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Comparison of PASL, PCASL and background suppressed 3D PCASL in Mild Cognitive Impairment

Sudipto Dolui^{1,2,3}, Marta Vidorreta^{2,3}, Ze Wang⁴, Ilya M. Nasrallah¹, Abass Alavi¹, David A. Wolk², and John A. Detre^{1,2,3}

¹Department of Radiology, University of Pennsylvania, Philadelphia, Pennsylvania, USA

²Department of Neurology, University of Pennsylvania, Philadelphia, Pennsylvania, USA

³Center for Functional Neuroimaging, University of Pennsylvania, Philadelphia, Pennsylvania, USA

⁴Department of Radiology, Lewis Katz School of Medicine, Temple University, Philadelphia, Pennsylvania, USA

Abstract

We compared three implementations of single-shot arterial spin labeled (ASL) perfusion magnetic resonance imaging (MRI): two-dimensional (2D) pulsed ASL (PASL), 2D pseudo-continuous ASL (PCASL), and background suppressed (BS) 3D PCASL obtained in a cohort of patients with mild cognitive impairment (MCI) and elderly controls. Study subjects also underwent ¹⁸Fflurodeoxyglucose positron emission tomography (¹⁸F-FDG PET). While BS 3D PCASL showed the lowest (p<0.001) gray matter-white matter cerebral blood flow (CBF) contrast ratio, it provided the highest (p<0.001) temporal signal-to-noise ratio. Mean relative CBF estimated using the PCASL methods in posterior cingulate cortex (PCC), precuneus and hippocampus showed hypoperfusion in the MCI cohort compared to the controls consistent with hypometabolism measured by ¹⁸F-FDG PET. BS 3D PCASL demonstrated the highest discrimination between controls and patients with effect size comparable to that seen with ¹⁸F-FDG PET. 2D PASL did not demonstrate group differentiation with relative CBF in any ROI, whereas 2D PCASL demonstrated significant differences only in PCC and hippocampus. Mean global CBF values did not differ across methods and were highly correlated, however the correlations were significantly higher (p<0.001) when either the same labeling (PCASL) or the same acquisition strategy (2D) was used as compared to when both the labeling and readout methods differed. In addition, there were differences in regional distribution of CBF between the three modalities, which can be attributed to differences in sequence parameters. These results demonstrate the superiority of ASL with PCASL as well as BS 3D readout as a biomarker for regional brain function changes in MCI.

Disclosure/Conflict of Interest

There is no financial interest in relation to the work described in this paper.

Corresponding Author: John Detre, Departments of Neurology and Radiology, University of Pennsylvania, 3W Gates Pavilion, 3400 Spruce Street, Philadelphia, PA 19104, Office: 517 Goddard Labs, Phone: 215.349.8465, Fax: 215.349.5579, detre@mail.med.upenn.edu.

pulsed ASL; pseudo-continuous ASL; background suppression; 3D spiral acquisition; positron emission tomography; mild cognitive impairment

Introduction

Arterial Spin Labeled (ASL) perfusion MRI uses magnetically labeled arterial blood water as an endogenous tracer to measure regional cerebral blood flow (CBF) (Alsop, et al., 2015; Detre, et al., 1992; Williams, et al., 1992), which is tightly coupled to regional brain metabolism (Raichle, 1998). A number of different strategies exist for magnetic labeling of arterial blood and for measuring the effects of labeling on brain signals, and combinations of these techniques can give rise to various acquisition protocols. Among the labeling approaches, pseudo-continuous ASL (PCASL) (Dai, et al., 2008) is currently the recommended strategy (Alsop, et al., 2015) because of its superior efficiency and compatibility with modern MR hardware. On the other hand, pulsed ASL (PASL) (Wong, et al., 1998) has been available for almost 2 decades and is more widely employed due to its ease of implementation. ASL MRI has frequently been carried out using a two-dimensional (2D) single-shot echo planar imaging (EPI) readout because of its speed and sensitivity, though the effects of ASL can be measured with any method. Three-dimensional (3D) readout schemes such as 3D GRASE (Gunther, et al., 2005) and 3D spiral (Dai, et al., 2008) can be optimally combined with background suppression (BS) techniques of static brain water (Fernandez-Seara, et al., 2008; Garcia, et al., 2005; Maleki, et al., 2012; Ye, et al., 2000) to increase ASL temporal signal-to-noise ratio (Vidorreta, et al., 2013; Ye, et al., 2000) by reducing physiological noise (Wu, et al., 2009).

ASL MRI has demonstrated sensitivity to regional alterations in brain function in patients with Alzheimer's disease (AD) (Alsop, et al., 2008; Alsop, et al., 2000; Chao, et al., 2009; Chen, et al., 2011b; Dai, et al., 2009; Du, et al., 2006; Wang, et al., 2013; Xu, et al., 2007). Significant hypoperfusion has been demonstrated in posterior cingulate cortex (PCC), precuneus, inferior parietal, and lateral prefrontal cortices and the effects are largely independent of underlying gray matter atrophy (Hu, et al., 2010; Johnson, et al., 2005; Mattsson, et al., 2014; Wolk and Detre, 2012). Hypoperfusion has also been reported in patients with amnestic mild cognitive impairment (MCI) (Johnson, et al., 2005; Wang, et al., 2013; Xekardaki, et al., 2015), which is a heterogeneous clinical category consisting of individuals with prodromal AD, and a significant minority of subjects who have cognitive deficits from other etiologies and will not demonstrate pathological evidence of AD in the future. However, hypoperfusion in this population is less pronounced than in AD, and a few studies have also reported increased perfusion in the medial temporal lobes of patients with MCI and early AD (Alsop, et al., 2008; Dai, et al., 2009). Areas of hypoperfusion largely overlap with regions of hypometabolism reported for ¹⁸F-fluorodeoxyglucose Positron Emission Tomography (¹⁸F-FDG PET) (Chen, et al., 2011b; Verfaillie, et al., 2015), leading to the notion that ASL MRI might provide a less invasive and less costly substitute for ¹⁸F-FDG PET as a measure of regional brain function in AD.

