# Helium-3 MR *q*-Space Imaging With Radial Acquisition and Iterative Highly Constrained Back-Projection

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An undersampled diffusion-weighted stack-of-stars acquisition is combined with iterative highly constrained back-projection to perform hyperpolarized helium-3 MR q-space imaging with combined regional correction of radiofrequency- and T<sub>1</sub>-related signal loss in a single breath-held scan. The technique is tested in computer simulations and phantom experiments and demonstrated in a healthy human volunteer with whole-lung coverage in a 13-sec breath-hold. Measures of lung microstructure at three different lung volumes are evaluated using inhaled gas volumes of 500 mL, 1000 mL, and 1500 mL to demonstrate feasibility. Phantom results demonstrate that the proposed technique is in agreement with theoretical values, as well as with a fully sampled two-dimensional Cartesian acquisition. Results from the volunteer study demonstrate that the root mean squared diffusion distance increased significantly from the 500-mL volume to the 1000-mL volume. This technique represents the first demonstration of a spatially resolved hyperpolarized helium-3 q-space imaging technique and shows promise for microstructural evaluation of lung disease in three dimensions. Magn Reson Med 63:41-50, 2010. © 2009 Wiley-

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In hyperpolarized (HP) helium-3 (He-3) MRI, diffusionweighted (DW) methods have proven useful in evaluating the progression of degenerative lung diseases such as emphysema (1,2), as well as for detecting early emphysematous changes in asymptomatic smokers (3). Due to breathhold time limitations, the conventional approach to DW HP He-3 imaging is to acquire one DW image and one non-DW image and to calculate the apparent diffusion coefficient (ADC). A potential drawback of this approach is that the ADC depends on the level of diffusion weighting applied. q-Space imaging (QSI) (4) is a generalization of ADC imaging in which images acquired at multiple diffusion weightings are used to compute diffusion probability profiles. This provides a measurement of the level of restriction independent of the diffusion weightings used. However, due to the longer scan times needed to acquire

multiple DW images and sensitivity to motion, QSI has typically been limited to neuro-MRI applications such as for resolving the orientations of crossing white matter tracks in the brain (5).

Recently, QSI has also been applied to HP He-3 NMR spectra acquired over the whole human lung (6). A bigaussian function was shown to fit the measured displacement probability profile. It was noted that the measured root mean squared (RMS) values of smaller and larger gaussian terms correlated well with previously reported average diameters of alveoli and terminal and respiratory bronchioles, respectively. Moreover, a recent proposed modification to the QSI diffusion model (7) that accounts for the short diffusion time condition ( $\Delta \approx \delta$ ) has been shown to give accurate measurements in the case of free gaussian diffusion and improved measurements compared to the unmodified analysis in the case of restricted diffusion. This suggests that it may be possible to obtain measurements of microscopic lung structures with QSI. However, the constraints of imaging within a breath-hold and with finite HP magnetization have prevented this technique from being applied to spatially resolved HP He-3 imaging of the lung in vivo.

To meet these challenges, a radially undersampled stack-of-stars acquisition, combined with an iterative constrained reconstruction, is proposed. Undersampling in k-space allows the acquisition of fewer k-space lines per reconstructed image (8). This reduces both scan time and the amount of magnetization needed per image. However, as the level of undersampling increases, the amount of streak artifact is also expected to increase.

The highly constrained back-projection reconstruction (HYPR) was shown to mitigate streak artifact in the presence of angular undersampling in sparse image data such as dynamic contrast-enhanced angiography (9). The success of HYPR in accelerated angiography using undersampled projection acquisition is the motivation for applying it to HP He-3 QSI of the lung. However, DW HP He-3 MRI of the human lung presents a greater challenge to the HYPR reconstruction than angiography due to decreased spatial and parametric (i.e., time or diffusion weighting) sparsity. Unlike angiographic images, which have relatively few vascular structures in the field of view that enhance gradually in time, DW HP He-3 images contain signal throughout the lungs that may change abruptly with each diffusion weighting. The iterative HYPR (I-HYPR) reconstruction (10) has been shown to be more robust to decreased sparsity. Therefore, in the present work, I-HYPR is used in order to reconstruct more accurate quantitative parametric images from the undersampled data.

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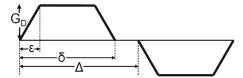


FIG. 1. The diffusion weighting gradient is a bipolar trapezoidal gradient with height,  $G_D$ , ramp time,  $\epsilon$ , lobe time,  $\delta$ , and diffusion time,  $\Delta$ .

In this work, the undersampled q-space methodology is tested in computer simulations and phantom experiments. Feasibility is demonstrated in a healthy human volunteer for three-dimensional (3D) QSI in a 13 sec breath hold that includes simultaneous correction for signal loss due to radiofrequency (RF) and  $T_1$  over the duration of the breath hold.

