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Diagnostic Performance of a Rapid Magnetic Resonance Imaging Method of Measuring Hepatic Steatosis

Michael J. House1*, Eng K. Gan2,3, Leon A. Adams2,4, Oyekoya T. Ayonrinde2,3,7, Sander J. Bangma5, Prithi S. Bhathal6, John K. Olynky2,3,7,8, Tim G. St. Pierre1

1 School of Physics, The University of Western Australia, Crawley, Australia, 2 School of Medicine and Pharmacology, The University of Western Australia, Crawley, Australia, 3 Department of Gastroenterology, Fremantle Hospital, Fremantle, Australia, 4 Liver Transplant Unit, Sir Charles Gairdner Hospital, Nedlands, Australia, 5 Resonance Health Ltd, Claremont, Australia, 6 Department of Pathology, The University of Melbourne, Melbourne, Australia, 7 Curtin Health Innovation Research Institute, Curtin University of Technology, Bentley, Australia, 8 Institute for Immunology & Infectious Diseases, Murdoch University, Murdoch, Australia

Abstract

Objectives: Hepatic steatosis is associated with an increased risk of developing serious liver disease and other clinical sequelae of the metabolic syndrome. However, visual estimates of steatosis from histological sections of biopsy samples are subjective and reliant on an invasive procedure with associated risks. The aim of this study was to test the ability of a rapid, routinely available, magnetic resonance imaging (MRI) method to diagnose clinically relevant grades of hepatic steatosis in a cohort of patients with diverse liver diseases.

Materials and Methods: Fifty-nine patients with a range of liver diseases underwent liver biopsy and MRI. Hepatic steatosis was quantified firstly using an opposed-phase, in-phase gradient echo, single breath-hold MRI methodology and secondly, using liver biopsy with visual estimation by a histopathologist and by computer-assisted morphometric image analysis. The area under the receiver operating characteristic (ROC) curve was used to assess the diagnostic performance of the MRI method against the biopsy observations.

Results: The MRI approach had high sensitivity and specificity at all hepatic steatosis thresholds. Areas under ROC curves were 0.962, 0.993, and 0.972 at thresholds of 5%, 33%, and 66% liver fat, respectively. MRI measurements were strongly associated with visual (r² = 0.83) and computer-assisted morphometric (r² = 0.84) estimates of hepatic steatosis from histological specimens.

Conclusions: This MRI approach, using a conventional, rapid, gradient echo method, has high sensitivity and specificity for diagnosing liver fat at all grades of steatosis in a cohort with a range of liver diseases.

Introduction

Fatty liver disease is associated with an increased risk of carcinoma, cardiovascular death, cirrhosis, reduced effectiveness of antiviral treatments, and is implicated in the development of diabetes [1]. A visual estimate of the fat content in a liver biopsy specimen by a histopathologist is considered the gold standard for clinically assessing liver fat. Liver fat levels are generally estimated by the percentage of hepatocytes containing intracellular fat vacuoles graded categorically on a scale from 0 to 100% [2]. Visual estimates of liver fat in biopsy samples are subjective, have poor reproducibility [3], are potentially unrepresentative of the whole liver and require an invasive procedure with associated risks to obtain the sample [4]. Some patients with fatty liver disease are biopsied as part of their routine clinical assessment to determine the severity of fibrosis or inflammation. However, with obesity now common in developed countries there is increased interest in non-invasive methods of quantifying liver fat content for research purposes, diagnosis, and monitoring intervention programs.

Magnetic resonance imaging (MRI) and spectroscopy (MRS) have been used for decades to qualitatively and quantitatively assess liver fat [5,6]. Spectroscopy is currently considered the most accurate quantitative MR method for measuring liver fat, but imaging approaches based on the Dixon method offer the advantage of assessing larger regions of the liver in a comparatively short time and with simpler image processing. Furthermore, use of
MRS is largely limited to research settings due to the technical expertise and cost required to routinely deliver it.

