Combining SENSE and reduced field-of-view for high-resolution diffusion weighted magnetic resonance imaging

Jisu Hu1†, Ming Li2†, Yakang Dai1†, Chen Geng1, Baotong Tong1, Zhiyong Zhou1, Xue Liang2, Wen Yang2 and Bing Zhang2*

Abstract
Background: In diffusion-weighted magnetic resonance imaging (DWI) using single-shot echo planar imaging (ss-EPI), both reduced field-of-view (FOV) excitation and sensitivity encoding (SENSE) alone can increase in-plane resolution to some degree. However, when the two techniques are combined to further increase resolution without pronounced geometric distortion, the resulted images are often corrupted by high level of noise and artifact due to the numerical restriction in SENSE. Hence, this study is aimed to provide a reconstruction method to deal with this problem.

Methods: The proposed reconstruction method was developed and implemented to deal with the high level of noise and artifact in the combination of reduced FOV imaging and traditional SENSE, in which all the imaging data were considered jointly by incorporating the motion induced phase variations among excitations. The in vivo human spine diffusion images from ten subjects were acquired at 1.5 T and reconstructed using the proposed method, and compared with SENSE magnitude average results for a range of reduction factors in reduced FOV. These images were evaluated by two radiologists using visual scores (considering distortion, noise and artifact levels) from 1 to 10.

Results: The proposed method was able to reconstruct images with greatly reduced noise and artifact compared to SENSE magnitude average. The mean g-factors were maintained close to 1 along with enhanced signal-to-noise ratio efficiency. The image quality scores of the proposed method were significantly higher (P < 0.01) than SENSE magnitude average for all the evaluated reduction factors.

Conclusion: The proposed method can improve the combination of SENSE and reduced FOV for high-resolution ss-EPI DWI with reduced noise and artifact.

Keywords: Diffusion-weighted imaging, Reduced field-of-view, Single-shot echo-planar imaging, Parallel imaging, SENSE, g-Factor
Background
Diffusion-weighted magnetic resonance imaging (DWI) has shown great value both in clinical diagnosis and scientific studies. Single-shot echo-planar imaging (ss-EPI), which is widely used in current clinical DWI applications, has the advantages of short acquisition time, high signal-to-noise ratio (SNR) efficiency and less sensitivity to motion. Nevertheless, it also suffers from geometric distortion, image blurring and signal loss caused by the local field inhomogeneity of the static external magnetic field B₀, for which imaging resolution is highly restricted in practice, even when parallel imaging is applied, such as sensitivity encoding (SENSE) [1] and generalized auto-calibrating partially parallel acquisitions (GRAPPA) [2].

To increase in-plane resolution without pronounced geometric distortion, reduced field-of-view (FOV) imaging has been developed by reducing the FOV along the phase-encoding (PE) direction and thus shortening the echo train length [3–8]. This technique is especially useful for high resolution diffusion imaging of specific anatomical structures like prostate [9–11], spinal cord [4, 12–16], optic nerve [17–19] and etc. For this goal further, an intuitive strategy is to combine parallel imaging with reduced FOV, which has been successfully applied in 7 T applications [20–22]. However, due to less field inhomogeneity in low field systems, the coil sensitivity maps are smoother than those at 7 T [23]. Moreover, the even smaller variance in the reduced FOV worsens the ill-conditioning of SENSE encoding matrix. Therefore, the combination of SENSE and reduced FOV often results in unacceptable noise and artifact level at low field strengths [24–26], which cannot be corrected by just averaging over the magnitudes of individual SENSE results.

In this paper, we propose a method for improving the combination of SENSE and reduced FOV imaging in ss-EPI DWI to produce images with reduced noise and artifact level. The phase variations among repeated acquisitions were incorporated into the coil sensitivities and all the imaging data can be jointly used to reconstruct the final image. In vivo experiment was carried out and the proposed method was applied for the acquired imaging data. Results show that the proposed method is effective in reducing noise and artifact level compared to the conventional SENSE magnitude average.

