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Quantitative and clinical impact of MRI-based attenuation correction methods in [18F]FDG evaluation of dementia



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Abstract

Background: Positron emission tomography/magnetic resonance imaging (PET/MRI) is a promising diagnostic imaging tool for the diagnosis of dementia, as PET can add complementary information to the routine imaging examination with MRI. The purpose of this study was to evaluate the influence of MRI-based attenuation correction (MRAC) on diagnostic assessment of dementia with [¹⁸F]FDG PET. Quantitative differences in both [¹⁸F]FDG uptake and z-scores were calculated for three clinically available (DixonNoBone, DixonBone, UTE) and two research MRAC methods (UCL, DeepUTE) compared to CT-based AC (CTAC). Furthermore, diagnoses based on visual evaluations were made by three nuclear medicine physicians and one neuroradiologist (PET_{CT}, PET_{DeepUTE}, PET_{DixonBone}, PET_{UTE}, PET_{CT} + MRI, PET_{DixonBone} + MRI). In addition, pons and cerebellum were compared as reference regions for normalization.

Results: The mean absolute difference in z-scores were smallest between MRAC and CTAC with cerebellum as reference region: $0.15 \pm 0.11 \, \sigma$ (DeepUTE), $0.15 \pm 0.12 \, \sigma$ (UCL), $0.23 \pm 0.20 \, \sigma$ (DixonBone), $0.32 \pm 0.28 \, \sigma$ (DixonNoBone), and $0.54 \pm 0.40 \, \sigma$ (UTE). In the visual evaluation, the diagnoses agreed with PET_{CT} in 74% (PET_{DeepUTE}), 67% (PET_{DixonBone}), and 70% (PET_{UTE}) of the patients, while PET_{CT} + MRI agreed with PET_{DixonBone} + MRI in 89% of the patients.

Conclusion: The MRAC research methods performed close to that of CTAC in the quantitative evaluation of [18 F]FDG uptake and z-scores. Among the clinically implemented MRAC methods, Dixon_{Bone} should be preferred for diagnostic assessment of dementia with [18 F]FDG PET/MRI. However, as artifacts occur in Dixon_{Bone} attenuation maps, they must be visually inspected to assure proper quantification.

Keywords: PET/MRI, Attenuation correction, z-scores, dementia

Background

Magnetic resonance imaging (MRI) is today the preferred imaging modality in the clinical workup of suspected neurodegenerative disease due to the high spatial resolution and high soft tissue contrast. MRI can identify atrophy in dementia and exclude other diseases like vascular disease, cerebral amyloid angiopathy, brain tumors, and traumatic as well as inflammatory brain changes [1]. Positron emission

tomography (PET) with fluorodeoxyglucose ([¹⁸F]FDG) is however increasingly used to support the clinical diagnosis of patients with suspected dementia, as hypometabolism in certain brain regions can help identify specific types of dementia, including Alzheimer's disease (AD) and frontotemporal dementia (FTD) [2]. PET has a higher sensitivity for detecting early metabolic changes, which takes place prior to the morphological changes visible on MRI [1]. Hybrid PET/MRI systems have opened up the opportunity for simultaneous PET/MRI acquisitions, enabling fast and convenient examinations for patients with dementia. The information from PET and MRI is complementary, and detection of dementia with the combination of [¹⁸F]FDG

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PET and MRI is more accurate than with either of the imaging modalities alone [3].

As a complement to the visual assessment of hypometabolism in PET images performed by nuclear medicine physicians, PET data can be compared to databases of age-matched healthy controls. Z-score maps are then calculated, which represents the number of standard deviations (σ) separating the [¹⁸F]FDG uptake of the patient and the average of the healthy controls, where moderate hypometabolism is defined as a z-score between -2σ and -3σ , and severe hypometabolism for a z-score below -3σ [4]. A prerequisite for using such quantitative comparisons clinically is quantitatively accurate PET images, which are heavily dependent on attenuation correction (AC). AC is one of the most important corrections that needs to be performed on PET images, but is still challenging when using a PET/MRI system.

For PET/computed tomography (CT) systems, AC is based on CT images (CTAC), which is scaled by a bilinear function to represent the linear attenuation coefficients (LACs) of the 511 keV photons. For PET/MRI systems, alternative methods had to be developed in order to calculate attenuation maps from MRI data since there is no direct relation between the MRI signal and the electron density of tissue. Several proposed brain MRI-based AC (MRAC) methods have demonstrated a small and acceptable bias from CTAC (regional difference within $\pm 5\%$) [5]. Most of these promising methods are however not implemented in clinical systems, except for Dixon with bone model that recently became available on the Siemens PET/MRI system (software VE11P). A few studies have compared clinically implemented and research MRAC methods with CTAC in the evaluation of cognitive impairment. Cabello et al. [6] compared Dixon-based (without bone) AC and ultrashort echo time (UTE) AC with four novel MRAC methods. They concluded that Dixon- and UTE-based AC were inferior to the research MRAC methods, both when measuring [18F]FDG uptake and z-score accuracy to identify regions with reduced metabolism, compared to CTAC. These findings need to be re-evaluated after the recent software upgrade with modifications to the Dixon and UTE sequences.