Previous studies have compared various ASL implementations. PCASL was shown to have higher SNR than PASL and higher tagging efficiency than continuous ASL (CASL) (Wu, et al., 2007). It was also reported to have superior test-retest reliability than CASL and PASL (Chen, et al., 2011a). Vidorreta et al compared 2D PCASL with 3D PCASL methods (GRASE and spiral acquisitions, with and without BS), demonstrating that BS 3D methods provide higher temporal and spatial SNR in healthy subjects (Vidorreta, et al., 2013). More recently, PCASL with 3D GRASE demonstrated better reproducibility than 2D EPI acquisition for scans acquired 4 weeks apart, though 2D EPI demonstrated better, although not statistically significant spatial correlation with 15 O H₂O PET CBF (Kilroy, et al., 2014).

Comparisons have also been made between ASL with ¹⁸F-FDG PET. Perfusion measured by 2D PCASL was found to correlate well with glucose metabolism measured by PET in a healthy cohort, albeit with significant regional variability (Cha, et al., 2013). In patients with AD, good agreement in hypoperfusion and hypometabolism measured by 2D PCASL and ¹⁸F-FDG PET, respectively, was demonstrated (Chen, et al., 2011b; Musiek, et al., 2012). High agreement was also reported between ¹⁸F-FDG PET hypometabolism and BS PCASL with 3D spiral acquisition hypoperfusion patterns in patients with frontotemporal dementia and Alzheimer's disease compared to controls (Verfaillie, et al., 2015).

In this study we compared ASL MRI data acquired from patients with MCI and elderly control subjects using single-shot PASL acquired with 2D EPI using parameters matched to the ASL data in the Alzheimer's Disease Neuroimaging Initiative (ADNI, http://adni.loni.usc.edu), PCASL with 2D EPI readout using parameters previously demonstrated to be sensitive to regional CBF changes in Alzheimer's disease (Chen, et al., 2011b; Musiek, et al., 2012; Xie, et al., 2016; Zhang, et al., 2012), and PCASL with BS 3D spiral imaging, from here on referred to as 2D PASL, 2D PCASL and BS 3D PCASL, respectively. The goal of this study was to assess the effects of different labeling and acquisition strategies on individual subject data as well as the sensitivity of the methods in detecting group differences between MCI patients and controls. The methods were also compared to ¹⁸F-FDG PET, which represents a more established method for distinguishing between control and patient groups based on regional neural activity.

Material and Methods

Cohort

A sample of 24 amnestic MCI patients and 22 controls were recruited from the Penn Memory Center specifically for this study. As part of their evaluation, each participant underwent an extensive evaluation, including medical history and physical examination, neurological history and examination, and psychometric assessment. All patients had at least the following neuropsychological measures: Mini-Mental State Examination (MMSE) (Folstein, et al., 1975) (reported in Table I); Digit Span subtest of the Wechsler Adult Intelligence Scale III (Wechsler, 1987); Category fluency (animals) (Spreen and Strauss, 1998); Consortium to Establish a Registry for Alzheimer's Disease (CERAD) Word List Memory (WLM) test (Morris, et al., 1989); Trail Making Test (TMT) A and B (Reitan, 1958); and a 30-item version of the Boston Naming Test (BNT) (Kaplan, et al., 1983). Additionally, relevant blood work and brain imaging studies were evaluated. Almost all

patients had structural imaging available, usually MRI, that was used as part of the clinical decision-making, largely to rule out other causes of MCI. Clinical diagnosis was determined by review of the above data at a consensus conference attended by neurologists, psychiatrists, and neuropsychologists. The scans obtained as part of the study presented here were not used for making the clinical diagnosis.

The diagnosis of amnestic MCI was made following the criteria outlined by Petersen and others (Petersen, 2004; Petersen, et al., 2009; Winblad, et al., 2004). Patients had to have a memory problem, generally intact cognitive functioning and activities of daily living, objective evidence of memory impairment on cognitive testing, and not qualify for a diagnosis of dementia. There was no strict cut-off for the degree of memory impairment as clinical judgment accounting for the premorbid status of the patient and performance on other cognitive tests weighs into decisions of objective impairment (Petersen, 2004). However, in general these patients performed around 1.5 standard deviations (SDs) below age-adjusted means on verbal and/or non-verbal memory tests. Controls were defined by an absence of significant cognitive complaints, normal performance on age-adjusted cognitive measures, and consensus conference designation as 'normal'. Inclusion criteria included age between 50–85, greater than 7 years of education, and English speaking at an early age. Participants were excluded if they had a history of clinical stroke, significant traumatic brain injury, alcohol or drug abuse/dependence, prior electroconvulsive therapy, and any significant disease or medical/psychiatric condition that was felt to impact neuropsychological performance.

The human subjects' research in this study was performed in compliance with the Code of Ethics of the World Medical Association (Declaration of Helsinki) and the standards established by the Institutional Review Board and the National Institutes of Health. All subjects provided informed consent for this study.

MRI and PET data acquisition

¹⁸F-FDG PET and MRI scans of 19 patients and 6 controls were obtained on the same day. One patient and one control had their MRI scan obtained 6 days after and 13 days before their PET scans, respectively. ASL MRI from the remaining 4 patients and 15 controls were obtained with mean \pm SD intervals of 430.5 \pm 36.7 days and 411 \pm 41.3 days, respectively, after their PET scan.