Methods
Theory
For Cartesian sampling trajectory, let m denote the number of folding pixels and c the number of coils, SENSE [1] can be formulated as \( \mathbf{b} = \mathbf{S} \mathbf{p} \), where \( \mathbf{b} \) is the column vector of acquired aliased image pixels from all coil elements with length c, \( \mathbf{S} \) the \( c \times m \) sensitivity matrix, and \( \mathbf{p} \) the vector of the \( m \) pixels in the to-be-reconstructed image. For simplicity, noise correlation is not considered here. The unfolded image can then be reconstructed through a pseudo-inverse by \( \mathbf{p} = (\mathbf{S}^{H} \mathbf{S})^{-1} \mathbf{S}^{H} \mathbf{b} \) for each position in the aliased coil images. In matrix theory, the instability of matrix inversion is measured by the condition number. Correspondingly in SENSE, the g-factor is introduced and formulated as \( g_{p} = \sqrt{\text{tr}([\mathbf{S}^{H} \mathbf{S}]^{-1})} \), which accounts for the SNR penalty imposed by the ill-conditioned matrix inversion. Hence, the g-factor is determined by the independency of coil sensitivity maps [26]. In reduced FOV, as the aliased pixel locations are closer than in the full FOV, the coil sensitivities in each row of \( \mathbf{S} \) become similar, especially in low
field systems. Consequently, matrix $S$ will be very rank-deficient, which in turn results in high g-factors. In this case, noise in the sampled data can be much more amplified in the reconstructed images of reduced FOV than those of full FOV. This intrinsic problem makes the combination of reduced FOV and SENSE difficult in low field systems.

To solve this issue, it is proposed in this work to jointly reconstruct the final image from all the imaging data in the framework of SENSE. Due to the sensitivity to motion in DWI, images of repeated acquisitions usually have motion induced phase differences while the magnitudes generally remain unchanged. For this reason, conventionally these images are reconstructed individually and only the magnitudes are averaged for the final image (SENSE magnitude average, the default vendor implementation). In the proposed method, though, the phase information is incorporated to construct the new formulation for all NSAs (number of signals averaged) like SENSE,

$$
\bar{b} = \begin{bmatrix}
S_1(r_1) \exp(i\phi_1(r_1)) & \cdots & S_1(r_m) \exp(i\phi_1(r_m)) \\
\vdots & \ddots & \vdots \\
S_c(r_1) \exp(i\phi_1(r_1)) & \cdots & S_c(r_m) \exp(i\phi_1(r_m)) \\
\vdots & \ddots & \vdots \\
S_1(r_1) \exp(i\phi_a(r_1)) & \cdots & S_1(r_m) \exp(i\phi_a(r_m)) \\
\vdots & \ddots & \vdots \\
S_c(r_1) \exp(i\phi_a(r_1)) & \cdots & S_c(r_m) \exp(i\phi_a(r_m))
\end{bmatrix}
\rho = \bar{S}\rho
$$

where $\bar{b}$ is the column vector of aliased image pixels from all coils and all NSAs with length $ca$, $S$ the original coil sensitivity profiles, $\phi$ the phase of each NSA, $\bar{S}$ the new $ca \times m$ sensitivity matrix with $a$ counting the NSA, and $r$ the pixel location. In this regard, data from all NSAs can be treated as sampled from the virtual coil elements with the new set of modulated coil sensitivity profiles defined by

$$
\bar{S}_i(r) = S_{(i-1) \mod c+1}(r) \exp(i\phi_{\text{floor}((i-1)/c)+1}(r)), \quad i = 1, 2, \ldots, \bar{c}
$$

where $i$ counts through the virtual coil elements. Therefore, the number of virtual coil elements $\bar{c}$ in Eq. (2) is $a$ times of $c$. With the same unknowns but more equations, the linear equation system in Eq. (1) becomes more stable than the original SENSE for each individual acquisition and the g-factors can be kept in a reasonable level.