While several reports have compared histological liver fat estimates with quantitative MRI approaches [7–25], there are fewer studies [7,9,15,22,23,26,27] which have assessed the diagnostic performance of MRI using a pathologist’s visual estimate of fat content as the reference standard to identify the clinically relevant thresholds as defined by the Nonalcoholic Steatosis Liver Disease [15]. The largest of these diagnostic studies [15] investigated a cohort of potential living donors without any documented liver disease. Some studies have also supplemented the histopathologist’s visual estimate of fat with computer assisted image processing techniques to objectively measure the area and hence volume fraction occupied by vesicular fat in the biopsy histological section. These image-processing approaches generally use either stereology counting [20,21] or image segmentation [16] with shape analysis [18,19,29] to quantify the area of fat in a biopsy sample. Only two studies [18,29] have compared a quantitative MRI method to fat measurements based on a histopathologist’s visual estimate and those based on morphometric image analysis of the biopsy histological section.

The general approach for measuring liver fat using MRI is to reduce or correct for confounding factors so that there is, as close as possible, a linear, one-to-one relationship between the measured MRI signal and the volume fraction of liver fat. While potentially improving the direct relationship between the MRI signal and liver fat, the acquisition adjustments to minimise confounding factors lead to poor signal to noise in the acquired images, reducing the generic diagnostic quality and potentially the diagnostic sensitivity. Rather than attempt to obtain a measured signal that accurately represented the proton density ratio of fat to water, we took a calibration-based approach that purposefully distorted the measured signal away from the fat-water proton density ratio in order to increase the sensitivity of the measurement to the presence of fat and maintain diagnostic image quality. Hence, the primary aim of this study was to test the ability of this approach, which comprised a triple gradient echo, single breath-hold MRI method, to diagnose clinically relevant grades of hepatic steatosis in a diverse cohort using a histopathologist’s visual estimate of fat content as the reference standard according to the NASH CRN thresholds [28]. A secondary aim was to measure the diagnostic performance of MRI to grade liver fat when a reference standard based on quantitative computer assisted morphometric image analysis of histological sections was used instead of a histopathologist’s visual assessment.

Materials and Methods

Ethics Statement

Written informed consent was obtained from each subject and the study protocol conformed to the ethical guidelines of the 1975 Declaration of Helsinki as reflected in approval by the Fremantle Hospital Human Research Ethics Committee and the Sir Charles Gairdner Hospital Human Research Ethics Committee.

Subjects

Ten healthy controls and 65 patients were enrolled in the study. The patients were recruited from the hepatology outpatient clinics at Fremantle and Sir Charles Gairdner Hospitals, Western Australia. The patient inclusion criteria were: age between 18 and 75 years and written informed consent. Control subjects were not biopsied and required a body mass index (BMI) less than 25 and no history of liver disease. The control subjects were included to provide a baseline of normal liver fat levels as measured by this MRI technique. Exclusion criteria were: contraindications for MRI, pregnancy or lactation. One patient was also excluded after their history indicated fluctuations in weight and alcohol consumption during the study period. An additional five cases were excluded for incorrect MRI acquisition or unavailability of images for morphometric analysis leaving 59 patients that entered the study (Table S1 in File S1).

Liver Histology and Quantification of Liver Fat

The patients underwent percutaneous liver biopsy with ultrasound guidance as part of their routine clinical management. Each liver biopsy specimen was reviewed by an experienced hepatopathologist (PB), blinded to the patient’s identity and MRI data, who visually estimated the percentage of hepatocytes containing fat on a continuous scale between 0% and 100% (HIS-VIS). The METAVIR scoring system was used to stage the amount of fibrosis.

Computer Assisted Morphometric Image Analysis of Liver Fat Area

The fraction of vesicular fat vacuoles in Masson’s trichrome stained histological sections was assessed by computer assisted morphometric image analysis. Histological sections of the biopsies were scanned in colour using an Aperio Scanscope XT (Aperio Technologies, Inc., California, USA) automated slide scanner and ImageScope software. Using ImageJ 1.42 (NIH, USA) software, the fat vacuoles were automatically identified and areas measured in a multi-stage process using thresholding, size, and circularity criteria. Fat vacuoles, holes, tears or vessels in the biopsy section appear as high intensity (white) areas (Figure 1A). These high intensity regions were automatically segmented by applying a threshold intensity of 220 (out of 256) on the green image band, which on inspection provided the best contrast between fat vacuoles and liver parenchyma. The threshold image was converted into a binary image such that the fat vacuoles and any other high intensity regions were given a value of 255 and non-fat tissue a value of 0 (Figure 1B). The fat vacuoles were then automatically identified and delineated using the Analyze Particles tool in ImageJ with a circularity index between 0.5 and 1 and a size threshold between 100 and 10000 pixels (equivalent to diameters from 5.6 to 56 μm). The circularity index ranges from 0 (infinitely elongated polygon) to 1 (perfect circle) and is defined as \((\pi \times \text{Area})/\text{Perimeter}^2\).

