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# Arterial Spin Labeling at 3T in semantic dementia: perfusion abnormalities detection and comparison with FDG-PET

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**Abstract.** Arterial Spin Labeling (ASL) is a non invasive perfusion imaging technique which has shown great diagnosis potential in dementia. However, it has never been applied to semantic dementia (SD), a rare subtype of frontotemporal lobar degeneration characterized by the gradual loss of conceptual knowledge, which is actually explored by a now well established marker of SD: <sup>18</sup>fluorodeoxyglucose-positron emission tomography (FDG-PET) imaging. Although ASL and FDG-PET respectively measure perfusion and metabolism, they have been shown to be strongly correlated. We propose to explore the ability of ASL to detect perfusion abnormalities in SD in comparison with FDG-PET. Using patients and healthy subjects data from an ongoing clinical study, we apply our analysis framework starting with visual comparison of ASL and FDG-PET, and focusing on ASL data preprocessing and statistical analysis at the individual and group level. Preliminary results yield concordant observations between ASL and FDG-PET as well as expected hypoperfusions in SD, namely in the left temporal lobe, thus suggesting the potential of ASL to assess perfusion impairments in SD.

**Keywords:** ASL, FDG-PET, MRI, semantic dementia.

## 1 Introduction

Arterial spin labeling (ASL) measures brain perfusion by using blood water protons as an endogenous tracer. An inversion pulse labels the inflowing blood and after a delay called inversion time, a labeled image of the volume of interest is acquired. The subtraction of the labeled image from a control image, i.e., non labeled, reflects the quantity of spins that have perfused the imaged volume, producing a perfusion-weighted (PW) image. The signal to noise ratio of the PW map is increased by acquiring repeated pairs of label/control images. Cerebral blood flow (CBF in ml/100g/min units) is estimated by application of a quantification model.

As it is a non invasive, reproducible and quantitative technique, ASL has been used in numerous clinical applications [1]. Its diagnosis potential has namely been shown in dementia, for instance to differentiate Alzheimer’s disease (AD) from frontotemporal dementia or to predict cognitive decline and conversion from Mild Cognitive Impairment to AD [2,3]. Previous studies have also compared ASL with  $^{18}\text{F}$ fluoro-deoxyglucose positron emission tomography (FDG-PET), which remains the reference technique. Indeed, FDG-PET imaging is particularly sensitive in detecting disease-related functional changes [4] and it has been well established as a marker of degenerative dementia. Though ASL and FDG-PET image different physiological processes, perfusion on the one side and glucidic metabolism on the other side, they are generally tightly coupled at rest [5] and they were shown to identify similar abnormality patterns in AD [6].

To our knowledge, ASL has never been applied to semantic dementia (SD). SD is a rare subtype of frontotemporal lobar degeneration [7] characterized by a gradual loss of conceptual knowledge and of word meaning. The SD syndrome arises out of the temporal lobe, usually showing atrophy often bilateral but mostly left-sided on structural MRI, as well as hypometabolism, also predominantly on the left side, in FDG-PET [8].

In this paper we explore the ability of ASL to detect perfusion abnormalities in SD in comparison with FDG-PET, relying on data extracted from an ongoing study. Section 2 describes the population, the imaging protocols as well the methodological framework for data processing and analysis. Section 3 presents preliminary results before discussion in Sect. 4.

## 2 Materials and Methods

### 2.1 Participants

Fifteen subjects have been included in the study so far: 6 patients fulfilling the SD diagnostic criteria [9] (2 males, 4 females, mean age =  $63.3 \pm 2.4$  yo), and 9 healthy control subjects (3 males, 6 females, mean age =  $60.8 \pm 3.9$  yo), matched for sex and age (Mann-Whitney U test with  $p = 0.28$ ). All the subjects underwent an MRI scan, only patients underwent a FDG-PET examination. In addition, cognitive performance was assessed for all the subjects. Note that these neuropsychological data are not included in the analysis presented in this paper.