The phase information of each NSA can be obtained either from the image itself [27–29] or the navigator echo [30, 31]. In this work, total variation (TV) denoising [32] was applied for phase estimation formulated as

$$
\arg \min_{x \in \mathbb{C}^n} \left\{ 0.5||y - x||_2^2 + \lambda ||\nabla x||_1 \right\}
$$

where the first term measures the sum of squares error of the noisy signal $y$ and the denoised $x$, and the second term represents the L1 norm of the gradient of $x$ with the regularization term $\lambda$. Traditional SENSE was performed for each NSA followed by TV denoising according to Eq. (3), and the phase component was extracted from the
denoised image. Then the virtual coil sensitivity profiles were constructed according to Eq. (2) for the proposed reconstruction.

**In vivo experiment**

As a specific example of reduced FOV imaging, the previously proposed Improved Zoom FOV imaging (iZoom) [3] was used for data acquisition in this work. This method reduces the FOV along the PE direction with two dimensional selective excitation techniques and provides a new algorithm to improve the image uniformity. Hence, we name the proposed reconstruction as ZOOM-SENSE.

The in vivo experiment was designed to validate the effectiveness of the proposed method in reducing noise and artifact level compared to SENSE magnitude average at 1.5 T (Philips Healthcare, Multiva, Suzhou, China) using a 16-channel head-spine coil. Sagittal images of human cervical spinal cord from ten subjects were acquired after obtaining written informed consent and all human studies were performed under Institutional Review Board (IRB) approval from the local institution. The following parameters were kept the same in all the diffusion acquisitions: in-plane resolution = 1.4 × 1.4 mm², slice thickness = 5 mm, 10 slices, PE direction = anterior/posterior (AP), TR = 3500 ms, NSA = 6, b = 0 s/mm² and 3 diffusion encoding directions with b = 500 s/mm², and scan time = 2.5 min. Specifically, the traditional full FOV (230 × 230 mm²) ss-EPI DWI with reduction factor (R) of 2 was first performed for comparison purpose. Then iZoom ss-EPI DWI was scanned using FOV of 230 × 80 mm² with R = 1/1.5/2/2.5. Since the echo train lengths varied among these scans, the effective TE in full FOV DWI was 140 ms. For reduced FOV, the effective TEs for R = 1/1.5/2/2.5 were 124/99/86/79 ms. Finally, the corresponding T2-weighted images were obtained as the reference for evaluating the distortion levels of individual diffusion images.

All the diffusion imaging data were reconstructed using SENSE magnitude average first. The sensitivity maps were obtained by dividing the individual phased-array coil images by the corresponding body coil image in a reference scan. Then for reduced FOV data sampling with R = 1.5/2/2.5, ZOOM-SENSE was performed and the regularization term in TV denoising was set to be 0.1 empirically. The g-factor maps of both SENSE and ZOOM-SENSE were calculated along with the reconstructions to visualize the noise amplification. All the reconstructions were implemented offline in MATLAB (MathWorks, Natick, MA).

The relative SNRs (rSNR) of SENSE magnitude average or ZOOM-SENSE to full Fourier encoding [33] in reduced FOV for R = 1.5/2/2.5 were evaluated for each pixel according to

\[
\text{rSNR} = \exp \left( \frac{\Delta TE}{T_2} \right) \frac{1}{g \sqrt{R}}
\]  

(4)

where the numerator accounts for the signal increase with the TE reduction by \(\Delta TE(T_{\text{full-sampling}} - T_{\text{reduced-sampling}})\), whereas the denominator includes the square root of R due to reduced data sampling and the g-factor penalty. The T2 relaxation time of white matter at 1.5 T was assumed to be 80 ms here. Both the spatial mean and minimum of the relative SNRs across the image were calculated for each R and reconstruction.
The diffusion images were also visually scored in terms of image quality, geometric distortion and artifact levels in a 10-point scale (10 for the best result). For each subject, the following methods were compared as detailed in Table 2. Hence, a total of 80 images were displayed in a random order and evaluated by two radiologists both with 3-year experience in MRI diagnosis. They were blinded to the acquisition parameters and reconstruction methods. The corresponding T2-weighted images were given for evaluating the distortion levels. The qualitative visual score measurements of SENSE magnitude average and ZOOM-SENSE results with the same acquisition parameters were compared using a Wilcoxon signed-rank test with a significance threshold of $P < 0.05$ in SPSS (SPSS Inc., Chicago, IL).

