

fMRI Using GRAPPA EPI with High Spatial Resolution Improves BOLD Signal Detection at 3T

Dionyssios Mintzopoulos^{1,2}, Loukas G. Astrakas^{1,2}, Masahiko Hiroki¹, Azadeh Khanicheh³, Lawrence L. Wald¹, Bruce R. Rosen¹ and A. Aria Tzika^{*,1,2}

¹Athinoula A. Martinos Center for Biomedical Imaging, Department of Radiology, Massachusetts General Hospital, Harvard Medical School, Charlestown, MA 02129, USA

²NMR Surgical Laboratory, Department of Surgery, Massachusetts General Hospital and Shriners Burn Institute, Harvard Medical School, Boston, MA 02114, USA

³Department of Mechanical and Industrial Engineering, Northeastern University, Boston, MA 02115, USA

Abstract: We compared GRAPPA parallel MRI (pMRI) to regular MRI (non-GRAPPA) for BOLD fMRI while keeping all other parameters fixed. We acquired both GRAPPA and non-GRAPPA images using a high resolution as well as a low resolution EPI matrix. We found significantly larger values of percent BOLD signal when comparing higher resolution acquisitions to lower resolution ones, independently of whether data were acquired with GRAPPA EPI or with regular EPI. We also demonstrated no loss of functional activation or BOLD signal when comparing GRAPPA to non-GRAPPA while keeping the same spatial resolution. We propose to use pMRI to gain the ability to perform whole-brain acquisition at higher spatial resolution with TE in the 30-40 ms range for optimal BOLD detection at 3T, or at faster scan times. To this end, we conclude that GRAPPA pMRI is advantageous for BOLD fMRI and whole-brain EPI acquisition at high spatial resolution.

Keywords: fMRI, EPI, BOLD, GRAPPA, human brain mapping, 3T.

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INTRODUCTION

The underlying principle of functional magnetic resonance imaging (fMRI) is that neuronal activation is accompanied by changes in regional blood flow, volume, and oxygenation [1-7]. An advantage of fMRI is that brain anatomy and activation can be measured simultaneously. The most widely applied fMRI technique for mapping brain activity is Blood Oxygenation Level Dependent (BOLD) imaging, which uses endogenous deoxyhemoglobin as a contrast source [8]. Cortical activation causes alterations in blood oxygenation, which create changes in microscopic susceptibility measured using T2*-weighted sequences [9-11].

Accelerated MRI using parallel magnetic resonance imaging (pMRI) techniques has recently been used in research and clinical procedures in order to improve MRI resolution [12, 13]. In the context of high magnetic fields, pMRI addresses high-field specific problems such as the need to reduce susceptibility artifacts [14]. Parallel imaging techniques are particularly advantageous in high and ultra high field fMRI, allowing high resolution, high accuracy, and high contrast-to-noise imaging [15]. Advances in both parallel imaging acquisition and reconstruction hold promise for improving the performance of parallel imaging sequences,

for reducing artifacts and for increasing spatiotemporal resolution using high acceleration factors [16-18]. The most commonly used parallel imaging techniques relate to *k*-space domain methods (for example, GeneRalized Autocalibrating Partially Parallel Acquisitions, or GRAPPA [13]) or to real-space domain methods (for example, Sensitivity Encoding, or SENSE [12]). SENSE can be more sensitive than GRAPPA to numerical errors or errors in sensitivity profiles while GRAPPA has been found to perform more robustly [19-21] and to have better noise distributional properties [22-24]. Overall, the *k*-space-based reconstruction method of GRAPPA is well suited for fMRI [25]. Indeed, a recent study reported clear superior performance of GRAPPA compared to modified SENSE (mSENSE) in fMRI [26].

pMRI techniques are relatively new and have not been extensively applied in fMRI. pMRI can be used to increase spatial resolution or temporal resolution (i.e. the single-volume acquisition time, TR). However, because of under-sampling, pMRI images have lower signal-to-noise ratio (SNR) and there is concern that detection of the weak BOLD signal may suffer as well. Studies that have employed GRAPPA pMRI for fMRI have reported encouraging results [26-29]. Recently, Preibisch *et al.* found no significant difference in timeseries Signal-to-Noise Ratio (SNR) when using GRAPPA [26]. Little *et al.* [28] found that BOLD activations in the occipital cortex were not reduced and for several subjects the detected volume was larger using GRAPPA than without. Lütcke *et al.* [27] did not detect significant signal difference between non-GRAPPA EPI and

*Address correspondence to this author at the Director of NMR Surgical Laboratory, Department of Surgery, Massachusetts General Hospital and Shriners Burn Institute, Harvard Medical School, 51 Blossom Street, Room 261, Boston, MA 02114, USA; Tel: (617) 371-4871; Fax: (617) 720-3544; E-mail: atzika@hms.harvard.edu

GRAPPA EPI when the spatial resolution and TE were kept the same in both imaging protocols, but interpreted their results as implying that pMRI is not advantageous for fMRI. Further, Preibisch *et al.* found that application of SENSE pMRI to fMRI resulted in a substantial increase in speed and/or spatial resolution and in significantly reduced distortions and blurring, while, at the same time, statistical power and the temporal signal to noise ratio (tSNR) of the fMRI timeseries were not affected [29].

In this work, we report on the use of parallel MR methods (GRAPPA) for fMRI of a motor (hand-squeezing) task that has been useful in movement therapy in hemiplegia [30] and in clinical brain mapping of human motor deficits [31]. Using that fMRI protocol and a detailed noise modeling, we investigated the effect of GRAPPA pMRI and higher spatial resolution on the detection of BOLD signal.

MATERIALS AND METHODS

Study Design

BOLD fMRI was performed while subjects executed a block-design motor task (hand-squeezing). Each block consisted of 30 s of squeezing and 30 s of rest. Each task presentation was comprised of three sequentially presented blocks, resulting in total task duration of 180 s. The arrangement of the acquisitions described below was randomized in order to avoid effects of habituation. Since we employ a block-design paradigm, a TR of 3 s is adequate; therefore, we chose to focus on optimizing the spatial resolution rather than the temporal resolution (TR) of our experiment in order to exploit fMRI at higher spatial resolution.

Two experiments were performed. In the first experiment, subjects performed the task while four types of images were acquired: GRAPPA EPI images at high spatial resolution, non-GRAPPA EPI (regular EPI) images at high spatial resolution, GRAPPA EPI images at low spatial resolution, and non-GRAPPA EPI images at high spatial resolution. These sets can be arranged, for example, in a 2×2 matrix with high and low spatial resolution along the two rows and GRAPPA / non-GRAPPA along the two columns. The single-volume acquisition time (TR), echo time (TE), field of view, and number of slices were the same for all images. In particular, TE was long (TE = 42 ms) in order to keep the other parameters the same. The in-plane acquisition matrix was the same in the two high spatial resolution images and the same in the two low resolution images, but differed between high and low resolution. Subjects performed the task at the same percentage (60%) of their maximum effort level.

In the second experiment, two types of images were acquired: high spatial resolution GRAPPA and low spatial resolution non-GRAPPA. In contrast to the first experiment, the acquisitions in the second experiment are not directly comparable to each other because they have different parameters. Keeping the same parameters for comparison purposes (as was done in the first experiment) prevents the full exploitation of the GRAPPA advantages. Therefore we conducted this second more realistic experiment allowing full optimization of speed and resolution both in the parallel and non-parallel fMRI. This allowed us to test the performance of the two approaches in pulse sequence configurations closer to real clinical applications. To this end, we set TE =

30 ms, and subjects performed the task at three different percentages (15%, 45%, and 60%) of their maximum effort level.