MR imaging and processing

MR images were acquired on a 3-Tesla Siemens Trio MRI scanner (Erlangen, Germany) equipped with a product 32-channel head coil. We did not control for modulators of CBF such as coffee, smoking or time of the day. This could be a potential confound, although is unlikely to impact comparisons across ASL methods since they were performed in the same scanning session. High-resolution sagittal T1 images were collected using 3D magnetization-prepared rapid gradient echo (MPRAGE) with the following parameters: repetition time (TR)=1.9s, echo time (TE)= 2.89ms, inversion time=900ms, flip angle=9°, bandwidth=170Hz/px, voxel size= $1 \times 1 \times 1 \text{mm}^3$, Field-of-view (FOV)= $256 \times 256 \text{mm}^2$, 176 slices. ASL data were collected with the three selected methods after instructing subjects to

remain still to the best of their ability. Each ASL scan had an approximate duration of 6 minutes. Details of the ASL sequences are as follows:

2D PASL—PASL data was acquired following the ADNI protocol (http://adni.loni.usc.edu/ methods/documents/mri-protocols/) using a FAIREST scheme (Lai, et al., 2001) combined with QUIPSSII (Luh, et al., 1999) with TI/TI1=1.9s/700ms for accurate bolus definition. A 2D EPI readout was employed with TR/TE=3.4s/12ms, FOV=256×256mm², matrix size=64×64, in plane resolution=4×4mm², bandwidth=2368 Hz/px, phase partial Fourier=6/8, EPI factor=64. 24 slices of 4mm-thickness and a distance factor of 25% were acquired. 52 label-control pairs were acquired. Unlike in ADNI, a separate M0 image was not acquired, and hence the mean of the control images (images acquired with control labeling) was used as M0 image for CBF quantification.

2D PCASL—The PCASL labeling was implemented using a labeling time=1.5s, post labeling delay (PLD)=1.5s, and labeling plane offset=9cm. 2D EPI readout parameters for 2D PCASL were: TR/TE=4s/18ms, FOV=220×220mm², matrix size=64×64, in plane resolution=3.4×3.4mm², bandwidth=3004 Hz/px, EPI factor=64. 18 slices with 6mm thickness and a 20% distance factor were acquired. 45 label-control pairs were acquired. Similar to 2D PASL data, the mean of the control images was used as M0 image.

BS 3D PCASL—The PCASL labeling was implemented using a labeling time=1.5s, PLD=1.5s, and labeling plane offset=8cm. 3D–PCASL data were obtained with TR/ TE=4.5s/11ms, FOV=220×220mm², matrix size=64×64, voxel size=3.4×3.4×5 mm³. 26 nominal partitions were acquired with a centric encoding scheme and slice Partial Fourier=6/8, using a stack of 2-interleaf spirals with maximum slew rate=120T/m/s and maximum gradient amplitude=36mT/m. The background suppression scheme consisted of a presaturation module at the beginning of each TR, followed by a slice-selective C-FOCI pulse (Ordidge, et al., 1996) and two additional non-selective hyperbolic-secant pulses played during the PLD time, with timings optimized to suppress the static tissue magnetization signal to 10% of its equilibrium value (Vidorreta, et al., 2013). 40 label-control pairs were acquired. A separate M0 image was obtained without any magnetization preparation for CBF quantification.

Image processing was performed using Statistical Parametric Mapping (SPM8, Wellcome Department of Imaging Neuroscience, London, UK), ASL toolbox (Wang, et al., 2008) and custom MATLAB (The Mathworks Inc., Natick, MA) scripts. First, the raw time series volumes for each method were realigned to correct for head motion using the method proposed by Wang (Wang, 2012). Briefly, the method consists of estimating a 6-parameter rigid-body transformation for each volume to align it to the first volume of the series. Thereafter the spurious motion component caused by systematic label/control switching is regressed out from the estimated parameters before applying the transformation to the volumes. Next, the mean control images were automatically coregistered to the high-resolution T1 images using SPM8 with normalized mutual information objective function, and the registrations were manually checked and corrected using a boundary-based approach (Greve and Fischl, 2009) in the case of errors. Misregistrations were mostly noted in the case of 3D PCASL (~43%) and were much less common in 2D PCASL (~20%). The PASL mean

control images showed poor tissue contrast and unidentifiable boundaries, making manual modification challenging. Instead, results of the automated coregistration algorithm were used. The resulting transformations were then applied to the corresponding label and control images. In parallel, the structural images were segmented into gray matter (GM), white matter (WM) and cerebro-spinal fluid (CSF) tissue probability maps (TPMs) using the "New Segment" tool in SPM8. The GM and WM TPMs were subsequently resliced to the native space, smoothed by an isotropic Gaussian kernel with FWHM=5mm and then thresholded to 0.75 to construct binary masks to be used to extract mean CBF in GM and WM. In addition, to create a mask to extract whole-brain CBF, the sum of smoothed gray matter and white matter TPMs was thresholded to 0.75. Finally, the Diffeomorphic Anatomical Registration using Exponentiated Lie algebra (DARTEL) method (Ashburner, 2007) in SPM8 was implemented to create a group specific template combining each subject's segmented gray and white matter TPMs. The created template was subsequently registered to the MNI152 space using a linear affine transformation. The two transformations were later used to map the CBF maps and gray matter TPM to the MNI space.

