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# **Assessment of a High-SNR chemical-shift-encoded MRI with complex reconstruction for Proton Density Fat Fraction (PDFF) estimation overall and in the low-fat range**

**Charlie C. Park, MAS**1, **Catherine Hooker, BS**1, **Jonathan C. Hooker, BS**1, **Emily Bass, MA**1, **William Haufe, BS**1, **Alexandra Schlein, BS**1, **Yesenia Covarrubias, MS**1, **Elhamy Heba, MD**1, **Mark Bydder, PhD**1, **Tanya Wolfson, MA**2, **Anthony Gamst, PhD**2, **Rohit Loomba, MD MHSc**3,4, **Jeffrey Schwimmer, MD**5,6, **Diego Hernando, PhD**7, **Scott B. Reeder, MD PhD**7,8,9,10,11, **Michael Middleton, MD PhD**1, **Claude B. Sirlin, MD**1, and **Gavin Hamilton, PhD**<sup>1</sup>

<sup>1</sup> Liver Imaging Group, Department of Radiology, University of California at San Diego, La Jolla, California

<sup>2</sup>Computational and Applied Statistics Laboratory, San Diego Supercomputer Center, University of California – San Diego, San Diego, California

<sup>3</sup>Division of Gastroenterology, Department of Medicine, University of California at San Diego, La Jolla, California

<sup>4</sup>Division of Epidemiology, Department of Family Medicine and Preventive Medicine, University of California at San Diego, La Jolla, California

<sup>5</sup>Division of Gastroenterology, Hepatology, and Nutrition, Department of Pediatrics, University of California San Diego School of Medicine, La Jolla, California

<sup>6</sup>Department of Gastroenterology, Rady Children's Hospital San Diego, California

<sup>7</sup>Department of Radiology, University of Wisconsin - Madison, Madison, Wisconsin

<sup>8</sup>Department of Medical Physics, University of Wisconsin - Madison, Madison, Wisconsin

<sup>9</sup>Department of Biomedical Engineering, University of Wisconsin - Madison, Madison, Wisconsin

<sup>10</sup>Department of Medicine, University of Wisconsin - Madison, Madison, Wisconsin

<sup>11</sup>Department of Emergency Medicine, University of Wisconsin - Madison, Madison, Wisconsin

# **Abstract**

**Background—**Improving signal-to-noise ratio of chemical-shift-encoded MRI acquisition with complex reconstruction (MRI-C) may improve the accuracy and precision of non-invasive proton density fat fraction (PDFF) quantification in patients with hepatic steatosis.

Correspondence to: Gavin Hamilton, MR Research Physicist, Department of Radiology, University of California San Diego, 9500 Gilman Drive, La Jolla, CA 92093, MC 0888, Tel: (858) 246-2194, Cell: (619) 804-3752, ghamilton@ucsd.edu.

Conflict of Interest: Dr. Sirlin consults, advises, and is on the speakers' bureau for Bayer. He has received grants from GE Healthcare. All other authors report no other relevant conflict of interests.

**Purpose—**To assess the accuracy of high SNR (Hi-SNR) MRI-C versus standard MRI-C acquisition to estimate hepatic PDFF in adult and pediatric non-alcoholic fatty liver disease (NAFLD) using MR Spectroscopy (MRS) sequence as the reference standard.

**Study Type—Prospective** 

**Population/Subjects—**231 adult and pediatric patients with known or suspected NAFLD

**Field Strength/Sequence—**PDFF estimated at 3T by three MR techniques: standard MRI-C; a Hi-SNR MRI-C variant with increased slice thickness, decreased matrix size, and no parallel imaging; and MRS (reference standard).

**Assessment—**MRI-PDFF was measured by image analysts using a region of interest coregistered with the MRS-PDFF voxel.

**Statistical Tests—**Linear regression analyses were used to assess accuracy and precision of MRI-estimated PDFF for MRS-PDFF as a function of MRI-PDFF using the standard and Hi-SNR MRI-C for all patients and for patients with MRS-PDFF<10%.