Twelve healthy subjects participated in the experiment. Six subjects completed the first experiment, and six subjects completed the second experiment.

All studies were approved by the Institutional Review Board at Massachusetts General Hospital and were performed on a 3T scanner (Siemens TIM Trio) at the Athinoula A. Martinos Center for Biomedical Imaging.

Methods for Hand-Squeezing

The subjects squeezed the handles of a novel MR-compatible robotic device (MR_CHIROD) [32, 33]. Their maximum squeezing strength was found by adjusting the resistive force necessary to close the handles of the MR_CHIROD until the subjects could barely do so.

The block design paradigm consisted of three alternating action and resting epochs of 30 s each, lasting a total of 180 s. This design aimed to minimize subject fatigue while still allowing good signal detection. The subjects' squeezing rate was guided by a visually projected metronome stimulus, which projected a cue circle oscillating radially at a frequency of 1 Hz on a neutral-background screen. A fixation cross was projected during rest. The stimulus was implemented using the Psychophysics Toolbox in Matlab.

Specifically, the task presentation lasted 3 min 27 s. The first 27s were used for the GRAPPA reference calibration scans (acquired once initially) and to insure the magnetization had equilibrated. The subjects rested during the first 27 s. The magnetization equilibration scans were subsequently discarded, and only the 60 volumes corresponding to the 180 s were used for analysis.

The subjects' arms were kept extended at their sides and extra padding was used to minimize elbow flexion and further reflexive motion, and to minimize translational and rotational head motion. Typical translational head motion ranged from 0.1-0.4 mm during scans, as reported from the motion correction function in Statistical Parametric Mapping (SPM). All subjects were able to complete the task without difficulty.

Performing the experiment at scaled (percent) levels of each subject's own maximum force of squeezing compensates for performance confounds by constraining between-subjects performance to make it approximately the same [34]. Subject training typically required approximately 10-15 min, and took place before scanning.

The sequence of acquisitions was randomized in order to avoid effects of habituation.

Methods for Imaging

First Experiment

Four imaging protocols were used: high spatial resolution GRAPPA and high spatial resolution non-GRAPPA (regular EPI), and low spatial resolution GRAPPA and low spatial resolution non-GRAPPA (regular EPI). For high spatial resolution whole-brain EPI acquisition for BOLD fMRI, voxel size was 2mm × 2mm × 3mm, (in-plane matrix 96 × 96)

TR/TE = 3000/42 ms, 32 axial slices, 40% distance factor. Flip Angle (FA) = 90° at TR = 3000 ms. The TR/TE = 3000/42 ms were necessary to accommodate both GRAPPA and non-GRAPPA protocols with all other parameters being equal. The high resolution GRAPPA protocol (termed HRG3) had GRAPPA acceleration factor $R = 3$, 92 GRAPPA autocalibration lines (the maximum allowed number), bandwidth per pixel 754 Hz/px (1.43 ms echo spacing). The high resolution non-GRAPPA protocol (HRnG) had bandwidth per pixel 1443 Hz/px (0.76 ms echo spacing). The low spatial resolution protocols had 33 slices at 20% distance factor (the skip factor must be adjusted between high and low resolution protocols to account for the different slice thickness) and $3.1 \text{ mm} \times 3.1 \text{ mm} \times 5 \text{ mm}$ voxel size, which is typical for an fMRI experiment. The GRAPPA protocol (LRG3) had 61 GRAPPA autocalibration lines (the maximum number), 752 Hz/px bandwidth. The non-GRAPPA protocol (LRnG) had 988 Hz/px bandwidth. GRAPPA EPI reconstruction was done with Sum of Squares (SoS) as implemented by the vendor.

Second Experiment

In this experiment, GRAPPA was used to reach the highest possible spatial resolution with whole brain acquisition at TE of 30 ms and TR of 3000 ms. The GRAPPA and non-GRAPPA sequences were not directly comparable. The rationale was to obtain the highest spatiotemporal resolution possible while imaging the whole brain for fMRI, in order to minimize partial-volume effects that reduce the spatial sensitivity of BOLD fMRI [35-39]. GRAPPA was chosen to implement pMRI in order to reach these goals [19-24]. The number of GRAPPA autocalibration lines (reference lines for GRAPPA calibration) was the maximum possible, in order to ensure the best-quality GRAPPA reconstruction. Single-volume acquisition time TR was the minimum possible at the highest spatial resolution; bandwidth per pixel was minimized to improve signal-to-noise ratio (SNR); and echo time TE was chosen to be 30 ms, following theoretical and empirical arguments suggesting an optimal TE of 30 ms (with a range for TE from 30 ms to 40 ms) for fMRI at 3T [40, 41]. Typical acquisition parameters for the high spatial resolution GRAPPA protocol were as follows: TR/TE = 3000 ms/30 ms; GRAPPA acceleration factor $R = 3$; voxel size of $1.6 \text{ mm} \times 1.6 \text{ mm} \times 3.0 \text{ mm}$, 128×128 in plane matrix, $200 \text{ mm} \times 200 \text{ mm}$ field of view (FOV); 48 slices (5% skip) covering the entire brain with a tilted axial orientation; 85 GRAPPA autocalibration lines; 1.5 kHz bandwidth per pixel. In comparison, the same spatial resolution without GRAPPA requires sparse fMRI time resolution (TR = 6 s) and very long TE (approximately 60 ms) and results in concomitant loss of contrast. GRAPPA EPI reconstruction was done with Sum of Squares (SoS) as implemented by the vendor.

The non-GRAPPA sequence had voxel size of $3.1 \text{ mm} \times 3.1 \text{ mm} \times 5.0 \text{ mm}$, and bandwidth per pixel 3.32 kHz. TR/TE = 1500/30 ms and a flip angle of 75° was chosen to maximize gray matter signal at TR = 1500 ms, assuming that T1 of gray matter at 3 T ranges approximately from 1000 ms to 1300 ms [42].

Anatomical Reference Scans

A high resolution 3-dimensional, T1-weighted, MP-RAGE (magnetization-prepared rapid gradient echo) image was acquired for anatomical reference and optimal gray-white matter contrast (Sagittal orientation, flip angle = 7° , TE = 4.73 ms, TR = 2530 ms. TI = 1100 ms, voxel resolution 1mm isotropic, acquisition matrix = $352 \times 352 \times 192$).

Methods for Signal to Noise Ratio (SNR) Measurement

Both the time-independent spatial SNR (sSNR) and the temporal or timeseries SNR were estimated from unprocessed ("raw") EPI images.

Image, or spatial, SNR (sSNR) was calculated by dividing image intensity values at each voxel, by the (scalar) image standard deviation $SD(x)$ [43]. $SD(x)$ was calculated by fitting the intensity histogram from a noise region of interest (ROI). The noise ROI was drawn near the image edge to contain voxels of the image background. The intensities of the background voxels are assumed to be composites of fluctuations drawn from zero-mean Gaussians. Noise estimated in this manner does not have physiological contributions and is assumed to be additive and solely thermal in origin. GRAPPA sequences had the same TE as their non-GRAPPA counterparts; therefore SNR is proportional to square-root voxel volume and inverse square-root bandwidth per pixel [44] with an additional penalty $R^{1/2}$ for GRAPPA. The sSNR metric relies on low-intensity pixels to estimate image noise, but pMRI image unfolding and reconstruction steps suppress the signal outside the brain and introduce spatial correlations that locally alter the noise distribution and can leave residual artifacts near the center of the brain, where coil sensitivities are lower [45, 46]. Such spatial correlations are not taken into account by this calculation of sSNR. For this reason, in order to avoid the possible pitfalls of sSNR estimation, we proceeded to work with the temporal or timeseries SNR (tSNR).

