Erratum to Kellman P, McVeigh ER. Image Reconstruction in SNR Units: A General Method for SNR Measurement. Magn Reson Med. 2005; 54:1439–1447.

## Summary of revision

In MRM 54:1439–1447, a noise pre-whitening approach was used prior to image reconstruction. In this approach, the noise correlation matrix becomes the identity defined with unit variance for real and imaginary noise components. Hence, the noise correlation  $\mathbf{R}_n$  did not appear explicitly in Eq. [7] as required. In this erratum we re-write Eq. [7] to include the noise correlation. This step required modifying the definition of noise correlation (Eq. [1]) to be consistent with parallel imaging literature [5], which led to further modification of Eq [5-6].

The results are not affected since the experimental validation used the alternative noise pre-whitening approach. The updated equations include the pre-whitened form as Eq. [8]. The revised Eq. [1] and re-written subsection Array Combining follow.

## Noise Measurement

Eq. [1] is revised to conform with the more familiar conjugate form.

$$R_{ij} = (1/N) \sum_{k=1}^{N} n_i(k) n_j^*(k)$$

and includes noise power contribution from real and imaginary components.

<u>Array Combining (this replaces Array Combining section in its entirety)</u>

Roemer, et al. [7] formulated equations for phased array combined image reconstruction for both root-sum-of squares (RSS) magnitude and optimum B<sub>1</sub>-weighted combining, and Pruessmann, et al. [5] formulated equations for parallel imaging using the image domain SENSE method.

$$SNR_{RSS} = \sqrt{2(\mathbf{p}^{H}\mathbf{R}_{n}^{-1}\mathbf{p})}$$

$$SNR_{B1-weighted} = \sqrt{2} |\mathbf{b}^{H}\mathbf{R}_{n}^{-1}\mathbf{p}| / \sqrt{\mathbf{b}^{H}\mathbf{R}_{n}^{-1}\mathbf{b}}$$
[6]

$$SNR_{SENSE} = \sqrt{2} \left| \mathbf{u}^T \mathbf{p} \right| / \sqrt{\mathbf{u}^T \mathbf{R}_n \mathbf{u}^*}$$

The equations [5] and [6] for SNR scaled images follow Roemer's formulation, where SNR is the pixel intensity in SNR units,  $\bf p$  is the vector of complex image values for each coil,  $\bf b$  is the vector of complex coil sensitivities, and  $\bf R_n$  is the noise correlation matrix. The SNR estimates for the resultant magnitude images are very good at high SNR and may be further corrected to provide a good estimate at low SNR [1,2]. The factor of  $\sqrt{2}$  is used to account for the SNR definition in terms of the real channel noise component, consistent with MRI literature [1,2].

In the case of SENSE [5], described by Eq. [7],  $\mathbf{u}$  represents the  $N_{\text{coils}}x1$  column vector of unmixing coefficients which are a reformatting of the unmixing matrix  $\mathbf{U} = \left(\mathbf{S}^H \mathbf{R}_n^{-1} \mathbf{S}\right)^{-1} \mathbf{S}^H \mathbf{R}_n^{-1}$  which contains the optimum noise weighting (S is the coil sensitivity matrix, superscript  $^H$  denotes the Hermitian or conjugate transpose operation) (i.e., column vector  $\mathbf{u}$  is transpose of k-th row of  $\mathbf{U}$  for the k-th subimage). The unmixing vector is reformatted to be applied as a phased

