence of elevated ascorbate and myo-inositol in the PCC as markers that distinguish participants with mild to moderate AD from controls. While elevated myo-inositol may be a surrogate marker of neuroinflammation, the unexpected elevation of the antioxidant ascorbate may reflect infiltration of ascorbate-rich leukocytes.

Keywords: Ascorbate, myo-inositol, neurochemical profile, neuroinflammation, posterior cingulate cortex, ultra-high field, ultra-short echo time

INTRODUCTION

Noninvasive magnetic resonance spectroscopy (MRS) has been used to measure disease-associated neurochemical concentration differences in human Alzheimer’s disease (AD). Most existing reports are focused on four major neurochemicals contributing the strongest signals to in vivo spectra. N-acetylaspartate (NAA), typically reported together with N-acetylaspartylglutamate (NAAG), i.e., tNAA, is a neuronal metabolite [1] associated with neuronal viability [1], density [2], or dysfunction [2–4]. Lower
tNAA has been reported in AD [2, 3]. The concentration of creatine-containing compounds (tCr, creatine plus phosphocreatine), markers of energy utilization, has been found to be the same in people with AD and healthy controls [5], and is often used in AD literature as an internal reference. Glycerophosphorylcholine and phosphorylcholine are part of cellular membrane, but contribute to the total choline (tCho) signal when membranes break down, thus tCho is thought of as a membrane marker [6]. Some report increases in tCho in AD while others report no change [3]. Myo-inositol (mIns) is a glial osmolyte [3, 5, 7] traditionally seen as a marker of glial cell content [2] or activation [3, 7], but recently proposed as a surrogate marker of neuroinflammation in AD [1]. In the posterior cingulate cortex (PCC), a brain region impacted by AD and readily studied with MRS, elevated mIns/tCr is present at preclinical stages of AD, whereas low tNAA/tCr and elevated tCho/tCr are not generally detected until diagnosis of mild cognitive impairment (MCI) or later stage AD [1, 5, 8]. The magnitudes of the mIns elevation and tNAA reduction increase as AD progresses [1, 5]. These disease-associated concentration differences in mIns and tNAA are thought to reflect increasing glial activation and neuronal loss. mIns elevation may be related to glial proliferation and microglial activation such as occurs when Aβ plaques are surrounded by microglia and astrocytes in AD.

Concentrations of neurochemicals contributing weaker signals in MRS are more challenging to quantify. There are few reports on their noninvasively measured concentrations in AD. We recently validated an ultra-short echo time, ultra-high field MRS approach to quantify fourteen neurochemicals in the context of aging, and found that the concentration of the antioxidant vitamin C (Asc, ascorbate) was lower in the brains of older adults [9]. In AD, Asc has several potential protective mechanisms associated with its antioxidant capacity in addition to scavenging free radicals directly [10]. Asc modulates amyloid accumulation via a mechanism associated with the oxidative stress pathway and protects the endothelia from reactive oxygen species [10]. Additionally, Asc moderates glia-mediated inflammation which is activated by oxidative stress [11].

The main goal of the present work was to study more extensively the neurochemical profile than has been done previously in clinical AD with particular interest in antioxidant concentrations. In addition to extending the neurochemical profile, the use of ultra-high field (≥7 T) improves sensitivity and reliability for quantifying the entire neurochemical profile [12, 13]. Therefore, a secondary goal of this study was to observe whether scanning at ultra-high field increases sensitivity of the neurochemical profile to AD. The study was designed and powered to compare dichotomous groups, well-characterized participants with mild to moderate AD versus cognitively normal, asymptomatic, neurologically intact age-matched controls.

