

## Accelerated Water-fat Separation Using Parallel Imaging, Compressed Sensing, and Multiscale Cubic B-splines

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**INTRODUCTION:** Chemical shift-encoded water-fat separation techniques are used in both research and clinical settings because they are highly robust to off-resonance. A tradeoff when using these techniques is a longer scan time since data must be collected at multiple echo-times. Previous works have proposed to use either parallel imaging (PI) [1] or compressed sensing (CS) [2,3] to shorten the scan time. In this work, we introduce a joint parallel imaging and compressed sensing approach for water-fat separation. The proposed approach is compared to an existing PI/water-fat separation [4-6] method. We show 3.4x 1D acceleration with the proposed approach, while the existing method is limited to 1.9x 1D acceleration.

**THEORY:** We model the noisy undersampled k-space measurements ( $k$ ) as a function of the unknown coil sensitivities ( $C$ ), unknown coil-independent water and fat images ( $\rho$ ), unknown B0 field map ( $\Psi$ ), known k-space sampling ( $F_u$ ), and known chemical-shift modeling ( $A$ ) (Eq. 1). The coil maps are derived from the SPIRiT [7] k-space kernel [8,9]. The water and fat images are estimated by Eq. 2 where  $\lambda$  is the regularization parameter and  $W$  is the Daubechies-4 wavelet transform. The B0 field map is updated by Eq. 3 where  $\Delta\Psi$  is the field map update term,  $\Delta\rho$  is the water-fat error term,  $r$  is the residual,  $x$  is the linear function relating  $\Delta\Psi$ ,  $\Delta\rho$  to  $r$ , and  $B_m$  is the cubic B-splines (CBS) basis at the  $m^{\text{th}}$  scale (Fig. 1). The reconstruction procedure is summarized below:

- 1) derive coil sensitivities from 16 central k-space lines
- 2) initialize B0 field map to zero and choose the coarsest-scale B-splines ( $m = 1$ )
- 3) estimate the water and fat images (Eq. 2)
- 4) update field map estimate using B-splines at the  $m^{\text{th}}$  scale (Eq. 3)
- 5) IF  $\max |\Delta\Psi| < 1\text{Hz}$  THEN go to step 6 ELSE go to step 3
- 6) IF  $m = m_{\text{max}}$  THEN done ELSE  $m = m + 1$ , go to step 3  
( $m_{\text{max}}$  defines finest-scale CBS basis)

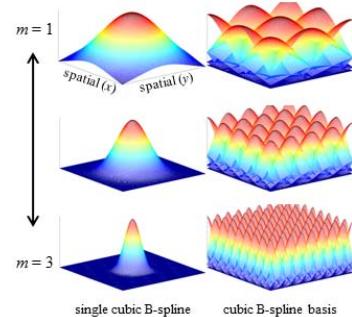
**METHODS:** Fully-sampled axial liver data were collected at three echo-times ( $TE_1 = 2.184\text{ms}$ ,  $\Delta TE = 0.794\text{ms}$ ) on a GE Signa EXCITE HDx 3T system (GE Healthcare, Waukesha, WI) using an investigational 3D IDEAL SPGR sequence and eight-channel torso array. The acquisition matrix size was  $256 \times 256$  with 8 slices. The fully-sampled data were undersampled by net factor of 3.4x using a jittered grid pattern [10] and reconstructed using the proposed PI-CS-CBS method. For reference, the fully-sampled data were reconstructed using IDEAL with region-growing (IDEAL-RG) [5,6]. For comparison, the data were uniformly undersampled by net factor of 1.9x ( $= 2x + 16$  central k-space lines) and 3.4x ( $= 4x + 16$  central k-space lines) and were reconstructed using ARC [4] and then IDEAL-RG (ARC/IDEAL-RG).

**RESULTS:** Fig. 2 shows the estimates from the reference 1x IDEAL-RG, 3.4x ARC/IDEAL-RG, and 3.4x PI-CS-CBS. The 1.9x ARC/IDEAL-RG estimates were similar to the reference (not shown). The 3.4x proposed PI-CS-CBS method yields similar estimates as the reference whereas the 3.4x ARC/IDEAL-RG estimates exhibit unresolved aliasing and noise amplification.