CBF quantification for each control-label pair was obtained using

$$\text{CBF}(\text{ml}/100\,\text{g/min}) = \frac{60 \times 100\,\lambda\Delta\text{M} \text{ e}^{\omega/T_{1,\text{blood}}}}{2\,\alpha\tau M_0}$$

and

$$\text{CBF}(\text{ml}/100\,\text{g/min}) = \frac{60 \times 100\,\lambda\Delta\text{M}}{2\,\alpha\,T_{1,\text{blood}}M_0(1 - e^{-\tau/T_{1,\text{blood}}})}$$

for PASL and PCASL, respectively. In these equations, M is the control-label difference, λ is the brain/blood partition coefficient, ω is the post labeling delay, $T_{1\text{blood}}$ is the T_1 of blood, α is the labeling (tagging) efficiency, M_0 is the equilibrium magnetization of the brain and τ is the labeling duration. Assumed parameter values followed the recommendations in Alsop et al. (Alsop, et al., 2015). In particular, $\lambda = \lambda_{\text{mean}} = 0.9 \text{ ml/g}$ (mean λ for gray and white matter), T_{1blood} =1650 ms. For PASL, $\omega = 1.9$ s, $\tau = 700$ ms and a = 0.98 (Alsop, et al., 2015), and M_0 was made equal to the mean of the control images. For the PCASL techniques, $\omega = \tau = 1.5$ s. The labeling efficiencies for 2D PCASL and BS 3D PCASL were assumed to be 0.85 (Alsop, et al., 2015) and 0.72 (Vidorreta, et al., 2013) respectively. Similar to PASL, M_0 was made equal to the mean of the control images for 2D PCASL. In the case of BS 3D PCASL, the separately acquired M_0 was coregistered to the mean control image. In each case, the M_0 images were smoothed using a 5mm isotropic kernel to avoid misregistration errors (Alsop, et al., 2015) before being used in the CBF calculation. The resulting CBF time series data was then averaged over time to obtain the mean CBF map. A binary mask comprising of gray matter, white matter and ventricular CSF was created and applied to the mean CBF map to set out-of-brain voxels to zero. The CBF maps and the GM TPMs were mapped to the MNI space using the DARTEL template and affine transformation. Images in MNI space were smoothed with a FWHM=8mm isotropic Gaussian kernel to better account for errors due to voxel mismatching with normalization

across subjects. The MNI-normalized smoothed GM TPM was thresholded to 0.4 to create a gray matter mask and region of interest (ROI) analyses for each subject were performed only within these masks to restrict ROI values to voxels comprised of predominantly gray matter.

To account for incomplete T1 recovery (Alsop, et al., 2015) for the 2D acquisition strategies and T2*/T2 decay (Cavusoglu, et al., 2009) in the case of 2D/3D acquisitions, respectively, extracted ROI CBFs were further multiplied by

$$\kappa = (1 - e^{-\mathrm{TR}/T_{1,\mathrm{tissue}}}) \times \frac{\lambda_{\mathrm{tissue}}}{\lambda_{\mathrm{mean}}} \times \frac{e^{\mathrm{TE}/T_{2,\mathrm{blood}}^*}}{e^{\mathrm{TE}/T_{2,\mathrm{tissue}}^*}}$$

for 2D acquisitions and

$$\kappa {=} \frac{\lambda_{\rm tissue}}{\lambda_{\rm mean}} \times \frac{e^{{\rm TE}/T_{2,\rm blood}}}{e^{{\rm TE}/T_{2,\rm tissue}}}$$

for 3D acquisitions. The goal of this is to correct for variable TR and TE across ASL sessions. Tissue specific parameters were chosen based on whether the ROIs were in GM or WM. We used $\lambda_{GM}/\lambda_{WM}$ 0.98/0.82 (Herscovitch and Raichle, 1985), $T_{1,GM}/T_{1,WM} = 1331/832$ ms, $T_{2,GM}/T_{2,WM}/T_{2,blood} = 80/110/186$ ms $T_{2}^*,_{GM}/T_{2}^*,_{WM}/T_{2}^*,_{blood} = 44.2/44.7/43.6$ ms (Cavusoglu, et al., 2009; Chen and Pike, 2009; Wansapura, et al., 1999). For global CBF measurements, average of the parameters for GM and WM were used.

PET imaging and processing

¹⁸F-FDG PET/CT was acquired on a Philips Gemini TF PET/CT scanner (Amsterdam, Netherlands). Participants were instructed not to eat for at least 4 hours prior to their PET scan. Blood glucose levels were tested and individuals with blood glucose level greater than 180 mg/dl were excluded from the study. Participants then received an intravenous injection of 5.0 ± 0.5 mCi of ¹⁸F-FDG as they rested quietly in a dimly lit room with their eyes open but with minimal sensory stimulation. Approximately 30 minutes after the injection, a 30-minute 3D emission scan was obtained (256mm FOV, 128×128 matrix, 2×2×2mm voxel size). Line-of-response row-action maximum likelihood algorithm reconstruction using sharp setting was performed followed by CT attenuation correction.

PET images were coregistered to the T1 images using SPM8 with normalized mutual information objective function similar to the ASL data and the quality of the coregistration was manually verified. Thereafter the images were normalized to the MNI space using the DARTEL template following a procedure similar to the mean CBF maps described above. This involves smoothing by an isotropic Gaussian kernel with FWHM=8mm and use of subject specific gray matter masks to restrict the analysis within the gray matter. Finally, SUVR maps were generated using whole cerebellum as reference tissue. Regional uptake was measured from SUVR maps using the same methods applied to the CBF maps.

Details of statistical Analysis

Comparison of temporal signal-to-noise ratio (tSNR)

For each subject, the mean CBF within the whole-brain mask was computed for each CBF volume of the time series. The tSNR was computed as the ratio of the mean and standard deviation of these values. The tSNR measures for each acquisition method and for each group were compared statistically using a two-way mixed ANOVA with the within-subject factor of 'ASL modality' and the between-subject factor of 'diagnosis'. Post-hoc comparisons were carried out using Bonferroni correction. The statistical analysis was performed using SPSS Statistics, version 23 (IBM Corp). To assess the effect of subject motion on tSNR, average translation and rotation was estimated from the parameters of the rigid-body realignment of the image time series for each subject and Spearman's correlation coefficient of the quantities with tSNR were computed and reported. Motion in controls and patients was compared using a mixed ANOVA model with 'ASL methods' and 'type of motion' (translation or rotation) as repeated measures and 'diagnosis' as between-group variable.