MATERIALS AND METHODS

Participants and study design

Older adults with clinical AD were identified through an IRB approved database of patients cared for at the Geriatric Research, Education and Clinical Center (GRECC) Memory Loss Clinic at the Minneapolis Veterans Affairs Medical Center. The GRECC clinic employs an interdisciplinary team to rigorously assess patients referred for cognitive evaluation. All patients are reviewed and diagnosed through consensus diagnosis with team members, including a neurologist, internist, geropsychiatrists, and neuropsychologists. Participants enrolled in this study met criteria for a diagnosis of probable or possible AD based on NINCDS-ADRDA criteria [14]. Those with possible AD were not recruited if non-AD medical conditions were felt to potentially affect regions of interest (posterior cingulate and occipital lobe). One participant was recruited from a well-established, interdisciplinary community dementia clinic, Health Partners Center for Memory and Aging (CMA) and the diagnosis of probable AD was confirmed by the study neurologist (JRM), the Medical Director of the Minneapolis GRECC. Participants with AD had mild to moderate dementia, based on the Montreal Cognitive Assessment (MoCA [15]) ≥12/30 or Mini-Mental State Exam (MMSE [16]) ≥17/30 and independence in basic activities of daily living. All participants with AD met the same exclusion criteria as controls (listed below) except for cognitive complaints (allowed), and a neurological diagnosis that could affect cognition (only AD allowed).

Thirty three controls completed a medical history questionnaire validated for recruitment of healthy controls for cognitive aging research [17] and cognitive and psychological screens (MoCA [15] and Beck Depression Inventory-II (BDI-II [18]). For participants who passed screens, the study neurologist reviewed with each their medical history and
performed a neurological examination to confirm their status as a neurologically healthy older adult. The controls were prospectively age- and gender-matched with the older adults with AD.

Exclusion criteria for all participants were claustrophobia, non-removable metallic objects of concern for MRI, smoking, weight $>$113 kg accompanied by BMI $>$35, and head circumference $\geq$23.25 inches due to the limited coil size. Medical history on the day of the MRI scan addressed the following exclusion criteria for controls: cognitive complaints, unstable medication regimen, currently symptomatic for a major psychiatric disorder, history of alcohol or substance abuse or dependence within the past two years, serious medical illness (including creatinine clearance <40 mL/min [19]), medical condition or neurological diagnosis that potentially could affect cognition, history of seizure, stroke, traumatic brain injury, or hospital stay related to head injury, or use of investigational drugs that may impact cognition within 30 days or 5 half-lives, whichever is longer. Examination exclusion criteria on the day of the MRI scan included a score <24 on MoCA [15], score $\geq$20 on BDI-II [18], or any abnormality on the neurological examination that could be due to a lesion or disease of the brain.

Sixteen older adults with clinical AD (15 males, 1 female; age 78 $\pm$ 7 (mean $\pm$ standard deviation) years; age range: 69 to 89 years; MoCA ($n$ = 11) = 16 $\pm$ 4; range: 12 to 26; MMSE ($n$ = 5) = 19 $\pm$ 2; range: 17 to 21) and thirty three controls (25 males, 8 females; age: 76 $\pm$ 6 years; age range: 67 to 88 years; MoCA = 27 $\pm$ 2; range: 24 to 29; creatinine = 78 $\pm$ 24 mL/min; range: 42 to 128 mL/min) provided informed consent according to procedures approved by the Human Subjects’ Protection Committee at the University of Minnesota, Institutional Review Board, and underwent an MRI exam at 7 T. Six of the thirty three controls underwent two additional retest scans (for a total of three scans at approximately weekly intervals). Sixteen controls have been part of previously published work [9] and seventeen controls were newly recruited (de novo).

