NOISE IN PARALLEL MRI

How to determine whether single-coil assumptions still hold (they don't)



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BACKGROUND

ABSTRACT

- Existing image processing tools require assumptions about MRI noise [1]–[3]:
 - 1. Stationary PDF
 - 2. Gaussian (or other known) PDF
 - 3. Spatially Uncorrelated
- Assumptions may not hold for Parallel MRI, rendering existing tools invalid
- Need an exploratory analysis framework to test these assumptions

PARALLEL MRI

- Conventional MRI is notoriously slow (vs. CT)
- Parallel MRI promises $2 \times$ to $4 \times$ reduced scan time
- Subsample Fourier space using multiple spatially-sensitive readout coils [1]
- Using coil sensitivity profiles, the subsampled data can be [4]:
 - Unwrapped in the spatial domain (e.g. SENSE, Figure 1)
 - Interpolated in Fourier space (e.g. GRAPPA)
- PMRI for all future scanning protocols: thoracic, fMRI, any clinical

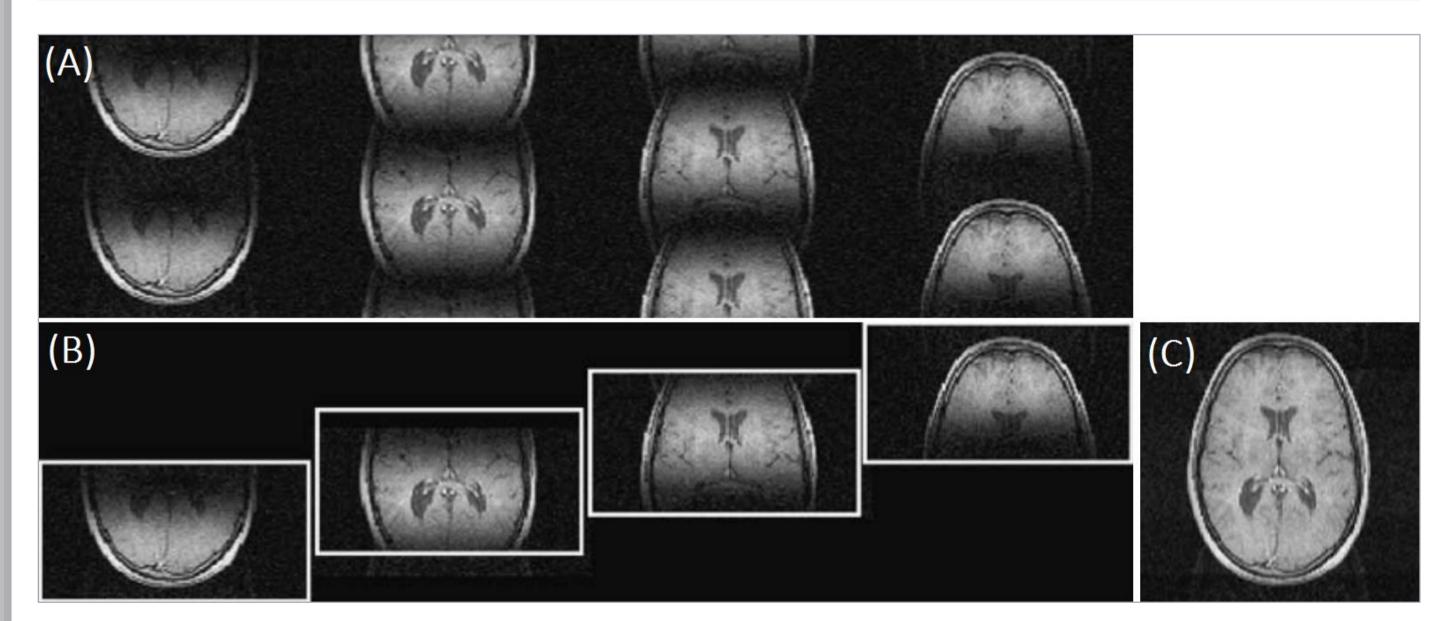


Figure 1: PRMI reconstruction (A) original spatial aliasing due to Fourier space subsampling (B) region selection based on coil sensitivity profiles (C) reconstructed image (adapted from [4])