Comparison of presence of outliers

While tSNR assesses the overall variability in the CBF time series, it does not indicate the number of outliers in the series that are markedly different from the remaining volumes and therefore likely dominated by artifacts. Hence, in addition to computing tSNR, we counted the number of outliers present in the CBF time series and compared the results across methods. A volume was considered an outlier if its mean CBF was 1.5 interquartiles below the first quartile or above the third quartile of the time series (Tukey, 1977). The goal of this analysis was two-fold: First, we wanted to assess if non-BS techniques are more susceptible to have outlier label/control pairs compared to BS techniques. The second goal was to assess the rate of outliers in the BS 3D implementation, since that provides an indication of potential challenges in multishot 3D acquisitions, where there is less flexibility of discarding contaminated volumes. The number of outliers for each acquisition method was compared using Friedman test. Post hoc pairwise comparisons were conducted using Wilcoxon signed rank test.

Comparison of Gray Matter to White Matter Contrast Ratio

The mean CBF within gray matter and white matter was extracted from the mean CBF maps in native space and their ratio was computed. Similar to tSNR, the contrasts for each method were also compared with a two-way mixed ANOVA.

Comparison of global CBF obtained using each ASL method

Since all ASL methods are designed to quantify the same physiological parameter (CBF), we compared the global CBF values obtained using each method. Mean CBF values for each method and diagnostic condition were reported and compared using a two-way mixed ANOVA similar to tSNR. Bland-Altman plots were also shown for pairwise comparisons of the global CBFs obtained by the three methods. Additionally, pairwise correlations were computed and reported. Statistical tests for determining the significance of the difference in

correlation coefficients were performed using the method described by Steiger (Steiger, 1980) and evaluated using the software developed by Lee and Preacher (Lee and Preacher, 2013). The method converts each correlation coefficient into a z-score using Fisher's r-to-z transformation, computes the asymptotic covariance of the estimates and subsequently uses the quantities in an asymptotic z-test. We also compared the regional distribution of CBF between the three modalities based on the control subjects, since prior studies, e.g. (Lovblad, et al., 2015), showed that differences in labeling can lead to regional variability and difference in signal intensity.

Comparison of sensitivity to differentiate MCI patients from controls between the three ASL methods and PET

Standardized uptake value ratio (SUVR) for PET images were computed by dividing each voxel's SUV by the mean SUV within the cerebellum. In the case of ASL, cerebellum is often not in the area of coverage and, hence, the mean CBF within the cortical GM was used for computing the relative CBF. The mean relative CBF and mean SUVR within PCC, precuneus, and hippocampus were extracted for ASL and PET for each subject. These ROIs have most consistently been shown to be sensitive to AD (Wolk and Detre, 2012). Statistical significances of group differences were evaluated using non-parametric Mann-Whitney U-tests. Effect sizes were calculated by converting the Mann-Whitney U to a z-score and then dividing by the square root of the total number of subjects.

RESULTS

Demographic characteristics of the study cohort are summarized in Table I. There were no significant differences in age or education between MCI and control subjects. Figure 1 shows the mean CBF maps of a representative control and MCI patient obtained using the three ASL methods. For comparison, the SUVR maps of the same subjects are also shown.

Comparison of temporal signal to noise ratio (tSNR)

TSNR values for the three ASL methods and for the two cohorts are reported in Table II. BS 3D PCASL had an almost 2.5-fold increase in tSNR as compared to the 2D methods. A twoway mixed ANOVA demonstrated that there was a statistically significant main effect of method (F(1.43,62.79)=152.4, p<0.001, using Greenhouse-Geisser correction). Post-hoc tests using Bonferroni correction for multiple comparisons showed that BS 3D PCASL had significantly higher (p<0.001) tSNR compared to the two 2D methods, and 2D PCASL tSNR was also significantly higher (p=0.022) compared to 2D PASL. There was also a significant main effect of diagnosis on tSNR (F(1,44)=4.328, p=0.043) with higher tSNR in control subjects than MCI for all the three methods. There was no significant interaction between method and diagnosis. Mean translations in 2D PASL, 2D PCASL and BS 3D PCASL were all significantly anti-correlated with corresponding tSNRs, with Spearman's correlation coefficients of -0.42 (p=0.004), -0.53 (p<0.001) and -0.36 (p=0.013) respectively. However, the mean rotation was only significantly correlated with tSNR in the case of 2D PCASL with Spearman's correlation coefficient of -0.40 (p=0.006). Translations and rotations in patients were higher than in controls, but the differences did not reach statistical significance (p=0.137).

Comparison of presence of outliers

The mean±SD, as well as the range of detected outlier CBF volumes are reported in Table II for each method and diagnostic condition. There was a statistically significant difference in number of outliers, χ^2 =19.141, p<0.001. Post-hoc analyses with Wilcoxon signed rank test and Bonferroni correction demonstrated that the 2D PASL and 2D PCASL had significantly higher number of outliers compared to BS 3D PCASL (p=0.004 and p=0.029, respectively). Both the PCASL methods exhibited a similar range (0–4) of outliers, whereas the range was higher (0–6) in 2D PASL.

Comparison of Gray Matter to White Matter CBF Contrast Ratio

The GM-WM CBF ratios are shown in Table II. The GM-WM CBF ratio was lowest in BS 3D PCASL almost certainly due to markedly increased blurring, which is evident in Figure 1. The blurring is the result of the T2 magnetization decay in time resulting in broadening of the point spread function in kz space. This blurring also reduces apparent CBF in basal ganglia and thalamus. A two-way mixed ANOVA demonstrated a significant main effect of method (F(1.45,60.21)=75.8, p<0.001, with Greenhouse-Geisser correction). Post-hoc tests showed that GM-WM contrast in 2D PCASL and 2D PASL was significantly higher than BS 3D PCASL (p<0.001). There was no effect of diagnosis or any interaction between method and diagnosis.