**Results—**Two hundred seventy-one exams from 231 patients were included (mean MRS-PDFF: 12.6% [SD:10.4]; range: 0.9–41.9). High agreement between MRI-PDFF and MRS-PDFF was demonstrated across the overall range of PDFF with regression slope of 1.035 for the standard MRI-C and 1.008 for Hi-SNR MRI-C. Hi-SNR MRI-C, compared to standard MRI-C, provided small but statistically significant improvements in the slope (respectively, 1.008 vs 1.035,  $P=0.004$ ) and mean bias (0.412 vs 0.673,  $P<0.0001$ ) overall. In the low-fat patients only, Hi-SNR MRI-C provided improvements in the slope  $(1.058 \text{ vs } 1.190, P=0.002)$ , mean bias  $(0.168 \text{ vs } 0.368, P=0.002)$  $P = 0.007$ ), intercept (-0.153 vs -0.796, P $\lt 0.0001$ ), and borderline improvement in the R<sup>2</sup> (0.888) vs  $0.813, P=0.01$ ).

**Data Conclusion—**Compared to standard MRI-C, Hi-SNR MRI-C provides slightly higher MRI-PDFF estimation accuracy across the overall range of PDFF and improves both accuracy and precision in the low PDFF range.

#### **Keywords**

hepatic steatosis; MRI-C; Hi-SNR; MRS; proton density fat fraction; PDFF; optimization; accuracy; quantitative imaging biomarker

# **Introduction**

Proton density fat fraction (PDFF) is a quantitative, confounder-corrected magnetic resonance (MR)-based biomarker for the non-invasive assessment of fat in the liver (1-3). It can be acquired by magnetic resonance imaging (MRI) and magnetic resonance spectroscopy (MRS) techniques, and, although MRS is the current gold standard for PDFF measurement, its use is impractical in clinical settings due to technical requirements and spatial coverage limitation (4-7). Meanwhile, chemical-shift-encoded MRI with complex reconstruction (MRI-C) permits accurate PDFF estimation across the entire liver over a dynamic range of 0 to 100% using a spectrally corrected reconstruction; furthermore, this imaging modality has been demonstrated to be highly accurate with respect to MRS (8-11).

To obtain accurate PDFF estimation, MRI-C corrects or minimizes multiple confounders, including T1 and noise bias (12), spectral complexity (13,14),  $B_0$  field inhomogeneity (15,16), phase errors from eddy currents (9), and T2\* decay (14,17-19). Nevertheless, MRI-C sequences estimate PDFF with a small bias of about 1% with respect to MRS. This may have clinical implications for the early detection of nonalcoholic fatty liver disease (NAFLD), as PDFF cutoffs ranging from 3.5 to 6.9% have been proposed for the accurate discrimination of any steatosis (defined as grades 1–3 on NASH Clinical Research Network criteria) versus no steatosis (grade 0) diagnosed on biopsy (3,20-22). This relatively wide range for identifying steatosis partly reflects differences in disease severity between study cohorts, but it may also reflect imprecision in the low-fat range of the MRI-PDFF estimation methods. Additionally, many clinical trials in NASH have used PDFF as an eligibility criterion; typically, PDFF values of  $5\%$  are used for this purpose (23,24). Therefore, further improvement in the accuracy and reliability of MRI-based PDFF estimation techniques may be important to narrow and validate the PDFF thresholds for clinical use and for clinical trials.

A recent mixed phantom-and-in-vivo study by Motosugi et al. proposed increasing the signal-to-noise ratio (SNR) of MRI-C for improved PDFF quantification by increasing imaging voxel size, reducing bandwidth, and removing parallel imaging (25). This high-SNR MRI-C sequence (Hi-SNR MRI-C) in healthy adult patients demonstrated improved precision of PDFF quantification across the entire range of fat. However, to date, no study has assessed the accuracy and precision of the Hi-SNR MRI-C sequence specifically in the low-fat fraction range. Moreover, this sequence has not been validated in adults and children with NAFLD.

The purpose of this study was to assess and compare the accuracy and precision of the Hi-SNR MRI-C and standard MRI-C for PDFF estimation with respect to MR spectroscopy as the reference standard. We assessed these acquisitions across the entire range of PDFF and at the low-fat fraction range of MRS-PDFF < 10% in pediatric and adult patients with known or suspected NAFLD.