**Results**

Figure 1 shows all the diffusion images obtained using different acquisition parameters and reconstructions from one subject with slight cervical disc herniation (diffusion images show no damage in the spinal cord) along one diffusion direction. The red contour of the spine geometry is obtained from the corresponding T2-weighted image (Fig. 1a) and is overlaid onto all the presented diffusion images to evaluate the distortion levels. It can be seen that in the full FOV DWI with $R = 2$ (Fig. 1b) the spinal cord is most distorted from the true structure geometry among all the diffusion images. When using iZOOM excitation to reduce the FOV in the PE direction to 80 mm, the distortion level
is reduced using fully sampled data (Fig. 1c) compared to the full FOV case and is further lowered gradually as \( R \) increases (Fig. 1d–f). Note that with \( R = 2 \) and 2.5, the geometry distortion is almost negligible and the spinal cord is well aligned with the red contour.

Figure 1 also highlights the performance comparison of SENSE magnitude average and ZOOM-SENSE reconstructions in reduced FOV DWI. In SENSE magnitude average results, there is a distinct rise in the noise and artifact level as \( R \) increases (Fig. 1d–f). Especially at \( R = 2.5 \), the imaged object is covered by terribly heavy noise and artifact and the images are clinically unusable (Fig. 1f). In contrast, the anatomical details are shown more clearly in the corresponding ZOOM-SENSE results even using high reduction factors (Fig. 1g–i). The corresponding g-factor maps of SENSE and ZOOM-SENSE reconstructions are presented in Fig. 2, and the g-factors and relative SNRs are detailed in Table 1 for the three reduction factors. It can be observed

![Figure 2](image-url)

**Table 1** Comparison of SENSE magnitude average and ZOOM-SENSE in terms of g-factors and rSNRs

| \( R \) | SENSE magnitude average | ZOOM-SENSE |
|-------|-------------------------|------------|
|       | g-Factor | rSNR | g-Factor | rSNR |
|       | Mean | Max | Mean | Min | Mean | Max | Mean | Min |
| 1.5   | 1.08 | 1.97 | 1.0057 | 0.5662 | 1.01 | 1.4 | 1.0902 | 0.7948 |
| 2     | 1.43 | 4.35 | 0.8028 | 0.2612 | 1.04 | 1.68 | 1.0472 | 0.6762 |
| 2.5   | 2.18 | 8.97 | 0.5599 | 0.1238 | 1.11 | 2.22 | 0.9745 | 0.5001 |
that the maximum and mean g-factors of SENSE can reach up to 8.97 and 2.18 at R = 2.5, which will result in undesirable noise amplification, while those of ZOOM-SENSE can be kept no more than 2.22 and 1.11 in the same case. The relative SNRs demonstrate that the highly elevated g-factors outgrow the T2 effect in SENSE magnitude average reconstruction, whereas the SNR efficiency can be enhanced or at least maintained even at high reduction factors using ZOOM-SENSE.

Figure 3 shows the phase maps of initial SENSE results before and after denoising for 4 representative NSAs. Due to the problem of individual SENSE reconstruction in reduced FOV, the noise level is higher in the directly derived phase maps. After denoising, smoother phase maps can be obtained for the proposed reconstruction. More importantly, it is shown that the phase variations are obvious among different NSAs, which guarantees that imaging data from all NSAs can be combined effectively to perform ZOOM-SENSE reconstruction for reduced g-factors and better image quality.