#### **THEORY**

## q-Space

QSI is performed by acquiring images of an object with multiple diffusion weightings. The degree of diffusion weighting can be expressed by a *b*-value and depends on the shape and size of the diffusion sensitizing gradient pulse. For a bipolar trapezoidal pulse (Fig. 1), the *b*-value, *b*, is given by (11):

$$b = \gamma^2 G_D^2 \cdot \left[ \delta^2 \left( \Delta - \frac{\delta}{3} \right) + \epsilon \left( \delta^2 - 2\Delta \delta + \Delta \epsilon - \frac{7}{6} \delta \epsilon + \frac{8}{15} \epsilon^2 \right) \right]$$
 [1]

where  $\gamma$  is the gyromagnetic ratio,  $G_D$  is the diffusion gradient strength,  $\epsilon$  is the ramp time,  $\delta$  is the pulse duration of a single lobe, and  $\Delta$  is the time between the beginning of the first lobe and the beginning of the second, also known as the diffusion time. Alternatively, diffusion weighting for a bipolar trapezoidal pulse can be specified by a vector,  $\tilde{q}$ , given by (12):

$$\vec{q} = \frac{\gamma G_D}{2\pi} (\delta - \varepsilon) \hat{G}_D$$
 [2]

where  $\hat{G}_D$  is a unit vector pointing in the direction of the diffusion weighting gradient. Let the expected value of the measured signal at a particular  $\vec{q}$  and for a given diffusion time,  $\Delta$ , normalized by the non-DW signal be  $E[\bar{q},\Delta]$ . Now  $E[\bar{q},\Delta]$  is given by the integral:

$$E[\vec{q}, \Delta] = \int DPP[\vec{R}, \Delta] \exp[i2\pi \vec{q} \cdot \vec{R}] d\vec{R}$$
 [3]

where the probability that a spin will undergo a displacement  $\vec{R}$  in the diffusion time,  $\Delta$ , is defined as the diffusion probability profile,  $DPP[\vec{R},\Delta]$ , and the frequency coefficient of the complex term depends on the projection of the displacement vector onto  $\vec{q}$ . Note that  $\vec{q}$  has units of in-

verse length and is the Fourier conjugate variable to the average displacement vector  $\vec{R}$ . Thus, the DPP can be calculated by performing the Fourier transform on the measured DW signal.

The key assumption needed for Eq. 3 to hold is that the width of the diffusion pulse be much smaller than the diffusion time ( $\delta \ll \Delta$ ). This ensures that nearly all of the measured diffusion takes place between the diffusion pulses and not during them. This assumption, in the interest of performing QSI in a short breath-hold scan, is clearly violated for the bipolar trapezoidal pulse where  $\delta =$  $\Delta$  (Fig. 1). Violating this assumption results in a measured DPP that is narrower than the ideal DPP of the system (13-15). A recent proposed modification to the QSI methodology has been shown to give accurate measured DPP's in the case of free gaussian diffusion and improved measured DPP's compared to the unmodified analysis in the case of restricted diffusion (7). Two modifications are given. First, the q-values are multiplied by a constant factor, depending on the relative values of  $\Delta$  and  $\delta$  accord-

$$\vec{q}_{\mathrm{mod}} = \vec{q} \sqrt{\frac{\Delta - \delta/3}{\Delta + \delta}}$$
 [4]

Second, the results should be interpreted at the experimental diffusion time,  $t_{exp}$ , given by,

$$t_{\rm exp} = \Delta + \delta \tag{5}$$

In this work, computations will be performed using the modified values given by Eq. 4 and Eq. 5.

#### I-HYPR

Details of the I-HYPR reconstruction have been published previously (10). Briefly, I-HYPR applies constrained reconstruction based on a maximum-likelihood approach to reconstruct a parametric series of images from radially undersampled projection data. The radially undersampled reconstruction problem is addressed by using a constraining image reconstructed from data gathered from all or part of the parametric space. To enforce consistency with the measured data, the measured projections are normalized by synthesized projections of the constraining image at the identical projection angles to form a ratio sinogram. This ratio sinogram is reconstructed with unfiltered back-projection (FBP) and the resulting image is multiplied by the constraining image to produce a new estimate of the image. The new estimate is taken to be the constraining image for the next iteration and the process is repeated.

## MATERIALS AND METHODS

To demonstrate the accuracy of the I-HYPR reconstruction on DW data, it was first tested using a simulation synthesized from data acquired using a DW Cartesian sequence. The acquisition and reconstruction methods were then tested in phantom experiments for a full q-space experiment using multiple b-value weightings acquired within a rapid 3D acquisition designed for single breath-hold imag-

ing. Finally, the experimental sequence and reconstruction were tested in a normal human subject undergoing different inflation lung volumes to establish quantitative agreement with previous studies and demonstrate expected qualitative changes in alveolar and small airway structures as a function of lung volume.

#### Simulation

To demonstrate the feasibility of calculating quantitative parametric maps in the lungs using I-HYPR, a radial DW He-3 MR data set was simulated from He-3 MR images acquired with a Cartesian trajectory. Sinogram data were synthesized by performing the radon transform on the weighted and unweighted images from a single  $128 \times 128$ pixel slice positioned in a region of known gas trapping, using 200 angles for the unweighted image and trials of 4, 8, 16, 32, and 200 angles for the weighted image. To simulate realistic levels of noise, independent realizations of zero mean gaussian noise with varying standard deviation were added in quadrature to the sinogram. The noisy sinogram was then calculated by taking the square root of the sum of the two squared channels. The unweighted image was calculated by FBP. The weighted images were calculated with I-HYPR and FBP for comparison. ADC maps were calculated by:

$$ADC = -\frac{1}{b} \ln \left( \frac{S_{DW}}{S_0} \right)$$
 [6]

where  $S_{DW}$  and  $S_0$  are the diffusion weighted and unweighted signals respectively and b is the b-value (1.6 sec/cm<sup>2</sup>). To compare the noise levels in the simulations, the noise level was quantified by dividing the standard deviation of a region in the background of the sinogram by the mean value of the entire sinogram and multiplying by 100. These values are reported as percentage of noise.

