Regularized Estimation of Magnitude and Phase of Multiple-Coil B1 Field via Bloch-Siegert B1 Mapping

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Introduction: Parallel excitation pulse design usually requires accurate magnitude and phase maps of the B₁ field produced by each coil. Bloch-Siegert (BS) B₁ mapping [1] has been shown to be fast and accurate; however, the B₁ map produced by this phase-based method may suffer from low SNR in low magnitude regions having insufficient excitation or low spin density. This problem has been mitigated in [2] by using combinations of multiple coils for imaging excitation. However, it does not help for low spin density regions and insufficient excitation is still possible as accurate B₁ maps are unknown; furthermore, estimation of B₁ phase needs another set of scans, which is time-consuming and information redundant. In this work, we propose a regularized method to jointly estimate the magnitude and (relative) phase of multi-coil B₁ maps from BS B₁ mapping data without using additional scans for phase estimation. By utilizing the prior knowledge that B₁ maps are smooth [3], the regularization terms can help improve quality of the B₁ maps in low magnitude regions. The method was demonstrated by phantom experiments.

Theory: We propose to acquire the standard BS B₁ mapping data [1] that needs 2*R scans (R = the number of coils). In each scan, the same coil combination is used for the BS pulse and its corresponding slice excitation pulse. The composite B_1^+ field $E_r(x)$ produced at each time r is described in (1), where $\alpha_{r,j}$ is a complex weight that indicates how the coils are combined at each time, $B_i(x)$ is the magnitude of the B_1 map produced by the *jth* coil with a unit input current, and $\phi_j(x)$ is the corresponding B₁ phase map. A convenient choice of $\alpha_{r,j}$ is the "all-but-one" strategy, where $\alpha_{r,r}=0$ and $\alpha_{r,j}=1$ when $j \neq r$. The signal models for the BS data (reconstructed images) of the *rth* pair of scans are described in (2), where r = 1, 2, ..., R; the superscripts +/- denote the scan with $+\omega_{RF}$ or $-\omega_{RF}$ BS pulse, $I_r^{\pm}(x)$ is the image of each scan, μ is the ratio between the actual flip and $|E_r(x)|$, $m_r^{\pm}(x)$ is the magnitude related to spin density, T_1 , T_2 , T_R , T_E , flip angle, receive sensitivity, magnetization transfer (MT) effect, etc., $\phi_h(x)$ is the corresponding background phase, and $K_{RS}^{\pm}(x)$ is the BS pulse constant that incorporates the B₀ field map $\omega_0(x)$ [1]. We simplify (2) into (3) by changing variables: $z_r(x) \triangleq$ $E_r(\mathbf{x})e^{i\phi_b(\mathbf{x})}$, $\tilde{\phi}_r(\mathbf{x}) \triangleq \angle z_r(\mathbf{x})$, $\tilde{B}_r(\mathbf{x}) \triangleq |z_r(\mathbf{x})|$, and $M_r^{\pm}(\mathbf{x}) \triangleq sin(\mu|z_r(\mathbf{x})|) m_r^{\pm}(\mathbf{x})$. Thus we can obtain the magnitude and relative phase of the B₁ maps by only estimating $\tilde{B}_r(x)$ and $\tilde{\phi}_r(x)$. $M_r^{\pm}(x)$ is a set of nuisance parameters that needs to be jointly estimated, but they are fortunately linear terms that can be easily estimated.

 $E_r(\mathbf{x}) = \sum_{j=1}^n \alpha_{r,j} B_j(\mathbf{x}) e^{i\phi_j(\mathbf{x})}$ (1) $\int I_r^+(x) = \sin(\mu |E_r(x)|) e^{i \angle E_r(x)} m_r^+(x) e^{i \phi_b(x)} e^{i K_{BS}^+(x) |E_r(x)|^2}$ (2) $I_r^-(x) = \sin(\mu |E_r(x)|) e^{i \angle E_r(x)} m_r^-(x) e^{i \phi_b(x)} e^{i K_{BS}^-(x) |E_r(x)|^2}$ $\int I_r^+(\mathbf{x}) = M_r^+(\mathbf{x}) e^{i[K_{BS}^+(\mathbf{x})\tilde{B}_r(\mathbf{x})^2 + \tilde{\phi}_r(\mathbf{x})]}$ (3) $I_r^-(x) = M_r^-(x)e^{i[K_{BS}^-(x)\tilde{B}_r(x)^2 + \tilde{\phi}_r(x)]}$ $\Psi\left(\tilde{B}(x), \tilde{\phi}(x), M(x)\right) = \sum_{r=1}^{N} \sum_{\varsigma=+,-} \left\| I_r^{\varsigma}(x) - M_r^{\varsigma}(x) e^{i\left[K_{BS}^{\varsigma}(x)\tilde{B}_r(x)^2 + \tilde{\phi}_r(x)\right]} \right\|$ $\begin{aligned} &+\beta_{1} \sum_{r=1}^{R} \left\| C \tilde{B}_{r}(\mathbf{x}) \right\|^{2} + \beta_{2} \sum_{r=2}^{R} \left\| C e^{i \left[\tilde{\phi}_{r}(\mathbf{x}) - \tilde{\phi}_{1}(\mathbf{x}) \right]} \right\|^{2} \\ & \left[B_{1}(\mathbf{x}) e^{i \phi_{1}(\mathbf{x})} \right] & \vdots & \ddots & \vdots \\ & \vdots & \ddots & \vdots \\ & B_{R}(\mathbf{x}) e^{i \phi_{1}R(\mathbf{x})} \end{aligned} = \begin{bmatrix} \alpha_{1,1} & \cdots & \alpha_{1,R} \\ \vdots & \ddots & \vdots \\ \alpha_{R,1} & \cdots & \alpha_{R,R} \end{bmatrix}^{-1} \begin{bmatrix} \tilde{B}_{1}(\mathbf{x}) e^{i \tilde{\phi}_{1}(\mathbf{x})} \\ \vdots & \vdots \\ \tilde{B}_{n}(\mathbf{x}) e^{i \tilde{\phi}_{1}(\mathbf{x})} \end{aligned}$ (4) (5)