Table 2 presents the imaging scores evaluated by the two reviewers for different combinations of scanning parameters and reconstruction methods. The first reviewer gives the highest mean score to the ZOOM-SENSE results at R = 1.5, while the second reviewer believes that the overall image quality is best in ZOOM-SENSE results at R = 2. For reduced FOV with R = 1.5/2/2.5, the image scores of ZOOM-SENSE are significantly higher than those of the corresponding SENSE magnitude average results (P = 0.007, 0.004 and 0.004 for the first reviewer, and P = 0.004, 0.005 and 0.004 for the second reviewer), which confirms the improvement of ZOOM-SENSE over SENSE magnitude average by radiologists.

Fig. 3 Phase maps from initial SENSE results before (upper row) and after (lower row) denoising for 4 representative NSAs for R = 2.
Discussion and conclusion

In this study we proposed exploiting the phase inconsistency among NSAs to jointly perform SENSE-like reconstruction to improve the combination of reduced FOV excitation and parallel acquisition in ss-EPI DWI. To increase in-plane resolution without pronounced geometric distortion, the echo train length can be shortened intuitively by combining reduced FOV imaging and SENSE. However, compared to the full FOV case, closer locations of aliased pixels in reduced FOV can easily result in ill-conditioned inverse of SENSE encoding matrix and the corresponding g-factors will be much more elevated. In that case, even a small amount of noise in the sampled data can be amplified in the final image to an unacceptable extent. This happens when FOV is sufficiently small along with just slightly higher reduction factors. Moreover, due to the phase inconsistency among NSAs, individual SENSE results cannot be complex averaged. The in vivo experiment results confirm that the proposed ZOOM-SENSE reconstruction outperforms SENSE magnitude average in reducing noise and artifact level even when the SENSE results are severely corrupted. Note that the reduced TE by parallel acquisition will increase the SNR in the imaging data, but this advantage is hampered by the strong g-factor penalty introduced in traditional SENSE reconstruction. In ZOOM-SENSE, however, the overall g-factors are maintained close to 1, for which some anatomical details (e.g., the intervertebral discs) are better revealed with under-sampled data. More generally, although iZoom is used in this study for reduced FOV excitation, the proposed reconstruction can theoretically be combined with other small FOV imaging techniques. As well, it can potentially turn the combination of reduced FOV and SENSE into immediate practical use without additional hardware requirements or sequence modifications.

Despite the above advantages, it should be pointed out that the performance of ZOOM-SENSE is closely related mainly to two factors. The first is the extent of phase inconsistency among different NSAs. If more physiological motion is introduced during data acquisition, the phase maps will be very different from NSA to NSA, which brings stronger independency of the modulated coil sensitivity profiles in ZOOM-SENSE and results in more decreased overall g-factors. On the contrary, if ideally without NSA-to-NSA phase variations, the g-factor map in ZOOM-SENSE will remain the same as in traditional SENSE and performing ZOOM-SENSE will

| FOV along PE direction (mm) | R | Reconstruction          | Scores by reviewer 1 (mean ± STD) | Scores by reviewer 2 (mean ± STD) |
|-----------------------------|---|-------------------------|----------------------------------|----------------------------------|
| 230                         | 2 | SENSE magnitude average | 5.80 ± 1.14                      | 5.30 ± 0.48                      |
| 80                          | 1 | SENSE magnitude average | 6.50 ± 1.35                      | 5.70 ± 0.48                      |
| 80                          | 1.5 | SENSE magnitude average | 7.00 ± 0.67                      | 6.30 ± 0.48                      |
| 80                          | 1.5 | ZOOM-SENSE              | 9.20 ± 0.79                      | 8.30 ± 0.48                      |
| 80                          | 2  | SENSE magnitude average | 3.60 ± 0.84                      | 4.10 ± 1.20                      |
| 80                          | 2  | ZOOM-SENSE              | 8.70 ± 0.82                      | 8.90 ± 0.57                      |
| 80                          | 2.5 | SENSE magnitude average | 1.60 ± 0.84                      | 2.30 ± 0.48                      |
| 80                          | 2.5 | ZOOM-SENSE              | 7.80 ± 0.42                      | 7.80 ± 0.42                      |
almost be identical to complex average of individual SENSE results. The second is the accuracy of phase estimation after initial SENSE results. Although in this study denoising is successfully used to get smooth phase maps for ZOOM-SENSE, the particularly high noise and artifact level is likely to hinder the phase estimation in some circumstances. When using higher reduction factors, the image object may be even buried in the noise and artifact in the initial SENSE results, making the phase estimation by denoising unreliable and leaving residual artifact on the images (e.g., the ZOOM-SENSE results for $R = 2.5$). In this case, an alternative way to obtain the phase information may be to acquire navigator echoes during the scan [30]. This might potentially outperform the phase estimation in this study but needs further investigations.