To quantitatively evaluate the errors in the DW images reconstructed with I-HYPR and the corresponding ADC maps, they were compared to the fully sampled noise-free FBP-reconstructed DW image and the corresponding ADC map. The percentage of RMS difference between the DW images reconstructed with I-HYPR and the fully sampled FBP image was computed over the region of support of the object for each noise level and number of simulated angles. Restricting the calculation to the region of support eliminates the influence of pixels in the noise background with very low signal values. The RMS difference in the ADC images was also calculated in this way. Additionally, the mean and standard deviation in two separate  $10 \times 10$  pixel regions of interest (ROI) were computed for each noise level and number of simulated angles in both the DW and ADC images.

## **Phantom Construction**

The phantom experiments were carefully designed to test the ability of the technique to measure expected diffusion dimensions for different levels of restriction ranging from unrestricted, partially restricted, to fully restricted. Two phantoms were constructed from 5-cm-inner-diameter PVC pipe sealed at both ends with end caps. One end cap

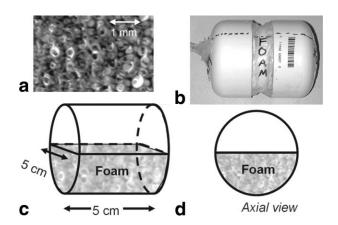


FIG. 2. Phantom 1 consisting of unrestricted and partially restricted regions. **a:** A high-resolution optical scan of the porous foam material used to construct the phantom. **b:** A photograph of the exterior of the phantom, with schematic diagrams (**c,d**) depicting the free-diffusion region (**c,d** top) and the foam-containing region (**c,d** bottom).

on each of the phantoms had a hole fitted with 1/8-inchinner-diameter Tygon tube (US Plastic Corp., Lima, OH) to allow evacuation and filling. Phantom 1 contained a piece of synthetic foam cut in the shape of a half-cylinder to occupy half of the space in the phantom for a net volume of 180 mL. The average separation of structures in the foam was estimated with high-resolution optical scanning by measuring the approximate diameter of 20 of the circular structures in the scanner image (Fig. 2a). The approximate separation was found to be 270  $\pm$  60  $\mu$ m (Fig. 2). It should also be noted that the pores in the foam material are highly connected and do not provide complete restriction of the gas. Phantom 2 contained a stack of 290-µm-thick acetal plastic plates separated by one plate thickness (Fig. 3). However the plate separation was not completely uniform due to deflection of the thin plates (Fig. 3c) and was measured with high-resolution optical scanning to be  $280 \pm 100 \ \mu m$ .

## **HP Gas Production**

HP  $^3$ He gas was polarized using a prototype commercial spin-exchange optical pumping polarizer (HeliSpin; GE Healthcare, Milwaukee, WI) to polarize  $^3$ He to 30-40%. For human studies, a "dose" of HP  $^3$ He was prepared by drawing  $^3$ He into a Tedlar plastic bag (Jensen Inert Inc., Coral Springs, FL) from the polarizing unit. The Tedlar bag was purged and rinsed with nitrogen prior to filling to prevent oxygen contamination. The desired volume was obtained by adding nitrogen to the bag. For phantom studies, the  $^3$ He was drawn from the polarizing unit directly into the phantom via an attached 1/8-inch-diameter Tygon tube (US Plastic Corp.). Phantoms were evacuated and purged with nitrogen several times before filling with  $^3$ He to 1.0  $\pm$  0.1 atm.

## HP He-3 MRI

DW HP He-3 MRI was performed on the phantoms and a healthy volunteer using a 1.5-T MR scanner (Signa HDx; 44 O'Halloran et al.

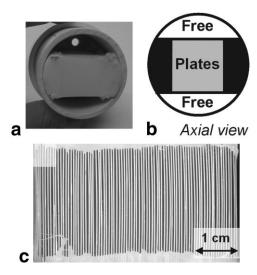


FIG. 3. Phantom 2 consisting of unrestricted and highly restricted regions. **a:** A photograph of the inside of the phantom before the second end cap was attached. **b:** A schematic diagram of the phantom corresponding to the view in the photograph in **(a). c:** A high-resolution optical scan of the parallel plates used in the phantom.

GE Healthcare) with a transmit/receive chest RF coil tuned to the He-3 resonant frequency. Scans were performed over a 3D volume consisting of eight axial slices at a 3-cm thickness, field of view = 42 cm,  $\pm$  62.5 kHz bandwidth, nominal flip angle of 2° and in-plane resolution of 3.3 mm<sup>2</sup>, using a fast-GRE 3D stack-of-stars sequence with radial acquisition in the axial plane and conventional phase encoding in the superior-inferior (z-axis) direction (8). All phase encodes for a given projection were acquired before proceeding to the next projection. Phase-encoding was chosen over two-dimensional (2D) multislice excitation in order to reduce the scan time using a short-duration nonselective RF pulse. A total of 256 unique projection angles were acquired per phase encode, with 16 projections acquired per each of the eight nonzero q-values and 16 projections without diffusion weighting acquired between each set of progressively decreasing DW projections (Fig. 4). Nonweighted projections were acquired at intervals evenly interspersed throughout the breath hold to correct the diffusion-weighted data for signal loss due to flip angle and  $T_1$ -decay, as described later. Diffusion-encoding pulses were bipolar trapezoidal pulses oriented along the z-axis (Fig. 1).