### 2.2 Data Acquisition

**Anatomical and perfusion MRI Data** Patients and control subjects were scanned on a 3T whole body Siemens MR scanner (Magnetom Verio, Siemens Healthcare, Erlangen, Germany) with a 32-channel head coil. Anatomical data included a high resolution 3D T1-weighted MPRAGE sequence (3D T1) with the following imaging parameters: TR/TE/TI = 1900/2.27/900 ms,  $256 \times 256 \text{ mm}^2$  FOV and 176 sagittal slices,  $1 \times 1 \times 1 \text{ mm}^3$  resolution, parallel imaging GRAPPA2.

Perfusion data were acquired using a pulsed ASL PICORE Q2TIPS sequence with total scan time of approximately 6 minutes. The imaging parameters were:

TR/TE = 3000/18 ms, flip angle  $90^\circ$ , matrix size  $64 \times 64$ ,  $TI_1/TI_1s/TI_2 = 700/1500/1800$  ms. The labeling slice was 10 cm thick and was placed 3 cm below the acquisition volume. Crusher was set to achieve a cutoff velocity of 4 cm/s. Fourteen axial slices were acquired sequentially from inferior to superior in the AC-PC plane, with  $3 \times 3$  mm<sup>2</sup> in-plane resolution, 7 mm slice thickness and 0.7 mm gap. One control volume plus 60 repetitions, i.e., label/control pairs, (121 volumes), finally composed the ASL data series.

**PET Data** Only patients underwent the FDG-PET protocol imaging which was conducted on a discovery ST PET/CT Imaging system, GE, Healthcare, Waukesha, WI. PET/CT imaging was initiated 30 minutes after intravenous injection of 148 MBq of <sup>18</sup>FDG and was conducted at rest in a dark quiet room. The in-plane PET image resolution was  $2.34 \times 2.34$  mm<sup>2</sup>, with 3.27 mm slice thickness. Non-contrast CT acquisitions were performed (2 mm slice thickness) for anatomical co-registration, attenuation and scatter correction of the PET images.

### 2.3 Data Pre-Processing

Pre-processing of the image data was performed with an inhouse pipeline using MATLAB (v. 7.12.0, The MathWorks Inc., 2011) and the SPM8 toolbox (Wellcome Department of Imaging Neuroscience at University College London, UK) as follows:

**Anatomical data** The anatomical 3D T1 was first denoised using the NL-means algorithm by [10] and corrected for intensity inhomogeneity using SPM. The images were then segmented into grey matter (GM), white matter (WM) and cerebro-spinal fluid (CSF) probability maps using the MNI template tissue probability as an a priori for brain tissue classification. Estimated by this same unified segmentation model SPM routine, spatial normalisation parameters were applied to register the 3D T1 volume to the MNI space.

**ASL data** The ASL data series was motion corrected by a rigid body transform minimising the sum of squared differences cost function. A two-pass procedure first realigned all the control and label volumes onto the first volume of the series, then registered the series to the mean of the images aligned in the first pass.

The motion-corrected ASL series was co-registered to the 3D T1 using a rigid transform. The latter was estimated by maximizing normalised mutual information between the mean control image, i.e., the average of all the realigned control volumes, and the 3D T1 GM map.

Letting aside the first volume of the series,  $pM_0$ , the co-registered ASL images were pairwise subtracted (control - label images) to produce a series of perfusion weighted maps, which was subsequently averaged to produce a perfusion weighted (PW) map.

The PW map was quantified to a CBF map by applying the standard kinetic model [11,12] :

$$f = \frac{\lambda \Delta M}{2\alpha T I_1 \exp(-T I_2 + i d x_{sl} * T I_{sl}) / T_{1b} M_0} \quad (1)$$

where  $f$  is the CBF map,  $\Delta M$  is the perfusion weighted map,  $\lambda = 0.9 \text{ ml.g}^{-1}$  is the blood/tissue water partition coefficient,  $M_0$  represents the equilibrium magnetization of arterial blood and is approximated by the first volume of the series  $pM_0$ ,  $\alpha = 0.95$  measures the labeling efficiency,  $T I_2 = 1800 \text{ ms}$  is the inversion time of the ASL sequence,  $i d x_{sl}$  is the slice index, starting from 0 for the first acquired slice,  $T I_{sl} = 45 \text{ ms}$  is the duration of acquisition of one slice,  $T I_1 = 700 \text{ ms}$  is the temporal width of the bolus,  $T_{1b} = 1500 \text{ ms}$  is the  $T_1$  of blood [13].