Comparison of global CBF obtained using each ASL method

The global CBF values from the three methods are listed in Table II. There was no statistically significant difference in CBF between methods (p=0.124), diagnosis (p=0.194) or interaction between methods and diagnosis (p=0.292). Variability in CBF derived from BS 3D PCASL was the highest across the cohorts. Figure 2 shows the Bland-Altman plots for pairwise comparisons of the global CBFs obtained with the three methods. The differences in CBF values obtained using 2D PASL and BS 3D PCASL had higher variance compared to the differences obtained using either the same labeling or same acquisition strategies. Mean CBF values for whole brain obtained using the three methods were also highly correlated. The correlation between the two PCASL methods (2D and 3D) was highest with Pearson's correlation coefficients of 0.79, p<0.0001. The correlation between the two 2D methods (PASL and PCASL) was also high, with a coefficient of 0.75, p<<0.0001. The correlation between BS 3D PCASL and 2D PASL was considerably lower (0.49, p<0.001). When the correlations were compared, both BS 3D PCASL vs 2D PCASL and 2D PCASL vs 2D PASL correlations demonstrated significantly higher correlation compared to BS 3D PCASL vs 2D PASL (p<0.001). Figure 3 shows the differences in relative CBFs in controls between the three modalities. 2D PASL had higher relative CBF in the cerebellum, occipital cortex and basal ganglia, and lower perfusion in the frontal lobe compared to the other two modalities. On the other hand, the distribution of relative CBF with BS 3D PCASL was opposite. The distribution of 2D PCASL was somewhere in between the two modalities.

Comparison of sensitivity to differentiate MCI patients and controls between the three ASL methods and PET

Figure 4 shows the average CBF maps across the cohort separately for controls and MCI patients as well as their differences. The bottom row of the same figure shows the average SUVR in controls and MCI patients and their difference. Glucose metabolism in the MCI cohort was lower in the posterior cingulate cortex, bilateral and medial parietal lobe, precuneus, and lateral prefrontal cortices. The hypometabolism in PCC was consistent with the hypoperfusion in both the PCASL methods whereas that in the medial parietal and the precuneus was consistent only with hypoperfusion pattern in 3D PCASL. The hypometabolism in the frontal lobe in the MCI cohort was not observed with the PCASL methods, but was reflected by PASL (although not statistically significantly).

Figure 5 shows box plots of the mean relative CBF values for each ASL methods and mean SUVR for PET, and Table III summarizes the effect sizes in each of the three a-priori ROIs: PCC, precuneus and hippocampus. ¹⁸F-FDG PET displayed significant differences in metabolism between control and MCI groups for all three regions. Likewise, both PCASL methods displayed relative hypoperfusion in the MCI cohort compared to controls. Group differences in BS 3D PCASL were statistically significant in all three ROIs, while only the PCC and hippocampus were significant with 2D PCASL. No regions displayed a significant difference with 2D PASL. In fact, the precuneus was the only region in which MCI was associated with lower perfusion than the controls in absolute terms. As evident from the box plots, the variability of the relative mean CBF values in BS 3D PCASL was much lower compared to the 2D methods, and was comparable to PET.

Discussion

Background-suppressed 3D PCASL provided superior sensitivity for detecting group differences in regional CBF between MCI patients and elderly controls than the other ASL methods tested, with effect sizes comparable to those seen with ¹⁸F-FDG PET imaging. PCASL methods demonstrated higher temporal signal-to-noise ratios than PASL, likely due to improved labeling in the case of 2D PCASL, along with BS in the case of 3D PCASL. The sequence performance results obtained in this clinical population are generally consistent with those previously reported in young controls (Vidorreta, et al., 2013), though, as expected, the tSNR values were lower, which is likely attributable to lower CBF and higher physiological noise in patients and elderly controls considered for this study. Across control and patient cohorts, there was good agreement between global CBF values obtained from each method. However, agreement on an individual subject level for different ASL schemes were found to be higher when using either the same labeling strategy (PCASL) or the same imaging strategy (2D EPI), and was significantly reduced when both the labeling and the data acquisition strategy differed. Thus, while data acquired with different ASL implementations are in general agreement, as they all measure the same physiological parameter (i.e. CBF), differences in labeling and imaging strategies cumulatively impact the degree of agreement. Multisite and multiplatform studies using ASL as a biomarker for neurodegeneration would clearly benefit from closely matched or identical ASL implementations, as has previously been suggested for normative data (Mutsaerts, et al.,

2015). In particular, absolute CBF from significantly different ASL protocols should be compared, combined and interpreted with caution.

BS 3D PCASL showed higher variability in CBF across subjects. A possible explanation is that the labeling efficiency at the labeling offset of 8 cm used in 3D may be more variable than the offset of 9 cm used in the 2D protocol. More recent data suggested that a lower labeling location provides more consistent results (Vidorreta, et al., 2017). Note that this higher variability does not affect the group discrimination analysis, which was performed using relative CBF.

Observed hypoperfusion in PCC, precuneus and hippocampus in the MCI cohort using the PCASL methods was consistent with the levels of hypometabolism as measured by PET, although this difference did not reach significance with 2D PCASL in the precuneus. Overall, PCC seemed to be the most sensitive ROI for differentiating MCI patients and elderly controls with PCASL. A similar observation was made previously (Xekardaki, et al., 2015), where PCC predicted cognitive deterioration in controls destined to develop MCI. In the case of PASL, ROI data failed to show group differences within the sample, likely reflecting the relatively poor sensitivity of this ASL variant, potentially compounded by coregistration errors due to poor tissue contrast that precluded manual correction. On the other hand, BS 3D PCASL demonstrated the largest effect sizes amongst the ASL variants (Table III). The dispersion of the relative CBF values in BS 3D PCASL from figure 5 appeared lower compared to 2D PCASL and 2D PASL methods and comparable to PET. Effect sizes were mostly medium (0.3–0.5), which is likely due to the low severity of disease in the MCI cohort studied. Indeed, effect sizes of the BS 3D PCASL were very similar to the "gold standard" ¹⁸F-FDG PET measurements.