Two different Signa HDx 1.5-T (GE Healthcare) scanners were used to perform the experiments. Due to differences in the maximum achievable gradient strength on the two scanners, the human volunteer and phantom 1 were scanned with different parameters than phantom 2. For the volunteer and phantom 1, the maximum diffusion gradient strength,  $G_d$ , was 3.2 G/cm, resulting in q-values ranging from 0.18 to 1.5 mm<sup>-1</sup> ( $q_{mod} = 0.10$  to 0.84 mm<sup>-1</sup>), $\epsilon = 300$   $\mu$ s,  $\delta = 1700$   $\mu$ s,  $\Delta = 1700$   $\mu$ s, pulse repetition time/echo time = 6.1/4.2 ms, and total scan time 13 sec. For phantom 2, the maximum diffusion gradient strength,  $G_d$ , was 1.9 G/cm, resulting in q-values ranging from 0.15 to 1.2 mm<sup>-1</sup> ( $q_{mod} = 0.088$  to 0.71 mm<sup>-1</sup>),  $\epsilon = 300$   $\mu$ s,  $\delta = 2300$   $\mu$ s,  $\Delta =$ 

 $2300~\mu s,$  pulse repetition time/echo time = 7.2/5.4 ms, and a total scan time of 15 sec.

To verify the accuracy of the 3D acquisition and reconstruction in comparison to a conventional fully sampled Cartesian method, an additional set of eight 2D fast-GRE DW HP He-3 MRI scans of phantom 1 was acquired. Each of the eight 2D images was acquired at a q-value corresponding to the eight q-values acquired in the 3D scan. The 2D scans were performed on a single 3-cm-thick slice with 20-cm field of view, 256  $\times$  256 image matrix, and 10° flip angle.

The human subject experiment was approved by our institutional review board and performed under IND#64867. To investigate the effect of varying the inhaled volume on *q*-space results, the human volunteer underwent MRI following inhalation of a gas mixture consisting of 400 mL He-3 at 30% polarization, mixed with different volumes of nitrogen (100 mL, 600 mL, and 1100 mL), for a total of three scans with total inhaled gas volumes of 500 mL, 1000 mL, and 1500 mL.

#### Reconstruction

Each set of 16 consecutively acquired angles was reconstructed using I-HYPR, resulting in eight unweighted and eight diffusion-weighted images for each of the eight slices, for a total of 128 images per scan. Iteration was terminated after the RMS difference between the sinograms of the reconstructed images and the measured data ceased to decrease or after 50 iterations, whichever condition was met first. The constraining images were reconstructed using all 256 unique projections acquired for a given slice.

## RF and $T_1$ Correction

To obtain quantitative diffusion measurements, it is necessary to correct for the confounding effect of signal loss

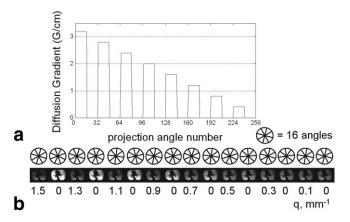


FIG. 4. **a:** The maximum of the diffusion gradient as a function of projection number for each slice. Each group of 16 consecutively acquired angles shares a common diffusion weighting and was reconstructed to form a unique image. Note that there are eight such groups with 0 diffusion weighting. These were reconstructed to form eight unique unweighted images that were fitted to obtain the flip angle maps. Another representation of this **(b)** depicts the reconstructed image, indexed by a *q*-value beneath the representation of the acquired angle set used to reconstruct that image.

due to the applied RF pulse and  $T_1$ . This correction must be performed pixelwise due to the inhomogeneity of the amplitude of radiofrequency field for the flexible RF-coil design used, which wraps snugly around the chest. To obtain a correction map, the signal in each pixel from the eight unweighted images,  $S_n$ , was fit to the model (6):

$$S_n = S_0 \cdot K^n \tag{7}$$

where  $S_0$  is the signal in the first image and n is the number of RF pulses applied between  $S_0$  and  $S_n$  and K is a constant that can be modeled as

$$K = \cos(\alpha)\exp(-TR/T_1)$$
 [8]

where  $\alpha$  is the flip angle and  $T_1$  is the relaxation time, dominated by the effect of paramagnetic molecular oxygen. The signal intensities for all images were then corrected by inverting Eq. 7.

#### Data Analysis

Using the modified q-values,  $q_{mod}$ , the normalized signal intensity  $E[\vec{q}_{mod}, \Delta]$  was calculated by dividing the diffusion-weighted images by the unweighted images, where the unweighted images were the average of the eight flipangle-corrected unweighted images for each slice. The  $E[\vec{q}_{mod}, \Delta]$  was then fit to the gaussian model:

$$E[\vec{q}_{\text{mod}}, \Delta] = Z \exp(-2\pi X_{RMS}^2 \vec{q}_{\text{mod}}^2)$$
 [9]

where  $X_{RMS}$  is the RMS width of the DPP. This is equivalent to taking the Fourier transform of  $E[\vec{q}_{mod}, \Delta]$  to obtain the DPP and fitting to:

$$DPP(x) = Z \exp\left(-0.5 \cdot \left(\frac{x}{X_{RMS}}\right)^2\right)$$
 [10]

Fitting to Eq. 9, however, is preferred because this approach is more robust to apodization error caused by the limited region sampled in *q*-space. Note that in the present work a single gaussian model is used instead of the bigaussian model used in the previous work by Shanbhag et al. (6). This is because the four-parameter bi-gaussian model would not provide a robust fit to the eight data points used for the fitting in this study.

Theoretical  $X_{RMS}$  values for unrestricted gaussian diffusion were calculated using the experimental diffusion time,  $t_{exp}$ , and the relation,

$$X_{RMS} = \sqrt{2Dt_{\rm exp}}$$
 [11]

with the diffusion coefficient of pure helium, D = 1.9 cm²/sec (16) for comparison with phantom experiments and the diffusion coefficient of helium diluted in air, specifically D = 0.88 cm²/sec, for comparison with the volunteer scans. For restricted diffusion in the phantoms and lung parenchyma in the human volunteer,  $X_{RMS}$  values were compared with the known average dimensions of the acetate plates, the foam, and values for alveolar and small airway dimensions in the literature (17).