The criteria for thresholding intensity, size, and circularity were established by inspection of the results of different thresholds, sizes and circularities on the effective simultaneous exclusion of large vessels, ducts or other large areas of high intensity and inclusion of fat vacuoles. The analysis produced the size and circularity of each individual fat vacuole and the total area of all the fat vacuoles within the threshold ranges. To compute the total area of the biopsy sample, the binary image was reversed and the Fill Holes tool used to produce a biopsy image without holes. The total area of the biopsy sample was measured and the areal fat fraction (HIS-MORPH) computed from the ratio of fat area to total biopsy tissue area.

MRI

Data acquisition. All MRI measurements were made on Siemens 1.3 T Avanto scanners (Siemens Medical Systems, Erlangen, Germany) at Fremantle Hospital, St John of God Murdoch Hospital, and Hollywood Private Hospital, Western Australia. The median time between biopsy and MRI was 57 days.
Phased-array torso coils were centred over the liver of the subjects. MRI acquisition comprised an opposed-phase, in-phase, opposed-phase gradient echo sequence (TEs 2.38, 4.76, 7.14 ms, TR 88 ms, 1 excitation, flip angle 70 degrees, bandwidth 500 Hz). Data from three axial slices, positioned through the widest part of the liver, were acquired in a single breath-hold. The slice thickness was 4 mm and the matrix was 256 x 256 with a field of view 300 x 300 mm. Liver iron concentrations (LIC) were measured using a validated non-invasive MRI method (FerriScan®) [30–32].

Image processing. A single analyst (MJH), blinded to the identity and medical histories of the subjects, reviewed and processed all images using ImageJ 1.42 (NIH, USA). On each of the three slices a circular region of interest (ROI) about 580 mm² was delineated within the right lobe of the liver, avoiding large intrahepatic vessels and any obvious motion-affected regions (Figure 2). The average image intensities within the ROIs, for all three echoes and slices, were used to calculate a parameter, \( \alpha \), (Eqn. 1). The final value of \( \alpha \) reported was the average for the three slices. The parameter \( \alpha \) can take any value between 0 and 1 and is related to the liver fat concentration, but is also dependent on the MRI pulse sequence parameters.

The parameter \( \alpha \) in the liver ROI was calculated for each slice using the following equation:

\[
\alpha = \{ IP - OP1[\exp((TE1 - TE2)/T_2^*)]\}/2IP
\]

where IP is image intensity in the ROI of the first in-phase image, OP1 is image intensity in the ROI of the first opposed-phase image, TE1 is the first opposed-phase echo time, and TE2 is the in-phase echo time [33]. A \( T_2^* \) estimate for correcting the \( T_2^* \) signal decay of the OP1 echo was derived from the two OP images using the equation:

\[
T_2^* = (TE3 - TE1)/[\ln(OP1/OP2)]
\]  

(2)

where OP1 and OP2 are the signal intensities in the ROI of the first and second opposed phase images, TE1 is the first opposed-phase echo time, and TE3 is the second opposed-phase echo time.

Analysis. The relationship between \( \alpha \) and the volume fraction of fat in the liver is not expected to be linear owing to both the difference in proton density between fat and the surrounding tissue and the difference in T1 (the MRI related parameter known as the longitudinal relaxation time constant) between fat and the surrounding tissue. The relationship between \( \alpha \) and volume fraction (f) of fat in the liver is expected to be of the form:

\[
\alpha = kf/(1 + kf - f)
\]

(3)

where \( k \) is a constant [33]. Since the fat vacuoles are approximately spherical and the histological sections are thin compared to the size of the fat vacuoles, the area fraction of the histological section accounted for by fat vacuoles (HIS-MORPH) will be equal to the volume fraction of fat in the liver tissue.