**MR acquisition**

All MR data were acquired using the same methodology as was used in a previous study [9]. Briefly, studies were performed on a 7-T whole-body Siemens MAGNETOM scanner (Siemens, Erlangen, Germany) using a 16-channel transmission line head array radiofrequency coil with $B_1^+$ shimming [20]. Magnetization-prepared rapid gradient-echo (MPRAGE) images were acquired to position 8 cm$^3$ volumes-of-interest (VOIs) in the PCC and the occipital cortex (OCC), and to obtain gray matter (GM), white matter (WM), and cerebrospinal fluid (CSF) content. VOIs were positioned symmetrically in both hemispheres. The PCC VOI was rotated to align with the corpus callosum and placed in the parietal lobe, 2 mm posterior to the marginal sulcus, and 2 mm superior of the corpus callosum. The OCC VOI was rotated to align with the boundaries of the occipital lobe and placed in the center of the occipital lobe vertically while the back corner was placed 4 mm anterior to the back of the occipital lobe. Proton-density (PD) images were also acquired to correct for the intensity field bias in $T_1$-weighted MPRAGE images [21]. MR spectra were measured using a stimulated echo acquisition mode (STEAM) sequence with ultra-short echo time (repetition time ($T_R$) = 5 s; echo time ($T_E$) = 8 ms; mixing time ($T_M$) = 32 ms; 128 averages in PCC and 64 averages in OCC) and with 3D outer volume suppression interleaved with variable power and optimized relaxation delay (VAPOR) water suppression [22]. The chemical shift displacement of the VOI was 4% of the voxel dimension per ppm. The $B_1$ field and water suppression pulses were calibrated in both VOIs for each individual. Each free induction decay was acquired with 2048 complex data points using a spectral width of 6 kHz. A non-suppressed water spectrum was acquired in each brain region for eddy current correction and absolute quantification ($T_R$ = 5 s; $T_E$ = 8 ms; $T_M$ = 32 ms; 1 average). For some participants, MPRAGE images [23] ($T_R$ = 2.4 s; $T_E$ = 2.24 ms; field of view: 256 $\times$ 256 mm$^2$; matrix: 320 $\times$ 320; slice thickness = 0.8 mm, 208 slices, acceleration factor = 2, acquisition time: 6 min 38 s) were acquired at 3 T to obtain GM, WM, and CSF content.

**Image processing and analysis**

Image processing and analysis were performed using the same methodology as was used in a previous study [9]. Briefly, MPRAGE images acquired at 7 T or 3 T for all participants after processing were segmented into GM, WM, and CSF content using Freesurfer [24]. For all controls except two from the earlier study, 7 T images were used to obtain tissue content. For the remaining two controls from the earlier study, de novo controls and participants with AD, 3 T images were used to obtain tissue content.
Spectral processing and quantification

The acquired spectra were processed in Matlab (MathWorks Inc., Natick, MA). Eddy-current effects were corrected using the phase information from non-suppressed water spectrum. Single-shot spectra were frequency and phase aligned using a cross-correlation algorithm.

All spectra were analyzed using LCModel [25] 6.3-1J (Stephen Provencher, Inc., Oakville, ON, Canada) using the same methodology as was used in a previous study [9]. The following metabolites were included in the basis set: Asc, aspartate (Asp), creatine (Cr), γ-aminobutyric acid (GABA), glucose (Glc), glutamine (Gln), glutamate (Glu), glutathione (GSH), glycero-phosphorylcholine (GPC), lactate (Lac), mIns, NAA, NAAG, phosphocreatine (PCr), phospho-rycholine (PCho), phosphorylthanolamine (PE), scy-illo-inositol (sIns), and taurine (Tau). Experimentally measured, age-specific macromolecule spectra were also included in the basis set. In addition, standard lipid resonances provided by LCModel were included in the analysis as described previously [9]. A resonance at 3.155 ppm was also included to fit the resonance of methylsulfonylmethane, a dietary sup-plement [26]. No baseline correction, zero-filling, or line broadening were applied to the in vivo data prior to the analysis. Spectra were fitted between 0.5 and 4.1 ppm using a knot spacing (DKNTMN) of 5 to impose a flat spline baseline [27].

Quantification was performed using the unsup-pressed water signal obtained from the same VOI as the neurochemical signals for normalization [28]. This water signal was corrected for GM, WM, and CSF content. The water content for CSF and WM were assumed to be 0.97 [29] and 0.71 [4, 30, 31]. For GM, the water content was assumed to be 0.79 for older women [30] and 0.76 for older men [30]. T1 and T2 relaxation time constants of water used in the calculation of the attenuation factors were taken from published reports [T1(GM) = 2130 ms, T1(WM) = 1220 ms, T1(CSF) = 4425 [32]; T2(GM) = 50 ms, T2(WM) = 55 ms, T2(CSF) = 141 ms [33]] and included a 10% reduc-tion in T2 of tissue water [34] [T2(GM) = 45 ms, T2(WM) = 49.5 ms]. The T1 and T2 relaxation time constants of neurochemicals were not taken into consideration as attenuation factors since their T1 are shorter and their T2 are longer than water, thus the effect would be very small given the use of repetition time of 5 s and echo time of 8 ms.