The most relevant clinical issue is whether MRAC have an impact on clinical neurodegenerative diagnosis. Werner et al. [7] found that the pattern of hypometabolism remained largely unchanged with Dixon and that the clinical impact was negligible compared to CTAC. Franceschi et al. [8] found similar performance for Dixon and the prototype of Dixon with bone model in visually identifying hypometabolism without z-scores, and also concluded that even Dixon is acceptable for routine clinical evaluation of dementia. Still, the

quantitative errors should be further reduced and MRAC methods better imitating CTAC is warranted.

Another factor that potentially can impact the presence of hypometabolism is the choice of reference region. In the comparison to the database of healthy controls, the [¹⁸F]FDG uptake is normalized to a reference region, which should be unaffected by the disease. The most commonly used reference regions in dementia evaluations are cerebellum and pons, and incorrect AC in these regions can induce a bias in the [¹⁸F]FDG uptake affecting z-scores throughout the brain. The accuracy of the MRAC methods in the reference region is thus important and should be investigated further.

The aim of this study was to assess the quantitative and clinical impact of the implemented MRAC methods in [18F]FDG PET evaluation of dementia (Dixon, Dixon with bone model, UTE) on the Siemens Biograph PET/MRI scanner. Two research MRAC methods (DeepUTE and UCL) presented in the literature were also included for comparison, in addition to CTAC as reference. Secondary aims were to investigate how the choice of reference region influenced the z-scores quantitatively.

Materials and Methods

Patients

Twenty-seven consecutive patients with suspected dementia were referred to brain PET/CT and PET/MRI examinations. Nine patients were excluded from this study due to incorrect anatomical position during the PET/CT examination (n = 5), misregistration of bone in $Dixon_{Bone}$ MRAC (n = 2) (the artifacts could not be removed manually and a new Dixon acquisition was not acquired), aliasing in MRI scans (n = 1), and problems with co-registration of PET images to the MNI PET template (n = 1). The 18 patients included had a mean age of 69 ± 9 years and a mean weight of 75 ± 16 kg. Patient characteristics and the proposed diagnosis made by a nuclear medicine physician and a neuroradiologist based on PET/CT and MR imaging and clinical referral text is given in Table 1. The study was approved by the Regional Committee for ethics in Medical Research (REC Central) (ref. number: 2013/1371) and all patients gave written informed consent.

Image acquisition and reconstruction

Image acquisition was performed on a Biograph mCT PET/CT system (software version VG51C), and subsequently on a Biograph mMR PET/MRI system (software version VE11P) (Siemens Healthcare GmbH, Erlangen, Germany). All patients fasted at least 6 h prior to intravenous injection of [18 F]FDG (210 \pm 46 MBq). The patients were kept blindfolded in a quiet room during the uptake phase prior to the PET/CT examination, which was performed 35 \pm 1 min post injection (p.i.),

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Table 1 Patient characteristics

Patient	Age (years)	Gender	Proposed diagnosis ^a
1	72	М	Non-specific
2	70	F	FTD
3	49	F	Normal
4	78	F	Normal
5	74	М	FTD
6	64	М	Normal
7	83	F	AD/FTD ^b
8	54	F	Non-specific
9	61	М	Normal
10	82	М	Normal
11	71	F	Non-specific
12	75	F	Normal
13	68	М	Non-specific
14	66	F	Normal
15	72	F	Non-specific
16	63	М	AD
17	69	М	Non-specific ^c
18	70	F	Normal

AD Alzheimer's disease, FTD frontotemporal dementia, Non-specific other subtypes of dementia, and other patterns of hypometabolism that cannot be explained by image artifacts

followed by the PET/MRI examination, performed 64 \pm 9 min p.i.

Only the low-dose CT scan and the corresponding attenuation map were used (as reference) from the PET/ CT examination. The PET (20 min) and MRI (17 min) acquisitions were performed simultaneously, and the MRI protocol consisted of the same sequences as in the clinical MRI protocol for patients with suspicion of dementia (sagittal 3D T1 MPRAGE, coronal T2, transversal FLAIR, GRE T2* (microhemorrhage), and diffusion weighted imaging) in addition to the sequences for MRAC; a high-resolution two-point Dixon VIBE and a UTE sequence. All PET reconstructions were performed on the mMR system using 3D OSEM reconstruction (three iterations and 21 subsets, 344 × 344 image matrix, 4 mm Gaussian filter) and corrections for scatter, randoms, detector normalization, decay, and attenuation.

Attenuation maps

PET data acquired at the PET/MRI system was reconstructed with the following five MR attenuation maps and a CT attenuation map (presented in Fig. 4) for each patient:

- Dixon_{NoBone}: Implemented at the mMR system. Segmentation-based method that relies on the two-point Dixon VIBE sequence (Brain HiRes), where air, fat, and soft tissue are segmented and assigned predefined discrete LACs (air: 0 cm⁻¹, fat: 0.0854 cm⁻¹, fat/soft tissue mix: 0.0927 cm⁻¹, and soft tissue: 0.1000 cm⁻¹).
- 2. Dixon_{Bone}: Implemented at the mMR system (product in the latest software, VE11P). Similar to Dixon_{NoBone}, but includes continuous bone information from an integrated bone atlas by registration of MR images of the subject to MR images of the atlas [9, 10]. The atlas contains sets of pre-aligned MR image and bone mask pairs with bone densities as LACs in cm⁻¹ at the PET energy level of 511 keV.
- 3. UTE: Implemented at the mMR system. Segmentation-based method that relies on the two images from the UTE sequence with different echo times (TE₁ and TE₂), and segments the image into air (0 cm⁻¹), soft tissue (0.1000 cm⁻¹), and bone (0.1510 cm⁻¹).
- 4. UCL: Atlas-based method using a database of 41 paired T1-weighted MRI and CT data sets [11–13]. All MRI data sets of the atlas are non-rigidly registered to the patient's MRI data and normalized correlation coefficients are calculated at each voxel. A pseudo CT is then calculated from averaged weights of the CT data sets based on the correlation coefficients. In this study, T1-weighted MPRAGE was used as input in a web-based tool, after bias correction with FMRIB Software Library (FSL, Oxford Centre for Functional MRI of the Brain, UK), as recommended by the distributor. The returned UCL attenuation map in Houndfield units (HU) was converted to LACs [14] and smoothed with a 4 mm Gaussian filter.
- 5. DeepUTE: Artificial intelligence approach to MRAC, using a deep learning algorithm [15]. Briefly, the method uses a modified 3D U-net architecture [16] for image-to-image learning of paired UTE and CT data. Compared to [15], the network was here trained using data from 832 adult examinations.
- 6. CT: Attenuation map generated by converting a low-dose CT scan on the mCT scanner to LACs [14]. The bed and head holder was excluded from the CT attenuation map by making a semi-automatic head mask (CT head mask) with the software MRIcron [17], and the attenuation map was multiplied by 10,000 to get the same order of magnitude as the MRAC maps at the mMR system. The CT attenuation maps did not cover the neck region sufficiently for attenuation correction of the PET data from the PET/MRI system due to differences in the

^aDiagnosis based on PET/CT + MRI and clinical referral text

^bAmbiguous clinical information as well as imaging data, but clearly neurodegenerative

^cSuspicion of normal pressure hydrocephalus (later confirmed clinically and operated with ventricular shunt)

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axial field of view of the PET-detectors. The area outside the CT head mask was therefore substituted with the ${\rm Dixon_{Bone}}$ attenuation map for each patient. The CT image was rigidly registered to the Dixon in-phase image and the same transformation was performed on the CT attenuation map.

The same voxels that were substituted by Dixon_{Bone} in the CT attenuation maps were also substituted by Dixon_{Bone} in all evaluated MR attenuation maps. In order to perform this voxel substitution, the UTE TE₂ image and the T1w MPRAGE image was registered to the Dixon in-phase image, and the resulting transformations were used on the respective attenuation maps. All registrations were performed with Aliza Medical Imaging 1.35.3 (Bonn, Germany) (using elastix version 4.8 [18, 19]) [20]. To enable import of the modified attenuation maps at the PET/MRI system, all attenuation maps used the header file of Dixon_{Bone} with exchange of the pixel data.

Quantitative analysis

Bone artifacts

After software upgrade (from VB20P to VE11P) of the PET/MRI system, bone artifacts have been observed in the ${\rm Dixon_{Bone}}$ and UTE attenuation maps. Two of the most severe artifacts seen in the attenuation maps are misplacement of bone segments from other parts of the body found in the ${\rm Dixon_{Bone}}$ and bone present inside the brain nearby the anterior ventricles in the UTE attenuation maps. The ${\rm Dixon_{Bone}}$ and UTE attenuation maps were therefore visually inspected for these artifacts.

[18F]FDG uptake

The [18F]FDG uptake in all PET reconstructions were measured in 15 brain regions that were chosen to match the brain regions in the software used for zscore analysis and visual assessment. The regions were in MNI space and taken from the Harvard-Oxford Cortical Structural Atlas, MNI Structural Atlas, and Talairach Daemon Labels in FSL (Oxford Centre for Functional MRI of the Brain, UK). The PET images of the patients were converted to MNI space by co-registration to a dementia-specific [18F]FDG-PET template [21, 22]. The PET_{DixonBone} was first registered with elastix to the PET template in a two-step process (rigid and non-rigid registration), and the resulting transform was used on the other five PET images of the same patient for transformation to MNI space. Relative difference (RD) was calculated in each brain region, and was defined as

$$RD(\%) = \frac{\overline{PET}_{MRAC} - \overline{PET}_{CTAC}}{\overline{PET}_{CTAC}} \times 100, \tag{2}$$

where \overline{PET}_{MRAC} and \overline{PET}_{CTAC} is the average activity measured in a brain region in PET_{MRAC} and PET_{CTAC} , respectively. The results are presented by using the boxplot function in MATLAB (R2017b). Absolute RDs were also calculated and averaged over patients and brain regions as \overline{RD}_{abs} .

Z-scores

The visual evaluations were performed with the software Cortex ID (GE Healthcare, Waukesha WI, USA), where z-scores were calculated in 26 brain regions. The database constitutes of 294 healthy controls divided in six age groups, imaged with [18 F]FDG PET and using a transmission scan of 68 Ge for attenuation correction. Both cerebellum and pons were used as reference regions in the quantitative analysis. Quantitative comparison of z-scores between PET_{MRAC} (PET_{DixonBone}, PET_{UTE}, PET_{UCL}, PET_{DeepUTE}) and PET_{CTAC} were performed by calculating the difference, D, and absolute difference, $D_{\rm abs}$, in each brain region, where

$$D = Z_{MRAC} - Z_{CTAC}$$
, and

$$D_{abs} = \mid Z_{MRAC} - Z_{CTAC} \mid$$
,

and D_{abs} was averaged over patients and brain regions as $\overline{D_{abs}}$. The boxplot function in MATLAB was used to present the differences in z-scores.