Statistical Analysis

Descriptive clinical and demographic characteristics were compared using Chi-squared analysis (for categorical data), the
Measuring Hepatic Steatosis with MRI

Results

Demographic and Clinical Data

The control subjects were younger and had lower BMIs than the patients (Table 1). BMI values were normally distributed. Thirty-two patients had no/mild fibrosis (METAIR 0 or 1) and 27 patients had moderate/severe fibrosis (METAIR 2, 3 or 4). Five patients had liver iron concentration levels above the normal maximum level of 1.8 mg[Fe]/g dry tissue. The median liver fat level of the 59 patients was 5% (range 0 to 98%) from HIS-VIS and 0.029 (range 0.002 to 0.216) from HIS-MORPH. The median value of α for the control subjects (0.024, range 0.013 to 0.105) was significantly lower compared with the patients (0.072, range 0.01 to 0.41, p = 0.012). The 95% prediction interval of α for the control subjects was 0.008 to 0.094. This prediction interval can be viewed as an estimate of the reference range of MRI α values for healthy subjects without liver problems. Interestingly, one of the ten control subjects without any recognised liver condition was outside this reference range.

Performance of MRI for Predicting Histological Fat Levels

The ROC curve analyses are summarised in Tables 2 and 3. The analyses showed that the MRI α values had high AUCs, sensitivities and specificities at all three liver fat thresholds (Table 2, 3). Based on the ROC curve analysis, the clinically relevant NASH CRN steatosis grade boundaries of 5%, 33% and 66% correspond to MRI α values of 0.067, 0.135, and 0.171, respectively (Table 2). There were no significant differences between the AUC’s from morphometric image analysis compared with the pathologist’s visual estimate (Table 2, 3).

Relationship between MRI and Histology

Figure 3 shows a fit of Equation 3 to the α versus HIS-MORPH data. The coefficient of determination, $r^2$, for this fit was 0.84. Linear regression analysis showed that there were significant coefficients of determination between the pathologist’s visual estimates of fat (HIST-VIS) and α from the MRI measurement ($r^2 = 0.83$, Fig. 4) and between the pathologist’s visual estimates of fat (HIST-VIS) and the morphometric fat fraction HIS-MORPH ($r^2 = 0.72$).

Limits of Agreement between Liver Fat Measurements

The MRI alpha results were transformed to equivalent HIST-MORPH values using Eqn. [3] and these values (MRI HIST-MORPH fat fraction) are plotted against the HIST-MORPH measurements (Figure 5a). A plot of the difference against the mean of the two measurements indicated that there was an increase in the variance of the difference between the measurements as the magnitude increased. Hence, the measurements were not amenable to Bland and Altman [34] analysis without transformation. The natural logarithms of each measurement were therefore calculated and the difference between the natural logarithms of the two fat measurements plotted against the mean of the two natural logarithms of the fat measurements (Figure 5b).

Student’s t test (for continuous parametric data), or the Mann-Whitney test (for non-parametric data). Data were tested for normality using the Komolgorov-Smirnov test. Non-Gaussian distributions were summarised by their median value and range. For non-Gaussian distributions the 95% prediction interval was calculated on the log-transformed data. Relationships between continuous parameters were assessed using the coefficient of determination ($r^2$). The performance of the MRI technique for predicting the histologically measured fat grades was assessed using receiver operating characteristic (ROC) curve analysis. The area under the ROC curve (AUC) was used to assess the diagnostic performance of the MRI method against the biopsy observations. The thresholds of α were identified by the cut-off that produced the highest combined sum of sensitivity and specificity for distinguishing histological fat scores above and below the standard thresholds. Two ROC curve analyses were performed. The first analysis used the NASH CRN histological scoring system cut-off values of ≥5%, >33%, and >66% liver fat [2], as determined by the visual estimation of the hepatopathologist (PB), to define the diagnostic groups above and below the three cut-offs. A second ROC curve analysis used cut-offs in HIS-MORPH fractions of ≥0.014, >0.043, and >0.077. These morphometric cut-offs were derived from the regression line of HIS-MORPH against HIS-VIS using the visual histopathologist cut-offs of 5%, 33% and 66% as input into the regression equation. The methods of Bland and Altman [34] were used to assess the 95% limits of agreement between MRI and biopsy (HIST-MORPH) fat estimates. A p value less than 0.05 was considered statistically significant.