Due to its inherent low resolution, the intensity in a single voxel results from a mixture of the GM, WM, CSF measured signals and ASL suffers from severe partial volume effect. These partial volumes effects (PVE) were corrected as described in [14]. This approach considers the perfusion signal in a voxel as a weighted sum of the GM and WM perfusion signals where the weights are given by the 3D T1 brain tissues probability maps. A total variation optimization algorithm estimates the respective contributions of GM/WM perfusion signals yielding high resolution perfusion maps, henceforth called PVE CBF maps.

In order to perform inter-subject voxelwise comparison, the PVE CBF maps, already aligned to the 3D T1 image, were registered to MNI space using the spatial normalisation parameters estimated on the anatomical data.

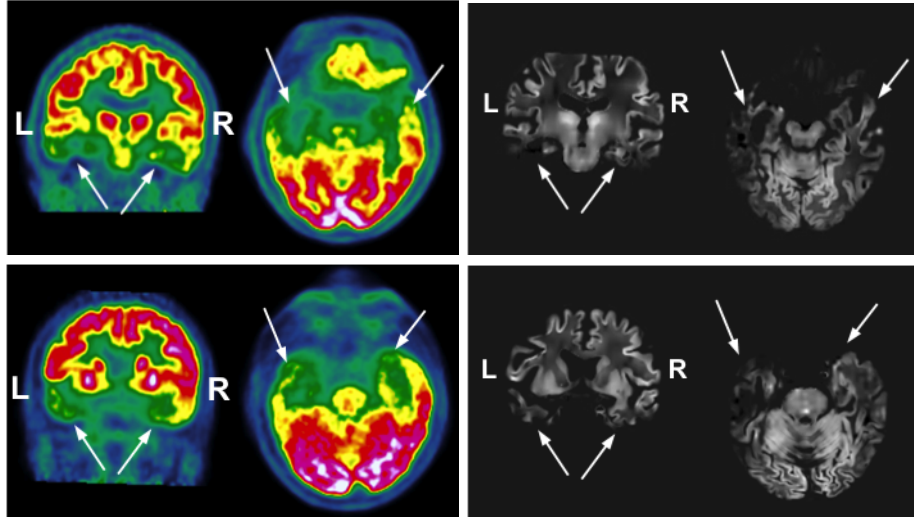
As CBF varies greatly across individuals, the PVE CBF maps were intensity normalised to account for a global interindividual variability. The normalisation factor was set to the mean of the PVE CBF values over the voxels having a grey matter probability superior to 0.7.

**PET Data** The attenuation corrected PET images were coregistered to the 3D T1 image using the normalised mutual information cost function. Spatial normalisation to the MNI space was then performed using the parameters estimated for the 3D T1.

## 2.4 Data Analysis

Data analysis is performed in 3 steps: first, the data were visually evaluated, with a focus on the concordance between perfusion and metabolism as observed in the images. Then ASL data were quantitatively analysed at an individual level to identify perfusion defects in each patient based on a template approach. Last, group analysis was done voxelwise and in regions of interest.

Note that only voxels where the brain tissues probability exceeded 0.5 were tested in the subsequent statistical analyses. For voxelwise comparisons, the data were smoothed with a  $8 \times 8 \times 8 \text{ mm}^3$  Gaussian kernel.