Visual evaluation

To limit the number of images in the visual evaluation, MRAC methods were chosen based on the z-score analysis. The best and worst of the clinically implemented MRAC methods were included, in addition to the best research MRAC method. The PET_{CT} was used as reference. Three nuclear medicine physicians (brain PET experience; reader 1: 3 years, reader 2: 10 years, reader 3: 1 year) performed the visual assessments individually. Based on PET images and z-scores, the patients were either categorized as normal or diagnosed with AD, FTD, or non-specific pathology (other subtypes of dementia, like DLB, and other patterns of hypometabolism that cannot be explained by image artifacts). The physicians were blinded for AC method and patient ID, and had no information regarding patient history or MRI.

A second reading was made based on both PET images, z-scores, and MR images by a nuclear medicine physician (reader 3) and a neuroradiologist (4-year experience in neuroradiology and European Diploma in NeuroRadiology (EDiNR)) in conjunction. The best clinically implemented MRAC method based on the z-score

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analysis was chosen for this second visual evaluation, and PET_{CT} was used as reference. The first and second visual evaluation was done 2 months apart.

PET images and z-scores were evaluated in Cortex ID, while MR images were assessed with the hospitals Picture Archiving and Communication System (PACS; Sectra IDS 7). Cerebellum was chosen as reference region in all visual evaluations.

Statistical analysis

κ-statistics were calculated (with Stata/MP 15.1, Stata-Corp LLC, USA) to determine the agreement between PET_{CT} and each PET_{MRAC} in the visual evaluations after correction for the agreement expected by chance. The inter-reader agreement was also calculated for each AC method in the evaluations with three readers. A κ value of 0 indicates no agreement better than chance, and the values were interpreted according to Landis et al. [23] (poor: < 0, slight: 0.00–0.20, fair: 0.21–0.40, moderate: 0.41–0.60, substantial: 0.61–0.80, almost perfect: 0.81–1.00).

Results

Quantitative analysis

Bone artifacts

Bone artifacts were observed in 22 % (4/18) of the Dixon_{Bone} attenuation maps, while no artifacts were seen in the corresponding Dixon images. New Dixon sequences were acquired for two patients with large bone segments from other parts of the body infiltrating the head (Fig. 1a), resulting in artifact-free attenuation maps (Fig. 1b). Artifacts positioned outside the head (Fig. 1c) were manually removed (Fig. 1d) for the last two patients. Hence, only artifact free Dixon_{Bone} attenuation maps were included in the study. Furthermore, in 89% (16/18) of the UTE attenuation maps, minor bone artifacts were observed inside the brain close to the anterior ventricles (Fig. 1e). The UTE artifacts were not removed.

[18F]FDG uptake

The mean absolute relative difference ($\overline{RD_{abs}}$) in [^{18}F]FDG uptake compared to PET_{CT} was the smallest for PET_{DeepUTE} and the largest when omitting bone information in PET_{DixonNoBone} (Table 2). PET_{DixonBone} performed similar to the research MRAC methods, but had slightly larger range of RD. The relative differences in [^{18}F]FDG uptake for the different brain regions are presented in Fig. 2. Patient 3, with abnormal anatomy (an arachnoid cyst in the posterior fossa), caused most of the outliers seen in Fig. 2. The attenuation maps with corresponding PET images for all reconstructions are demonstrated for this patient in Additional file 1: Figure S1.

Z-scores

The mean absolute difference $(\overline{D_{abs}})$ in z-score between CTAC and MRAC was minimized with the research methods (PET_{DeepUTE} and PET_{UCL}), which also had the smallest range. Among the clinically implemented methods, PET_{DixonBone} performed best, closely followed by PET_{DixonNoBone}. The largest $\overline{D_{abs}}$ was found with PET_{UTE} (Table 3). For all MRAC methods, smaller differences were found for cerebellum than for pons as reference region.

Figure 3 shows that the difference in z-scores between CTAC and the MRAC methods were more stable across brain regions for the research methods than for the clinical methods. PET_{DeepUTE} slightly overestimated and PET_{UCL} slightly underestimated the z-scores compared to PET_{CT} for most brain regions for both reference regions (Fig. 3). The clinical MRAC methods (PET_{DixonBone}, PET_{DixonNoBone}, and PET_{UTE}) yielded lower z-scores than PET_{CT} with pons as reference region, and both over- and underestimated z-scores with cerebellum as reference region (Fig. 3). Examples of z-score maps for one patient with dementia are presented in Fig. 4, where increased hypometabolism is especially pronounced for PET_{UTE} with pons as reference region.



Fig. 1 Typical bone artifacts found in attenuation maps from PET/MRI. **a** Dixon_{Bone} attenuation map with large infiltrative bone segment. Attenuation maps like this were only included in the study if the patient had a second acquisition yielding **b** an artifact-free attenuation map. **c** Dixon_{Bone} attenuation map with artifact in the upper right corner, which could be **d** manually removed. **e** UTE attenuation map with smaller bone segments inside the brain nearby the anterior ventricles. These were not removed

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Table 2 The mean absolute relative difference $(\overline{RD_{abs}})$ and the range of RD in 18 F[FDG] uptake for the PET_{MRAC} methods compared to PET_{CTAC}.