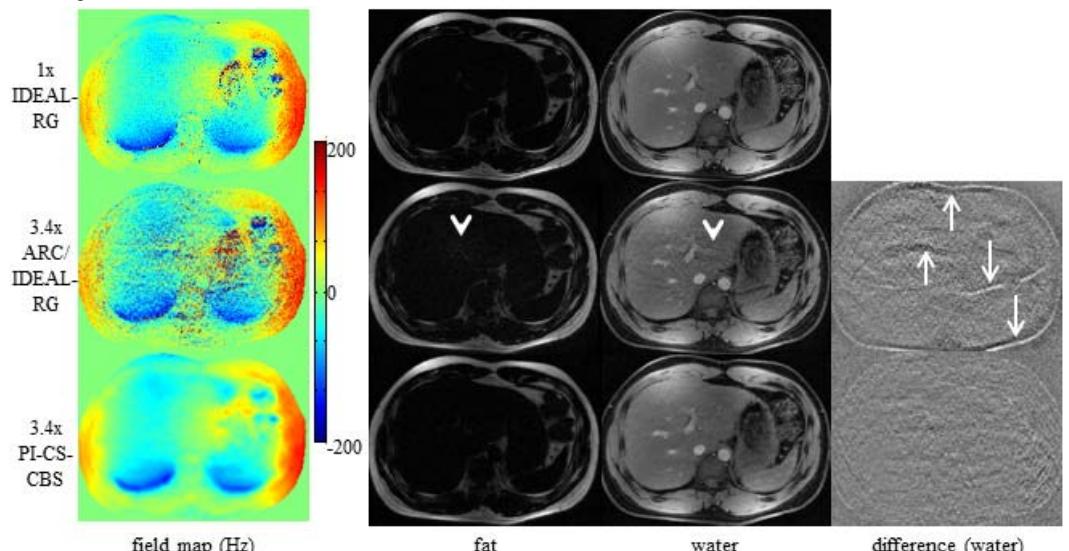
**CONCLUSION:** We have demonstrated uniform water-fat separation using a joint PI-CS approach. The proposed approach achieves 3.4x 1D undersampling, which outperforms ARC/IDEAL-RG and more than makes up for the additional scan time of a three-point acquisition.

**REFERENCES:** [1] Ma et al. ISMRM 2003 p.1069 [2] Doneva et al. MRM 2010;64:1749-59 [3] Sharma et al. MRM (early view) [4] Beatty et al. ISMRM 2007 p. 1749 [5] Reeder et al. MRM 2005;54:636-44. [6] Yu et al. MRM 2005;54:1032-39 [7] Lustig et al. MRM 2010;64:457-71 [8] Lustig et al. ISMRM 2011 p. 479 [9] Lai et al. ISMRM 2011 p.65 [10] Cook. ACM Trans Graph 1986;5:51-72

$$\begin{aligned}
 (1) \quad & k = F_u C \Psi A \rho + N(0, \sigma^2) \\
 (2) \quad & \min_{\rho} \|k - F_u C \Psi A \rho\|_2 + \lambda \|W \rho\|_1 \\
 (3) \quad & \min_{\Delta\Psi, \Delta\rho} \|r - x(\Delta\Psi, \Delta\rho)\|_2 + \lambda \|W \Delta\rho\|_1 \\
 & \text{s.t. } \Delta\Psi \in \text{span}\{B_m\}
 \end{aligned}$$



**Fig. 1:** cubic B-spline (left) and corresponding basis (right); field map update at the  $m^{\text{th}}$  scale is restricted to be in the space that is spanned by the  $m^{\text{th}}$  cubic B-spline basis



**Fig. 2:** water, fat, and B0 field map estimates using the three reconstruction methods; the difference images are shown on the same window/level, using 1x IDEAL-RG as the reference; the ARC/IDEAL-RG estimates exhibit unresolved aliasing (arrows) and noise amplification (arrowheads) that do not appear in the proposed PI-CS-CBS reconstruction