# **Materials and Methods**

#### **Study Design**

This was a prospective cross-sectional, single-center, secondary analysis of patients with suspected NAFLD who underwent contemporaneous Hi-SNR MRI-C, standard MRI-C, and MRS acquisition; imaging was performed as part of institutional prospective clinical research studies between February 2014 and June 2015. For these clinical research studies, a total of 231 patients with histologically confirmed or suspected NAFLD were recruited at the NAFLD Research Center (adults) or by the Pediatric Fatty Liver Clinic at Rady Children's Hospital-San Diego (children) and referred for MR examination. Suspicion of NAFLD was based on the presence of at least one of the following: (1) elevated liver transaminases in the presence of obesity, (2) presence of diabetes mellitus, (3) family history of NAFLD, or (4) unexplained elevation of liver transaminases. Exclusion criteria were (1) history of liver disease other than NAFLD, (2) history of chronic alcohol consumption or

We included in our study all consecutive pediatric and adult patients with confirmed or suspected NAFLD who underwent standard and Hi-SNR MRI-C, and MRS, as part of clinical research studies at our institution during the study period. If patients underwent more than one exam during the study period, all exams were included to increase sample size for improved assessment and validation. Exams of 14 pediatric and 14 adult patients meeting inclusion criteria were later excluded from our analysis for the following reasons: (1) the presence of imaging artifact(s) at the MRS voxel location on the MR image, (2) the inability to place the required number of regions of interest (ROIs) at the MRS location due to anatomic factors, or (3) MRS technical failure.

This study was approved by our Institutional Review Board and was compliant with the Health Insurance Portability and Accountability Act. All adult patients provided written informed consent, and pediatric patients provided assent through parental informed consent.

#### **MR Examination**

Patients were instructed to fast for a minimum of four hours prior to imaging to minimize any confounding effects that might affect liver visualization on imaging. After safety screening, patients underwent a non-contrast MR examination on a 3T scanner (GE Signa EXCITE HDxt, GE Healthcare, Waukesha, WI, USA). Patients were positioned supine, with an 8-channel a/p torso phased-array receive coil centered over the liver. A dielectric pad was positioned between the coil and the abdominal wall. Each MR exam lasted about 45 minutes.

# **MRI Sequence & Analysis**

Hepatic fat quantification was performed using two, 6-echo MRI-C acquisitions; standard MRI-C (14,16) and Hi-SNR MRI-C. The investigational Hi-SNR MRI-C sequence was designed by increasing the slice thickness from 8 mm to 10 mm, increasing the voxel size by decreasing the acquisition matrix from either  $192 \times 160$  to  $128 \times 128$  or from  $256 \times 128$  to  $128 \times 128$ , reducing the bandwidth from 125 KHz to 100 KHz, and eliminating parallel imaging. Based on these changes, we calculated that the Hi-SNR protocol had an approximately 3.44-fold increase in SNR compared to the standard MRI-C acquisition using the following formula:

Relative SNR Ratio = 
$$
1/((X_1 * Y_1)/(X_2 * Y_2)) * (\sqrt{BW_1})/(\sqrt{BW_2}) * (ST_1)/(ST_2) * 1/((\sqrt{R_1})/(\sqrt{R_2}))
$$

Where,  $X$ , Y: voxel dimensions,  $BW$ : bandwidth,  $ST$ : slice thickness,  $R$ : parallel imaging acceleration factor/

Our version of the MRI-C sequence software did not permit user control of the repetition time (26) or echo spacing; these were set automatically by the scanner computer, and did not limit the functionality of our validation testing. Hence, the changes in slice thickness and

matrix introduced slight differences in TR and echo spacing between the two sequences, as listed in Table 1. The standard MRI-C sequence was imaged over a single breath-hold; however, due to the elimination of parallel imaging, the Hi-SNR sequence was acquired over 2–3 overlapping single breath-holds to image the whole liver.

Postprocessing of both standard and Hi-SNR MRI-C sequences was performed automatically by the scanner computer with a specialized reconstruction algorithm (8,14) that generated T2\*-corrected multi-fat-peak model parametric PDFF maps online. The resulting parametric PDFF maps were transferred offline for analysis.

All image analyses were performed by three trained image analysts (6–18 months experience) using Osirix imaging software (Pixmeo, Geneva, Switzerland). To prevent reader-related bias, each reader scored both MRI-C sequences in any given patient. For each sequence, a reader manually placed three circular 1-cm radius regions of interest (ROIs) that were co-localized to the MRS voxel location and on the image superior and the image inferior. Readers used the fifth-echo image of the source magnitude images for ROI placement due to the relative ease of hepatic anatomy delineation provided (Figure 1). In addition, placing the ROIs on the source images rather than on PDFF maps prevented feedback bias. All three ROIs were then propagated onto corresponding parametric PDFF maps, and the mean PDFF value was recorded from the average of the three ROIs. Additionally, a T1 bias correction method was applied to the PDFF values obtained from the maps using previously published T1 values at 3 Tesla for liver water (822 ms) and liver fat (312ms) (27), as described by Kuhn el al (28).