tSNR includes a component of physiological noise due to breathing, pulsation, and presence of draining vessels [47, 48]. Generally, tSNR will bear the signature of any residual correlations that alter the timecourse of a voxel. tSNR at each voxel is defined from a series of images as the mean (over time) intensity divided by the standard deviation,

$$tSNR = \langle S(t) \rangle_t / SD(t).$$

The error model for tSNR hinges on the assumption that voxel intensity is a constant value plus additive random noise. During tSNR calculation, a mask was created from the mean intensity image in order to exclude the voxels outside the brain. The tSNR empirical distributions (histograms) were calculated, normalized to unit area, and fitted by a Gaussian profile. The mean value and standard deviation of the tSNR were estimated from the two Gaussian fit parameters, location $\hat{\mu}$ and full-width-half-maximum $\hat{\sigma}$, of the histograms,

$$mean(tSNR) = \hat{\mu}$$

$$sd(tSNR) = \hat{\sigma}$$

For the purposes of estimating tSNR, unprocessed EPI images were aligned to the first volume in the timeseries. No other processing steps, in particular no preprocessing steps for fMRI, were carried out, i.e. images were not normalized to stereotactic space, and were not smoothed, in order to avoid preprocessing steps that may alter the statistical properties of the timeseries. SPM (SPM2, Wellcome Department of Cognitive Neurology, London, UK, <http://www.fil.ion.ucl.ac.uk>) was used for image alignment. MRICRO (<http://www.sph.sc.edu/comd/rorden/mricro.html>) was used for image visualization and presentation. Numerical calculations were performed in Matlab (Version 7.0 R14), The Mathworks, Natick, MA, USA) using custom-written code and standard functions for nonlinear curve fitting. Statistical tests (Shapiro-Wilk for normality, and T-tests or non-parametric Mann-Whitney U-tests for comparison of means) were performed in R (R version 2.6.1, <http://www.r-project.org/>).

fMRI Analysis

First-level fMRI analysis was performed with the SPM2 software package (SPM2, Wellcome Department of Cognitive Neurology, London, UK, <http://www.fil.ion.ucl.ac.uk>). EPI images were corrected for slice-timing, aligned, normal-

ized to stereotactic space, reformatted to a standard SPM representation at $2\text{ mm} \times 2\text{ mm} \times 2\text{ mm}$, and smoothed. Activations were determined with T-tests in the whole brain, and threshold was set to $P = 0.05$ corrected for multiple comparisons. BOLD signal, or more precisely percent BOLD signal, was calculated from

$$\text{Percent BOLD} = 100 \times \beta_1 / \beta_0$$

where β_0 and β_1 are the estimated parameters from the General Linear Model fit to the fMRI timeseries, $\mathbf{Y} = \beta_0 + \beta_1 \mathbf{X}$. Activated areas shown in the Results were selected by Boolean conjunction of the BOLD maps with the clusters of voxels that are above threshold ($P < 0.05$ corrected) and contain more than 100 voxels per cluster. Images were normalized to stereotactic space and smoothed with a Gaussian kernel of $4 \times (\text{voxel dimension})$. The $4 \times$ kernel was chosen as an empirical optimal choice between maximizing sensitivity through use of large smoothing kernels [49] and retaining ample spatial specificity of the original functional images. A temporal (4 s FWHM) Gaussian filter was applied to account for temporal auto-correlations in the timeseries.

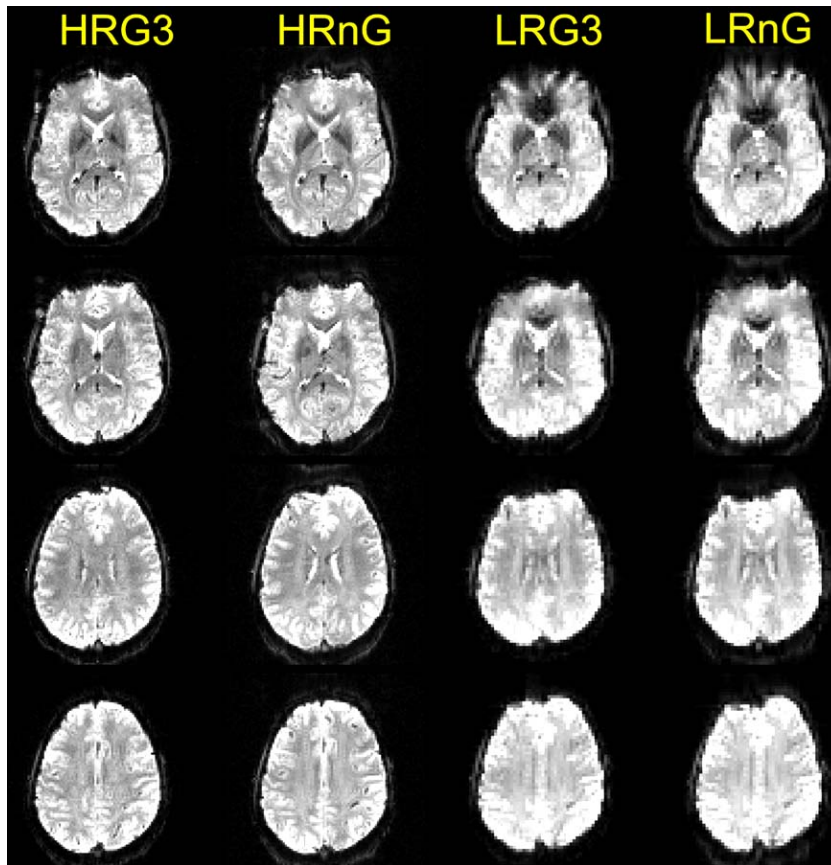


Fig. (1). Comparison of GRAPPA and non-GRAPPA EPI images at high and low resolution. Representative images show the increased blurring and distortions of non-GRAPPA EPI relative to GRAPPA EPI. Four axial slices are arranged from top to bottom rows. Images are from one subject.

Abbreviations: HRG3, high resolution GRAPPA (R=3), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA (R=3), LRnG, low resolution non-GRAPPA, (TE = 42 ms).

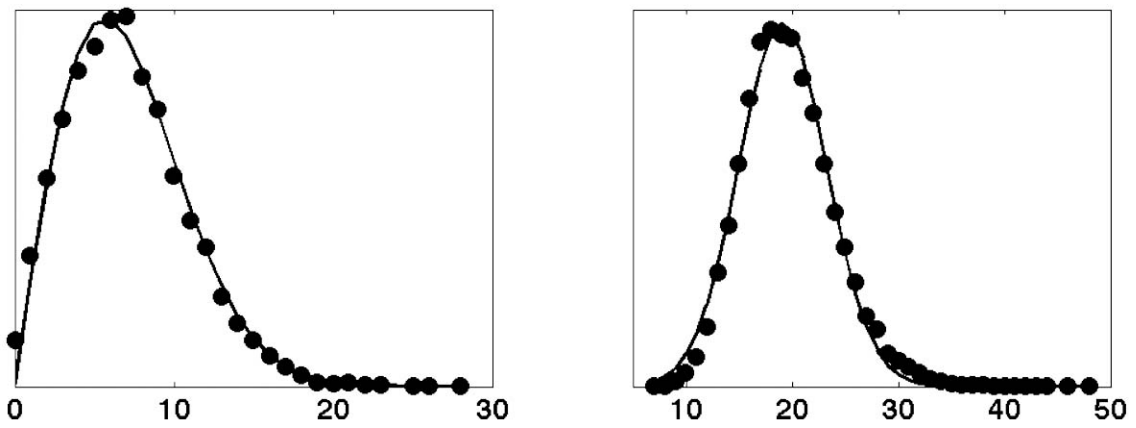


Fig. (2). Typical noise histograms for image SNR (sSNR) calculations. Histograms show the typical distribution of intensities from noise Regions of Interest (ROI) near the edge of the field of view. Left, a non-GRAPPA noise ROI fitted with a Rician distribution. Right, a GRAPPA noise ROI fitted with a non-central Chi-Square distribution.

