

Dark blood T2-weighted imaging of the human heart with Al-assisted compressed sensing: a patient cohort study

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Background: Dark blood T2-weighted (DB-T2W) imaging is widely used to evaluate myocardial edema in myocarditis and inflammatory cardiomyopathy. However, this technique is sensitive to arrhythmia, tachycardia, and cardiac and respiratory motion due to the long scan time with multiple breath-holds. The application of artificial intelligence (AI)-assisted compressed sensing (ACS) has facilitated significant progress in accelerating medical imaging. However, the effect of DB-T2W imaging on ACS has not been elucidated. This study aimed to examine the effects of ACS on the image quality of single-shot and multi-shot DB-T2W imaging of edema.

Methods: Thirty-three patients were included in this study and received DB-T2W imaging with ACS, including single-shot acquisition (SS-ACS) and multi-shot acquisition (MS-ACS). The resulting images were compared with those of the conventional multi-shot DB-T2W imaging with parallel imaging (MS-PI). Quantitative assessments of the signal-to-noise ratio (SNR), tissue contrast ratio (CR), and contrast-to-noise ratio (CNR) were performed. Three radiologists independently evaluated the overall image quality, blood nulling, free wall of the left ventricle, free wall of the right ventricle, and interventricular septum using a 5-point Likert scale.

Results: The total scan time of the DB-T2W imaging with ACS was significantly reduced compared to the conventional parallel imaging [number of heartbeats (SS-ACS:MS-ACS:MS-PI) =19:63:99; P<0.001]. The $SNR_{myocardium}$ and $CNR_{blood-myocardium}$ of MS-ACS and SS-ACS were higher than those of MS-PI (all P values <0.01). Furthermore, the $CR_{blood-myocardium}$ of SS-ACS was also higher than that of MS-PI (P<0.01). There were significant differences in overall image quality, blood nulling, left ventricle free wall visibility, and septum visibility between the MS-PI, MS-ACS, and SS-ACS protocols (all P values <0.05). Moreover, blood in the heart was better nulled using SS-ACS (P<0.01).

Conclusions: The ACS method shortens the scan time of DB-T2W imaging and achieves comparable or even better image quality compared to the PI method. Moreover, DB-T2W imaging using the ACS method can reduce the number of breath-holds to 1 with single-shot acquisition.

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Introduction

Dark blood T2-weighted (DB-T2W) imaging with fat saturation is routinely used for cardiac morphological evaluation, which is valuable for the diagnosis of myocardial edema (1-3). The 2-dimensional (2D) dual inversion recovery fast spin echo sequence is conventionally used for cardiac DB-T2W morphological imaging (4), which acquires k-space data in multi-shot acquisition across multiple cardiac cycles. However, this technique is sensitive to arrhythmia, tachycardia, and cardiac and respiratory motion, which can lead to myocardial signal loss, reduced contrast, and motion artifacts (5). To resolve these limitations, parallel imaging (PI) can be used to decrease scanning time to 11 heartbeats per slice or improve the spatial resolution with high acceleration factors (6-8); however, this can be degraded by noise amplification and undersampling artifacts and leads to lower image quality and a decreased signal-to-noise ratio (SNR) (9).

Alternatively, compressed sensing (CS) facilitates rapid acquisition without perceptible deterioration of the SNR by performing a nonlinear iterative reconstruction (10-12). This has been used in a wide variety of biomedical research and clinical applications, including DB-T2W imaging (13). In the reconstruction, image information is recovered from undersampled k-space data by exploiting the sparsity in an appropriate transform domain. Combined with the techniques of PI and half Fourier (14-16), CS reconstruction can even achieve higher acceleration with a shorter scan time. However, insufficient sparseness may lead to noise-like aliasing artifacts with a very high acceleration factor.