#### **MRS Sequence & Analysis**

Single-voxel MRS was performed using Stimulated Echo Acquisition Mode (STEAM) (29).  $A 20 \times 20 \times 20$  mm<sup>3</sup> voxel was placed on the right hepatic lobe (Couinaud segment VI or VII) avoiding large blood vessels, biliary ducts, and liver edges. Following automated shimming during free breathing, five TR 3500 ms spectra were collected at echo times of 10, 15, 20, 25, and 30 msec in a single (∼21 sec) breath-hold to permit T2 estimation while minimizing fat-peak-j-coupling (29). The TR and echo spacing for each technique were automatically set by our scanner computer, but were close to the optimal echo spacing for complex reconstruction per Reeder et al (16).MRS acquisition parameters are listed in Table 1.

MR spectral analysis was performed by a single MR spectroscopist with >15 years of experience blinded to MRI-C acquisitions and clinical data. Analysis was performed using custom prior knowledge using the Advanced Method for Accurate, Robust, and Efficient Spectral fitting of MRS data (AMARES) included in Java-based magnetic resonance user interface software package (30,31). T2-corrected areas of the water (4–6 ppm) and the fat (0–3 ppm) were estimated as there is insufficient spectral resolution in vivo to accurately characterize the individual fat peaks, or to distinguish water from two nearby fat peaks (32). The contribution to the water peak from the neighboring fat peaks was corrected using a previously derived fat spectrum post-T2 correction, which reassigns these fat peaks from water to the fat signal  $(32)$ .

#### **Statistical Analysis**

All statistical analyses were performed using the R software package (R Foundation for Statistical Computing, Vienna, Austria, 2016) by a staff biostatistician, supervised by a faculty biostatistician (each with more than 20 years of experience). All summaries and primary analyses were performed on all patients, and on the subgroup of patients with MRS-PDFF < 10%. Bootstrap extension with by-patient resampling was applied to all confidence intervals (CI's) and tests of significance to adjust for patients with multiple scans.

### **Cohort age and BMI were summarized descriptively**

Intra-class correlation coefficient (ICC) and Bland-Altman analysis were used to assess agreement and systematic disagreement between standard MRI-C and Hi-SNR MRI-C sequences, and 95% CI's were computed around ICCs. Bland-Altman plots were generated, and 95% Limits of Agreement (LOA) and Bland-Altman bias were computed, with significance of the bias assessed using a bootstrap extension of a paired t-test.

To assess and compare the accuracy and precision of standard MRI-C and Hi-SNR MRI-C sequences using MRS-PDFF as the reference standard, we performed univariate regression analyses modeling MRS-PDFF as a function of standard MRI-C PDFF and, separately, of Hi-SNR MRI-C PDFF. Regression accuracy metrics (intercept and slope of the regression line, average bias of the regression, and  $R^2$ ) were computed for each model. Bootstrap-based 95% CI's were computed around all regression metrics. Bootstrap-based tests were used to compare the accuracy metrics of the two MRI-C sequences. Bonferroni correction was applied to each set of accuracy metric comparisons (for all patients and for patients with MRS-PDFF < 10%) to ensure a family-wise alpha level of 0.05.

## **Secondary Analyses**

We assessed the performance of standard MRI-C and Hi-SNR MRI-C sequences to classify MRS-based PDFF cutoffs of 3%, 4%, 5%, 6%, and 7% using the same thresholds. Performance parameters: sensitivity, specificity, positive predictive value (PPV), negative predictive value (NPV), and total accuracy with 95% bootstrap-based CI's were computed for each threshold.