Abbreviations: HRG3, high resolution GRAPPA (R=3), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA (R=3), LRnG, low resolution non-GRAPPA.

RESULTS

First Experiment

Overall Image Quality

Typical EPI images (GRAPPA and non GRAPPA, from a single subject), are shown in Fig. (1). Visual inspection shows no apparent artifacts in GRAPPA images. The high resolution images show reduced blurring. GRAPPA images show fewer susceptibility distortions, especially in the frontal areas. This difference in susceptibility distortion persists in the higher slices (bottom row of Fig. 1), although it is not as evident as in the lower slices (top row of Fig. 1).

SNR

Fig. (2) shows characteristic fits to the noise histograms, used for sSNR calculation. Noise intensities were measured from ROI near the edge of the field of view. The histogram resulting from the non-GRAPPA noise ROI was fit very well by a Rician distribution [50]. The histogram resulting from the GRAPPA noise ROI was fit by a central chi-square distribution [51]. The sSNR metric may be problematic for image analysis (see Methods and Discussion) and was not subsequently employed [45, 46].

Summary results for tSNR are listed in Table 1. Representative image-wise tSNR calculations from a single subject are shown in Fig. (3), depicting the mean intensity and standard deviation images calculated from raw EPI images, the tSNR image, and the tSNR empirical distribution (histograms). The tSNR histograms are fitted very well by Gaussian curves (fourth row of Fig. 3). The tSNR mean value and standard deviation were estimated distributionally from the tSNR histogram. Representative fit curves are shown superimposed on the tSNR histograms (fourth row of Fig. 3) and also plotted together (Fig. 4). Values for tSNR are listed in Table 1. The values for tSNR for the high resolution non-GRAPPA (HRnG) images are slightly higher compared to GRAPPA (HRG3) (Table 1 and bottom two panels of Fig. 6). The difference reached statistical significance with paired T-test uncorrected for multiple comparisons (Table 2) and also when the Bonferroni correction for multiple comparisons is employed (Table 3). There was no statistically significant difference in tSNR between the low resolution LRG3 and LRnG images, nor between HRG3 and low resolution acquisitions (LRG3, LRnG) (bottom two panels of Fig. 6).

Table 1. Summary Statistics from fMRI Activated Clusters in the Motor/Somatosensory Region (First Experiment)

	HRG3	HRnG	LRG3	LRnG
Maximum percent BOLD	5 ± 1	4.6 ± 0.6	3.5 ± 0.7	3.3 ± 0.7
Mean percent BOLD	1.8 ± 0.1	1.7 ± 0.2	1.4 ± 0.2	1.3 ± 0.2
Number of activated voxels ^a	2600 ± 1000	2600 ± 1200	3500 ± 1700	5000 ± 3000
Number of activated voxels ^b	1549	1350	1027	2583
Maximum T-score	19 ± 4	20 ± 6	19 ± 6	22 ± 8
tSNR	16 ± 4	20 ± 5	16 ± 3	16 ± 2

^aSummary Statistics from first-level data.

^bSecond-level results (group activation, random effects).

Abbreviations: HRG3, high spatial resolution GRAPPA (R=3), HRnG, high spatial resolution non-GRAPPA, LRG3, low spatial resolution GRAPPA (R=3), LRnG, low spatial resolution non-GRAPPA.

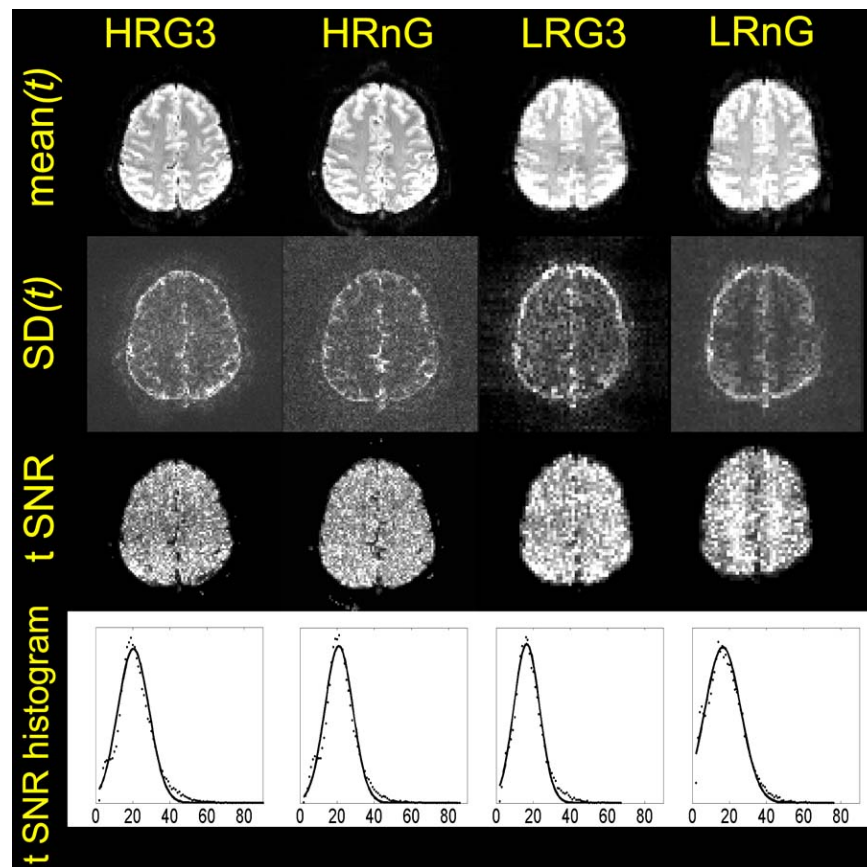


Fig. (3). Typical timeseries SNR (tSNR) calculations and results. Top row depicts the mean intensity image computed from an epoch of EPI images from each of the four protocols, HRG3, HRnG, LRG3, and LRnG. The second row depicts the standard deviation ($SD(t)$) of the timeseries. The third row shows the timeseries SNR (tSNR) computed from the first and second rows, mean and SD images. The fourth row shows the distribution of image intensities of the tSNR images. In each case, the histograms were normalized to unit area and fitted with a Gaussian. These Gaussian fit distributions are shown together in Fig. (4). Images are from a single subject.

Abbreviations: HRG3, high resolution GRAPPA ($R=3$), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA ($R=3$), LRnG, low resolution non-GRAPPA.

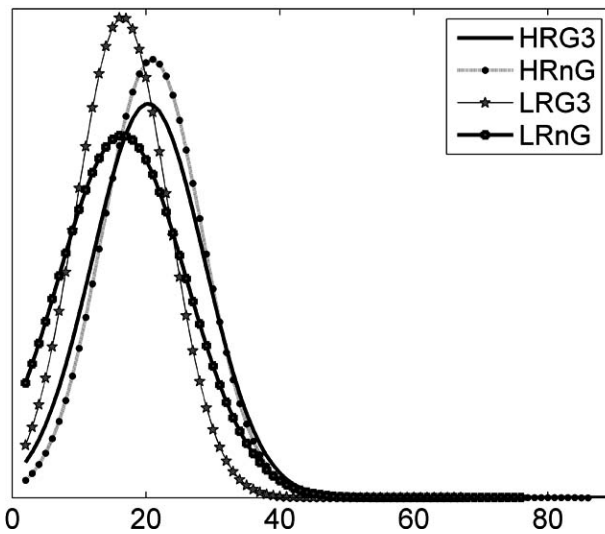


Fig. (4). Comparison of typical tSNR intensity distributions. The four Gaussian fit distributions of Fig. (3) are shown superimposed together. Each distribution is normalized to unit area. Abbreviations: HRG3, high resolution GRAPPA ($R=3$), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA ($R=3$), LRnG, low resolution non-GRAPPA.