Statistical analyses

Statistical analyses were done with JMP Pro v. 13.1.0 (SAS Institute, Cary, NC). Neurochemical concentrations in the two cohorts (AD and controls) were compared using a mixed linear model with fixed effects group (AD versus control), region (PCC versus OCC), and their interaction, with group being a between-participant effect and region a within-participant effect. The random effects were participant and replication within participant (i.e., retest, for participants who had them). Pre-specified comparisons included comparing AD versus control in each region. The analysis used the restricted like-lihood method. The present study’s sample size was chosen to provide adequate power for the dichoto-mous primary comparison of AD versus control. After the present study’s data were collected, we real-ized that a previous study had collected identical data on healthy elderly subjects, and included re-test data for some of those subjects, so that these previously collected data could be added to the present study’s control group to increase statistical power at no cost. To check whether the controls from the earlier study differed from the de novo controls, we compared the two groups of controls using the analysis described above (with the two groups of controls in place of AD versus controls). Tests showed that the two groups of healthy older adults were indeed similar, so the analyses presented combine these two groups into one control group. To test the impact of including only one woman with AD in the study, the same analysis was re-done including only males. To control for multiple testing (14 neurochemicals and two brain regions), the threshold of significance was 0.0018 = 0.05/28. To check whether including the controls from the earlier
study changed the comparison AD versus controls, we re-did the primary analysis omitting them.

The bootstrap forests method (random forests as implemented in JMP) was used to explore whether the 28 neurochemical-region combinations differentiate participants with AD from controls. The outcome was each person’s group, with the 28 neurochemical-region combinations as predictors. Other random-forests specifications were: bootstrapping by simple random sampling with replacement on the whole dataset; seven predictors randomly selected to be considered for each split; and each split minimizes the \textit{p}-value among available splits. This analysis rates each neurochemical for its ability to differentiate the groups, using the G^2 measure (for which higher is better) and gauges the ensemble’s ability to differentiate the groups using two measures, generalized \( R^2 \) (higher is better) and root mean squared error (RMSE; lower is better). These performance measures are described in detail below. We also report same-sample sensitivity and specificity of the fitted random forest. This is known to overstate the sensitivity and specificity that would be obtained in an out-of-sample validation.

To explore the importance of Asp in differentiating AD from control, this analysis was repeated without Asp in the dataset. To explore the relative importance of biochemical and structural measures in differentiating AD from control, the random forest analysis was also repeated adding GM, WM and CSF content for each region as predictors.

The \( G^2 \) measure is used to describe each potential predictor’s predictive power. For each potential predictor, the \( G^2 \) measure is first computed for each of the 1000 trees in the ensemble then summed across the trees to give the predictor’s \( G^2 \) measure for the random-forests ensemble. For a single tree, a predictor’s \( G^2 \) measure is computed first for each node in the tree that splits on that predictor, and then summed across nodes to give the \( G^2 \) measure for the whole tree. For a single node splitting on the predictor, the contribution to the tree’s \( G^2 \) is \( G^2_{\text{parent node}} = (G^2_{\text{left node}} + G^2_{\text{right node}}) \), where for each node, the node’s \( G^2 \) is computed as \( 2 \times \sum \text{over subjects included in the node of } -\ln(p) \), where a participant’s \( p \) is their probability (as estimated in the given node) of being in the group (AD or control) that they are actually in.

