Improving Parallel Imaging by Jointly Reconstructing Multi-Contrast Data

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Synopsis

We propose a general joint reconstruction framework to accelerate multi-contrast acquisitions further than currently possible with conventional parallel imaging. Our joint parallel imaging techniques simultaneously exploit similarities between echoes/phase-cycles/contrasts, virtual coil concept, partial Fourier acquisition, complementary sampling across images along with limited support and smooth phase constraints. These permit highly accelerated 2D, Simultaneous MultiSlice and 3D acquisitions as well as improved calibrationless parallel imaging from multiple contrasts. Our algorithms, JVC-GRAPPA and J-LORAKS, provide over 2-fold improvement in reconstruction error compared to conventional GRAPPA, with improved mitigation of artifacts and noise amplification.

Introduction

MRI exams routinely involve multi-contrast acquisition to obtain complementary information. Multi-echo acquisition finds applications in relaxation time mapping \cite{1-3} and water/fat imaging \cite{4-7}. Phase-cycled bSSFP requires multiple images with different phase offsets to shift the location of B0-related artifacts. These images are then combined to eliminate banding artifacts \cite{8-10}. Unfortunately, high-resolution multi-contrast/echo/cycle imaging with whole-brain coverage leads to excessive scan times.

For faster acquisitions, previous joint reconstruction techniques explored k-space interpolation across time \cite{11-12}, echoes \cite{13-14}, or excitation modes \cite{15}. Herein, we propose a general framework to accelerate multi-contrast acquisitions further than currently possible. We introduce Joint Virtual Coil (JVC-) GRAPPA, which allows all coils from all contrasts to contribute to reconstruction of a particular channel. This employs VC concept \cite{16-18} and complementary k-space sampling across contrasts.

We extend joint parallel imaging to exploit limited support and smooth phase constraints \cite{19-21} through Joint (J-) LORAKS. J-LORAKS achieves a more parsimonious low rank representation in k-space by considering multi-contrast images as additional coils. It seamlessly incorporates partial Fourier (PF) and permits improved calibrationless reconstruction.

Matlab code/data: \url{http://bit.ly/2sY1FJT}

Reconstruction Algorithms

All experiments used 16 SVD compressed channels \cite{22-23} and were performed at 3T. PF experiments employed POCS processing \cite{24-26} following GRAPPA reconstruction \cite{27}.

\textbf{JVC-GRAPPA}: creates additional channels by treating data from other contrasts as extra coils. To increase size of available k-space for calibration of large number of kernels, an initial Joint-GRAPPA was performed from ACS data. VC concept was then employed to double the number of channels. This way, the entire k-space of the interim reconstruction became available for calibration. We run 3 iterations to refine the reconstruction.

\textbf{J-LORAKS}: also stacks data from all contrasts in the channel axis, and employs VC. It enforces local k-space neighborhoods, now extended across contrasts, to have low rank. The parameters associated with this constraint, neighborhood size and rank of local k-space matrices, were optimized to reduce RMSE.

Data Acquisition and Comparison Cases

\textbf{2D Phase-Cycled bSSFP}: was acquired with four phase-cycles, FOV=380×380mm\textsuperscript{2}, matrix=160×160, TR/TE=3.3/1.54ms, using 34-channel reception and 20 ACS lines.

GRAPPA and JVC-GRAPPA were performed at R=7x1 undersampling. Keeping the net acceleration the same, R=6x1-fold undersampling with 7/8 PF was explored using J-GRAPPA & POCS and J-LORAKS.
**3D ME-MPRAGE:** was collected at 1mm³, FOV=256×240×192mm³, TR/TI=2530/1100ms, TE’s=1.7/3.6/4.7/5.3ms, 32-channel reception and 24x24 ACS region.

GRAPPA and JVC-GRAPPA were performed with R=4x4 undersampling. At the same acceleration, R=4x3 with 6/8 PF was explored using J-GRAPPA & POCS, as well as J-LORAKS.

**SMS ME-TSE:** was acquired using FOV=240×240, matrix=256×256, slice gap=12.8mm, TR=4sec, TE’s=12/25/50/62/87/99ms, 32-channel reception and 24 ACS lines.

Slice-GRAPPA & POCS was employed at MultiBand-10 acceleration with 6/8 PF. Joint reconstructions were performed with Joint Slice-GRAPPA & POCS and J-SMS-LORAKS.

**Calibrationless 3D ME-GRE:** was collected with FOV=240×240×192, matrix=160×160x128, TR=23 ms, TE’s=3/7/11/15/19ms using 32-channel reception. R=4-fold calibrationless Poisson sampling was utilized.

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## Results

**bSSFP [Fig1]:** GRAPPA broke down with 19.0% RMSE, while JVC-GRAPPA demonstrated better artifact mitigation with 10.7% error. POCS completed k-space appeared underestimated in J-GRAPPA with PF. J-LORAKS mitigated this problem with the best RMSE (8.0%).

**ME-MPRAGE [Fig2]:** GRAPPA demonstrated severe artifacts with 14.8% RMSE. JVC-GRAPPA partially mitigated these with 7.8% error. Although J-GRAPPA & POCS further reduced artifacts, partially sampled k-space suffered from underestimation. J-LORAKS addressed these to provide a cleaner reconstruction (7.9% RMSE).