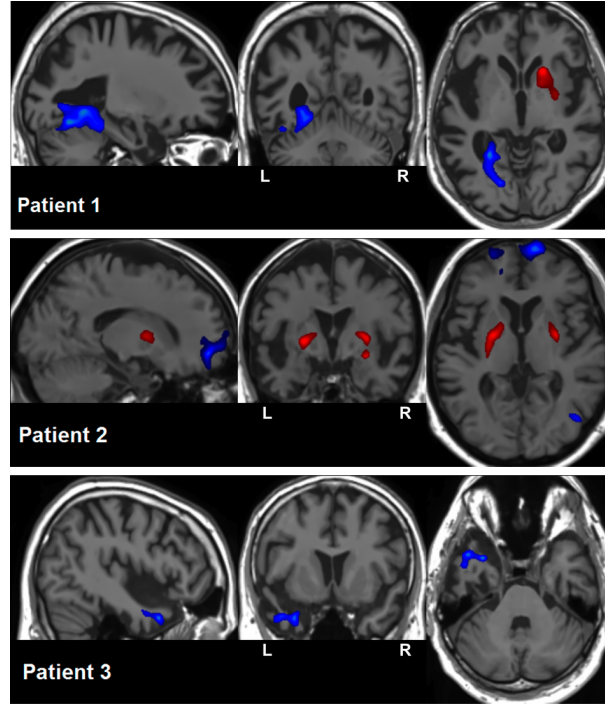
**Fig. 1.** Visual comparison of ASL and FDG-PET illustrated for 2 patients. Left, PET data with metabolism decreasing from red to green; right, PVE CBF maps. The white arrows indicate the hypometabolic and hypoperfused areas.

**Qualitative analysis** For each patient, a radiologist visually inspected the PVE CBF and PET maps seeking to detect any metabolic and perfusion abnormalities. The reading focused on the superior frontal, basifrontal, anterior and posterior temporal, parietal and occipital areas for the right and left hemispheres.

**Voxelwise Individual Statistical Analysis of ASL Data** Hypo- and hyper-perfusions were sought at an individual level by comparing each patient ASL map to an ASL template. Assuming a Gaussian distribution in each voxel, such a template is defined as the mean and associated standard deviation map estimated from a set of input maps. Using the set of smoothed PVE corrected, spatially and intensity normalised CBF maps from the control group as inputs, the ASL template was built as described in [15] and is considered as a reference for brain normal perfusion for an age- and sex-matched population with respect to the patient group.

Voxelwise comparisons between the individual maps and the ASL template were run based on one-tailed t-tests<sup>1</sup> as implemented in the SPM General Linear Model (SPM). Differences were considered significant at the cluster level at a threshold of  $p < 0.05$  (Family-wise-error (FWE) corrected for multiple comparisons at the cluster level with a cluster defining threshold of  $p < 0.001$ ).

<sup>1</sup>  $\frac{v - \hat{\mu}}{\hat{\sigma} \sqrt{1/N + 1}} \sim T^{N-1}$  where  $v$  is the voxel value to be tested, and  $\hat{\mu}$ ,  $\hat{\sigma}$  and  $N$  are respectively the sample mean, variance and size.

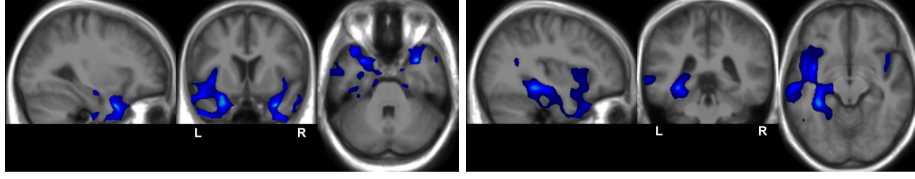


**Fig. 2.** Voxelwise individual statistic analysis: for 3 patients, detected hypoperfusions (blue) and hyperperfusions (red) superimposed on their anatomical map.

**Voxelwise Group Statistical analysis** The smoothed PVE corrected, spatially and intensity normalised CBF maps of both control and patient groups were fed to GLM one-tailed two-sample t-tests analysis. Between group comparisons were considered significant at the cluster level at a threshold of  $p < 0.05$ , FWE corrected, with a cluster defining threshold of  $p < 0.001$ .