We compared the ratios of tSNR of GRAPPA to tSNR of non-GRAPPA versus theoretical ratios calculated assuming thermal noise [44]. We first compared GRAPPA to non-GRAPPA at the same resolution (HRG3 to HRnG and LRG3 to LRnG). In that case, theoretical SNR is inversely proportional to the square root of bandwidth per pixel [44] with an additional penalty $R^{1/2}$ for GRAPPA, with all other parameters (TE, field of view and acquisition matrix, and number of averages) being equal. Using the values of bandwidth per pixel from Methods and acceleration factor $R = 3$ gives the twin theoretical estimates $SNR(HRG3)/SNR(HRnG) = 0.800$ and $SNR(LRG3)/SNR(LRnG) = 0.662$. The experimental ratio at high resolution is $tSNR(HRG3)/tSNR(HRnG) = 0.8 \pm 0.2$. The low resolution experimental ratio is $tSNR(LRG3)/tSNR(LRnG) = 1.0 \pm 0.1$. We then compared high resolution acquisitions to low resolution ones. In this case, theoretical SNR (assuming thermal noise) scales in proportion to the square root of voxel volume and in inverse proportion to the bandwidth for thermal noise. The acceleration factor does not influence the calculation of the theoretical ratios since it is the same for both high and low resolution EPI protocols. The theoretical ratios are $SNR(HRG3)/SNR(LRG3) = 0.52$, $SNR(HRnG)/SNR(LRnG)$

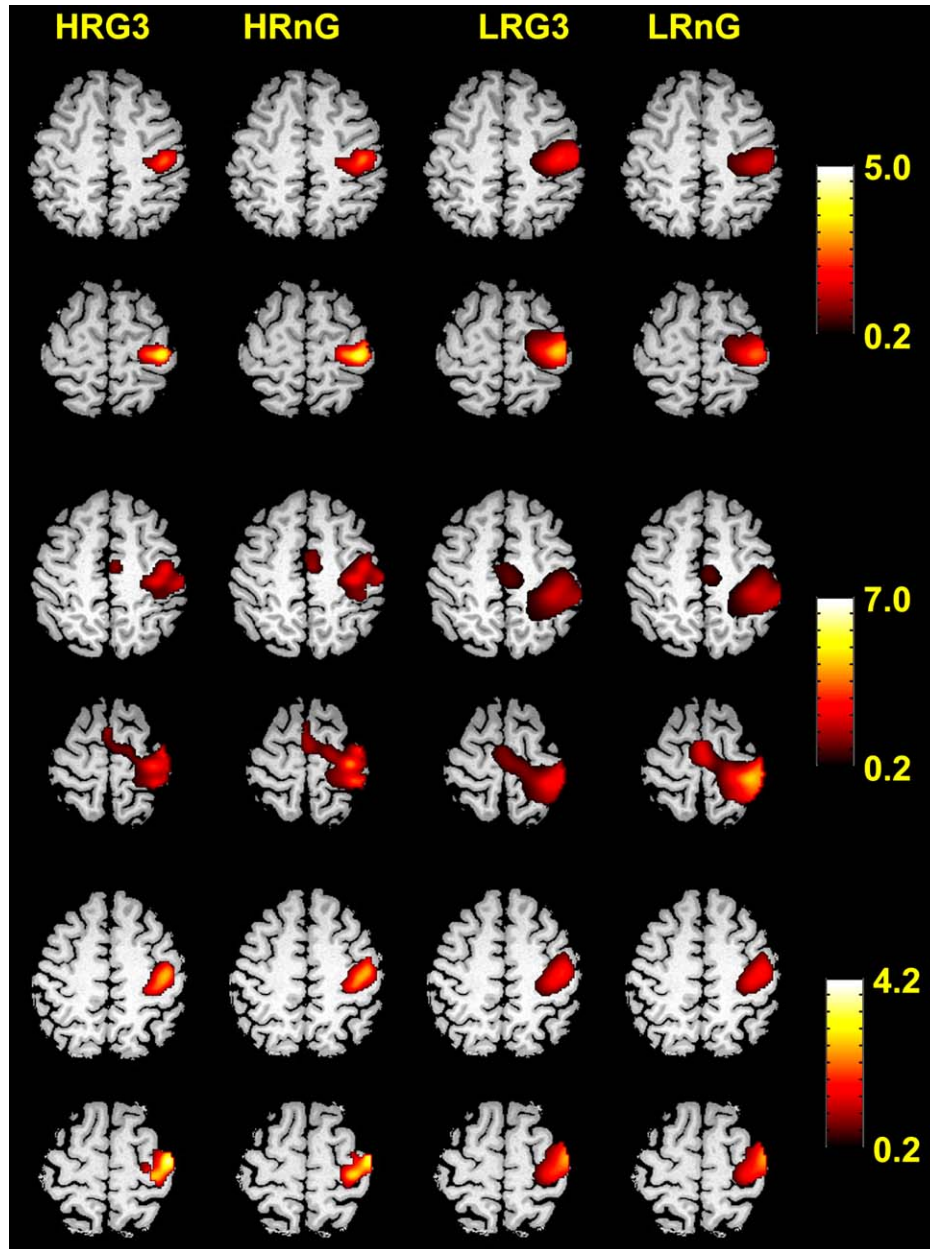


Fig. (5). Typical BOLD signal values and spatial activation profiles for each of the four acquisition protocols. Images are shown from three subjects (two axial slices from each subject at the level of the motor and somatosensory cortex). Subject 1, top two rows; subject 2, third and fourth rows; subject 3, fifth and sixth rows. The BOLD signal is more intense in the high resolution (HRG3, HRnG) than in the low resolution (LRG3, LRnG) images (BOLD is scaled to the same range). There is no essential difference in activated area between the GRAPPA and non-GRAPPA images (HRG3 compared to HRnG and LRG3 compared to LRnG). The low resolution images show more activation volume, but they are more blurred; the high resolution images show better cortical specificity than the low resolution images. The activated areas encompass the voxels that were significant at $P < 0.05$, corrected.

Abbreviations: HRG3, high resolution GRAPPA (R=3), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA (R=3), LRnG, low resolution non-GRAPPA.