MRAC method	RD _{abs} (%) mean ± std	RD (%) [min max]
PET _{DeepUTE}	2.2 ± 1.5	[- 10.6, 1.7]
PET _{UCL}	3.0 ± 1.4	[- 3.3, 7.3]
PET _{DixonBone}	2.5 ± 2.4	[- 13.0, 10.7]
PET _{DixonNoBone}	7.1 ± 3.7	[- 19.9, 7.4]
PET _{UTE}	4.1 ± 3.3	[- 12.7, 16.3]

Visual evaluation

PET_{DixonBone} and PET_{UTE} were chosen for the first visual evaluations (PET only) as these had the best and the worst results of the clinically implemented MRAC methods in the z-score analysis. The research MRAC methods performed relatively equal in the z-score analysis and PET_{DeepUTE} was chosen for the visual evaluation. Furthermore, PET_{CT} was included as reference. The results of the visual evaluations with PET only are presented in Table 4. The agreement in diagnosis between PET_{CT} and PET_{DeepUTE}, PET_{DixonBone} and PET_{UTE} was in average for the three readers 74%, 67%, and 70%, respectively (Table 5), and the κ-statistics indicated mostly *moderate* agreement between PET_{CT} and PET_{MRAC}. The inter-reader agreement was *fair* for PET_{CT} (κ = 0.30) and *slight* for PET_{DeepUTE} (κ = 0.17), PET_{DixonBone} (κ = 0.19), and PET_{UTE} (κ = 0.10).

In the second visual evaluation, which also included MRI, $PET_{DixonBone}$ was compared to PET_{CT} . When

MRI was included in the assessment, the agreement increased to 89% and the κ -statistics indicated *almost perfect* agreement ($\kappa = 0.82$) (Table 6) according to Landis et al. [23].

Discussion

The impact of MRAC on dementia assessment was evaluated in this study by comparing [$^{18}\mathrm{F}]\mathrm{FDG}$ uptake, z-scores, and clinical interpretation between PET_{MRAC} and PET_{CT}. The absolute mean quantitative differences in z-scores were small relative to the definition of hypometabolism for most MRAC methods with cerebellum as reference region, and especially for the research methods. Interpretation with PET alone yielded high uncertainties, while assessment with both PET and MRI resulted in almost perfect agreement between PET_{CT} and PET_DixonBone.

The bone artifacts found in the clinically available MRAC methods highlights the need for careful inspection of the attenuation maps in all brain examinations. In the Dixon_{Bone} attenuation maps, the artifacts were caused by misregistration between the Dixon images and the bone-template, misplacing large bone segments from other parts of the body in the brain. Due to the severity of these artifacts, they were removed by either acquiring a new Dixon acquisition free of this artifact, or manually when found outside the brain. Although not evaluated quantitatively, this artifact would likely induce large errors in the attenuation corrected PET images. The minor bone artifacts observed in most UTE attenuation maps

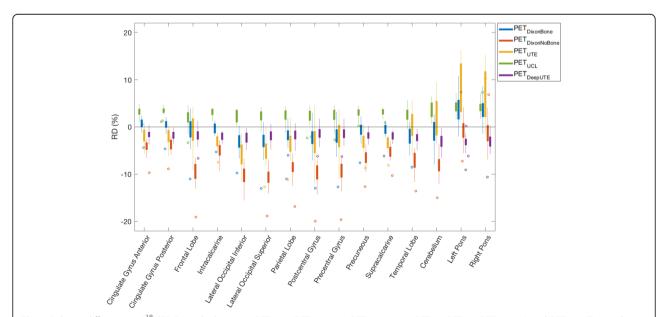


Fig. 2 Relative difference in [¹⁸F]FDG uptake between PET_{MRAC} (PET_{DixonBone}, PET_{DixonNoBone}, PET_{UTE}, PET_{UCL}, PET_{DeepUTE}) and PET_{CTAC}. Top and bottom edges of the boxes indicate 25th and 75th percentiles, while the whiskers extend to the most extreme data points, except for outliers that are marked as circles

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Table 3 The mean absolute difference $(\overline{D_{abs}})$ in z-score between PET_{MRAC} and PET_{CT} and the range of the difference (D), with pons and cerebellum as reference regions

MRAC method	D _{abs} (o) (pons)	D (o) (pons)	D _{abs} (σ) (cerebellum)	D (o) (cerebellum)
	mean ± std	[min max]	mean ± std	[min max]
PET _{DeepUTE}	0.19 ± 0.16	[- 0.43, 0.76]	0.15 ± 0.11	[-0.60, 0.60]
PET _{UCL}	0.21 ± 0.15	[- 0.69, 0.47]	0.15 ± 0.12	[-0.64, 0.44]
PET _{DixonBone}	0.48 ± 0.27	[- 1.52, 0.31]	0.23 ± 0.20	[- 1.14, 1.42]
PET _{DixonNoBone}	0.53 ± 0.35	[- 1.77, 0.74]	0.32 ± 0.28	[- 1.32, 1.67]
PET _{UTE}	1.13 ± 0.47	[-2.85, -0.02]	0.54 ± 0.40	[- 1.64, 2.49]

