

Fast Reconstruction for Regularized Quantitative Susceptibility Mapping

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TARGET AUDIENCE: Physicians and scientists interested in Quantitative Susceptibility Mapping (QSM) and phase imaging.
PURPOSE: QSM estimates the underlying magnetic susceptibility of tissues that give rise to changes in the magnetic field, and has applications in tissue iron quantification [1] and vessel oxygenation estimation [2]. ℓ_1 - and ℓ_2 -regularization have been proposed [3,4] to help solve the ill-conditioned dipole inversion in QSM. A high-resolution whole brain QSM reconstruction can take up to 20 min on a workstation, which poses a limit on QSM usability in clinical and research settings. Recently, the use of Split-Bregman (SB) formulation with ℓ_1 -regularization [5] and an efficient closed-form solution to the ℓ_2 -regularized problem [6] have been proposed to significantly decrease the computation cost of these problems. Herein, we introduce an improved SB ℓ_1 -regularized dipole inversion algorithm that offers 20× faster reconstruction relative to the standard nonlinear conjugate gradient (NCG) solver. This fast reconstruction renders estimation of regularization parameters with the L-curve heuristic feasible. Additionally, we extend SB ℓ_1 -regularization to admit magnitude-weighting that prevents smoothing across edges identified on the magnitude signal, and solve this more complicated problem 5× faster than the NCG approach. Further, we extend the previously proposed closed-form ℓ_2 -based inversion [6] to admit magnitude-weighting, and demonstrate 15× acceleration relative to NCG by employing a preconditioner that leads to faster convergence. Utility of the proposed methods is demonstrated in high-resolution (0.6 mm isotropic) 3D GRE data at 3T, as well as multi-echo Simultaneous Multi-Slice (SMS) EPI time-series at 7T, wherein processing of such large datasets would otherwise be prohibitive with conventional NCG.