TSNR in the control cohort was significantly higher than in the patient group, which we attribute to a combined effect of higher signal (CBF) in controls and higher physiological noise in patients, likely due to motion. Although motion differences across groups did not reach significance, the translations and rotation values were indeed higher in the MCI patients and the observed correlations between tSNR and motion parameters further suggests its impact on tSNR. Higher tSNR in controls versus patients has also been reported previously in stroke patients (Fernandez-Seara, et al., 2008). Outlier volumes were also detected in the CBF time series of each acquisition protocol, with greater numbers in the patient group, again likely due to increased physiological noise, and despite greatly increased tSNR in BS 3D PCASL data, outlier volumes were still present. While many 3D ASL methods use multishot acquisitions is the increased ease and efficiency of identifying and excluding outlier volumes in post-processing of ASL time series.

Manual coregistration was needed to correct errors in the automatic coregistration produced by SPM, particularly for 3D PCASL data. SPM8 relies on intensity-based methods for intermodal image registration using a normalized mutual information criterion. To provide optimal registration, the algorithm requires different tissues to have a distinct mean intensity. In EPI readouts, the between-tissue contrast is primarily driven by the amount of T2* contrast accumulated at the echo time (TE). In ASL, the typically employed TE in EPI is

shorter than in conventional blood oxygenation level dependent (BOLD) EPI acquisitions, resulting in lower contrast between tissues, which can occasionally cause errors in the registration output as observed in our data. In instances where tissue contrast was insufficient for successful automatic coregistration, registration was manually corrected using a boundary-based approach (Greve and Fischl, 2009). For 3D PCASL, the through-plane blurring presents additional challenges for the coregistration algorithm and hence required more manual correction. On the other hand, in PASL the effect of the FAIR inversion pulse together with the shorter TE (compared to 2D PCASL) resulted in poorly identified tissue boundaries precluding manual registration, so automatic coregistration results from SPM were used without additional correction. As a result, the PASL data may have had greater coregistration errors than the other methods, which could ultimately have influenced ROI results and the discrimination between controls and MCI patients.

We conducted our analyses using relative CBF to facilitate comparison with PET, for which standard analysis uses normalization of ¹⁸F-FDG uptake to cerebellum to account for differences in dose and overall cerebral delivery of ¹⁸F-FDG between subjects. In addition, mean CBF in gray matter or whole brain is highly variable even in healthy populations (Dolui, et al., 2016), and use of relative CBF eliminates this variability. Moreover, absolute CBF change within a specific ROI reflects both global and local CBF effects. While use of relative CBF obviously precludes detection of absolute CBF changes, which have also been observed with ASL in AD and MCI (Asllani, et al., 2008), it enables comparing regional changes in CBF in disease-specific patterns (Chen, et al., 2011b). In particular, reduction of relative CBF within specific ROI demonstrates lower metabolic function in that region that cannot be explained by global reduction in CBF. While cortical normalization of PET may have been more equivalent to the CBF measurements, this reference yielded a lower effect size for group differences within a priori ROIs compared to cerebellar normalization, as has been reported previously (Dukart, et al., 2010). Since ¹⁸F-FDG was being used in this context as a "gold standard," we chose the more optimized cerebellar normalization.

The nominal PLD in this study was shorter than the recently recommended PLD of 1.8-2 s in an ASL white paper (Alsop, et al., 2015). It should be noted that the recommended PLD was for BS ASL acquisitions, which have higher signal compared to their non-BS counterparts. In the case of non-BS 2D acquisitions, the choice of a shorter PLD represents a tradeoff between two factors: i) vascular transit artifacts and/or underestimation of CBF if the PLD is too short compared to the arterial transit time (ATT), and ii) loss of sensitivity because the signal strength decreases due to T1 decay of the spin label during the PLD. Further, since the 2D acquisition occurs slice by slice from inferior to superior direction, the last slice was acquired approximately 800ms after the initial post labeling delay, making the effective PLD in the superior slices much longer than the nominal value, with the mean PLD across the brain approximately 1.9s. Although 1.5s might be lower than the ATT for watershed regions of the brain that are supplied by the most distal branches of their arteries, particularly in elderly patients, those are not the regions of interest for this study. Instead, the ROIs we considered were located in the medial temporal and parietal regions, which are characterized by relatively shorter ATT (MacIntosh, et al., 2010). We did not observe any visible transit artifacts when inspecting the mean CBF maps from each subject. Finally, we based our analyses on mean CBF values in relatively large ROIs and the mean CBF within

an ROI would not change as long as the blood supplying the microvasculature has arrived within the ROI, even if intra-arterial (Qiu, et al., 2012).

The spatial distribution of CBF was different in the three methods. PASL showed higher CBF in the posterior circulation, cerebellum and basal ganglia and lower CBF in the frontal lobe whereas BS 3D PCASL had an exactly opposite distribution. Although much of this difference can be attributed to the different labeling and acquisition schemes (Lovblad, et al., 2015), differences in effective PLD and their mismatch with ATT also likely contributed. ATT can be measured experimentally, and would be highly beneficial for optimal accuracy in CBF quantification. Methods for rapidly measuring ATT have recently been described (Dai, et al., 2013; von Samson-Himmelstjerna, et al., 2016).

The study had few limitations. First, the ASL sequences evaluated not only differed with respect to the labeling and the presence of background suppression, but the voxel sizes were also different, which could have confounded comparisons between methods. Our choice of voxel dimensions was based on the previous literature and our own experience in ASL. The PASL resolution was matched to ADNI ASL protocol to facilitate interpretation of our findings with respect to ADNI data. However, the 2D PCASL replicated a protocol we had successfully applied to Alzheimer's disease in prior work (Chen, et al., 2011b; Musiek, et al., 2012; Xie, et al., 2016; Zhang, et al., 2012). For single shot 3D PCASL, the voxel size was chosen to be smaller than 2D PCASL in the z direction because of its characteristic blurring in that direction. Since our analysis focused on CBF in relatively large ROIs and not at individual voxels, the SNR should be largely independent of voxel resolution as smaller voxels have lower SNR but a larger number is required to cover the same spatial extent. However, some residual effects of voxel size may still be expected due to differences in partial volume contamination and physiological noise.