Two measures, generalized \( R^2 \), which is closely related to the \( G^2 \) measure described above, and RMSE, were used to describe the explanatory power of the whole bootstrap forests ensemble fit to these data. Both use the ensemble’s estimated probability that a participant is in the AD group, as a function of the participant’s characteristics. This ensemble probability is estimated by estimating the analogous probability separately for each of the 1000 trees in the bootstrap-forest ensemble, then averaging those 1000 probabilities to give a single estimate for the subject. Generalized \( R^2 \) is a generalization of the \( R^2 \) measure used in linear regression, applicable to binary outcomes like AD versus control. It is \( (1 - \left[ \frac{L(0) - L(\text{model})}{2/n} \right] / (1 - L(0)^{1/2}/n)) \), where \( n \) is the total sample size, \( L(\text{model}) \) is the sum over participants of \(-\ln(p)\), \( p \) being a participant’s estimated probability of being in the group that they are actually in, and \( L(0) \) is \(-\ln(n \text{ participant with AD/total } n)\). RMSE is \( \left( \sum \text{over subjects} [y - p]^2/n \right)^{0.5} \), where \( p \) is the estimated probability of being in the AD group and \( y \) is 1 for participants in the AD group and 0 for participants in the control group.

Linear regression analysis in Microsoft Excel v. 14 (Microsoft Corporation, Redmond WA) was used to assess correlations between Asp and the other 13 neurochemical concentrations as well as CSF content in two brain regions.

**RESULTS**

Neurochemical concentrations from a subset of controls have been reported previously [9]. This subset of controls did not differ significantly from the \textit{de novo} controls for any neurochemical, even without Bonferroni adjustment of \( p \)-value threshold (data not shown). High-quality spectra were consistently measured (Fig. 1) with strong, narrow resonances and flat baseline without evidence of contamination from unsuppressed water or lipids from outside the VOI. The tCr linewidths measured as the full width at half maximum of the resonance at 3.03 ppm were on average (± standard deviation) 11.9 ± 1.4 Hz (range: 8.7 to 15.9 Hz). The signal-to-noise ratios of the highest peak in the spectra reported by LCModel were 47 ± 7 (range: 23 to 60).

The VOIs in the PCC contained 42 ± 1.5% GM (mean ± standard deviation), 26.7 ± 1.3% WM, and 31.3 ± 1.9% CSF in the older adults with AD, and 46.8 ± 1.0% GM, 23.8 ± 0.9% WM, and 29.4 ± 1.3% CSF in the controls. The VOIs in the OCC contained 35.5 ± 1.5% GM, 35.8 ± 1.3% WM, and 28.9 ± 1.9% CSF in the older adults with...
AD, and $37.1 \pm 1.0\%$ GM, $36.3 \pm 0.9\%$ WM, and $26.5 \pm 1.3\%$ CSF in the controls.

Asc showed the greatest percent difference in neurochemical concentration between the two groups. CRLB for Asc were $0.20 \pm 0.04$ mM (mean ± SD, range 0.15–0.40 mM). The average CRLB for Asc differed less than 2% between participants with AD and controls for each brain region. There was no strong correlation reported by LCModel between Asc and any other neurochemical (or MM) for any individual or brain region. Table 1 lists all measured neurochemical concentrations by brain region and disease status. Asc, mIns, and tCho concentrations were markedly higher in the PCC ($p \leq 0.004$) in participants with AD, as was the concentration of Asc in the OCC in participants with AD ($p = 0.0007$). Removing the one female with AD and all control females changed the $p$-value of OCC Asc from 0.0007 to 0.0020, though the two analyses had similar effect sizes. $p$-values for PCC Asc, mIns, and tCho remained significant. Asc concentration was higher in both brain regions. While differences in mIns and tCho were specific to the PCC, no differences were specific to the OCC. Considering non-significant trends ($0.05/28 \leq p \leq 0.03$), PE concentration was lower in the PCC of participants with AD; in analyses comparing participants with AD to the subset of de novo controls, the estimated differences between AD and controls were as large or larger for the neurochemical-region combinations that tested significant in the primary analysis reported above, but due to the reduced control sample size, the $p$-values for these comparisons increased to slightly above our Bonferroni-adjusted threshold ($p = 0.0037, 0.0075, 0.0036, 0.0024$ for Asc PCC, Asc OCC, mIns PCC, and tCho PCC, respectively).