Figure 2. Axial magnetic resonance images of a patient with a typical ROI (solid grey circle) used for fat analysis. A) 2.38 ms (opposed phase), B) 4.76 ms (in phase), C) 7.14 ms (opposed phase). doi:10.1371/journal.pone.0059287.g002
There was no significant correlation between the differences and the mean of the liver fat logarithms (Spearman’s rank order correlation coefficient \( \rho = 0.1051, p = 0.4283 \)). The mean difference between the measurements (\( 20.103, \) standard deviation = 0.495) was not significantly different from 0 (\( p = 0.1167 \)). Furthermore, the mean differences between measurements for each of the three different scanners used in the study were not significantly different from zero (\( p = 0.65, 0.12, \) and 0.32). The upper and lower 95% limits of agreement between the two logarithms of fat measurements were found to be \(-1.073 (95\% CI -1.296 to -0.85)\) and \(0.868 (95\% CI 0.645–1.091),\) respectively (Figure 5b). These limits of agreement correspond to 95% of pairs of liver fat measurements being expected to have ratios (fat fraction measured by morphometry to fat fraction measured by MRI) between \(0.342 (95\% CI 0.274–0.427)\) and \(2.382 (95\% CI 1.907–2.976)\).

### Discussion

In this cohort of patients with heterogenous liver disease, a simple, rapid, quantitative MRI measurement based on in-phase and opposed-phase imaging can provide clinically relevant diagnostic information on liver fat levels. The MRI technique demonstrated high levels of sensitivity and specificity at all clinically relevant thresholds of liver fat (\(\geq 5\%), \(>33\%), \(>66\%\)) [2], using the gold standard of liver fat estimated by a pathologist from a biopsy sample. At these three cut-offs we achieved areas under ROC curves of 0.962, 0.993, and 0.972, respectively. Our results compare favourably to the small number of studies that have used a pathologist’s visual estimate of fat content as the reference standard to define thresholds and assess the diagnostic performance of comparable MRI approaches [7,9,15,22,26,27] (see Table S2 in File S1). The MRI technique presented here has higher areas under ROC curves for all liver fat thresholds described in the literature (Table S2 in File S1), except for two studies [7,15] investigating liver fat in non-diseased cohorts (potential living donor candidates). At 5% or more fat, our AUC was marginally lower (0.962 versus 0.987) than the result from Joe et al. [7] and, at 30% or more fat, Lee et al. [15] reported an AUC of 0.995 compared to our result of 0.992. For identifying subjects with histological fat \(\geq 5\%,\) McPherson et al. [22] calculated separate areas under ROC curves for subjects with METAVIR fibrosis stage 1 and below (AUC = 0.97) and subjects with METAVIR fibrosis stage 2 and above (AUC = 0.87). When analysed on a similar basis, our data gave AUCs of 0.95 and 0.99 respectively.

### Table 1. Clinical data of the study cohort.

<table>
<thead>
<tr>
<th>Characteristics</th>
<th>Controls</th>
<th>Chronic Liver Disease Patients</th>
<th>Patients with MRI and Morphometry</th>
</tr>
</thead>
<tbody>
<tr>
<td>N</td>
<td>10</td>
<td>65</td>
<td>59</td>
</tr>
<tr>
<td>Gender (female/male)</td>
<td>3/7</td>
<td>31/34</td>
<td>29/30</td>
</tr>
<tr>
<td>Age (years), median (range)</td>
<td>33.5 (24–47)</td>
<td>56 (20–72)*</td>
<td>56 (20–72)*</td>
</tr>
<tr>
<td>BMI (kg/m²), mean ± st. dev.</td>
<td>22.55±1.77</td>
<td>29.00±5.11(^\circ)</td>
<td>28.92±5.17(^\circ)</td>
</tr>
<tr>
<td>LIC (mg/g), median (range)</td>
<td>1.2 (0.4–1.8)</td>
<td>0.9 (0.3–4.8)</td>
<td>0.9 (0.3–4.4)</td>
</tr>
<tr>
<td>Diagnosis</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>AIH</td>
<td>3</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>ALD</td>
<td>3</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>HBV-HCV</td>
<td>18</td>
<td>16</td>
<td></td>
</tr>
<tr>
<td>NAFLD</td>
<td>11</td>
<td>10</td>
<td></td>
</tr>
<tr>
<td>NASH</td>
<td>19</td>
<td>17</td>
<td></td>
</tr>
<tr>
<td>NORM</td>
<td>3</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>PSC</td>
<td>4</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>OTHER</td>
<td>4</td>
<td>4</td>
<td></td>
</tr>
</tbody>
</table>