= 0.43, and $\text{SNR}(\text{HRG3})/\text{SNR}(\text{LRnG}) = 0.30$. All experimentally calculated ratios break this scaling relationship. Specifically, $\text{tSNR}(\text{HRG3})/\text{tSNR}(\text{LRG3}) = 1.0 \pm 0.2$, $\text{tSNR}(\text{HRnG})/\text{tSNR}(\text{LRnG}) = 1.2 \pm 0.2$ and $\text{tSNR}(\text{HRG3})/\text{tSNR}(\text{LRnG}) = 1.0 \pm 0.3$.

fMRI – Volume of Activation

While the low resolution acquisitions resulted in larger activation volume (Table 1), the increase barely reached statistical significance with paired T-tests uncorrected for

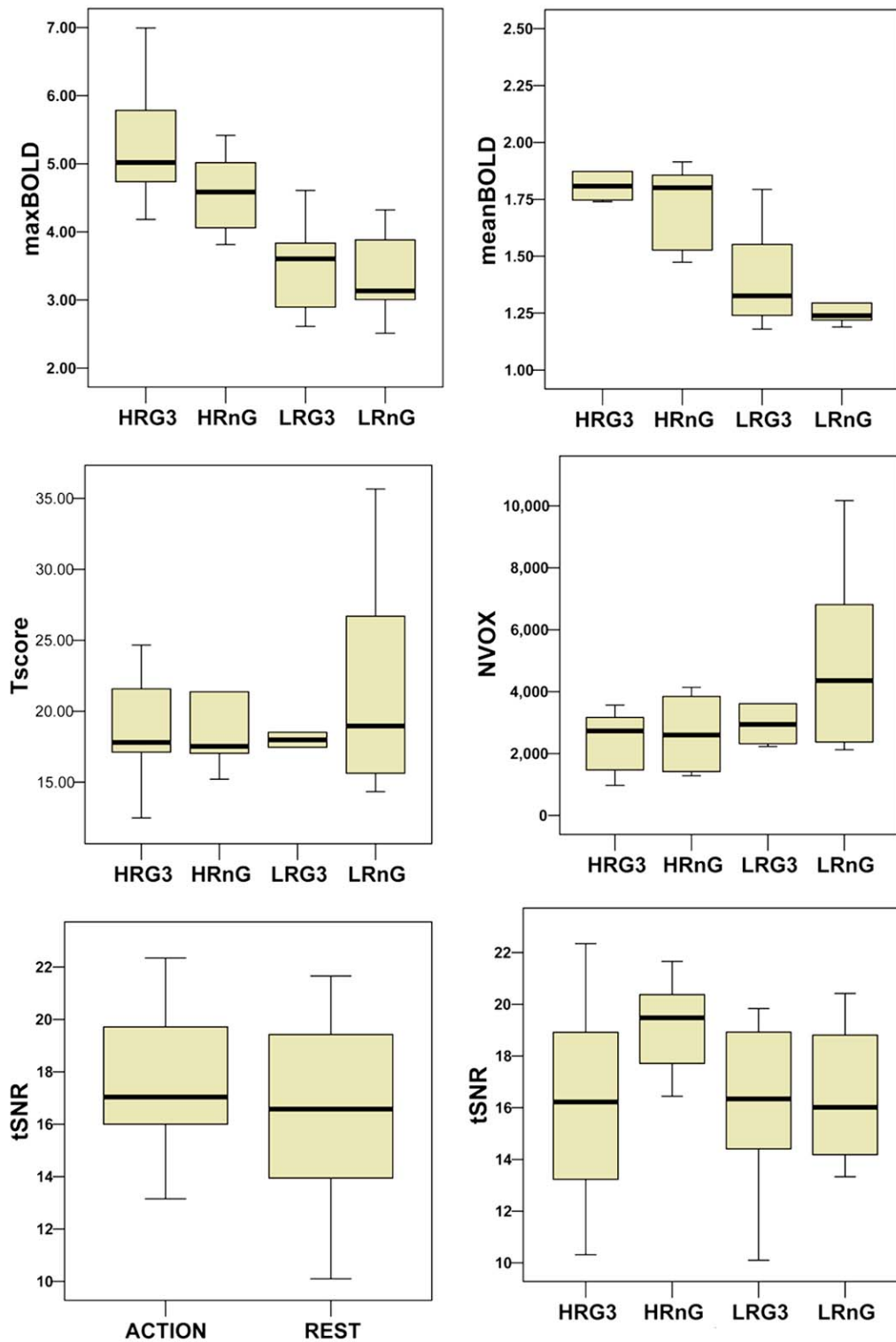


Fig. (6). Comparison of summary results. The box plots show the distributions of the maximal percent BOLD (maxBOLD), the mean percent BOLD (meanBOLD), the maximal T-scores (Tscore), the number of activated voxels (NVOX) and the tSNR. The two tSNR box plots are arranged in the following manner, one with the action / rest epochs as factor and the second with acquisition as factor.

Abbreviations: HRG3, high resolution GRAPPA (R=3), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA (R=3), LRnG, low resolution non-GRAPPA.

multiple comparisons (Table 2) and did not reach statistical significance after correction for multiple comparisons (Table 3). The dispersion of the activated volumes at low resolution

was larger than the dispersion of the high resolution acquisitions (Table 1 and Fig. 6). Fig. (7) shows group activation using image-wise second-level statistics (random effects,

Table 2. Comparison between High and Low Spatial Resolution Images and between GRAPPA and Non-GRAPPA^a

		HRnG	LRG3	LRnG
Number of activated voxels	HRG3	0.5818	0.2093	0.0508
	HRnG		0.2454	0.0646
	LRG3			0.2590
Maximum T-score	HRG3	0.4555	0.6517	0.3031
	HRnG		0.2774	0.5531
	LRG3			0.3929
tSNR	HRG3	0.0106*	0.7889	0.8144
	HRnG		0.0146*	0.0085*
	LRG3			0.9402
Maximum percent BOLD	HRG3	0.0799	0.0064*	0.0046*
	HRnG		0.0002*	0.0114*
	LRG3			0.5979
Mean percent BOLD	HRG3	0.0951	0.0083*	0.0018*
	HRnG		0.0357*	0.0020*
	LRG3			0.4402

^aP-Values, Paired T-test.

*Statistically significant.

Abbreviations: HRG3, high spatial resolution GRAPPA (R=3), HRnG, high spatial resolution non-GRAPPA, LRG3, low spatial resolution GRAPPA (R=3), LRnG, low spatial resolution non-GRAPPA.

Table 3. Comparison between High and Low Spatial Resolution Images and between GRAPPA and Non-GRAPPA^a

		HRnG	LRG3	LRnG
Number of activated voxels	HRG3	1.0000	1.000	0.2032
	HRnG		1.0000	0.2940
	LRG3			1.0000
Maximum T-score	HRG3	1.0000	1.0000	1.0000
	HRnG		1.0000	1.0000
	LRG3			1.0000
tSNR	HRG3	0.0351*	1.0000	1.0000
	HRnG		0.0640	0.0607
	LRG3			1.0000
Maximum percent BOLD	HRG3	0.7164	0.0038*	0.0013*
	HRnG		0.1528	0.0578
	LRG3			1.0000
Mean percent BOLD	HRG3	0.7054	0.0022*	0.0002*
	HRnG		0.0307*	0.0030*
	LRG3			1.0000

^aP-Values, ANOVA with Bonferroni Correction for Multiple Comparisons.

*Statistically significant.

Abbreviations: HRG3, high spatial resolution GRAPPA (R=3), HRnG, high spatial resolution non-GRAPPA, LRG3, low spatial resolution GRAPPA (R=3), LRnG, low spatial resolution non-GRAPPA.