SINGLE COIL NOISE ASSUMPTIONS

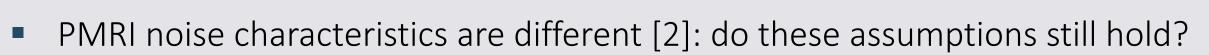
Conventional MRI noise distributions are modelled as additive field [1], [2]:

$$y(x_1, x_2) = f(x_1, x_2) + n(x_1, x_2),$$

 $(y(x_1, x_2) : observed image)$ $\{f(x_1,x_2): ideal\ signal\$ $n(x_1, x_2)$: noise image

FSIL

- 3 single coil noise assumptions: 1. Stationary; 2. Known PDF; 3. Spatially Uncorrelated
- These assumptions incorporated into popular analysis tools:
- Segmentation & coregistration [3]
- Gaussian Mixture Models
- Synthetic image volumes (BrainWeb)



EXPLORATORY NOISE ANALYSIS

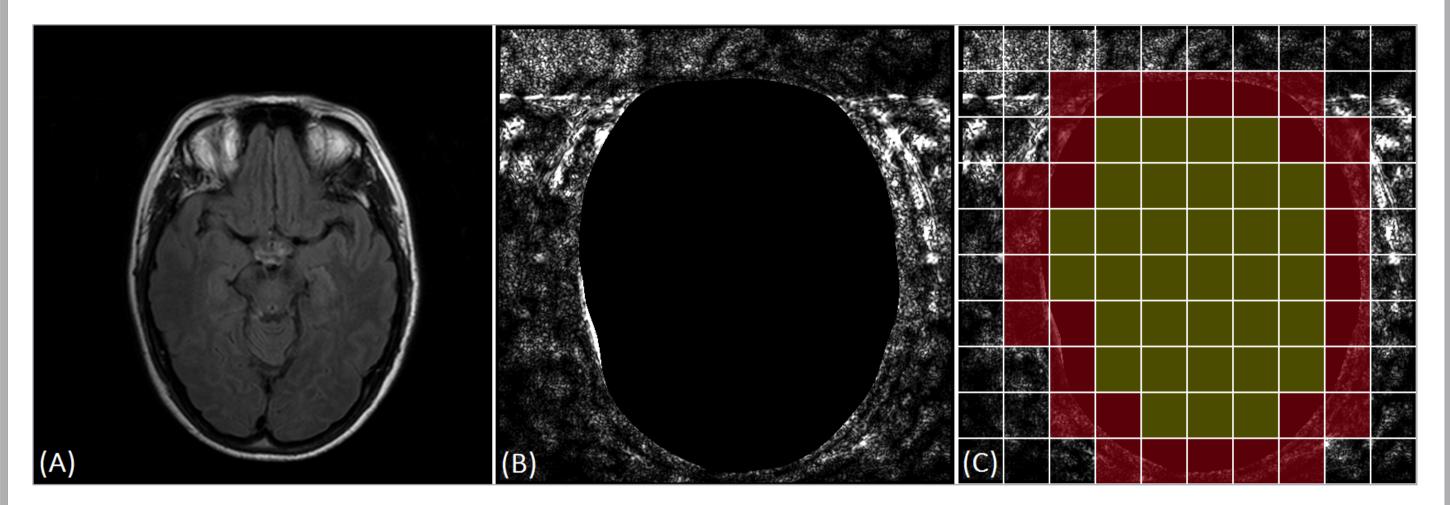
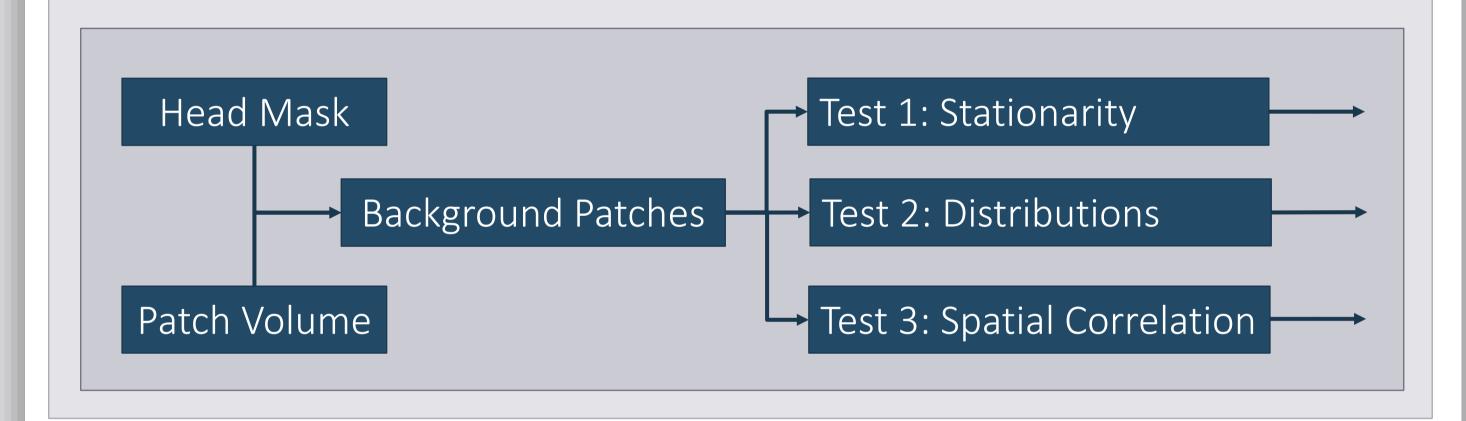