*Significantly different from controls, \(p<0.05\) Mann-Whitney test.
\(^\circ\)Significantly different from controls, \(p<0.05\) unpaired t-test. Abbreviations: AIH, autoimmune hepatitis; ALD, alcoholic liver disease; BMI, body mass index; HBV-HCV, chronic viral hepatitis B/C; LIC, liver iron concentration; MRI, magnetic resonance imaging; NAFLD, non-alcoholic fatty liver disease – simple steatosis; NASH, non-alcoholic steatohepatitis; NORM, normal; PSC, primary sclerosing cholangitis.

doi:10.1371/journal.pone.0059287.t001

### Table 2. Analysis of the area under the receiver operating characteristic curve using the histopathologist’s visual estimate of fat in the histological sections.

<table>
<thead>
<tr>
<th>Cut-off</th>
<th>MRI Cut off Value (a)</th>
<th>AUC</th>
<th>p value</th>
<th>Sensitivity</th>
<th>95% CI</th>
<th>Specificity</th>
<th>95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\geq 5% \text{ (Grade 0 vs 1–3)*})</td>
<td>&gt;0.067</td>
<td>0.9615</td>
<td>&lt;0.0001</td>
<td>90.91</td>
<td>75.67 to 98.08</td>
<td>96.15</td>
<td>80.36 to 99.90</td>
</tr>
<tr>
<td>&gt;33% (Grade 0–1 vs 2–3)*</td>
<td>&gt;0.135</td>
<td>0.9928</td>
<td>&lt;0.0001</td>
<td>100.0</td>
<td>85.18 to 100.0</td>
<td>97.22</td>
<td>85.47 to 99.93</td>
</tr>
<tr>
<td>&gt;66% (Grade 0–2 vs 3)*</td>
<td>&gt;0.171</td>
<td>0.9724</td>
<td>&lt;0.0001</td>
<td>100.0</td>
<td>79.41 to 100.0</td>
<td>88.37</td>
<td>74.92 to 96.11</td>
</tr>
</tbody>
</table>

*Cut-offs are defined according to the NASH CRN grading system. Abbreviations: AUC, Area under receiver operating characteristic curve; CI, confidence interval; MRI, magnetic resonance imaging.

doi:10.1371/journal.pone.0059287.t002
The diagnostic performance of the MRI technique was largely independent of the method of quantifying the fat content in biopsy specimens, consistent with the strong coefficients of determination between $\alpha$, HIS-VIS, and HIS-MORPH. The AUCs based on thresholds from the morphometric image analysis were similar to those based on the visual cut-offs (Table 2, 3). To our knowledge, no other study has used reference thresholds based on morphometric image analysis to assess the diagnostic performance of in-phase, opposed-phase MRI. The strong diagnostic performance of the MRI method using two independent reference standards suggests that it is robust. In our approach, the difference in $T_1$ between fat and tissue water is exploited, by use of a high flip angle, to distort the relationship between fat volume fraction and $\alpha$ away from a linear 1:1 relationship to a non-linear relationship (Figure 3). For tissues with a limited range of fat volume fraction, $f$, (e.g. from 0.00 to 0.40) the sensitivity of the MRI measurement $\alpha$ for predicting the fraction of protons in fat is potentially enhanced for two reasons. Firstly, the differential of the relationship between the MRI measurement $\alpha$ and the fraction of tissue protons in fat is increased and secondly, the random error on $\alpha$ can be decreased owing to better signal to noise ratios in the MRI data. The results from our study therefore indicate that the particular adaptation of the Dixon technique we used can be a useful non-invasive tool for diagnosing liver fat across a range of liver diseases with high sensitivity and specificity at all grades of steatosis.

This study also showed that the MRI $\alpha$ measurement was strongly associated with visual ($r^2 = 0.83$) and computer-assisted morphometric ($r^2 = 0.84$) estimates of liver fat from histological specimens. In comparison, several studies have also reported mostly strong correlations between MRI measurements and estimates or grades of liver fat from histology specimens [10–12,15–17,21–26,29,33,36]. These studies vary in the types and numbers of patients and the MRI acquisition and processing used (Table S3 in File S1), possibly explaining the large range in reported coefficients of determination (0.26–0.89). Our MRI versus biopsy results showed stronger coefficients of determination than all but two of these studies [17,36]. However, our study was larger and investigated a broader group of aetiologies compared to these studies (Table S3 in File S1, [17,36]). The strong coefficients of determination we observed between MRI and histological fat measurements suggest that our data could form the basis of a calibration between a quantitative MRI measurement sensitive to liver fat content and histological estimates of fat.