In conclusion, it is demonstrated that the proposed ZOOM-SENSE improves the combination of SENSE and reduced FOV at 1.5 T. By incorporating the phase information from all NSAs, all the imaging data can be treated as sampled from pseudo coil elements so that the matrix inversion becomes more stable than the original SENSE. Hence, the proposed method can produce images with decreased noise and artifact compared to SENSE magnitude average and it can be readily used in clinical applications.

Abbreviations

DWI: diffusion-weighted magnetic resonance imaging; ss-EPI: single-shot echo planar imaging; FOV: field-of-view; SENSE: sensitivity encoding; GRAPPA: generalized auto-calibrating partially parallel acquisitions; SNR: signal-to-noise ratio; rSNR: relative signal-to-noise ratio; PE: phase-encoding; NSA: number of signal average; TV: total variation; iZOOM: improve zoomed FOV imaging.

Authors’ contributions

Study concept and design JH, ML and BZ; drafting of the manuscript JH; designing and conducting the experiments JH, ML, XL and WY; revision of the manuscript YD, CG, BT and ZZ. All authors read and approved the final manuscript.

Author details

1 Suzhou Institute of Biomedical Engineering and Technology, Chinese Academy of Sciences, Suzhou, Jiangsu, China.
2 Department of Radiology, Affiliated Drum Tower Hospital of Nanjing University Medical School, 321 Zhongshan Road, Nanjing, Jiangsu, China.

Acknowledgements

The authors thank Feng Huang, Zhigang Wu, Wenxing Fang and Congyu Liao for the aid in experiments and helpful discussions.

Competing interests

The authors declare that they have no competing interests.

Availability of data and materials

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

Consent for publication

The data for experiments have obtained the consent of the volunteers and Affiliated Drum Tower Hospital of Nanjing University Medical School, and have been allowed for publication.

Ethics approval and consent to participate

All human studies have been approved by Medical Ethics Committee of Affiliated Drum Tower Hospital of Nanjing University Medical School.

Funding

This work was supported by National Natural Science Foundation of China (81571040, 81300925, 61501452), National and Provincial Postdoctoral Project (BE179, 1501076A), 6th Peak Talent Project (WSN-QS0), Key Project of Nanjing Health Bureau (ZXX14027), Key Research and Development Plan of Jiangsu (BE2016613, BE2016010-3, BE20160310-4, BE2016005, BE2017663), Virtual Reality Key Laboratory of Intelligent Interaction and Application Technology of Suzhou, Foundation of Key Laboratory in Science and Technology Development Project of Suzhou (SYG201606, SZS201609, SYG201706, SYG201707), SND Medical Plan Project (2016Z010), and National 863 Program of China (2015AA020514).

Publisher’s Note

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.
References

1. Pruessmann KP, Weiger M, Scheidegger MB, Boesiger P. SENSE: sensitivity encoding for fast MRI. Magn Reson Med. 1999;42(3):952–62.

2. Griswold MA, Jakob PM, Heidemann RM, Nittka M, Jellus V, Wang JM, Kiefer B, Haase A. Generalized autocalibrating partially parallel acquisitions (GRAPPA). Magn Reson Med. 2002;47(6):1202–10.

3. Wu Z, Fang WX, Huang F. B1 insensitive zoomed FOV imaging. In: Proceedings of the 23rd annual meeting of ISMRM, Tokyo, 2015. p. 4161.