Random forests distinguished participants with AD from controls with same-sample sensitivity and specificity of 88% and 97%, respectively, though out-of-sample values would be lower. Fourteen of 16 participants with AD were classified correctly, as were 32 of 33 controls. Generalized $R^2$ was 0.68, and RMSE was 0.28. The main predictors of clinical AD status were Asc ($G^2 = 4.4$) and mIns ($G^2 = 3.8$) in the PCC (for 49 persons). Other predictors were tCho in the PCC ($G^2 = 1.6$), and Asc ($G^2 = 2.0$), NAAG ($G^2 = 1.6$), and mIns ($G^2 = 1.4$) in the OCC. Adding structural information (i.e., volume of interest GM, WM, and CSF content) did not change the classification, with negligible changes in performance measures (generalized $R^2 = 0.67$, RMSE = 0.29), and no structural parameter was a major predictor ($G^2 < 1$).
The greatest concentration differences were Asc in the PCC (21%) and the OCC (17%). The next largest was mIns in the PCC (16%). The strongest random forests predictors of clinical AD status were Asc and mIns. Thus, the region-specific associations between Asc and mIns concentrations were examined (Fig. 2). Correlations between Asc and mIns concentrations were 0.76 in the PCC (0.76 in participants with AD and 0.71 in controls) and 0.50 in the OCC (0.30 in participants with AD and 0.55 in controls). The participant with AD who had the lowest mIns concentration in both brain regions was one of the false negatives classified by random forests. The other false negative had the lowest Asc concentration among participants with AD. The control participant who had the highest Asc concentration (and the highest mIns concentration in the PCC) was the false positive classified by random forests. There were no other strong Asc – neurochemical correlations ($r \leq 0.5$).

**DISCUSSION**

**Asc in AD**

Our findings reveal an increase in brain PCC Asc, mIns, and tCho in mild to moderate AD. While the findings on mIns and tCho are consistent with several former reports, difference in Asc concentrations has only been reported once [38]. Given that oxidative stress is part of AD pathology, high concentration of the antioxidant in AD is unexpected and counterintuitive. Rigor of the approach to measure Asc is evidenced by agreement of the neurochemical profile with existing literature.

While the main role of brain Asc in health and AD is protection against oxidative damage, Asc has many roles [10, 39, including as a cofactor and electron donor in the biosynthesis of neurotransmitters and myelin [10, 39–41]. Nevertheless, Asc is oxidized in all of these roles. Brain Asc concentration is under strong homeostatic control via the specific vitamin C transporter (SVCT2) [33, 34]. Therefore, even if there were a diminishment in the Asc-consuming process, which would be unexpected, this homeostasis would prevent accumulation of Asc. If Asc and Glc cotransport into cells were too slow [38], either intracellular Asc concentration would fall, or the SVCT2 would increase Asc transport. Either way, Asc concentration would not increase. Because extracellular concentration of Asc is substantially lower than intracellular concentration, the role that Asc plays in modulating neurotransmission [10, 39, 42] is unlikely to affect the Asc concentration measured in this study. Given that 30% CSF content was measured in this study, the basal Asc concentration in the CSF (0.1–0.4 mM) [43] would have to more than double to cause the 0.2–0.3 mM higher AD-associated concentrations that were measured. Furthermore, Asc concentration did not correlate ($r \leq 0.5$) with % CSF.