# **Results**

#### **Cohort Characteristics**

Two hundred and seventy-one exams from 231 patients (mean MRS-PDFF: 12.6% [SD: 10.4]; range: 0.9–41.9) were included in this study. A total of 144 exams from 120 patients had MRS-PDFF less than 10%. The cohort demographic is detailed in Table 2. We excluded 28 examinations from our analysis for the following reasons: (1) the presence of parallel imaging artifact at the MRS voxel location on the standard MRI-C image  $(n=2)$ ,  $(2)$  the inability to place all ROIs at the MRS location due to anatomic factors ( $n = 3$ ), (3) MRI acquisition technical failure ( $n= 2$ ), or (4) MRS technical failure ( $n= 21$ ). As the Hi-SNR sequence did not use parallel imaging, no exams were excluded due to parallel imaging artifacts on the Hi-SNR sequence.

#### **Assessment of Agreement Between Standard MRI-C and Hi-SNR MRI-C**

Bland-Altman plots are presented in Figure 2. PDFF measurements by the Hi-SNR sequence were minimally but significantly higher than the standard MRI-C values by  $0.18\%$  points ( $P$  $= 0.0018$ ) on average in the overall cohort. Agreement was high with an ICC of 0.995 (95%) CI 0.993–0.996); however, in the subset of patients with MRS-PDFF < 10%, no significant bias in the PDFF estimated by the MRI-C sequences was observed ( $P = 0.505$ ). ICC was 0.895 (95% CI 0.837–0.928).

#### **Accuracy of Standard Versus Hi-SNR MRI-C Sequences**

Univariate regression plots demonstrated a high correlation between PDFF estimated by the MRI-C sequences and MRS-PDFF across the overall range of PDFF (Figure 3). In the lowfat range (subset of patients with MRS-PDFF < 10%), standard MRI-C demonstrated relatively greater underestimation with a slope of 1.19, compared to that of Hi-SNR MRI-C (slope of 1.06). All MRI-C PDFF versus MRS-PDFF regression parameters are detailed in Table 3.

Bonferroni-adjusted bootstrap tests showed that the Hi-SNR MRI-C sequence, compared to the standard MRI-C, was more accurate overall based on regression bias significantly closer to 0 (0.67% for MRI-C vs. 0.41% for Hi-SNR MRI-C, difference of 0.26% [P < 0.0001]), and had and a slope closer to 1 (1.04 for MRI-C vs. 1.01 for Hi-SNR MRI-C, difference of 0.03  $[P = 0.004]$ ; these tests similarly showed the improved accuracy of Hi-SNR MRI-C in the low-fat range (bias of 0.37% for MRI-C vs. 0.17% for Hi-SNR MRI-C, difference of 0.20%  $[P = 0.007]$ , plus a slope closer to 1 (1.19 for MRI-C vs. 1.06 for Hi-SNR MRI-C, difference of 0.13  $[P = 0.002]$ .

In the low-fat range, all regression-based accuracy metrics were significantly better for the Hi-SNR MRI-C than for the standard MRI-C, with an intercept significantly closer to 0  $(-0.80\%$  for MRI-C vs.  $-1.53\%$  for Hi-SNR MRI-C, difference of 0.64% [ $P < 0.0001$ ]), and a significantly higher  $R^2$  (0.81 for MRI-C vs. 0.89 for Hi-SNR MRI-C, difference of 0.075 [P  $= 0.010$ ]) (Table 4).

# **Performance of Standard and Hi-SNR MRI-C PDFF in Classifying Equivalent MRS-based PDFF Cutoffs**

Table 5 summarizes the performance of the standard and Hi-SNR MRI-C sequences to approximate equivalent MRS-based PDFF cutoffs. Across MRS-PDFF cutoffs of 3–7%, the standard and Hi-SNR MRI-C sequences each demonstrated high accuracy for classifying equivalent MRS-PDFF cutoffs. The two MRI-C sequences demonstrated similar sensitivity, specificity, PPV, NPV, and accuracy at all MRS cutoffs.

# **Discussion**

In this study, a Hi-SNR MRI-C acquisition protocol provided slightly higher PDFF estimation accuracy in adult and pediatric NAFLD patients overall. In patients with low range of PDFF values (0–10%), we found that the Hi-SNR MRI-C sequence provided slightly higher PDFF estimation accuracy and precision. However, the observed increases in

accuracy and precision were small and possibly not clinically meaningful, suggesting that the current standard MRI-C sequence is already well optimized for PDFF estimation. Although the HI-SNR MRI-C sequence represents a slight improvement in PDFF estimation using MRS as the reference standard, this sequence sacrifices spatial resolution to increase the SNR and may reduce imaging's ability to delineate fine details within the liver. This potential limitation has not yet been assessed, and may prove to be a minor concern if the primary purpose of the imaging sequence is to quantify liver fat.