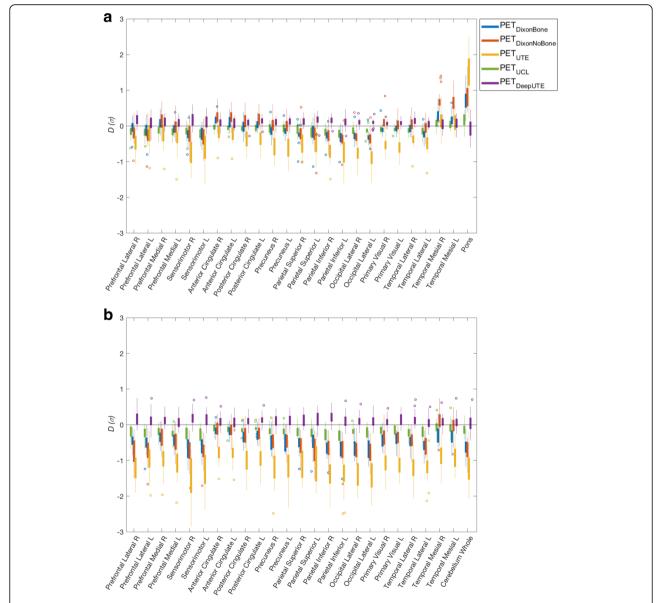


Fig. 3 Difference in z-score between PET_{MRAC} (PET_{DixonBone}, PET_{DixonNoBone}, PET_{UTE}, PET_{UCL}, PET_{DeepUTE}) and PET_{CTAC}, for different brain regions, with **a** cerebellum and **b** pons as reference region. Top and bottom edges of the boxes indicate 25th and 75th percentiles, while the whiskers extend to the most extreme data points, except for outliers that are marked as circles

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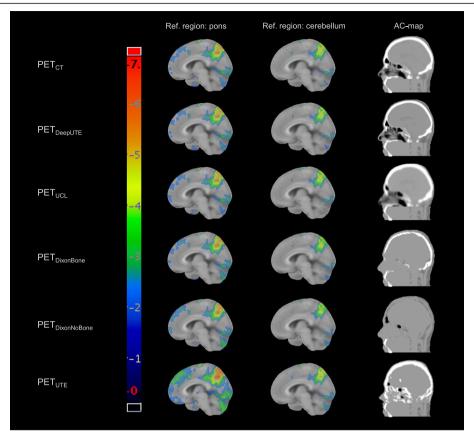


Fig. 4 Examples of z-score maps for one patient (number 16) for the included AC methods, with pons and cerebellum as reference regions, and the corresponding attenuation maps

were caused by changes in the UTE sequence and/or attenuation map algorithm after the software upgrade, and persisted even after acquiring new UTE images. Since the UTE attenuation maps are used clinically, they were not excluded from the current study. In clinical routine, a reliable and stable MRAC method is crucial and these problems need to be solved. The attenuation map errors have been reported to Siemens Healthcare, and will hopefully be solved in the near future. In the meantime, some of the artifacts can be avoided by implementing better procedures among radiographers to detect the artifacts and acquire new MR-based attenuation maps in such cases before the patient leaves the scanner table.

The research MRAC methods, as well as $PET_{DixonBone}$, all demonstrated small absolute differences compared to PET_{CT} regarding [^{18}F]FDG uptake, although the research methods had smaller RD range. Some outliers were observed in the analysis, and most of them were caused by a patient with an abnormal anatomy (arachnoid cyst in posterior fossa). DeepUTE gave least outliers for this patient with abnormal anatomy (2 15 brain regions), while Dixon_{Bone} and Dixon_{NoBone} yielded most outliers for this patient (10 15 brain regions). For

absolute differences in [¹⁸F]FDG uptake, the trend was the same as in previous studies [5–7, 24], with descending performance for PET_{UCL}, PET_{DixonBone}, PET_{UTE}, and PET_{DixonNobone} (DeepUTE has not been included in previous studies). PET_{UTE} yielded particularly large variations in the pons, probably due to misclassification of bone in that region, which makes pons not suited as a reference region with UTE AC. Furthermore, we found that the LACs for soft tissue were slightly higher with UCL AC and slightly lower with DeepUTE AC compared to the CTAC, which probably caused the general over- and underestimation of [¹⁸F]FDG uptake for the two methods, respectively.

For the z-score evaluation, the research MRAC methods yielded the best performance and the differences in z-scores between PET_{MRAC} and PET_{CT} were generally small compared to the definition of hypometabolism, except for PET_{UTE} , when using cerebellum as reference region. Of note, $PET_{DixonBone}$ and $PET_{DixonNo-Bone}$ yielded similar results for the z-scores, indicating that the missing bone information did not have a remarkable impact on z-scores. Despite small average differences in z-scores to PET_{CT} for most MRAC

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Table 4 Assigned diagnostic categorization for each patient made by three nuclear medicine physicians from PET images with three MRAC methods (PET_{DeepUTE}, PET_{DixonBone}, PET_{UTE}) and PET_{CT}. Cerebellum was used as reference region. Intra-reader discrepancies with PET_{CT} as reference are italicized