### Table 1

| Neurochemical | PCC                  | OCC                  |
|--------------|----------------------|----------------------|
|              | Concentration (mM)   | Concentration (mM)   |
|              | AD                   | Controls             | AD                   | Controls             |
| Asc          | $1.52 \pm 0.06$      | $1.26 \pm 0.04$      | $1.66 \pm 0.06$      | $1.42 \pm 0.04$      |
| Asp          | $2.12 \pm 0.10$      | $1.98 \pm 0.06$      | $2.01 \pm 0.10$      | $1.75 \pm 0.06$      |
| GABA         | $0.78 \pm 0.07$      | $0.87 \pm 0.04$      | $0.95 \pm 0.07$      | $1.04 \pm 0.04$      |
| Gln          | $3.52 \pm 0.11$      | $3.35 \pm 0.07$      | $3.18 \pm 0.11$      | $2.97 \pm 0.07$      |
| Glu          | $9.83 \pm 0.24$      | $10.13 \pm 0.16$     | $8.95 \pm 0.24$      | $8.95 \pm 0.16$      |
| GSH          | $1.21 \pm 0.03$      | $1.14 \pm 0.02$      | $1.12 \pm 0.03$      | $1.05 \pm 0.02$      |
| mIns         | $8.93 \pm 0.26$      | $7.69 \pm 0.18$      | $7.84 \pm 0.26$      | $7.06 \pm 0.18$      |
| NAA          | $10.49 \pm 0.28$     | $10.84 \pm 0.19$     | $11.81 \pm 0.28$     | $11.65 \pm 0.19$     |
| NAAG         | $1.02 \pm 0.05$      | $0.99 \pm 0.03$      | $0.93 \pm 0.05$      | $1.07 \pm 0.03$      |
| PE           | $0.97 \pm 0.07$      | $1.16 \pm 0.04$      | $0.75 \pm 0.07$      | $0.77 \pm 0.04$      |
| sIns         | $0.41 \pm 0.06$      | $0.45 \pm 0.04$      | $0.40 \pm 0.06$      | $0.45 \pm 0.04$      |
| Tau          | $1.71 \pm 0.07$      | $1.83 \pm 0.05$      | $1.73 \pm 0.07$      | $1.80 \pm 0.05$      |
| tCho         | $1.65 \pm 0.04$      | $1.47 \pm 0.03$      | $1.24 \pm 0.04$      | $1.16 \pm 0.03$      |
| tCr          | $9.86 \pm 0.25$      | $9.70 \pm 0.17$      | $9.84 \pm 0.25$      | $9.62 \pm 0.17$      |

Concentrations are expressed as mean ± standard error of the mean. The $p$-values for significantly different neurochemical concentrations for a given region are in bold face. To account for testing 14 neurochemicals and 2 brain regions, the threshold of significance was $0.0018 = 0.05/28$. 

- The antioxidant role of Asc in health and AD is protection against oxidative damage, Asc has many roles including as a cofactor and electron donor in the biosynthesis of neurotransmitters and myelin.
- The study revealed an increase in brain PCC Asc, mIns, and tCho in mild to moderate AD.
- Correlations between Asc and mIns concentrations were examined, with the strongest correlations found in the PCC and OCC.
- The control participant who had the highest Asc concentration among participants with AD was misclassified by random forests.
- Random forests were used to identify false negatives and positives.
- Extracellular Asc concentration is under strong homeostatic control via the specific vitamin C transporter (SVCT2), preventing accumulation of Asc in AD.
- The basal Asc concentration in the CSF would have to more than double to cause the 0.2–0.3 mM higher AD-associated concentrations that were measured.
- Asc concentration did not correlate with % CSF.
Therefore, a new source of Asc must be introduced to account for the measured higher Asc.

High brain Asc concentrations have been reported under other circumstances. During post-ischemic reperfusion, Asc levels increase as much as 300% [44, 45]. SVCT2 mRNA is found on neurons and in astrocytes [46], and SVCT2 expression has been shown to increase after ischemia-reperfusion [47]. Expression of SVCT2 in brain cells also increases under dietary depletion of Asc [48]. However, there are no reports of increased expression of SVCT2 on neurons in AD, and astrocytes do not change expression of SVCT2 in AD [49]. AD is the only condition under which supply associated increase in expression of SVCT2 is not a known aspect of the pathology yet brain Asc is elevated.

The intrinsic immune response to AD via microglial activation has potential to change brain Asc concentration [37], but is more likely to lower it. Activated microglia increase production of reactive oxygen species, which react with antioxidants, including Asc, thereby tending to lower Asc [50]. Activated microglia may also internalize SVCT2 making them less able to accumulate Asc [51]. Therefore, this aspect of disease pathology fails to provide an explanation for the higher brain Asc concentration measured in older adults with clinical AD in this study.