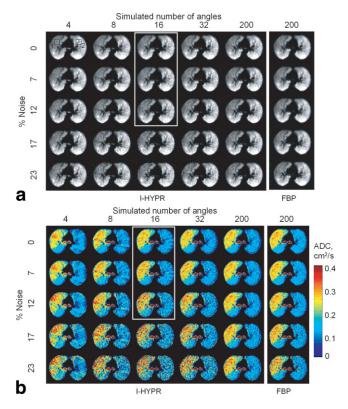


FIG. 5. **a:** DW images reconstructed with I-HYPR compared to the respective FBP images for various noise levels. **b:** ADC maps calculated from simulated radial data using I-HYPR on the weighted image for a range of undersampling and added noise compared to the ideal ADC map. Note that as the number of projections increases from left to right that the apparent streak artifact is reduced. Also the I-HYPR reconstruction appears to be less sensitive to noise than the FBP reconstruction. Black squares labeled "1" and "2" in the upper left of (a) depict the location of ROIs 1 and 2 used in the ROI calculations presented in Table 1. The white border indicates the operative range for volunteer studies.

#### **RESULTS**

## Simulation

The DW spin density images reconstructed using I-HYPR are compared in Fig. 5a to the DW images reconstructed using FBP for each respective noise level. Because the constraining image used is the unweighted image, the anatomy appears to be well represented in all cases, although structured artifacts resembling streaks are more apparent in the most extreme case of radial undersampling (four projections), particularly as noise level increases.

Fig. 5b shows the ADC maps computed with the diffusion-weighted images in Fig. 5a. The effects of undersampling and noise stemming from the reconstruction error are more apparent in the ADC maps than in the DW images. To evaluate the accuracy of the undersampled reconstruction, the ADC maps calculated from the images reconstructed using I-HYPR are compared with those using FBP reconstruction with 200 equally spaced projection angles. Two expected trends are apparent. The noise and artifact level in the reconstruction increases with the level of noise in the projection data from top to bottom, and the level of streak artifact decreases as the number of sampled angles

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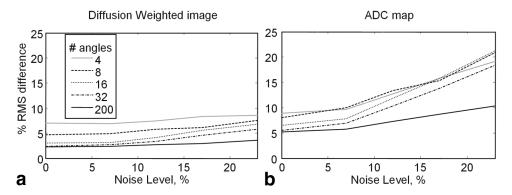


FIG. 6. **a:** The RMS difference plotted vs noise level for the DW images reconstructed with angularly undersampled I-HYPR (with four, eight, 16, 32, and 200 projection angles) and FBP using a fully sampled noise-free set of 200 angles. **b:** The RMS difference between the ADC images calculated using I-HYPR-reconstructed images and fully sampled noise-free FBP images. Note that the increased error in the ADC map is due to the logarithm operation. Note there is residual error even for full angular sampling due to the inexact nature of the I-HYPR reconstruction.

increases from left to right. Importantly, the I-HYPR maps are qualitatively the same as the exact FBP result using as few as 16 projection angles. This robustness of I-HYPR to angular undersampling can be exploited to accelerate acquisition speed while retaining accuracy for diffusion measures. Finally, note that a small amount of spatial blurring is also observed due to the linear interpolation used in the radon transform. The white box indicates the empirically selected level of angular undersampling chosen for the subsequent human volunteer studies.

In Fig. 6 the percentage of RMS difference between the DW images reconstructed with I-HYPR and the fully sampled noise-free FBP-reconstructed image is plotted vs the noise-level in varying numbers of angular undersampling (Fig. 6a). In all cases the error was below 10%. For noise levels below 15% and for 16 or greater simulated angles, the error was below 5%. Figure 6b shows a similar plot for the ADC maps calculated with I-HYPR-reconstructed DW images. The errors in the ADC images are greater than that in the DW images due to the logarithm operation performed in calculating the ADC (Eq. 6). Note that in the limiting case of zero added noise and full angular sampling there is still approximately 2% difference between the DW images reconstructed with I-HYPR and the DW image reconstructed with FBP. This is due to the fact that the I-HYPR reconstruction is fundamentally a statistical reconstruction and is not an exact reconstruction such as FBP. The approximate nature of the reconstruction manifests itself in the ADC maps as increased noise and, in cases of extreme undersampling, as structured artifacts.

The mean and standard deviations of both the DW image and the ADC map in the two ROIs considered in Figure 5 are given in Table 1. For both the DW image and the ADC map, the means of the I-HYPR reconstructions compare quite well with those in the fully sampled FBP image. Without noise, the standard deviations of the ROIs in the I-HYPR images are comparable to those in the FBP image without noise. As expected, the standard deviations were observed to increase with increased added noise.

## Phantom Studies

The phantom experiment results are summarized in Table 2 and Fig. 7. The free and restricted diffusion regions were

each segmented by visual inspection.  $X_{RMS}$  measured in the free diffusion region of phantom 1 for the conventional 2D Cartesian technique and the new proposed 3D stack-of-stars experiments were 1200  $\pm$  200  $\mu$ m and 1170  $\pm$  20  $\mu$ m, respectively (Table 2). These values are in good agreement with the theoretical values for  $X_{RMS}$  for unrestricted gaussian diffusion of 1140  $\mu$ m calculated with Eq. 11, using the experimental diffusion time  $t_{exp}=3400~\mu s$  and D = 2.05 cm²/sec. The measured signal for the free diffusion region in the conventional 2D and the 3D radial cases compares well with the theoretical expected values for the signal attenuation from unrestricted gaussian diffusion below about  $q_{mod}=0.5~\rm mm^{-1}$  (Fig. 7b, dotted line).