4. Saritas EU, Cunningham CH, Lee JH, Han ET, Nishimura DG. DWI of the spinal cord with reduced FOV single-shot EPI. Magn Reson Med. 2008;60(2):468–73.

5. Wilm BJ, Svensson J, Henning A, Pruessmann KP, Boesiger P, Kollia SS. Reduced field-of-view MRI using outer volume suppression for spinal cord diffusion imaging. Magn Reson Med. 2007;57(3):625–30.

6. Finsterbusch J. Improving the performance of diffusion-weighted inner field-of-view echo-planar imaging based on 2D-selective radiofrequency excitations by tilting the excitation plane. J Magn Reson Imaging. 2012;35(4):984–92.

7. Jeong E-K, Kim S-E, Guo J, Khomovsky EG, Parker DL. High-resolution DTI with 3D interleaved multislice reduced FOV single-shot diffusion-weighted EPI (3D ss-rFOV-DWEPI). Magn Reson Med. 2005;54(6):1575–9.

8. Wheeler-Kingshott CAM, Hickman SJ, Parker GJM, Ciccarelli O, Symms MR, Miller DH, Barker GJ. Investigating cervical spinal cord structure using axial diffusion tensor imaging. Neuroimage. 2002;16(1):93–102.

9. Korn N, Kupanowicz J, Banerjee S, Stanobinetos O, Saritas E, Novorolski S. Reduced-FOV excitation decreases susceptibility artifact in diffusion-weighted MRI with endorectal coil for prostate cancer detection. Magn Reson Imaging. 2015;33(1):56–62.

10. Tamada T, Ream JM, Doshi AM, Taneja SS, Rosenkrantz AB. Reduced field-of-view diffusion-weighted magnetic resonance imaging of the prostate at 3 T: comparison with standard echo-planar imaging technique for image quality and tumor assessment. J Comput Assist Tomogr. 2017;41(6):949–56.

11. Wamsdahl BA, Borisch EA, Kawashima A, Riederer SJ, Froemming AE. Conventional vs. reduced field of view diffusion weighted imaging of the prostate: comparison of image quality, correlation with histology, and inter-reader agreement. Magn Reson Imaging. 2018;47:67–76.

12. Alizadeh M, Fisher J, Saksena S, Sultan Y, Conklin CJ, Middleton DM, Finsterbusch J, Krisa L, Flanders AE, Faro SH, et al. Reduced field of view diffusion tensor imaging and fiber tractography of the pediatric cervical and thoracic spinal cord injury. J Neurotrauma. 2018;35(3):452–60.

13. Andre JB, Zaharchuk G, Saritas E, Komakula S, Shankaranarayan A, Banerjee S, Rosenberg J, Nishimura DG, Fischbein NJ. Clinical evaluation of reduced field-of-view diffusion-weighted imaging of the cervical and thoracic spine. Am J Neuroradiol. 2012;33(10):1860–6.

14. Crombe A, Alberti N, Hiba B, Uettwiller M, Doussset V, Tordius T. Cervical spinal cord DTI is improved by reduced FOV with specific balance between the number of diffusion gradient directions and averages. Am J Neuroradiol. 2016;37(11):2163–70.

15. Park EH, Lee YH, Jeong E-K, Roh YH, Suh J-S. Diffusion tensor imaging focusing on lower cervical spinal cord using 2D reduced FOV interleaved multislice single-shot diffusion-weighted echo-planar imaging: comparison with conventional single-shot diffusion-weighted echo-planar imaging. Magn Reson Imaging. 2015;33(4):401–6.

16. Sapkota N, Shi X, Shah LM, Bisson EF, Rose JW, Jeong E-K. Two-dimensional single-shot diffusion-weighted stimulated EPI with reduced FOV for ultrahigh-b radial diffusion-weighted imaging of spinal cord. Magn Reson Med. 2017;77(6):2167–73.