Patient	Reader	PET_CT	PET _{DeepUTE}	PET _{DixonBone}	PET _{UTE}
1	1	N	N	N	Ν
	2	Ν	Ν	N	NS
	3	Ν	Ν	N	NS
2	1	AD	N	AD	AD
	2	NS	FTD	FTD	NS
	3	FTD	NS	NS	NS
3	1	N	N	N	Ν
	2	N	N	NS	NS
	3	N	Ν	N	NS
4	1	N	Ν	AD	AD
	2	FTD	FTD	FTD	FTD
	3	N	Ν	N	Ν
5	1	FTD	FTD	FTD	FTD
	2	AD	NS	NS	AD
	3	NS	AD	AD	AD
6	1	N	Ν	N	AD
	2	N	Ν	N	Ν
	3	Ν	Ν	N	Ν
7	1	AD	NS	NS	NS
	2	AD	AD	NS	AD
	3	AD	AD	AD	AD
8	1	AD	N	Ν	AD
	2	FTD	FTD	FTD	FTD
	3	Ν	NS	NS	NS
9	1	Ν	N	N	NS
	2	NS	NS	Ν	NS
	3	NS	Ν	Ν	Ν
10	1	AD	Ν	Ν	AD
	2	Ν	Ν	N	Ν
	3	Ν	N	N	Ν
11	1	Ν	N	N	AD
	2	NS	FTD	NS	NS
	3	NS	NS	NS	NS
12	1	N	N	N	Ν
	2	N	N	N	Ν
	3	N	N	N	Ν
13	1	Ν	Ν	FTD	Ν
	2	NS	NS	NS	NS
	3	NS	NS	NS	NS
14	1	N	N	N	Ν
	2	N	FTD	FTD	FTD

Table 4 Assigned diagnostic categorization for each patient made by three nuclear medicine physicians from PET images with three MRAC methods (PET_{DeepUTE}, PET_{DixonBone}, PET_{UTE}) and PET_{CT}. Cerebellum was used as reference region. Intra-reader discrepancies with PET_{CT} as reference are italicized (*Continued*)

		C1			/
Patient	Reader	PET _{CT}	PET _{DeepUTE}	PET _{DixonBone}	PET _{UTE}
	3	Ν	N	NS	N
15	1	Ν	Ν	Ν	AD
	2	FTD	FTD	FTD	FTD
	3	Ν	NS	NS	NS
16	1	AD	AD	AD	AD
	2	AD	AD	AD	AD
	3	AD	AD	AD	AD
17	1	AD	AD	NS	AD
	2	AD	NS	AD	AD
	3	NS	NS	NS	NS
18	1	Ν	Ν	Ν	Ν
	2	NS	NS	NS	NS
	3	Ν	N	Ν	Ν

N normal, NS non-specific, AD Alzheimer's disease, FTD frontotemporal dementia

methods with cerebellum as reference region, large outliers were present for the clinical MRAC methods with a deviation from $PET_{CT} > 1~\sigma,$ which can have a considerable impact on a z-score assessment. Hence, it is highly desirable to implement the research methods at the clinical PET/MRI systems as soon as possible to avoid large biases in the z-score assessment.

Since the calculation of z-scores use a reference region for normalization, the accuracy of AC in the reference region is of particular importance as bias in this region will affect the hypometabolism globally. Available reference regions in the software used for visual evaluation in the current study were pons, cerebellum, and global cerebral cortex. The extent of hypometabolism can however be underestimated with global normalization [25]. Therefore, only cerebellum and pons were used in this study, and differences in z-scores between MRAC and CTAC were found to be smaller with cerebellum than pons as reference region for all MRAC methods. Although glucose metabolism in the pons have been found to be least affected by dementia among several reference regions [26], the small size makes this region prone to bias and the surrounding inhomogeneous bone affects both attenuation and scatter [25]. Cerebellum is larger and less prone to bias, and the cerebellar glucose metabolism is not significantly reduced for AD patients, except for severe AD [25, 27].

Based on our results, $Dixon_{Bone}$ with cerebellum as reference should be preferred among the clinically implemented MRAC methods when assessing z-scores.

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Table 5 κ -statistics for the agreement between PET_{CT} and PET_{MRAC} (PET_{DeepUTE}, PET_{DixonBone}, PET_{UTE}) for each reader. A κ value of 0 indicates no agreement better than chance, while 1.0 means perfect agreement

	PET _{CT} vs PET _{DeepUTE} Agreement (κ)	PET _{CT} vs PET _{DixonBone} Agreement (κ)	PET _{CT} vs PET _{UTE} Agreement (к)
Reader 1	77.8% (0.54)	66.7% (0.41)	66.7% (0.47)
Reader 2	72.2% (0.63)	66.7% (0.55)	83.3% (0.78)
Reader 3	72.2% (0.55)	66.7% (0.47)	61.1% (0.39)
Mean of readers	74.1%	66.7%	70.4%

However, for patients with abnormal anatomy and/or unusual tissue density, atlas-based methods should be used with caution [28]. In these cases, or when bone artifacts are present, $Dixon_{NoBone}$ could probably be used as an alternative for z-score assessment in the evaluation of dementia.