In addition, the Hi-SNR MRI-C sequence that we tested may be disadvantageous in clinical settings due to the multi-breath-hold acquisition (versus single-breath-hold acquisition of the standard MRI-C). However, the need for whole-liver coverage for PDFF estimation is unknown, and a single acquisition through the widest portion of the liver—typically requiring a single breath-hold—is likely to provide sufficient anatomic coverage to sample the liver adequately.

Finally, an unanticipated benefit of our Hi-SNR MRI-C technique was the elimination of parallel imaging artifacts, which occasionally corrupt PDFF parametric maps using the standard MRI-C sequence with the implemented parallel imaging acceleration factors. More robust parallel imaging methods would be required to address this limitation in standard MRI-C. Alternatively, compressed imaging and other methods to reduce acquisition time might prove helpful in the clinical setting.

Our techniques and findings are similar to those described in a recently published study by Motosugi et al., in which the High-SNR MRI-C sequence performed on 1.5 T incrementally improved the precision of PDFF estimation across the entire range of PDFF in a cohort of 20 adult volunteers and 28 adult patients (25). Our study demonstrated similar findings on 3.0 T and with a larger cohort of adult and pediatric NAFLD patients. In contrast with Motosugi et al., our study also demonstrated improvements in the regression bias (i.e., accuracy) using Hi-SNR MRI-C. We expect that this difference may be due to a larger sample size used in our study and is unlikely to be clinically significant.

While the effects of varying acquisition parameters such as voxel size and slice thickness may be of interest for further investigation, our result suggests that the effect of varying these acquisition parameters is small.

Prior studies have proposed various PDFF cutoffs ranging from 3.5 to 6.9 % to discriminate mild steatosis from no steatosis with respect to histology (3,20-22,33). As standard MRI-C and Hi-SNR MRI-C provide the same performance at these ranges of PDFF cutoffs, it is unlikely that the various cutoffs in the literature are caused by differences in the PDFF sequence parameters, but rather can be linked to differences in subject groups (e.g., different PDFF cutoff for adults and children). More technically, in a chemical-shift-encoded MRI sequence with complex reconstruction that utilizes fat spectral modeling, T2\* and T1 correction that is acquired using parameters that are within the boundaries of those reasonable to estimate PDFF, we found that the results are the same regardless of changing the parameters, and that accuracy is not limited by SNR.

That said, other methods (e.g., acquiring images at a high flip angle in the hepatobiliary phase following gadolinium administration) have been previously described for improving the SNR of acquisition while avoiding the loss of resolution (34,35). The method described in this paper represents one approach to increase the SNR for higher fat quantification accuracy without the need for contrast agent administration in the low-PDFF range.

A limitation of this study is that we did not assess the diagnostic performance of Hi-SNR MRI-C with respect to liver biopsy. However, we demonstrated that the standard MRI-C and Hi-SNR MRI-C modalities have similar performance for diagnosing steatosis using equivalent MRS-based PDFF cutoffs (from 3 to 7%) as proxy, suggesting that the two MRI-C PDFF sequences are also likely to demonstrate similar accuracy for diagnosing steatosis with respect to histology. Our study did not investigate optimal PDFF thresholds for diagnosing steatosis in adults or children with NAFLD. Another limitation was that the TR and echo spacing were altered automatically by our scanner computer. While all echo spacings were close to the optimal echo spacing for complex reconstruction as defined by Reeder et al. (16), we could not completely exclude effects due to changes in TR and echo spacing. As TRs were different between the standard and Hi-SNR MRI-C techniques, we applied T1 correction. Other limitations include: use of a single scanner from a single manufacturer, use of single field strength, and the highly-specialized, single-center setting of this study; all factors that might limit the generalizability of our findings in other clinical settings.

In conclusion, the Hi-SNR MRI-C acquisition provides an incremental improvement in the accuracy and precision of PDFF estimation in the low PDFF range. Further protocol optimization for Hi-SNR MRI-C may be warranted in order to optimize the applicability of this technique in a clinical setting.