 $X_{RMS}$  measured in the foam region of phantom 1 from the conventional 2D Cartesian and proposed 3D stack-of-stars acquisitions were 800  $\pm$  100  $\mu m$  and 840  $\pm$  10  $\mu m$ , respectively (Table 2). These values are substantially greater than the measured average pore size in the phantom of 270  $\pm$ 60 µm; however, this discrepancy is expected, given the fact that the pores in the foam material are highly connected and therefore do not represent a completely restricted diffusion boundary to the gas. As a comparison to conventionally fully sampled data, the log of the normalized diffusion-attenuated signal for the 2D and proposed 3D acquisitions is compared across the measured range of q-values for ROI's placed in the foam region (Fig. 7b) and in the free diffusion region (Fig. 7c). Measured signal values for the nine separate 2D conventional Cartesian acquisition (circles) lie within errors of the measured signal values for the single 3D undersampled radial stack of stars acquisition (line).

 $X_{RMS}$  measured in the free diffusion region of phantom 2 for the proposed 3D stack-of-stars experiments was 1390  $\pm$  80  $\mu$ m (Table 2). This value is in good agreement with the theoretical value for  $X_{RMS}$  for unrestricted gaussian diffusion of 1320  $\mu$ m calculated with Eq. 10 using the experimental diffusion time  $t_{exp}=4600~\mu s$  and D = 2.05 cm²/sec. Note that the theoretical value for free diffusion differs from that in phantom 1 due to the longer diffusion time,  $\Delta$ , required by the MR scanner gradients for the experiment with phantom 2.  $X_{RMS}$  measured in the restricted region of phantom 2 from the proposed 3D stack-of-stars acquisition was 126  $\pm$  40  $\mu$ m (Table 2). This compares well with the

Table 1
ROI Measurements for the Computer Simulation

	Noise %	I-HYPR		DW Image (a.u.)		FBP	
		Angles: 4	8	16	32	200	200
ROI 1	0	92 (10)	94 (11)	93 (11)	94 (11)	94 (11)	93 (10)
	7	91 (11)	93 (12)	92 (12)	93 (11)	93 (11)	93 (10)
	12	93 (12)	93 (14)	92 (16)	92 (13)	93 (12)	93 (10)
	17	90 (13)	90 (17)	93 (20)	91 (16)	93 (13)	93 (11)
	23	93 (18)	92 (22)	92 (18)	91 (21)	92 (14)	93 (12)
ROI 2	0	141 (31)	142 (33)	141 (32)	141 (33)	141 (33)	140 (29)
	7	141 (31)	141 (34)	140 (32)	140 (34)	140 (33)	140 (29)
	12	141 (32)	140 (34)	141 (36)	139 (34)	140 (34)	141 (30)
	17	139 (31)	140 (35)	138 (36)	140 (37)	140 (33)	140 (29)
	23	140 (33)	140 (36)	138 (40)	139 (38)	139 (33)	140 (29)
			ADO	C (cm²/sec)			
ROI 1	0	0.26 (0.03)	0.26 (0.03)	0.26 (0.03)	0.26 (0.02)	0.26 (0.02)	0.26 (0.02)
	7	0.27 (0.04)	0.26 (0.04)	0.27 (0.03)	0.27 (0.03)	0.26 (0.02)	0.26 (0.02)
	12	0.26 (0.04)	0.27 (0.05)	0.27 (0.06)	0.27 (0.05)	0.27 (0.03)	0.27 (0.03)
	17	0.28 (0.05)	0.28 (0.07)	0.27 (0.08)	0.28 (0.06)	0.27 (0.03)	0.26 (0.04)
	23	0.27 (0.07)	0.28 (0.09)	0.27 (0.09)	0.28 (0.08)	0.27 (0.04)	0.26 (0.05)
ROI 2	0	0.11 (0.02)	0.11 (0.02)	0.11 (0.02)	0.11 (0.02)	0.11 (0.03)	0.11 (0.02)
	7	0.11 (0.03)	0.11 (0.03)	0.11 (0.03)	0.12 (0.03)	0.11 (0.03)	0.11 (0.02)
	12	0.11 (0.03)	0.11 (0.03)	0.11 (0.05)	0.12 (0.04)	0.11 (0.03)	0.11 (0.02)
	17	0.12 (0.04)	0.12 (0.05)	0.12 (0.05)	0.12 (0.05)	0.11 (0.03)	0.11 (0.03)
	23	0.11 (0.05)	0.11 (0.07)	0.12 (0.06)	0.12 (0.06)	0.12 (0.04)	0.11 (0.04)

The mean value in the ROI for images shown in Figure 5 is reported with the standard deviation in parentheses. The top half of the table gives ROI measurements of signal intensity in the DW images and the lower half gives the ROI measurements in the ADC images.

ideal  $X_{RMS}$  value of 118  $\mu$ m predicted for an infinitely restricted 290- $\mu$ m-wide gap, as well as the measured value of 110  $\pm$  40  $\mu$ m that corresponds to the measured gap dimension of 275  $\pm$  100  $\mu$ m.

## Human Volunteer

In Fig. 8a, corrected maps of  $X_{RMS}$  are shown for each inhaled volume in the human volunteer examination. A decrease in  $X_{RMS}$  from anterior to posterior can be observed on the corrected  $X_{RMS}$  maps from each of the three inhaled volumes (Fig. 8a). This is the expected dependence due to the effect of gravity as the volunteer was in the supine position for imaging. A decrease in  $X_{RMS}$  from the apex to base of the lung is also apparent in each of the three volumes.