Regularization enforces prior knowledge to improve estimation. It is reasonable to assume that the magnitudes of the composite B_1 maps, $\tilde{B}_T(x)$, are spatially smooth. Although the absolute phase $\tilde{\phi}_T(x)$ is not necessarily smooth, the difference of it relative to a reference coil, e.g., $\tilde{\phi}_r(x) - \tilde{\phi}_1(x)$, should be smooth. Therefore, a finite differencing matrix C can be applied in regularization terms to penalize roughness. Since $\tilde{\phi}_r(x) - \tilde{\phi}_1(x)$ is likely to have phase wrap, we use the regularizer proposed in [4] that instead regularizes the roughness of $e^{i[\tilde{\phi}_r(x)-\tilde{\phi}_1(x)]}$. Our final cost function for estimating B1 is in (4), where $\tilde{B}(x) = [\tilde{B}_1(x), ..., \tilde{B}_R(x)]$, $\tilde{\phi}(x) = [\tilde{\phi}_1(x), ..., \tilde{\phi}_R(x)]$, $M(x) = [M_1(x), ..., M_R(x)]$, β_1 and β_2 are scalar regularization parameters. We estimate all the unknowns by minimizing $\Psi(\tilde{B}(x), \tilde{\phi}(x), M(x))$, during which $\tilde{B}(x), \tilde{\phi}(x)$ and M(x) are cyclically updated. We update M(x) by simply taking the real least square solution of (4) in each iteration. We use conjugate gradients with line search algorithm [5] to update $\tilde{B}(x)$ and $\tilde{\phi}(x)$, where backtracking line search [6] and monotonic line search [5] are used for $\tilde{B}(x)$ and $\tilde{\phi}(x)$ respectively. The standard approach [1] produces good initial guess for $\tilde{B}(x)$, and initial guess of $\tilde{\phi}(x)$ can then be solved analytically from equation (3) once $\tilde{B}(x)$ is initialized, without knowing M(x). Once $\tilde{B}(x)$ and $\tilde{\phi}(x)$ are estimated, the magnitude and relative phase of the original coils can be derived easily by (5), where $\phi'_r(x) = \phi_r(x) + \phi_b(x)$ which does not change the relative phase of the rth coil.

Methods and Results: The proposed method was tested by a phantom experiment on a 3T GE scanner equipped with an 8-coil custom parallel transmit/receive system [7]. Other than a routine B₀ mapping, we did a total of 16 scans with "all-but-one" coil combinations for the excitation and the BS pulse, where 8 ms Fermi BS pulses with ±4000 Hz off-resonance were followed by 2D spin-warp readout. Sequence parameters: 5 mm slice thickness, matrix 64*64, 26 cm FOV. Fig. 1 shows the results by the proposed method and the conventional magnitude is method (B_1) computed as in [1] and B₁ phase is computed by solving (3)).

Conclusions: The proposed method uses the same coil combinations for the excitation and BS pulses to jointly estimate B₁ phase and magnitude, which saves the set of R scans for phase estimation. This iterative

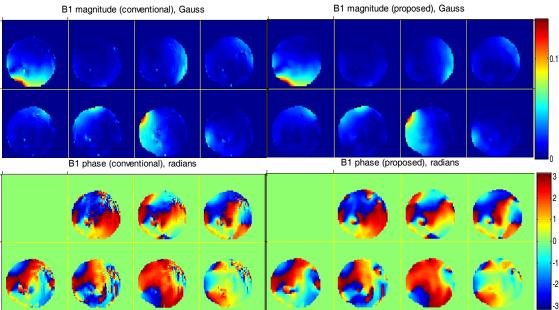


Fig. 1: B1 magnitude and relative phase of each individual coil

regularized estimation produces improved B1 magnitude and phase maps for low SNR regions. Future work will be to optimize the coil combinations, i. e., $\alpha_{r,j}$ in (1), to better reduce low excitation regions.

References: [1] Sacolick et al., MRM 63:1315-1322, 2010. [2] Sacolick et al., Proc. ISMRM 19:2926 (2011). [3] Funai et al., IEEE ISBI, 2007. [4] Zhao et al., Proc. ISMRM 19:2841 (2011). [5] Fessler et al., IEEE TIP, 8: 688-99, 1999. [6] K. Lange. Numerical analysis for statisticians, 1999. [7] System provided by group from Texas A&M University

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