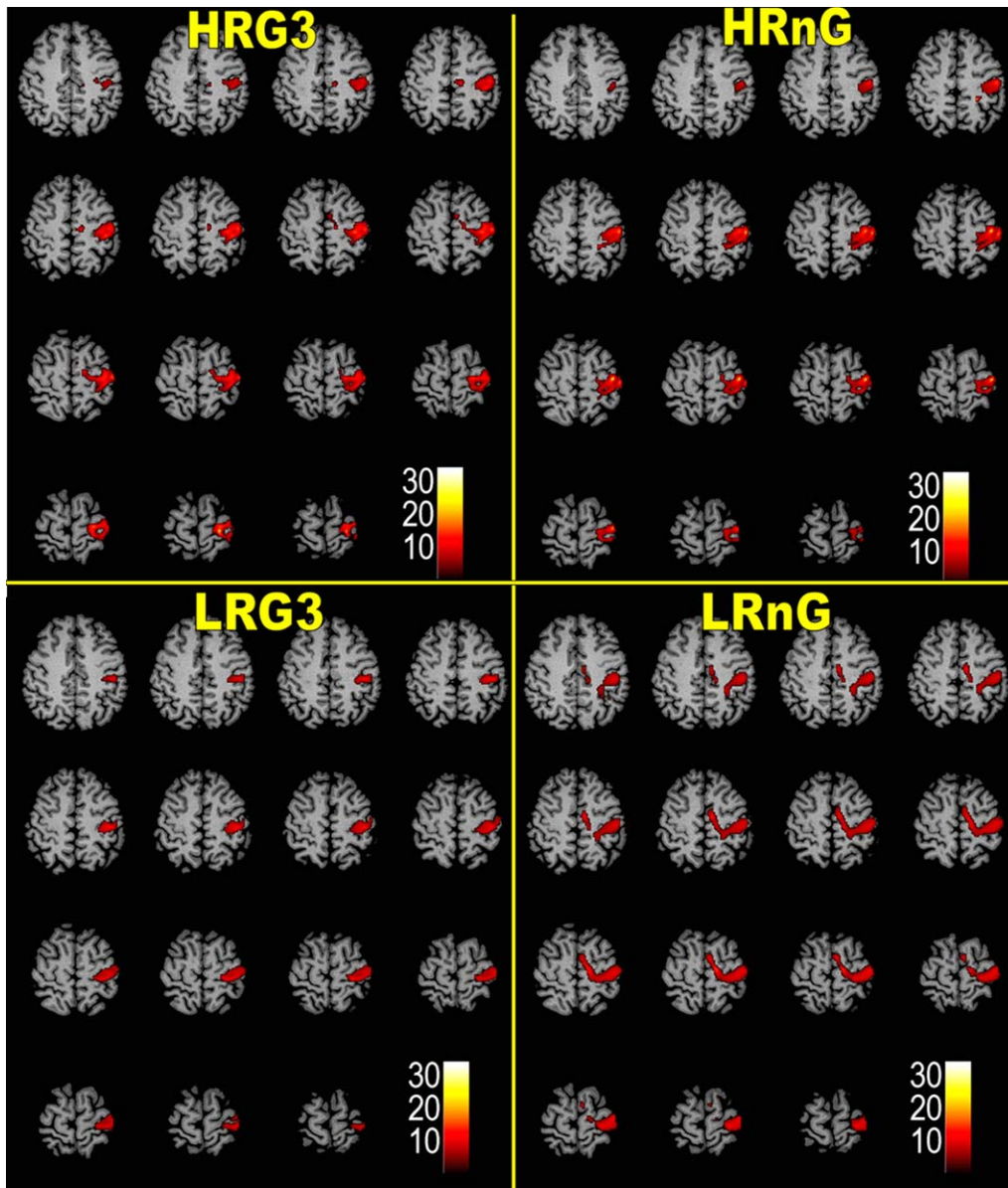


Fig. (7). Comparison of group activation patterns. The image shows group activations for each type of acquisition. The high resolution acquisitions show more detailed activation patterns in the cortical area, and the high resolution GRAPPA also shows activation in the supplementary motor area (N=6, random effects, $P < 0.001$ uncorrected). (Group comparison, color denotes statistical significance (T-scores)). Abbreviations: HRG3, high resolution GRAPPA (R=3), HRnG, high resolution non-GRAPPA, LRG3, low resolution GRAPPA (R=3), LRnG, low resolution non-GRAPPA.

$P < 0.001$, uncorrected). The group activation provides a measure of robustness of the measurements. The group activation in the low resolution is not significantly larger compared to the high resolution. The activated somatosensory cortex volume for LRG3 is approximately the same as that for the HR (HRG3, HRnG) values. The LRnG activated volume is larger, but the activation seems to connect the somatosensory area to the supplementary motor area (Fig. 7) and may be partly spurious.

fMRI – T-Scores

The variation in T-scores from the activated area in the contralateral somatosensory cortex was small (Tables 2 and

3) and statistical comparison of the T-scores themselves did not reveal any statistical significance (P-values shown in Tables 2 and 3). Specifically, the high resolution acquisition T-scores were compared to the low resolution T-scores (HRG3 compared to LRG3 and HRnG compared to LRnG), as well as comparison between the GRAPPA and the non-GRAPPA T-scores (for example, HRG3 compared to HRnG and LRG3 compared to LRnG).

fMRI – BOLD

Table 1 shows that the BOLD signal is approximately the same in GRAPPA compared to non-GRAPPA when spatial resolution is held constant. This holds true for both the

Table 4. Mean Percent BOLD Signal Detected with High Resolution GRAPPA EPI fMRI Compared to Low Resolution Non-GRAPPA EPI fMRI^a

Effort Level ^b	BOLD, high resolution GRAPPA ^c	BOLD, low resolution non-GRAPPA ^c	P-Value ^d	Ratio of high resolution GRAPPA to low resolution non-GRAPPA ^e
15%	1.7 ± 0.5	0.7 ± 0.3	0.0079	2.4 ± 0.5
45%	1.8 ± 0.6	0.9 ± 0.3	0.0305	2.0 ± 0.5
60%	1.8 ± 0.7	0.6 ± 0.3	0.0079	3.1 ± 0.8

^aSecond experiment, TE = 30 ms.

^bEffort measured as percentage of each subject's own maximum.

^cMean ± SD (N=6).

^dT-test, two-tailed.

^eRatio of mean BOLD values ± SD.

maximal value (peak value from the activated area in the contralateral somatosensory cortex) of percent BOLD as well as for the mean percent BOLD from the activated area in the contralateral somatosensory cortex (Table 1). Representative results for BOLD values from three of the six subjects are shown in Fig. (5). These results are further visualized in Fig. (6), which shows summary results from the first-level measurements (maxBOLD denotes the maximal percent BOLD values and meanBOLD the mean percent BOLD values).

The difference between high and low resolution BOLD values is significant and robust under statistical testing. The comparison between HRnG and LRG3 is highly significant when a paired T-test is used (Table 2) and is still significant but less so with P-value $P=0.021$ when a simple T-test is used. It was not significant under multiple comparisons with Bonferroni correction (Table 3). The comparisons between HRG3 and LR (LRG3, LRnG) retain high statistical significance both with paired T-test and under multiple comparison correction (Tables 2 and 3), while the HRnG – LRnG comparison is close to statistical significance ($P = 0.0578$) under correction for multiple comparisons (Table 3).

The difference in measured BOLD was also statistically significant in the second experiment (Table 4), where the subjects performed the experiment at various effort levels.

Second Experiment

Fig. (8) shows that activation was detected in the same brain areas and followed the same pattern using both the GRAPPA and non-GRAPPA EPI protocols.

Measured percent BOLD signal changes were consistently larger using the high resolution GRAPPA protocol (Table 4). This finding remained true as subjects performed the task across three different effort levels (levels of squeezing force). Table 4 compares group-average BOLD signal in the non-GRAPPA low resolution and in the GRAPPA high resolution protocols. The measured BOLD ratios (high resolution divided by low resolution) were averaged over the number of subjects, and the average figures were compared at each squeezing level using two-tailed T-tests. The increase in signal was found to be significant in all cases (P -values listed in Table 4). The ratio of activated volumes (low resolution non-GRAPPA to high resolution GRAPPA) was close to one, and the grand average over subjects and effort levels was 0.8 ± 0.4 . The ratios of activated volumes in this case did not significantly differ from 1 ($P = 0.2330$, T-test, two-tailed).