**ME-TSE [Fig3]:** Slice-GRAPPA followed by POCS returned 6.0% error. Despite the reduction in RMSE to 4.9%, some artifacts were present in Joint Slice-GRAPPA. J-SMS-LORAKS attained the best RMSE (3.7%), with some residual artifacts.

**Calibrationless ME-GRE [Fig4]:** J-LORAKS mitigated the high error in early echoes of conventional LORAKS, and alleviated the underestimation in the R2* map.

[Fig5]: Compares individual cycles/contrasts.

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## Discussion and Conclusion

Proposed joint reconstruction algorithms provided up to 2.4-fold reduction in RMSE over conventional parallel imaging. This was made possible by exploiting similarities between multi-contrast images, complementary sampling and VC concept. JVC-GRAPPA converted intensity and phase differences into extra spatial encoding, while J-LORAKS achieved lower rank k-space matrices due to added redundancy from multiple images. J-LORAKS also allowed improved calibrationless reconstruction and flexible PF sampling.

A drawback of joint reconstruction is the increased computation time. For Fig2, GRAPPA took 6sec, while JVC-GRAPPA and J-LORAKS required 6.8min and 5.4min. This can be mitigated by more coil compression achieved through e.g. GCC [28], restricting calibration to a smaller portion of k-space in JVC-GRAPPA, and reducing J-LORAKS iterations (currently 100).

Another limitation is motion between scans. While this will not affect multi-echo acquisitions, phase-cycled bSSFP could be impacted by potential mismatches. JVC-GRAPPA employs low-resolution kernels, and maybe resilient against small amounts of motion. For larger mismatches, an initial GRAPPA could enable motion correction, followed by JVC-GRAPPA or J-LORAKS reconstruction.

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Figures
**Fig 1.** Phase-cycled bSSFP with 7-fold total acceleration. Conventional GRAPPA broke down at such acceleration factor, and had 19.0% error with severe artifacts and noise amplification. JVC-GRAPPA mitigated some of these artifacts, but still yielded a large error of 10.7%. Combination of 6-fold uniform and 7/8 partial Fourier sampling provided the same 7-fold net acceleration. Here, virtual coil concept was not applicable in Joint-GRAPPA. Its combination with POCS led to 9.2% RMSE and signal underestimation in k-space (white arrow). J-LORAKS outperformed all methods with 8.0% error, and was more successful in completing the partially sampled k-space without the need for POCS.

**Fig 2.** Multi-echo MPRAGE reconstruction at 16-fold total acceleration. Conventional GRAPPA broke down at this high acceleration factor, yielding 14.8% error. JVC-GRAPPA managed to mitigate most of the structured artifacts and noise amplification, with 7.8% RMSE and some residual aliasing artifacts (yellow arrow). At the same net acceleration factor, the combination of R=4x3 uniform undersampling and 6/8 partial Fourier sampling was explored. Combination of J-GRAPPA and POCS had 8.5% error, underestimation in partially sampled k-space (white arrow), and structured artifact (yellow arrow). J-LORAKS was able to provide an improved reconstruction with 7.9% RMSE and more successfully completed partial k-space data.

**Fig 3.** Multi-echo Turbo Spin Echo with MultiBand=10 acceleration and 6/8 partial Fourier (displaying 5/10 slices). Combination of Slice GRAPPA and POCS had 6.0% RMSE with visible aliasing artifacts and k-space discontinuity at the partial Fourier transition. Joint Slice-GRAPPA was not able to utilize virtual coil concept, and required POCS post-processing to estimate partially sampled data. This led to 4.9% error with some artifacts and minor k-space discontinuity. Joint SMS-LORAKS did employ virtual coils, and incorporated partial Fourier without the need for POCS. This allowed 3.7% RMSE performance, while not fully mitigating aliasing artifacts at such high acceleration (yellow arrows).
**Fig.4.** Calibrationless multi-echo gradient-echo reconstruction at R=4-fold pseudo-random acceleration. Single-contrast calibrationless LORAKS yielded 4.2% error with underestimation in the $R^2*$ parameter map (yellow arrow). This is likely caused by the signal drop in the early echoes as can be better seen in the error maps. Joint calibrationless LORAKS had an improved RMSE performance of 3.1%, and mitigated the signal dropout problem in both the individual echoes and the estimated parameter map.

**Fig.5 (a)** Individual phase-cycles from bSSFP at 7-fold total acceleration. J-LORAKS employed staggered undersampling across cycles with partial Fourier, and dramatically reduced aliasing artifacts and noise amplification over GRAPPA. **(b)** Individual echoes in ME-MPRAGE at 16-fold total undersampling. GRAPPA suffered from severe aliasing artifacts and noise amplification. J-LORAKS mitigated these and utilized different 2D-CAIPIRINHA patterns, staggered acquisition and different partial Fourier sampling in each echo. **(c)** Individual echoes from one of the ten slices in SMS ME-TSE. Combination of Slice-GRAPPA with POCS suffered from noise amplification and artifacts. Joint SMS-LORAKS mitigated both issues, but minor aliasing artifacts were visible (arrows).

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