17. Seeger A, Schulze M, Schuettarf F, Ernemann U, Hauser T-K. Advanced diffusion-weighted imaging in patients with optic neuritis deficit—value of reduced field of view DWI and readout-segmented DWI. Neuroradiol J. 2018;31(2):272–7.

18. Wang L, Fan K, Zhang Y, Chen Y, Tian Q, Shi D. Quantitative assessment of optic nerve in patients with Leber’s hereditary optic neuropathy using reduced field-of-view diffusion tensor imaging. Eur J Radiol. 2017;93:24–9.

19. Zhang Y, Guo X, Wang M, Wang L, Tian Q, Zheng D, Shi D. Reduced field-of-view diffusion tensor imaging of the optic nerve in retinitis pigmentosa at 7 T. Am J Neuroradiol. 2016;37(8):1510–5.

20. Heidemann RM, Ivanov D, Trampel R, Fasano F, Meyer H, Pfeuffer J, Turner R. Isotropic submillimeter fMRI in the human brain at 7 T combining reduced field-of-view imaging and partially parallel acquisitions. Magn Reson Med. 2012;68(5):1506–16.

21. Wargo CJ, Gore JC. Localized high-resolution DTI of the human midbrain using single-shot EPI, parallel imaging, and outer-volume suppression at 7 T. Magn Reson Imaging. 2013;31(6):810–9.

22. von Morze C, Kelley DAC, Shepherd TM, Banerjee S, Xu DA, Hess CP. Reduced field-of-view diffusion-weighted imaging of the brain at 7 T. Magn Reson Imaging. 2010;28(10):1541–5.

23. Wiesinger F, Banerjee S, Pruessmann KP. Parallel imaging performance as a function of field strength—an experimental investigation using electrodynamic scaling. Magn Reson Med. 2004;52(5):953–64.

24. Wilm BJ, Gamper U, Buehrer M, Pruessmann KP, Luchinger R, Kollia SS, Boesiger P. The benefit of reduced FOV diffusion imaging with and without SENSE. In: Proceedings of joint annual meeting ISMRM-ESMRMB. Berlin, 2007.

25. Wiesinger F, Boesiger P, Pruessmann KP. Electrodynamic scaling and ultimate SNR in parallel MR imaging. Magn Reson Med. 2004;52(2):376–90.

26. Larkman DJ. The g-factor and coil design. Berlin: Springer; 2007. p. 37–48.

27. Chen NK, Guidon A, Chang HC, Song AW. A robust multi-shot scan strategy for high-resolution diffusion weighted MRI enabled by multiplexed sensitivity-encoding (MUSE). Neuroimage. 2013;72:41–7.
28. Liu CL, Moseley ME, Bammer R. Simultaneous phase correction and SENSE reconstruction for navigated multi-shot DWI with non-cartesian k-space sampling. Magn Reson Med. 2005;54(6):1412–22.
29. Liao C, Chen Y, Cao X, Chen S, He H, Mani M, Jacob M, Magnotta V, Zhong J. Efficient parallel reconstruction for high resolution multishot spiral diffusion data with low rank constraint. Magn Reson Med. 2017;77(3):1359–66.
30. Jeong HK, Gore JC, Anderson AW. High-resolution human diffusion tensor imaging using 2-D navigated multishot SENSE EPI at 7 T. Magn Reson Med. 2013;69(3):793–802.
31. Porter DA, Heidemann RM. High resolution diffusion-weighted imaging using readout-segmented echo-planar imaging, parallel imaging and a two-dimensional navigator-based reacquisition. Magn Reson Med. 2009;62(2):468–75.
32. Rudin LI, Osher S, Fatemi E. Nonlinear total variation based noise removal algorithms. Physica D. 1992;60(1–4):259–68.
33. Jaermann T, Crelier G, Pruessmann KP, Golay X, Netsch T, van Muiswinkel AMC, Mori S, van Zijl PCM, Valavanis A, Kollias S, Boesiger P. SENSE-DTI at 3 T. Magn Reson Med. 2004;51(2):230–6.