Figure 2: Patching pipeline (A) original PMRI FLAIR image (B) background signal at 100 x contrast scale (C) patched background signal: uncoloured patches are completely background and used for testing; yellow are completely head; red are a mixture.

NOISE ANALYSIS PIPELINE



TEST 1: STATIONARITY

- Question: Is the noise distribution consistent across the image plane?
- Test: 2-Sample Kolmogorov-Smirnov (KS) Test comparing patch distributions
- H_0 : Data in both patches come from the same distribution

Test statistic $\Delta_{1,2}$ of the KS Test $\Delta = \sup | CDF_1(Z) - CDF_2(Z) |$

TEST 2: DISTRIBUTIONS

- Question: Do the noise data follow a known distribution, e.g. Gaussian, Rician, Rayleigh...
- Test: 2-Sample KS Test comparing patch data to EM-optimally fit distribution
- H_0 : patch data comes from the distribution

Fitting distributions to patch data -Normal -Rician -Rayleigh Graylevel

TEST 3: SPATIAL CORRELATION

- Question: Are the data in a given patch spatially correlated?
- Test: 2D Spatial Correlation (2DSC) Test: compare observed M₂ statistic with M₂ from random permutations of same data
- H₀: patch data are spatially uncorrelated

Test statistic M_2 of the 2DSC Test

$$W_{i,j} = ||s_i - s_j||$$

$$U_{i,j} = |Z(s_i) - Z(s_j)|$$

$$M_2 = \sum_{i=1}^{N} \sum_{j=1}^{N} W_{i,j} U_{i,j}$$

RESULTS

REJECTING SINGLE COIL ASSUMPTIONS

- Test database: 12 SENSE-reconstructed FLAIR volumes:
- Volume size: $560 \times 560 \times 50$ / Patch size: $56 \times 56 \times 1 = 5000$ patches per volume
- Test 1: 96 % Nonstationary PDF
- ◆ Test 2: 90 % No Standard Data Distribution
- Test 3: 88 % Spatially Correlated
- Computational expense: infeasible to test all patches; random sampling sufficient