A possible explanation of high brain Asc concentration in older adults with mild to moderate AD is accumulation of cells rich in Asc. Asc concentration is high in neurons (10 mM) and glia (1 mM) [52], but neurons are unlikely to be accumulating in AD (rather, just the opposite), and the glial species most likely to be upregulated—the microglia—are likely to drive down the levels of Asc through their production of oxygen radicals. Leukocytes have Asc concentration as high as neurons and glia (up to 10 mM) [53, 54]. Leukocytes, including neutrophils, macrophages, and lymphocytes, are known to accumulate Asc, particularly when stimulated [53, 54]. Histological stains have demonstrated the presence of neutrophils and T cells (lymphocytes) in brain parenchyma of people with AD [55, 56]. The correlation of Asc with mIns measured in the current study suggests that peripheral leukocytes may be a complement to or an aspect of neuroinflammation.

**Distinctive neurochemistry of AD**

When the clinical severity of AD is not specified, PCC mIns/tCr and mIns are consistently reported as higher [2, 3], and PCC tNAA/tCr [2, 3] and tNAA [3] as lower in AD relative to controls. This study enrolled older adults with mild to moderate clinical AD, during which disease-associated higher mIns is typically apparent, whereas disease associated lowering in tNAA tends to occur at later stages [1, 5, 8]. Findings on PCC tCho/tCr are disparate, with a few indicating elevation, as was found in this study, and several indicating no difference [2].
In the OCC, one review reported higher mIns and lower tNAA and tNAA/tCr in AD relative to control [2]. A few studies reported no differences in OCC tNAA/tCr, tCho, or tCho/tCr [8, 57, 58] in AD relative to control. The OCC neurochemical concentrations measured in the current study, although not reaching significance, are in agreement with the consistent finding in the literature that mIns is higher in people with AD.

The method used in the current study is unique as spectra were acquired at ultra-short echo time, which reduces vulnerability of quantification to artifacts caused by disease-associated differences in T2 of neurochemicals. The T2 of NAA is substantially shorter in older adults [34]. When spectra are measured at a longer echo time, further shortening of T2 of NAA in AD would artificially cause tNAA concentration to appear lower in AD.

Differing impacts of aging and AD on brain mIns and Asc concentrations may demarcate aging from AD pathology. mIns, which is sensitive to glial activation, is elevated in the PCC in aging [9] and more so in AD. Asc concentration is lower in older than younger cognitively healthy adults, particularly in the OCC [9], but higher in older adults with AD than age-matched controls (PCC and OCC). A pro-inflammatory stage [50] would be consistent with elevated mIns and diminished Asc concentration in aging, whereas a chronic neuroinflammation stage with recruitment of leukocytes from peripheral blood would be consistent with elevated mIns and heightened Asc concentration in AD.

Conclusion

Compensatory upregulation in Asc transport or increased expression of the SVCT2 are ready explanations for the elevated brain Asc concentrations measured in older adults with mild to moderate clinical AD in this study. However, such accumulation of Asc is counterintuitive to the homeostatic mechanism of the SVCT2. The alternative explanation that elevated Asc arose from presence of recruited leukocytes warrants further investigation that includes systemic and histological biomarkers, given the contribution that research on involvement of peripheral immune cells could make toward understanding the underlying mechanisms of AD. A few studies have shown the presence of peripheral leukocytes in the human AD brain, and it has been suggested that such infiltration would be low grade and localized. This project provides in vivo evidence consistent with the presence of peripheral leukocytes in the brains of older adults with clinical AD. High mIns and Asc concentrations measured via ultra-high field, ultra-short echo time MRS may demarcate a transition from an age-associated pro-inflammatory stage to an AD-associated pathological neuroinflammatory response. As such, this method could potentially diagnose the presence of AD prior to neuronal deterioration. In that case, 7 T in vivo MRS may be useful to demarcate a stage when intervention may be effective in arresting or reversing disease progression. 7 T in vivo MRS may also be useful to monitor therapeutic response.

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