Improved Visualization of Middle Ear Cholesteatoma with Computed Diffusion-weighted Imaging

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Computed DWI (cDWI) is a mathematical technique that calculates arbitrary higher $b$ value images from at least two different lower $b$ values. In addition, the removal of high intensity noise with image processing on cDWI could improve cholesteatoma-background contrast-to-noise ratio (CNR). In the present study, noise reduction was performed by the cut-off values of apparent diffusion coefficient (ADC) less than 0 and $0.4 \times 10^{-3}$ s/mm$^2$. The cholesteatoma to non-cholesteatoma CNR was increased using a noise reduction algorithm for clinical setting.

Keywords: cholesteatoma, diffusion weighted imaging, magnetic resonance imaging, middle ear, signal-to-noise ratio

Introduction

Diffusion-weighted MR imaging (DWI) is the most widely used imaging technique for the detection of middle ear cholesteatoma.1-8 Turbo spin echo (TSE) DWI could reduce geometric distortions compared to echo-planar imaging (EPI) based acquisitions, and shows promise for the application in the middle ear region.3,4 On the other hand, it is known that the TSE–DWI sequence as well as higher $b$ value (>1000 s/mm$^2$) suffer from inherently lower signal-to-noise ratio (SNR).9,10

Computed DWI (cDWI) is a mathematical technique that calculates arbitrary higher $b$ value images from at least two different lower $b$ values. This may avoid eddy current distortions while keeping higher SNR.11,12 Recently, the utility of cDWI for prostate cancer or hepatic metastases detection using has been reported in the literatures.11,13-15

In contrast, the application of cDWI for middle ear diseases has not been explored yet. One potential drawback of cDWI is that voxels less than 0 s/mm² in ADC are practically present using voxel-by-voxel based calculation, which could reduce diagnostic quality. Removal of high intensity noise with image processing on cDWI could improve cholesteatoma to non-cholesteatoma contrast-to-noise ratio (CNR). Our purpose was to evaluate whether cDWI with a noise reduction algorithm increases CNR compare to that without noise reduction in middle ear cholesteatoma.

Materials and Methods

Case selection

This retrospective study was approved by Kyushu University Institutional Review Board for Clinical Research, and the requirement for written informed consent was waived. Consecutive patients diagnosed with suspected cholesteatoma who underwent preoperative MR imaging between October 2014 and August 2016 were eligible for inclusion. Two subjects were excluded from the study due to motion artifacts on DWI. Each patient underwent preoperative MRI on the day before surgery. All of the patients were confirmed the diagnosis of cholesteatoma at surgery.

Image acquisition

All images were obtained using a 3T MR imaging unit (Ingenia CX, Philips Medical Systems, Best, The Netherlands) with a 15-channel head array receiving coil for sensitivity encoding (SENSE) parallel imaging. The single-shot TSE–DWI scanning parameters were as follows: TR/TE = 4200/84 ms; flip
angle = 90°; echo train length = 46; refocusing control angle = 100°; b value = 0, 400 s/mm²; motion probing gradient = 3 orthogonal directions; effective diffusion time = 13.1 ms (diffusion gradient separation = 15.5 ms; diffusion gradient pulse duration = 7.1 ms); SENSE factor = 2; slice thickness/gap = 1.5/0 mm; slices = 18; acquisition matrix = 152 × 154; FOV = 230 mm; number of signal averages = 5. The total acquisition time was 2 m 48 s. DWIs with 1.5 mm isotropic voxels were generated for evaluation.

**Image evaluation**
Maps of ADC were calculated using the following formula:
\[ \ln (S/S_0) = -b \times ADC \]
where \( S_0 \) and \( S \) are the signal intensities for \( b \) values are 0 and 400 s/mm², respectively, and \( b \) itself is 400 s/mm². cDWI \((b = 800 \text{ s/mm}^2)\) was generated from two \( b \) values of 0 and 400 s/mm² by voxelwise fitting on a 3D workstation (Ziostation2, Ziosoft Inc., Tokyo, Japan) by using the following formula:

\[ S_b = S(0)e^{-b \times ADC} \]

It is well-known that ADC value displays positive value mathematically. In addition, previous reports have indicated that lower limit of ADC value in the musculoskeletal or endometrial malignant tumor was approximately \( 0.4 \times 10^{-3} \text{ mm}^2/\text{s} \). Therefore, noise reduction was performed with the cut-off values of ADC < 0 (ADC₀) and \( 0.4 \times 10^{-3} \text{ mm}^2/\text{s} \) (ADC₀₄) on the 3D workstation (Ziostation2). ROIs were manually placed by a neuroradiologist (KY with 16 years of experience) on the cholesteatoma, soft tissue adjacent to the cholesteatoma (non-cholesteatoma), and the contralateral cerebellum, respectively (Fig. 1). Careful attention was paid to avoid contamination of air for ROI positioning. The Ziostation2 software was used to draw the ROIs. The same ROIs were set up for ADC₀, ADC₀₄, and the control (without noise reduction algorithm).

**Image analysis**
The CNR was calculated by using the following formula:
\[ \text{CNR} = (\text{SI}_{\text{cholesteatoma}} - \text{SI}_{\text{non-cholesteatoma}})/\text{SD}_{\text{non-cholesteatoma}} \]
where SD_{non-cholesteatoma} is the standard deviation of the signal intensity (SI) within the ROI. The noise distribution is not homogeneous in parallel imaging, so it is good to estimate noise in close proximity to the site of SI measurement. In addition, the signal intensity ratio (rSI) of cholesteatoma to the contralateral cerebellum was calculated. The differences in the CNR and rSI were compared between ADC₀, ADC₀₄, and the control using one-way ANOVA followed by the Bonferroni correction for multiple comparison. Statistical analyses were performed using Graphpad Prism 5 (GraphPad Software Inc., San Diego, CA, USA). In the statistical analysis, the level of significance was set at \( P < 0.05 \).

