Efficient Variable Density SENSE Reconstruction

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Introduction

Variable density k-space sampling is sometimes used in parallel imaging to reduce aliased energy at low spatial frequencies (1) or to allow self-calibration using Nyquist-sampled data around the center of k-space (2). Variable density data can be reconstructed using SPACE RIP (1), Generalized SMASH (3), Generalized SENSE (4), GRAPPA (5) or by unwrapping only a uniformly undersampled subset, transforming back to k-space, and combining it with the Nyquist-sampled data (6). A method developed by Madore (7) for SENSE with self-calibration places the measured data at the correct sampling intervals in k-space and then fills in missing data with zeros. The data are multiplied by the appropriate density compensation factors for each k-space region. Fourier transformation is then followed by Cartesian SENSE unwrapping (8) assuming a reduction factor appropriate for the undersampled (outer) part of k-space. Madore's method has the advantage of being very computationally efficient but can result in ringing artifacts due to the abrupt transition in k-space density. **Methods**

The image quality of Madore's method can be improved with almost no reconstruction time penalty by a very simple extension. Consider a 1D example where the outer part of k-space has threefold undersampling and the center of k-space is Nyquist-sampled (Fig. 1, top). By appropriately choosing the sampling pattern, we can split the data into two interleaving sets, a high resolution set RH and a low resolution set RL (Fig. 1, center and bottom). Because of the uniform spacing of both the samples in RH and the "holes" in RL, both data sets can be reconstructed using uniform-density Cartesian SENSE. If RH and RL are multiplied by appropriate k-space windows H and L that compensate for the sampling density, the two unwrapped images can be added. The final reconstructed image I is

$$I = S\{FT[H \times RH + L \times RL]\}$$
[1]

where FT[] and S{} represent 1D Fourier transform and 1D SENSE unwrapping operators, respectively. For example consider the k-space window

$$L = \begin{cases} 1 & \text{in the Nyquist sampled region} \\ 0 & \text{otherwise} \end{cases}$$
[2]

with

 $H = 3 - 2 \times L$ For this choice of *L* the reconstruction is equivalent to Madore's method. However we can smooth the abrupt transitions in *H* and *L* to reduce the ringing. Consider

$$L = \left[1 + e^{(ky - c) / w}\right]^{-1} - \left[1 + e^{(ky + c) / w}\right]^{-1}$$
[4]

with *H* given by Eq. [3] and where *c* and *w* represent the radius of the Nyquist-sampled region and the transition width respectively. *L* and *H* are shown in Fig. 2 for c = 32, w = 8. The transition width *w* can be chosen to trade ringing for aliasing: increasing *w* decreases ringing but increases low frequency aliasing. Note that if a sharp transition is used (w = 0), then Eq. [4] becomes equivalent to Eq. [2]. It can be shown that this method requires the approximation that the SENSE unwrapping operator commutes with the convolution operator in image space. This approximation is also used in uniform-density SENSE reconstructions when k-space windowing is used to reduce Gibbs ringing or when low spatial resolution sensitivity data are used (9).

Results

Scan data were acquired on a commercial 1.5T scanner (GE Healthcare, Milwaukee, WI) using an eight-channel head coil (MRI Devices, Gainseville, FL). Fully sampled data (256x256 matrix) were decimated to give three-fold undersampling in the outer part of k-space with 64 Nyquist-sampled center lines (128 acquired ky lines, net reduction factor two). The Nyquist-sampled k-space lines were used to generate a low resolution coil sensitivity calibration. The data were then reconstructed using Eqs. [1], [3] and [4]. Figure 3 shows the results for two transition widths (w = 0, w = 8), demonstrating the reduction in ringing as the transition width increases. This method can be extended to reconstruct partial Fourier data by appropriately modifying H to overweight data conjugate to the additional missing ky lines. **Conclusions**

Variable density Cartesian SENSE data can be efficiently reconstructed by extending the method of Madore to reduce ringing artifacts introduced by the sharp kspace density changes. k-Space sampling is modified to allow the data to be segregated into interleaving low and high frequency subsets. Each subset is multiplied by a density compensation window that also smoothes the sampling density transitions. The data are added and reconstructed using the conventional uniform-density SENSE algorithm. The reconstruction time is almost the same as for uniformly-sampled data.

References

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х	0	0	х	0	0	х	0	0	Х	х	х	х	х	х	х	х	Х	х	0	0	Х	0	0	Х	0	0	Х	
x	0	0	X	0	0	X	0	0	Х	0	0	X	0	0	Х	0	0	х	0	0	Х	0	0	Х	0	0	х	RH
0	0	0	0	0	0	0	0	0	0	х	х	0	х	X	0	х	х	0	0	0	0	0	0	0	0	0	0	RL





Fig. 2. *H* (left) and *L* (right) computed using Eqs. [3] and [4] for 256 ky samples with c = 32, w = 8.



Fig. 3. Variable density scan with threefold undersampling and 64 center lines (c = 32) reconstructed with w = 0 (left), and w = 8 (right).