Figure 8b depicts the regional distribution of the correction factor, K, for each of the inhaled volumes. The correction maps are qualitatively similar for each volume (Fig. 8b), with

lower values of K observed in the central slices and higher values of K in the apical and basal slices. This dependence is due to the coil sensitivity profile. The small regional differences observable are to be expected because the sensitivity profile of the flexible coil depends on the position of the volunteer's chest and hence on the inhaled volume. Moreover, the regional distribution of oxygen partial pressure likely varies slightly between each inhaled volume.

Plots of  $X_{RMS}$  measured in ROIs covering the entire right and left lung parenchyma demonstrate the dependence of inhaled volume on apical to basal position (Fig. 9). The difference in the measured  $X_{RMS}$  between the 1500-mL volume and the 500-mL volume was approximately 22  $\mu$ m, while the difference between  $X_{RMS}$  in the 1500-mL volume and the 1000-mL volume did not appear to be significant. As an internal comparison to theory, the  $X_{RMS}$  values measured in ROI's containing the trachea were 780  $\pm$  40  $\mu$ m, 790  $\pm$  60  $\mu$ m, and 780  $\pm$  40  $\mu$ m for the

Table 2
Summary of Phantom Experiments

	2D	3D	Ideal
Phantom 1 (foam)			
X <sub>RMS</sub> foam (μm)	800 (100)	840 (10)	n/a <sup>a</sup>
$X_{RMS}$ free ( $\mu$ m)	1200 (200)	1170 (10)	1140
Phantom 2 (parallel plates)			
X <sub>RMS</sub> plates (μm)		126 (40)	118 <sup>b</sup>
X <sub>RMS</sub> free (μm)		1390 (80)	1320

<sup>&</sup>lt;sup>a</sup>Due to the complex structure of the foam phantom, no ideal  $X_{BMS}$  value can be determined.

<sup>&</sup>lt;sup>b</sup>This is the  $X_{RMS}$  value expected for a 290- $\mu$ m gap with infinitely impermeable walls.

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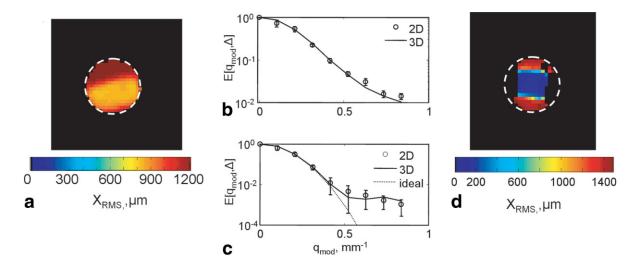


FIG. 7. **a:**  $X_{RMS}$  map from phantom 1 showing the free diffusion region (upper) and the region of partially restricted diffusion in the porous foam (lower). **b:** The measured normalized diffusion attenuated signal,  $E[q_{mod}, \Delta]$ , in the foam portion of phantom 1 compared for the Cartesian 2D (circles) and the undersampled 3D method of this work (line). **c:** The expected signal attenuation for free gaussian diffusion (dotted line) is compared to the 2D (circles) and 3D (solid line) measurements. Note the deviation from ideal when the noise floor is reached. **d:**  $X_{RMS}$  map from phantom 2 showing the highly restricted diffusion region of the parallel plates (center, blue) and the free diffusion regions (above and below, red). In both (a) and (d), a white dotted line indicates the approximate position of the inner wall of the phantom.

500-mL, 1000-mL, and 1500-mL inhaled volumes, respectively. These values are in good agreement with the theoretical  $X_{RMS}$  calculated for unrestricted gaussian diffusion of 774  $\mu$ m, the experimental diffusion time  $t_{exp}=3400~\mu$ s, and D = 0.88 cm²/sec using Eq. 11. Single pixel profiles though the  $X_{RMS}$  maps for each inhaled volume are compared in Fig. 9c. Figure 8a depicts the position of the profile (red dashed line). The decrease in  $X_{RMS}$  from anterior to posterior is apparent in each slice. Also note the

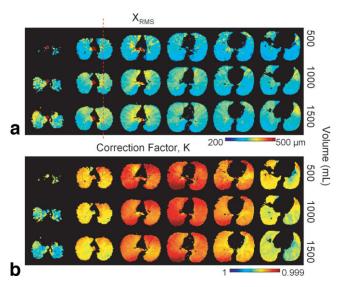


FIG. 8. Normal human volunteer study showing (a)  $X_{RMS}$  and correction maps from the three inhaled volumes are compared. Note the dependence of  $X_{RMS}$  on the apical to basal location and on the anterior to posterior location. b: The correction maps for all the images are compared and appear to be qualitatively similar. The differences are likely due to the effect of the volume change on the flexible coil. Note that the dashed line in (a) indicates the location of the profiles plotted in Fig. 9c

general increase in  $X_{RMS}$  value from the 500-mL volume (solid line) to the 1000-mL and 1500-mL volumes (dashed and dotted lines, respectively).

## **DISCUSSION**

This work demonstrates the feasibility of spatially resolved 3D QSI of human lung airspaces in vivo. This application of QSI in a 13-sec scan was enabled by radial undersampling and the I-HYPR-constrained reconstruction. Without accelerated acquisition, these measurements could not be performed in a reasonable breath-hold time. The I-HYPR reconstruction was tested in computer simulations and shown to provide accurate diffusion-weighted images, given the unweighted image as a constraining image. Phantom experiments demonstrated that the accelerated 3D QSI technique provides the same  $X_{RMS}$  values as a 2D Cartesian technique while enabling volume coverage. The RMS dimensions of the airspaces in a healthy human volunteer showed a dependence on inhaled volume, apical to basal, and anterior to posterior position in the lungs, and values agreed with previously published work (6).