**Results**
A total of 25 patients (M: F = 14:11; median age = 51 years; range, 14–78 years) with unilateral cholesteatoma were finally analyzed.

The average post-processing time was a few seconds.

All values are expressed as mean ± SD. Figure 2a shows the CNR of ADC₀₄ could be increased thanks to the noise reduction algorithm in the vast majority of subjects. The CNR of ADC₀₄ (7.24 ± 1.70) was significantly higher than those of the control (6.09 ± 1.80; \( P = 0.0023 \)) and ADC₀ (5.85 ± 1.20; \( P = 0.0021 \)). We observed no significant differences between the CNR of the control and that of ADC₀₄.

Figure 3 indicates that the rSI of ADC₀₄ (1.32 ± 0.31) tends to exhibit higher value compared with that of the control (1.14 ± 0.25; \( P = 0.08 \)) or ADC₀ (1.15 ± 0.26; \( P = 0.10 \), although no significant differences were found.

Figures 4 and 5 show representative cases.

**Discussion**
In the present study, we evaluated whether cDWI with a noise reduction algorithm would improve cholesteatoma to non-cholesteatoma CNR compared to cDWI without noise reduction algorithm. Our result show that the CNR could be increased for ADC₀₄ compared to the control. The noise reduction with the cut-off value of \( 0.4 \times 10^{-3} \text{ mm}^2/\text{s} \) is reasonable for clinical setting because it is assumed that the lower limit of ADC value in tumor tissue is approximately \( 0.4 \times 10^{-3} \text{ mm}^2/\text{s} \). Previous reports have shown that the improvement of diagnostic accuracy results from increasing suppression of the background signal. In addition, the average post-processing time was only a few seconds.
Fig. 2 Comparison of CNR among the control, ADC₀, and ADC₀.₄. (a) The line chart showing the CNR of ADC₀.₄ could be increased thanks to the noise reduction algorithm in the vast majority of subjects. (b) The boxplots reveal that the CNR of ADC₀.₄ are significantly higher than those of the control and ADC₀. No significant differences are found between the CNR of the control and that of ADC₀. ADC, apparent diffusion coefficient; CNR, contrast-to-noise ratio.

Therefore, cDWI with noise reduction algorithm is easily feasible in routine clinical practice.

The rSI of ADC₀.₄ tends to exhibit higher value compared with that of the control and ADC₀. It is not surprising that cDWI with noise reduction algorithm may be useful to distinguish cholesteatoma from adjacent granulation or fibrous tissue, whereas our study found that most subjects had only small amount of granulation or fibrous tissue around their cholesteatoma. Further experiments would be necessary in the near future.

The utility of cholesteatoma diagnosis using a non-EPI DWI technique has been reported in the literature.³ ⁴ ¹⁰ TSE–DWI is known to reduce curvilinear artifacts at the air–bone interface of the temporal bone because of its minimal image distortion compared with EPI-based DWI, and useful for the detection of even small cholesteatomas. DWI with high b value is necessary for the cholesteatoma detection, while an additional number of excitations is required to compensate for the SNR reduction, which tends to prolong the scanning time. Another drawback includes motion artifact due to prolongation of scanning time. In the present study, higher b value images (cDWI) were generated from the two different lower b values while maintaining adequate CNR. We hypothesized that the noise reduction algorithm was useful because of increase of high intensity noise (or artifact) due to the air–bone interface in the temporal bone region. Consequently, the usefulness of the noise reduction algorithm was proven in this study.

The b value of 800 s/mm² was computationally generated in the present study. The optimum b value has not yet been determined for the head and neck region. Although we should evaluate other b values, b values between 800 and 1000 s/mm² have been most commonly used.¹ ⁴ ¹⁰ ¹¹ In addition, the b value itself is not an important factor because ADC values strongly depends on the effective diffusion time.² ²² Thus, we believe that the effect of different b values distorted our results only minimally.

Our study has some limitations. First, the originally acquired DWI dataset (b = 800 s/mm²) was not evaluated. The effective diffusion time affects ADC value, which may reflect the restriction of water diffusion.² ²² Andica et al.² ²² reported that the different effective diffusion time resulted in the different ADC value of the epidermoid cysts. In our study, the diagnosis of cholesteatoma was confirmed during surgery in all cases. However, it should be taken into account that there is a difference of the diffusion time between the originally acquired and computed DWI. Second, the CNR of TSE–DWI was not compared with that of EPI-based DWI. Third, the optimum cut-off value was not evaluated in the present study. The voxels less than 0.4 × 10⁻³ mm²/s in ADC are not practically present. However, excessive noise reduction may cause inhomogeneous intensity in ADC₀.₄ images. The determination of the optimum cut-off value will be our next step and that is one of the current limitations.
Conclusion

The cholesteatoma to non-cholesteatoma CNR was increased using a noise reduction algorithm for clinical setting.

Conflicts of Interest

Yamato Shimomiya is an employee of Ziosoft, Atsushi Takemura is an employee of Philips. The remaining authors declare that they have no conflict of interest.

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