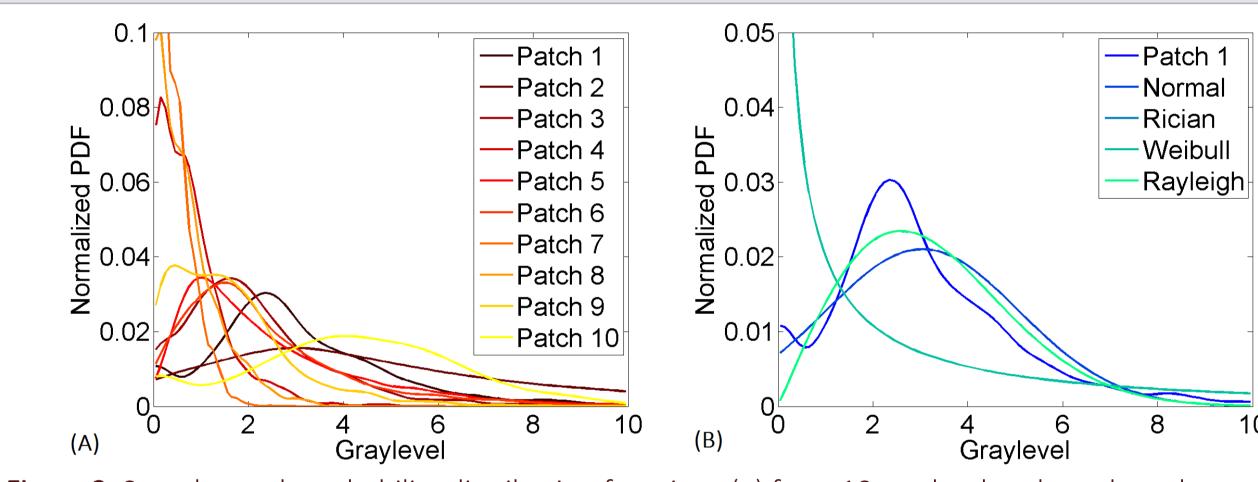


Figure 3. Sample patch probability distribution functions (a) from 10 randomly selected patches and (b) from a single patch, and the corresponding best fit distributions.

Table 1: Test results for 12 image volumes. Shown are the percentages (%) of patch tests for which the hypotheses of single-coil could not be rejected at α = 0.05. Null Hypotheses are Stationarity: Test 1; PDF Match: Test 2; Spatial Uncorrelation: Test 3. *1000 patches tested only

Volume No.		1	2	3	4	5	6	7	8	9	10	11	12	Mean
# Tested Patches		2050	2498	2409	2408	2500	1917	1871	2092	2093	2113	2459	2468	2240
% Stationary		3.8	5.2	6.7	2.8	1.7	4.2	5.6	1.4	8.5	5.5	4	2.7	4.34
	Gaussian	0	0	0.2	0	0	0	1.2	0	0.2	0.1	0	0	0.14
% PDF	Rician	0.6	0.2	0.3	0.6	0.7	0.1	1.3	0.5	0.8	0.5	0.3	0.2	0.51
Match	Rayleigh	0.3	0.1	0	0.4	0.3	0.1	1	0.4	0.5	0.2	0.2	0.2	0.31
	Weibull	11.3	7.5	5.6	12.3	14.8	9.4	10	18.1	6.6	7.7	10.7	11.2	10.4
% Spatially Uncorrelated*		11.3	12.9	14.2	10.3	9.4	13.4	11.6	9.7	17.7	13.4	12	10.9	12.23

FUTURE WORK

- Investigate implementation parameters:
 - KS Test alternatives for discrete distributions (e.g. uint8 volumes)
- Quantify impacts of patch size
- Apply testing framework to compare and contrast:
- PMRI Reconstruction methods (synthetic vs. SENSE vs. GRAPPA)
- Scanner manufacturers (GE vs. Siemens vs. Philips)
- Develop new model-free image analysis tools for when single-coil assumptions fail

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