A second limitation is that PET data of several subjects were separated from ASL acquisition by more than a year. Although this does not undermine the comparison between the concurrent ASL strategies, which was our primary goal, it could confound the comparison of ASL with PET. In particular, since the ASL data for 4 MCI patients were obtained later in the course of their disease than the PET data, group comparisons might be biased to show a greater effect size for ASL. We repeated the group analysis after excluding data from the 4 MCI patients whose ASL data was acquired one year later than the PET data. The restricted comparison did not appreciably change the findings.

Finally, no separate M0 image was acquired for the 2D acquisition techniques, so we used mean of the control images to quantify CBF from control-label perfusion difference, however we corrected for this error using the method suggested by Alsop et al. (Alsop, et al., 2015) to account for incomplete relaxation of tissue magnetization. In addition, CBF quantification was obtained with a fixed value of $T1_{blood}$ and its variability can introduce errors (Dolui, et al., 2016; Hales, et al., 2016; Wu, et al., 2010), though any such errors are unlikely to affect comparison significantly between the ASL variants.

In conclusion, both PCASL and background suppression improve the sensitivity of ASL over PASL acquisitions for detecting group differences between MCI and elderly controls.

In particular, despite increased blurring in the slice dimension, BS 3D PCASL showed the highest discrimination between MCI and control groups, and provided comparable discrimination to ¹⁸F-FDG PET. Thus, BS 3D PCASL methodology may provide an alternative to ¹⁸F-FDG PET that can be obtained during routine MRI in clinical practice and research, and accelerated 3D ASL acquisitions can be used to reduce readout times and blurring.

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Figure 1.

Triaxial view of mean CBF maps using 2D PASL (Row 1), 2D PCASL (Row 2) and BS 3D PCASL (Row 3) of a representative control subject on the left and MCI patient on the right. The bottom row shows the SUVR maps of the same subjects.



Figure 2.

Bland-Altman plots comparing the global CBF measurements obtained using (top) 2D PASL and 2D PCASL, (middle) 2D PCASL and BS 3D PCASL and (bottom) 2D PASL and BS 3D PCASL. The three lines in each case show the mean difference and the 95% confidence intervals.



Figure 3.

Difference in relative CBF in control subjects between (Left) 2D PASL and 2D PCASL, (Middle) 2D PCASL and BS 3D PCASL and (Right) 2D PASL and BS 3D PCASL.

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Figure 4.

Triaxial view of (Left to right) Average CBF maps of the controls, patients and their differences for 2D PASL (Row 1), 2D PCASL (Row 2), and BS 3D PCASL (Row 3). The bottom row shows the average SUVR maps of the controls, patients and their differences (left to right).

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Figure 5.

Box plots of mean relative CBFs and mean SUVR in different ROIs shown in blue for controls and red for MCI patients. Within each box, the three horizontal lines represent the first quartile, median and the third quartile (bottom to top). The two horizontal lines outside each box represent the minimum and maximum values ignoring outliers. Outliers, determined as 1.5 times the interquartile range above (or below) the third (or first) quartile, are shown separately with "+" symbols.

Table I

Demographic and Clinical Characteristics of the study sample (Values reported as Mean±Standard Deviation)

	Control	MCI
Number	22	24
Age, Years	70.9±7.0	74.0±7.4
Gender (F/M)	17/5	7/17
Education, Years	15.2±3.1	16.9±2.7
MMSE ^a	29.3±1.3	27.4±1.8

^astatistically significant difference, p<0.001

Table II

TSNR, #Outlier CBF volumes/subject, GM-WM Contrast Ratio and global CBF for the Control and MCI cohort

		2D PASL	2D PCASL	BS 3D PCASL
TSNR (mean±sd)	Control	2.28±0.96	2.83±1.56	6.44±2.12
	MCI	1.74±0.70	2.05±1.23	5.54±2.14
	Total	2.00±0.87	2.42±1.44	5.97±2.15
#Outliers/Subject (mean±SD (range))	Control	1.05±1.29 (0-4)	0.68±0.72(0-2)	0.14±0.47(0-2)
	MCI	1.83±1.66 (0-6)	1.29±1.12(0-4)	0.67±1.01(0-4)
	Total	1.46±1.53 (0-6)	1.00±0.99 (0-4)	0.41±0.83 (0-4)
GM-WM Contrast Ratio (mean±SD)	Control	2.42±0.72	2.29±0.48	1.39±0.13
	MCI	2.41±0.57	2.31±0.46	1.32±0.13
	Total	2.41±0.64	2.30±0.46	1.35±0.14
Global CBF (mean±SD)	Control	36.51±9.25	38.41±8.77	39.69±11.07
	MCI	34.83±7.62	33.28±7.69	35.68±10.32
	Total	35.63±8.39	35.73±8.53	37.60±10.76

TSNR: temporal signal to noise ratio, GM: grey matter, WM: white matter, CBF: cerebral blood flow, MCI: mild cognitive impairment

Table III

Table showing the effect sizes for Mann-Whitney U test for differentiating Controls and MCI patients in different ROIs.

	Effect Size	
PCC		
2D PASL	0.04	
2D PCASL	0.35	
BS 3D PCASL	0.42	
PET	0.47	
Precuneus		
2D PASL	0.19	
2D PCASL	0.20	
BS 3D PCASL	0.37	
PET	0.38	
Hippocampus		
2D PASL	0.01	
2D PCASL	0.29	
BS 3D PCASL	0.37	
PET	0.42	