DISCUSSION

The principal finding of this study is that a high resolution imaging protocol resulted in improved detection of BOLD signal, measured by higher values of percent BOLD

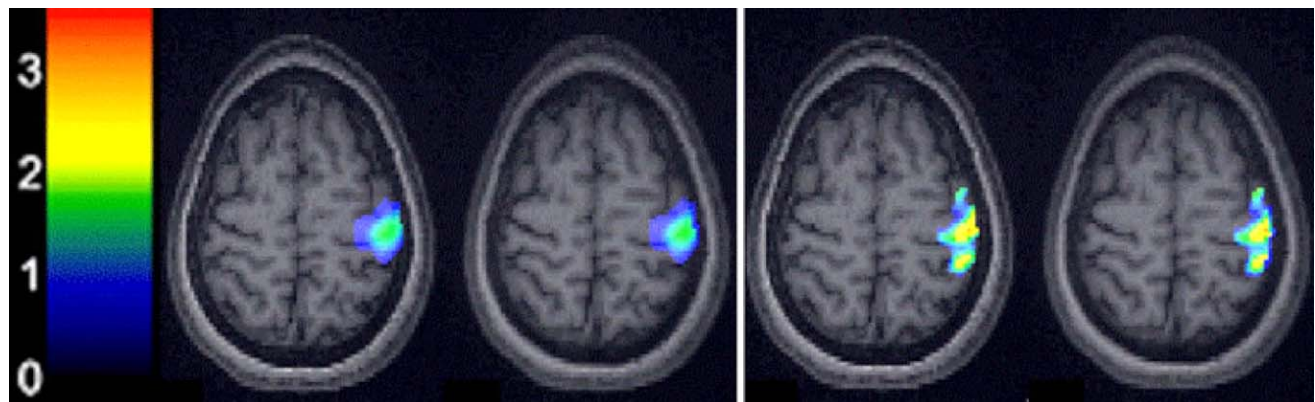


Fig. (8). Comparison of activation profiles and BOLD signal values between a high resolution GRAPPA fMRI experiment and a low resolution fMRI experiment. BOLD signal is higher from the high resolution GRAPPA experiment, and the spatial activation profile shows much better specificity. (Second experiment).

signal, with respect to a low resolution imaging protocol. In addition, we found that using GRAPPA is not detrimental to fMRI. Thus, comparison of GRAPPA to non-GRAPPA keeping all other parameters the same did not result in any significant loss of calculated BOLD signal and/or activation volume. However, the comparison of high resolution to low resolution resulted in statistical significance that was found to be robust under statistical testing, as it was true not only for comparisons with T-tests but also for comparisons corrected for multiple comparisons (Tables 2, 3, 6, and Fig. 6). Thus, our results support our conclusion that the significant improvement in BOLD detection observed herein is a true effect and not a random one (BOLD is a physiological measure and should be independent of the means of detection, in as far as the ratio β_1/β_0 removes any multiplicative factors); and that measurement of the physiological effect (BOLD) is not affected by use of GRAPPA. Instead, measurement of BOLD is improved through the use of GRAPPA pMRI to perform fMRI at higher spatial resolution.

Calculation of other metrics demonstrated that performing fMRI at higher spatial resolution and using GRAPPA for fMRI was not detrimental to the quality of the data. Although the activated volumes were larger with the low resolution acquisitions, so was the dispersion of those results (Table 1 and Fig. 6). This increase in activation volume of the low resolution scans (Table 1) barely reached statistical significance with paired T-tests uncorrected for multiple comparisons (Table 2) and did not reach statistical significance after correction for multiple comparisons (Table 3). This lack of statistical significance may be a consequence of the increased dispersion (Table 1 and Fig. 6), while activation images at the group level (Fig. 7) show very similar activations in all types of acquisitions. In the second experiment, which is complementary to the first, when GRAPPA was used to achieve high spatial resolution and optimal coverage for whole-brain BOLD fMRI at TR/TE = 3000/30 ms at 3 T, we found no difference in the other metrics. Furthermore, we found consistently higher BOLD values (Table 4), demonstrating that GRAPPA fMRI is useful to gain higher spatial resolution, without detriment to the fMRI measurements.

In this study we showed that acquisition at higher spatial resolution exhibited reduced physiological noise, because the scaling ratio between GRAPPA and non-GRAPPA was very close to what is expected assuming that the noise is thermal in origin. This was not the case for the low-resolution acquisitions where scaling between GRAPPA and non-GRAPPA according to the thermal noise model breaks down, which indicates that physiological noise is a much larger proportion of total noise. Physiological noise is proportional to overall signal [47, 52], and so increasing the spatial resolution decreases physiological noise as a proportion of total variance [52]. Although temporal SNR increases with spatial smoothing [53], very broad averaging will “wash out” finer features. Furthermore, all the experimentally calculated tSNR ratios between high resolution acquisitions and low resolution acquisitions also broke the scaling relationships that are expected for thermal noise, which implies that fMRI at higher resolution yields better than expected results. Our findings are in agreement with the literature [29, 47, 52]. Indeed in-

creased spatial resolution is beneficial to fMRI because it reduces the contribution of distant veins to the fMRI BOLD signal [37] and thus reduces the point-spread function [36] and BOLD signal displacement [54]. At a voxel size of approximately 8 mm^3 at 3 T, the partial voluming effect is reduced and thermal and physiological noise contributions to the tSNR [47] are approximately equal [52]. Such voxel size can be regarded as optimal [55] because lower resolution results in very small gains in tSNR while, in the opposite direction, increased spatial resolution results in decreased SNR. Smaller voxels at higher spatial resolution result in improved voxel homogeneity due to reduced partial voluming effect, and the effective T2* of each voxel more closely corresponds to the T2* of cortical gray matter and white matter, respectively. Larger voxels may suffer extra signal loss since the T2* may be shorter due to the partial voluming effect. Here, we used high spatial resolution GRAPPA to gain improved detection of BOLD signal.

The present study complements and extends previous reports by specifically examining the measured BOLD signal and by providing a double comparison between high and low resolution acquisitions as well as between GRAPPA and non-GRAPPA, performed in two different and complementary experiments. Our finding, in both of our experiments, that measured BOLD values did not decrease in GRAPPA compared to regular (non-GRAPPA) fMRI and are higher at higher resolution, demonstrated the utility of GRAPPA fMRI to real clinical applications. In agreement with our results shown in Tables 2 and 3, previous reports did not find adverse effect of GRAPPA or other pMRI techniques to fMRI. Indeed, we agree with Little *et al.* who performed a visual-fMRI study at 3T and reported that use of GRAPPA did not result in reduced BOLD signals in the occipital cortex, while for several subjects the detected volume was larger when using GRAPPA [28]. We also agree with Preibisch *et al.* who found no significant reductions in SNR or in statistical power at acceleration factor $R = 2$ in a motor-task fMRI study using SENSE (at 1.5T) [29]; we further agree with the results of Preibisch *et al.* who failed to find any significant reduction in tSNR in a comparison between GRAPPA and non-GRAPPA [26]. Finally, we agree with Lütcke *et al.* who did not detect any difference between GRAPPA and non-GRAPPA at the same T2*-weighting, although their overall assessment of the utility of parallel imaging for fMRI was negative [27].

In conclusion, we purport that use of GRAPPA is advantageous and not detrimental for fMRI at 3T. We therefore suggest the use of GRAPPA to attain higher resolution for fMRI without severely limiting the field of view and to use a TE in the range of 30-40ms at 3 T for optimal BOLD detection. Further improvements in pMRI may result in improved robustness and utility of pMRI for fMRI and other MR applications, and may blur the current distinction between k-space and real-space methods for pMRI [56-58]. Such applications may be ideally suited for the emerging field of pre-operative and surgical fMRI [59].

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