The Bland Altman analysis summarized in Figure 5 gives an indication of the precision of the MRI measurements for diagnosing liver fat.

### Table 3. Analysis of the area under the receiver operating characteristic curve using the morphometric image analysis estimate of fat in the histological sections.

<table>
<thead>
<tr>
<th>Cut-off</th>
<th>MRI Cut off Value ($\alpha$)</th>
<th>AUC</th>
<th>$p$ value</th>
<th>Sensitivity</th>
<th>95% CI</th>
<th>Specificity</th>
<th>95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\geq 0.014$</td>
<td>$&gt;0.060$</td>
<td>0.9639</td>
<td>$&lt;0.0001$</td>
<td>90.91</td>
<td>75.67 to 98.08</td>
<td>96.15</td>
<td>80.36 to 99.90</td>
</tr>
<tr>
<td>$&gt;0.043$</td>
<td>$&gt;0.141$</td>
<td>0.9925</td>
<td>$&lt;0.0001$</td>
<td>100.0</td>
<td>83.89 to 100.0</td>
<td>92.11</td>
<td>78.62 to 98.34</td>
</tr>
<tr>
<td>$&gt;0.077$</td>
<td>$&gt;0.188$</td>
<td>0.9869</td>
<td>$&lt;0.0001$</td>
<td>100.0</td>
<td>79.41 to 100.0</td>
<td>93.02</td>
<td>80.94 to 98.54</td>
</tr>
</tbody>
</table>

Abbreviations: AUC, Area under receiver operating characteristic curve; CI, confidence interval; MRI, magnetic resonance imaging. doi:10.1371/journal.pone.0059287.t003
measurement of liver fat insofar as the 95% limits of agreement are a measure of the scatter of the data about the calibration curve. Although some of the scatter will be attributable to sampling and analytical errors in the biopsy measurement, these 95% limits of agreement will enable comparison of the precision of other techniques when compared against the biopsy method in the future. While a separate prospective study with different scanners and a different cohort of subjects will be required to validate the accuracy of the technique for general use, there was no detectable bias in the calibration between the three scanners used in this study.

Our study has some limitations. Subjects did not have MRI and liver biopsy performed on the same day with potential for changes in hepatic steatosis levels between assessments. However, the strong associations and diagnostic performance we observed suggest that any changes in fat levels between biopsy and imaging were not great. While our cohort size was larger than many comparable studies, further validation studies on a larger independent patient group would be desirable. In-phase, opposed-phase MRI methods can potentially give erroneous liver fat estimates when the ratio of fat to water protons in the liver exceeds 0.5, leading to water/fat ambiguity issues [33]. In this study we assumed that the number of water protons in the liver always exceeded the number of fat protons. We made this assumption based on the results of the Dallas Heart Study [37], which reported that none of the 2207 subjects they studied with MRS exceeded 50% hepatic triglyceride content. Compared with MRS methods, whole liver imaging methods such as the one reported here are unable to give detailed spectroscopic information on the fat content of the liver.

In summary, we have shown that a rapid (single breath-hold), routinely available, triple gradient echo MRI method has a high degree of sensitivity and specificity to the amount of liver fat measured in a biopsy specimen, either by a pathologist’s visual estimate or by morphometric image analysis. Furthermore, this study indicates that this MRI approach provides noninvasive measurement of liver fat with high sensitivity and specificity for diagnosing liver fat at all grades of steatosis in a cohort with a range of liver conditions.

Supporting Information

File S1 Table S1. Data reported in this study. Table S2. Summary of studies reporting diagnostic performance of comparable MRI approaches using a pathologist’s visual estimate of fat content as the reference standard. Table S3. Summary of studies comparing histologic fat estimates to MRI. (DOC)

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Author Contributions

Clinical assessment and review of subjects: EKG LAA OTA JKO. Conceived and designed the experiments: MJH EKG LAA OTA SJB JKO TGS. Performed the experiments: MJH EKG PSB JKO. Analyzed the data: MJH EKG SJB TGS. Contributed reagents/materials/analysis tools: PSB. Wrote the paper: MJH EKG LAA OTA SJB PSB JKO TGS.
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