The visual evaluations with PET only yielded moderate agreement between PET_{CT} and PET_{MRAC} in general. Highest agreement was found for $PET_{DeepUTE}$, but the other MRAC methods performed quite similarly. Least false positive errors were found for $PET_{DeepUTE}$ compared to PET_{CT} , while false negative errors were highest for $PET_{DeepUTE}$ and $PET_{DixonBone}$. These results seem to

Table 6 Assigned diagnostic categorization made by one nuclear medicine physician (reader 3) and one neuroradiologist in conjunction for PET (PET $_{\text{CT}}$, PET $_{\text{DixonBone}}$) and MRI. Discrepancies from PET $_{\text{CT}}$ + MRI are italicized

Patient	$PET_{CT} + MRI$	PET _{DixonBone} + MRI
1	NS	NS
2	NS	NS
3	N	N
4	N	N
5	FTD	FTD
6	N	N
7	FTD	AD^{a}
8	N	N
9	N	N
10	N	N
11	Ns	NS
12	N	N
13	NS	NS
14	N	N
15	NS	NS
16	AD	NS
17	NS b	NS ^b
18	N	N

AD Alzheimer's disease, FTD frontotemporal dementia

be in agreement with the quantitative results, as PET_{UTE} underestimated z-scores, inducing false positive errors, while PET_{DeepUTE} slightly overestimated z-scores and tends to more often change pathology to normal than the opposite. However, the inter-reader agreement was low, which indicates that the visual assessment of [¹⁸F]FDG PET in dementia is difficult and subjective, and that these evaluations were influenced by additional factors than the different AC methods. Another study by Werner et al. [7], evaluating the clinical impact of different AC methods, demonstrated a higher agreement between the readers; however, the categorization of diagnosis was not the same as in our study, which could have caused less discrepancy in their results.

Due to the large discrepancies in the PET only evaluations, another assessment including MRI was performed. Adding MRI information yielded almost perfect agreement between MRAC and CTAC readings according to the κ -statistics, and in the two cases of discrepancies between PET_{CT} + MRI and $PET_{DixonBone}$ + MRI, the discrepancies were due to different subtypes of dementia. The improvement by including MRI was probably due to the ability to discard areas of hypometabolism due to other pathologies and normal variants (e.g., age-related atrophy, enlarged ventricles, and mega cisterna magna). Furthermore, information of neurodegenerative processes such as hippocampal atrophy (as seen in AD), focal cortical atrophy (as seen in FTD), and white matter hyperintensities (as seen in microvascular disease) was important complementary information to the PET findings. In a clinical setting with all clinical information and imaging available, the discrepancies between MRAC and CTAC would probably be further decreased, but this should be verified in studies with larger patient cohorts.

A limitation of this study is the small number of patients, and hence few patients having dementia and low diversity in diagnoses and severity. Furthermore, PET images suffer from partial volume effects due to the limited resolution that cause spill-out from one region to another. This was not corrected for and could cause a significant effect on hypometabolism from normal aging [29]. However, the aim of this study was to compare MRAC and CTAC, and not the exact diagnosis. Another factor that may affect the z-scores is that the PET

^aDefined as both FTD and AD in the proposed diagnosis based on PET/CT and MR imaging and clinical referral text (Table 1)

^bSuspicion of normal pressure hydrocephalus

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images in the database of Cortex ID were acquired and reconstructed differently than the PET images in this study. Still, the relative differences between CTAC and MRAC should be unaffected.

Conclusion

The quantitative differences in z-scores between CTAC and most MRAC methods were small relative to the definition of hypometabolism, with cerebellum as reference region. Although the research MRAC methods performed slightly better than the clinically implemented MRAC methods regarding calculations of the z-scores, the visual evaluations with PET and MRI demonstrated almost perfect agreement between Dixon_{Bone} and CTAC. Our results indicate that Dixon_{Bone} with cerebellum as reference region should be preferred among the clinically implemented MRAC methods when using Siemens PET/MRI system for dementia assessment with [¹⁸F]FDG PET/MRI. Although, inspection of the attenuation maps is a prerequisite for the use of PET/MRI in dementia evaluation.

Additional file

Additional file 1: Figure S1. Attenuation maps (top row) with corresponding PET images (bottom row) for patient number 3 with abnormal anatomy. (a) CT, (b) UCL, (c) DeepUTE, (d) Dixon_{Bone}, (e) Dixon_{NoBone} and (f) UTE. (PNG 1967 kb)

Abbreviations

AC: Attenuation correction; AD: Alzheimer's disease; CT: Computed tomography; CTAC: CT-based AC; [¹⁸F]FDG: Fluorodeoxyglucose; FTD: Frontotemporal dementia; HU: Houndfield units; LACs: Linear attenuation coefficients; MRI: Magnetic resonance imaging; MRAC: MRI-based AC; PET: Positron emission tomography; RD: Relative difference; UTE: Ultrashort echo time

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Authors' contributions

SKØ participated in the study design, generated attenuation maps, performed the image reconstructions, and data analysis. TK was responsible for patient recruitment and interpreted the PET images. EMB interpreted the MR images, while JFA and TS interpreted PET images. CNL generated the DeepUTE attenuation maps. AK participated in the study design, was responsible for the acquisition protocol, gave valuable technical support, and revised the manuscript critically. LE participated in the study design, applied for ethical approval and gathered informed consent, was responsible for the acquisition protocol, gave valuable support throughout the project, and revised the manuscript critically. All authors have read and approved the attached paper.

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Availability of data and materials

The datasets used and analyzed during the current study are available from the corresponding author on reasonable request.

Ethics approval and consent to participate

The study was approved by the Regional Committees for Medical and Health Research Ethics (REC) (ref. number: 2013/1371) and all patients gave written informed consent.

Consent for publication

Not applicable.

Competing interests

The authors declare that they have no competing interests.

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