**ROIs Group Statistical analysis** Regions of interest (ROIs) were extracted from the PVE corrected, spatially and intensity normalised CBF maps using the Automated Anatomical Labeling (AAL) template [16]. For each hemisphere, the segmented ROIs were: the brain lobes (frontal, temporal, parietal, occipital), the cingulate gyrus, the insula and the deep grey nuclei (putamen, caudate, thalamus). Contrary to the AAL lobe definition, the hippocampus, the parahippocampus and the temporal pole were not included in the limbic but in the temporal lobe as described in [17].

For each subject, the average CBF value was computed over each ROI, left and right, excluding voxels with GM probability less than 0.7. A two-tailed two-samples t-test tested the difference between the control group and the patients group at a significant level of  $\alpha = 0.05$ .



**Fig. 3.** Voxelwise group statistic analysis: superposition of the detected hypoperfusions (blue) to the average anatomy.

### 3 Results

**Qualitative Analysis** The visual inspection of the PVE CBF maps and FDG-PET data showed good agreement between the two modalities as illustrated in Fig. 1. For all the 6 patients, hypoperfusion and hypometabolism were observed bilaterally in the basifrontal as well as in the anterior temporal lobe though the left side was more severely affected. The posterior part of the temporal lobe was particularly impaired on the left side. For all the subjects but one, the left parietal lobe exhibited hypo- ASL and FDG-PET signals while its right counterpart presented normal imaging.

**Voxelwise Individual Statistical Analysis of ASL Data** Significant perfusion deviations from the ASL template were found for 3 out of the 6 patients, as illustrated in Fig. 2. Hypoperfusions could be observed in the left fusiform and lingual gyri for patient 1, in the right fusiform gyrus and in the bilateral basifrontal area for patient 2 and in the left anterior temporal area for patient 3. Hyperperfusions could also be observed in the lenticular nucleus, on the right side for patient 1 and on both sides for patient 2. As for the remaining 3 patients, no significant cluster was found.

**Voxelwise Group Statistical analysis** The voxelwise group statistical analysis detected hypoperfusions in the left temporo-parietal and temporo-occipital regions, and in the left and right temporal and insular regions. However, hypoperfusions globally predominate in the left anterior temporal region as shown in Fig. 3. No hyperperfusion was detected at the group level.

**ROIs Group Statistical analysis** The ROIs statistical analysis found significant differences between the patients and control group in the left temporal lobe ( $p = 0.0004$ ), in the left parietal lobe ( $p = 0.02$ ) and in the insula also on the left side ( $p = 0.002$ ).

### 4 Discussion

We have applied our framework for ASL data processing to SD, a rare neurodegenerative disease. Our preliminary results are encouraging as they globally



comply with the perfusion abnormalities expected in SD. First, the qualitative comparison of the PVE CBF maps and the FDG-PET data showed a good agreement between the two modalities with concordant regional hypoperfusion and hypometabolism. Second, the group analysis identified perfusion abnormalities in the frontal, insular and temporal regions, particularly pronounced on the left side as already reported in the literature. More surprising were the detected hypoperfusions in the left temporo-parietal and temporo-occipital regions. They might be related to the stage of the disease, and/or to an impairment of the fasciculi involved in SD [18] (arcuate and fronto-parietal longitudinal superior tracts). An analysis of the atrophy might also shed light on this result. Similarly, hyperperfusions of the lenticular globus detected at the individual level should be further investigated. Further analysis on a larger database might reveal more insight on these preliminary findings. Additional subjects are planned to be included in the next few months to increase the population of each group to at least 15.

This paper focused on the quantitative analysis of ASL data. Future work will include a quantitative analysis of PET data in which hypometabolism should be automatically detected, followed by a quantitative comparison between ASL and FDG-PET. Atrophy and cortical thickness will also be analysed from the 3D T1 structural data. Eventually, correlation of the imaging data with the neuropsychological scores will be studied. Ultimately, such a study might lead to demonstrate the ability of ASL to highlight cerebral abnormalities that give rise to cognitive deficit patterns in SD. This would allow improved clinical care and functional follow-up of SD patients as ASL is a non invasive and easily repeatable technique.

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