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# **Figure 1.**

Multi-echo Image Reconstruction of Hi-SNR MRI-C parametric PDFF Map, Six echo source real, imaginary, and magnitude images were reconstructed into a parametric PDFF map using complex-fitting algorithms that utilize fat spectral modeling and T2\* correction. A 1-cm circular region of interest was placed on the 5th echo magnitude image at MRS voxel location, then automatically propagated onto the PDFF map to obtain the mean PDFF. Example case of a 17-year-old patient with low-fat content with MRS-PDFF of 7.4%: a) 6.6% measured by standard MRI-C and b) 7.1% measured by Hi-SNR MRI-C. The higher SNR can be visualized in the Hi-SNR MRI-C parametric PDFF map compared to that of the standard MRI-C PDFF map.



## **Figure 2.**

Bland–Altman plots illustrating the difference between Standard MRI-C and Hi-SNR MRI-C PDFF across the entire range of PDFF (left panel) and at the low range of PDFF (right panel). Mean bias between the two MRI-C sequences is small but statistically significant in the all-patient group but was not significant in the low-PDFF patients only.



#### **Figure 3.**

Regression plots of MRS-PDFF as a function of standard MRI-C PDFF (top panels) and Hi-SNR MRI-C PDFF (bottom panels), for all patients (left panels) and low-fat patients (right panels). The line of best fit is shown in blue. The line of perfect MRS to MRI-C PDFF agreement (slope = 1, intercept = 0) is depicted in red.



# **Table 1 STEAM MRS, Standard MRI-C, and Hi-SNR MRI-C Parameters**

STEAM: Stimulated Echo Acquisition Mode; MRS: Magnetic Resonance Spectroscopy; MRI-C: Complex-based Magnetic Resonance Imaging; Hi-SNR: High Signal-to-Noise Ratio; kHz: TR: Relaxation time; TE: Echo time; TM: Mixing time; kHz: Kilohertz; FOV: Field-of-view

<b>Variable</b>	<b>All Patients</b>	Patients with MRS-PDFF < 10%
No. of Exams	231	120
Age, mean (SD)	31.1(21.6)	28.6(21.9)
Range	$8 - 78$	$8 - 78$
Age Group		
Adult	107(46.3)	45 (37.5)
Pediatric	124(53.7)	75(62.5)
BMI ( $\text{kg/m}^2$ ), mean (SD)	29.6(6.5)	27.9(6.8)
Range	$15.2 - 46.9$	$15.2 - 46.9$
Sex, $N$ $(\%)$		
Female	99 (42.8)	51 (42.5)
Male	132(57.1)	69 (57.5)
MRS-PDFF, % (SD)	12.62(10.36)	4.52(2.64)
Range	$0.89 - 41.88$	$0.89 - 9.84$
Standard MRI-PDFF, % (SD)	11.99 (9.91)	4.45(1.95)
Range	$1.45 - 39.79$	$1.45 - 9.52$
Hi-SNR MRI-PDFF, % (SD)	12.16 (10.22)	4.34(2.30)
Range	$1.15 - 39.57$	$1.15 - 9.91$

**Table 2 Cohort Demographics from 271 exams, 144 with MRS-PDFF < 10%**

MRS-PDFF: Magnetic Resonance Spectroscopy Proton Density Fat Fraction; Hi-SNR: High-Signal-to-noise-ratio

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MRI-C: Complex-based Magnetic Resonance Imaging; Hi-SNR: High Signal-to-Noise Ratio; PDFF: Proton Density Fat Fraction;

MRI-C: Complex-based Magnetic Resonance Imaging; Hi-SNR: High Signal-to-Noise Ratio; PDFF: Proton Density Fat Fraction;

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MRI-C: Complex-based Magnetic Resonance Imaging; Hi-SNR: High Signal-to-Noise Ratio; PDFF: Proton Density Fat Fraction; MRS: Magnetic Resonance Spectroscopy MRI-C: Complex-based Magnetic Resonance Imaging; Hi-SNR: High Signal-to-Noise Ratio; PDFF: Proton Density Fat Fraction; MRS: Magnetic Resonance Spectroscopy

\* : Significant

NS: Not significant<br>Because of the Bonferroni's correction, only p values <0.0042 are considered significant at family-wise 0.05 level Because of the Bonferroni's correction, only p values <0.0042 are considered significant at family-wise 0.05 level NS: Not significant

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PPV: Positive predictive value; NPV: Negative predictive value