METHODS: Tissue susceptibility χ relates to the measured field map ϕ via $\mathbf{DF}\chi = \mathbf{F}\phi$, where \mathbf{F} is Fourier transform operator and \mathbf{D} is the dipole kernel. **ℓ_2 -regularized QSM:** aims to solve $\min \|\mathbf{F}^{-1}\mathbf{D}\mathbf{F}\chi - \phi\|_2^2 + \beta \|\mathbf{W}\mathbf{G}\chi\|_2^2$ where \mathbf{G} is the gradient operator and \mathbf{W} is a binary mask derived from the gradient of the magnitude image. The optimizer is given by the solution of $(\mathbf{F}^{-1}\mathbf{D}^2\mathbf{F} + \beta\mathbf{G}^T\mathbf{W}^2\mathbf{G})\chi = \mathbf{F}^{-1}\mathbf{D}\mathbf{F}\phi$. Without magnitude weighting ($\mathbf{W} = \mathbf{I}$), it can be computed in closed-form [3] as $\chi = \mathbf{F}^{-1}(\mathbf{D}^2 + \beta\mathbf{E}^2)^{-1}\mathbf{D}\mathbf{F}\phi$, by expressing the gradient as $\mathbf{G} = \mathbf{F}^{-1}\mathbf{E}\mathbf{F}$ where \mathbf{E} is a diagonal matrix. Since the inversion involves only diagonal matrices, it requires only two FFTs. With magnitude weighting, the linear system is no longer diagonal. We propose to use the closed-form solution as preconditioner and iteratively solve the modified system $(\mathbf{D}^2 + \beta\mathbf{E}^2)^{-1} \cdot \{(\mathbf{D}^2 + \beta\mathbf{A}) \cdot \mathbf{F}\chi - \mathbf{D}\mathbf{F}\phi\} = \mathbf{0}$ where $\mathbf{A} = \mathbf{E}^H\mathbf{F}\mathbf{W}^2\mathbf{F}^{-1}\mathbf{E}$. As the weight matrix \mathbf{W} contains only the strongest edges, it is equal to identity \mathbf{I} except for ~5% of its entries [3]. This makes the approximation $(\mathbf{D}^2 + \beta\mathbf{E}^2)^{-1} \approx (\mathbf{D}^2 + \beta\mathbf{A})^{-1}$ valid, and renders the preconditioner useful. **ℓ_1 -regularized QSM:** we extend the SB formulation [7] to QSM by solving $\min 1/2 \|\mathbf{F}^{-1}\mathbf{D}\mathbf{F}\chi - \phi\|_2^2 + \lambda \|\mathbf{y}\|_1 + \mu/2 \|\mathbf{y} - \mathbf{W}\mathbf{G}\chi\|_2^2$. At iteration t , χ and \mathbf{y} are updated due to (i) $(\mathbf{D}^2 + \mu\mathbf{A})\mathbf{F}\chi_{t+1} = \mathbf{D}\mathbf{F}\phi + \mu\mathbf{E}^H\mathbf{F}\mathbf{W}^T\mathbf{y}_t$ and (ii) $\mathbf{y}_{t+1} = \max(|\mathbf{W}\mathbf{G}\chi_{t+1}| - \lambda/\mu, 0) \cdot \text{sign}(\mathbf{W}\mathbf{G}\chi_{t+1})$. Without magnitude weighting, (i) can be rapidly solved in closed-form, while (ii) is a simple soft-thresholding step. With the inclusion of \mathbf{W} , the preconditioner $(\mathbf{D}^2 + \mu\mathbf{E}^2)$ is employed for fast iterative solution via linear conjugate gradient. Using the susceptibility estimate from the previous iteration χ_t as initial guess further improves convergence. **Data Acquisition:** 3D GRE at 0.6 mm iso res was acquired on a volunteer at 3T (TR/TE=26/8.1ms, $R_{\text{inplane}}=2$, $T_{\text{acq}}=16\text{min}$), and a multi-echo SMS EPI dataset at 2 mm iso res was collected on a healthy volunteer at 7T (TR/TE₁/TE₂/TE₃/TE₄=2040/15/35/54/74ms, $R_{\text{inplane}} \times \text{MB}=3 \times 3$). Phase data were processed with Laplacian unwrapping [8] and Sharp filtering [9]. β , μ and λ values were chosen using L-curve.

RESULTS: 3D GRE: Reconstruction with closed-form ℓ_2 -regularization [6] took 0.9s, while the proposed ℓ_2 -based inversion with magnitude weighting was completed in 88s (Fig1, top). Proposed ℓ_1 -regularized QSM was finished in 60s and 275s without and with magnitude weighting (Fig1, bottom). Conventional NCG requires 1350s to reach the same convergence criterion of less than 1% change in the χ update (not shown). **SMS EPI:** ℓ_2 - and ℓ_1 -based dipole inversion with magnitude weighting are completed in 0.9s and 4s per frame in the time-series (Fig2 shows results for TE1 data).

DISCUSSION: The proposed dipole inversion algorithms dramatically reduce the processing time of ℓ_1 - and ℓ_2 -regularized QSM, while admitting prior information derived from the magnitude signal for edge-aware regularization. Such fast processing is made possible by efficient use of the closed-form ℓ_2 -based solver as preconditioner and the variable splitting method that decomposes ℓ_1 -penalty into a least-squares problem followed by a soft-thresholding step. While yielding up to 20× speed-up relative to conventional optimization methods, the proposed algorithms are further combined with fast phase unwrapping and background removal techniques to yield a rapid pipeline that might facilitate clinical application of QSM.

REFERENCES: [1] Langkammer C. et al. NIMG'12; [2] Fan A.P. et al. MRM'12; [3] Liu T. et al. MRM'10; [4] Liu J. et al. NIMG'12; [5] Chen Z. et al. Comp Assist Tomog'12; [6] Bilgic B. ISMRM'13; [7] Goldstein T. et al. SIAM'09; [8] Li W et al. NIMG'11; [9] Schweser F. et al. NIMG'11.