array combiner to the full FOV images reconstructed with zerofilling of undersampled data. In the case of SENSE [5], Eq. [7] may be written equivalently as  $SNR_{SENSE} = \sqrt{2} \left| \mathbf{u}^T \mathbf{p} \right| \sqrt{\left(\mathbf{S}^H \mathbf{R}_n^{-1} \mathbf{S}\right)_{k,k}} / g_k \text{ with the g-factor [5]}$  for k-th sub-image computed as  $g(x, y + kFOV / R) = g_k(x, y) = \sqrt{\left(\mathbf{S}^H \mathbf{R}_n^{-1} \mathbf{S}\right)_{(k,k)}^{-1} \left(\mathbf{S}^H \mathbf{R}_n^{-1} \mathbf{S}\right)_{(k,k)}^{-1}},$  0 < y < FOV/R, with field-of-view = FOV and k=0,1,...R-1, with R aliased images from uniform undersampling by R. In the preceding, the complex coil sensitivities  $\mathbf{s}_i(x,y)$  are assumed to be known. Sensitivity estimates may be made using in-vivo images aquired separately [5], or adaptively from the images in an auto-

Eqs. [5-7] describe the array combining which incorporate the noise weighting. Pixel intensities in SNR units are calculated by using the noise weighting as described along with scaling the signal to preserve the same effective gain as the noise. Equivalently, a noise pre-whitening step may be applied by combining channels to create virtual channels which are uncorrelated and have unit variance [11]. Pre-whitening may be realized by combining using the matrix L<sup>-1</sup> (i.e., p<sub>pre-whitened</sub>=L<sup>-1</sup>p), where the lower triangular matrix L is calculated from the Cholesky factorization of the noise correlation matrix, R<sub>n</sub>=L L<sup>H</sup> [11]. In the case of noise pre-whitening, SNR<sub>SENSE</sub> may be calculated as

calibrating manner [8-10].

$$SNR_{SENSE} = \sqrt{2} \left| \tilde{\mathbf{u}}^T \mathbf{p}_{pre-whitened} \right| / \sqrt{\tilde{\mathbf{u}}^T \tilde{\mathbf{u}}^*}$$
 [8]

where the noise correlation matrix for the whitened data becomes the identity, and  $\tilde{\mathbf{U}}$  represents the vector of unmixing coefficients which are a reformatting of the unmixing matrix  $\tilde{\mathbf{U}} = \left(\tilde{\mathbf{S}}^H \tilde{\mathbf{S}}\right)^{-1} \tilde{\mathbf{S}}^H$ , with the sensitivity matrix  $\tilde{\mathbf{S}} = \mathbf{L}^{-1} \mathbf{S}$ 

corresponding to the virtual channels, i.e., following noise prewhitening. In the case of pre-whitening, the denominator of Eq.

[8] 
$$\sqrt{\tilde{\mathbf{u}}^T \tilde{\mathbf{u}}^*}$$
 is equivalent to the g-factor, i.e.,  $SNR_{SENSE} = \sqrt{2} \left| \tilde{\mathbf{u}}^T \mathbf{p}_{pre-whitened} \right| / g_k$ , where it is assumed that

relative sensitivities are used such that  $diag(\tilde{\mathbf{S}}^H\tilde{\mathbf{S}})=1$ , i.e., normalized by RSS combined magnitude  $(\tilde{s}_i(x,y)/\sqrt{\sum |\tilde{s}_i(x,y)|^2}, \tilde{s}_i(x,y))$  is coil sensitivity profile for i-th virtual channel).

<u>Results</u>

The results used the pre-whitening form based on Eq [8]. The sentence regarding the image of Fig 8(a) should be revised according to the new notation. "The image of Fig. 8(a) is scaled by all noise factors with the exception of the SENSE g-factor  $(\sqrt{\tilde{\mathbf{u}}^T\tilde{\mathbf{u}}^*})$ ." Likewise for the SENSE implementation of Eq. [7], the output may be calculated as  $\sqrt{2} |\mathbf{u}^T\mathbf{p}| \sqrt{(\mathbf{S}^H\mathbf{R}_n^{-1}\mathbf{S})_{k,k}}$  with

separate scaling by the g-map, g<sub>k</sub>, to avoid local noise enhancement in cases of high g-factor.

The conclusions and significance of this paper are still valid. We apologize for any inconvenience caused by these errors.