Whole lung averages of the  $X_{RMS}$  for all inhaled volumes agree with the previously published results by Shanbhag et al. (6). They reported a weighted average of the measured  $X_{RMS}$  over the two fitted gaussian components in healthy adults to be 333  $\pm$  27  $\mu$ m. This compares well to the values in Table 3, which were calculated by averaging  $X_{RMS}$  values obtained from single gaussian fits over the entire lung. Note that the effect of changing the inhaled volume on the  $X_{RMS}$  is not significant at the whole volume level and can only be appreciated in the spatially resolved measurements.

Testing of the method in phantom experiments, ranging from unrestricted to intermediate and highly restricted diffusion, provided further support for the validity of the method. The modified q-space theory presented by Lori et

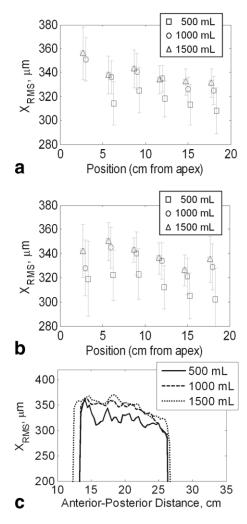


FIG. 9. Lung volume measures of average  $X_{RMS}$  values from the human volunteer for ROIs encompassing the entire left (a) and right (b) lungs depict the expected apical to basal dependence of structure size. Also the increase in  $X_{RMS}$  with inhaled volume can be seen. c: Anterior to posterior along dotted line in Fig. 8a profiles are compared for each inhaled volume. The anterior to posterior gradient is apparent for each inhaled volume.

al. (7) improved the accuracy of measurements of  $X_{RMS}$  based on modified q-values. Moreover, the concept of the experimental diffusion time,  $t_{exp}$ , from this same work aided in validating the present work by providing the correct diffusion time for the calculation of the expected free diffusion coefficients. The measured  $X_{RMS}$  for the free diffusion coefficient in the phantoms at two distinct diffusion times and the measured  $X_{RMS}$  in the trachea of the human volunteer were all in agreement with theory when calculated using  $t_{exp}$ . The  $X_{RMS}$  values measured in the restricted region of phantom 2, the acetal plastic plate phantom, are also in agreement with predicted values, providing further validation of the methodology in the opposite regimen of complete restriction along the measurement axis.

There are previously reported results in the literature on the effect of inhalation volume on lung microstructure size. In a rat model of emphysema, the ADC was shown to be greater after inhaling gas than at functional residual capacity (18). The observed effect was greater in control rats. Similarly it has been reported that the ADC in healthy human volunteers increased when the inhaled volume increased from 6% total lung capacity to 15% total lung capacity (19). However, a microscopy study of canine lung in vivo found that the number of alveoli in the microscope field increased with the inhaled volume, while the size of the alveoli remained fairly constant over a range of lung volumes from 20% to 80% of TLC (20). This led to the interpretation that as the inhaled volume increased, alveoli that were previously closed suddenly opened, accounting for the increased lung volume.

In the present study, an average increase in  $X_{RMS}$  over the whole lung of approximately 6.5% was observed when the inhaled volume increased from 500 mL to 1500 mL. If this increase is due strictly to balloon-like stretching of the airspaces, and a functional residual capacity of 3 L is assumed, then the expected percentage increase in alveolar length scale over this range is about 9%:

$$100 \cdot \left[ \left( \frac{3L + 1.5L}{3L + 0.5L} \right)^{\frac{1}{3}} - 1 \right] \approx 9\%$$
 [12]

Given that the observed increase in  $X_{RMS}$  of 6.5% is significantly lower than the 9% predicted by the expansion model, it is reasonable to speculate that alveolar recruitment may be taking place during inhalation. One possible explanation for this observation is that some stretching of the airspaces is occurring. Another is that additional paths of communication become available as neighboring alveolar units are recruited.

It is also interesting to note that most of the dependence is measured between 500 mL and 1000 mL, followed by very little change from 1000 mL to 1500 mL. This latter observation suggests that there may be a threshold lung volume beyond which there is limited change in the dimensions of the microstructural components of the lungs. Future work will explore these mechanics on a larger population of normal and asthmatic subjects, including pediatric patients.

#### **CONCLUSIONS**

The first 3D method for in vivo HP He-3 QSI of the human lung was presented and tested in phantom experiments and in computer simulations. The presented QSI method was completed within a 13-sec breath hold by using an accelerated acquisition with undersampled radial stack-of-stars k-space sampling in combination with the I-HYPR reconstruction. Measured  $X_{RMS}$  values were found to increase as the inhaled gas volume increased from 500 mL to

Summary of Human Experiments

	Human volunteer (inhaled volume)				
	500 mL	1000 mL	1500 mL	Ideal	
X <sub>RMS</sub> whole lung (μm)	320 (40)	340 (30)	340 (30)	_	
$X_{RMS}$ trachea ( $\mu$ m)	780 (40)	790 (60)	780 (40)	774	

1500 mL in a healthy volunteer. Future work includes examination of diseased populations with diseases such as asthma and emphysema with the QSI technique. In particular, the examination of small-scale structures in regions of gas trapping in asthma and fibrotic lung disease represents interesting future directions for this work.

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