In recent years, a convolutional neural network has been widely used in the field of artificial intelligence (AI) to reconstruct images with high quality, especially at a highly accelerated condition (17-21). Subsequently, a novel magnetic resonance (MR) acceleration solution, called AIassisted compressed sensing (ACS) (22), was proposed to further leverage the individual benefits of AI and CS and was approved by the US Food and Drug Administration (FDA) (23). The main architecture of ACS reconstruction is an integration of half Fourier, PI, CS, and AI in an iterative reconstruction procedure and has been validated in numerous applications, such as liver imaging (24) and kidney imaging (25). The feasibility of the DB-T2W imaging with the ACS method has also been reported to validate in previous abstracts (26-28). However, the effect of DB-T2W imaging with ACS remains to be explored.

In this study, we evaluated the effects of ACS on the image quality of single-shot (SS) and multi-shot (MS) DB-T2W imaging of edema by comparing it with conventional MS imaging with PI.

Theory

ACS reconstruction

The main architecture of ACS reconstruction is an integration of half Fourier, PI, CS, and AI in an iterative reconstruction procedure. The framework of this reconstruction is shown in Figure 1. Specifically, the AI module (Figure 1A) is trained using the ResNet network as reported previously (24), which can solve the problem of gradient disappearance while improving the performance and convergence of the network. The outputs of the AI module are then added into the iterative reconstruction as a new regularization term. Since its output is integrated into the iterative procedure, the uncertainty of the AI module is constrained by both mathematical constraints from the half Fourier, PI, and CS methods, as well as data consistency constraints from the original undersampled k-space. A diagram of the inline reconstruction is displayed in Figure 1B. Therefore, the ACS is defined as follows:

$$\arg\min_{x} \|Ex - y\|_{2}^{2} + \lambda_{1} \|\Phi x\|_{1} + \lambda_{2} \|x - x_{AI}\|_{1}$$
[1]

x denotes the image to be reconstructed, E denotes the production of Fourier encoding with a binary k-space sampling mask, y represents the acquired k-space data, Φ denotes the sparse transform (e.g., wavelet or total variation), and x_{AI} denotes the reconstructed image of the trained AI module, with the under-sampled k-space as input.

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Figure 1 Sequence diagram of dark blood T2-weighted imaging. (A) Multi-shot acquisition of the sequence. (B) Single-shot acquisition of the sequence. ECG, electrocardiogram; RF, radio frequency; IR, inversion recovery; Fat Sat, Fat Saturation.

Methods

DB-T2W sequence with MS acquisition

As shown in *Figure 1A*, a dual inversion recovery with MS fast spin echo readout acquisition (4) was implemented with ACS reconstruction to speed up the examination. An echo train length of 25 was applied in MS acquisition with ACS (MS-ACS) and with PI (MS-PI), but the MS-ACS method reduced the breath-holding time of 4 heartbeats for each slice due to the higher acceleration of ACS.

DB-T2W sequence with SS acquisition

As shown in *Figure 1B*, SS acquisition with the ACS method (SS-ACS) was designed to shorten the scan time of DB-T2W imaging to acquire full left ventricle (LV) coverage in a single breath-hold. In this method, a total of 49 phase encoding lines were acquired for each slice. To suppress the myocardial inhomogeneity and blurring due to the long acquisition, a variable flip angle strategy for refocusing pulses was designed as described previously (29) and was optimized according to

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Scanning parameters	MS-PI	MS-ACS	SS-ACS		
FOV, mm ²	400×320	400×320	400×320		
Voxel size, mm ²	1.95×1.56	1.95×1.56	1.95×1.56		
Slice thickness, mm	8.0	8.0	8.0		
TE, ms	69.68	63.44	82.8		
Refocus flip angle	180°	180°	Variable (70°~160°)		
Acceleration method	Pl	ACS	ACS		
Acceleration factor	2	3	3		
Shot number	Multi-shot	Multi-shot	Single-shot		
Acquisition slices	9	9	9		
Scan time per slice	11 heartbeats	7 heartbeats	-		
Total scan time	9×11 heartbeats	9×7 heartbeats	19 heartbeats		
Breath-hold, number	9	9	1		

CMR, cardiac magnetic resonance; MS-PI, multi-shot acquisition with parallel imaging; MS-ACS, multi-shot acquisition with artificial intelligence-assisted compressed sensing; SS-ACS, single-shot acquisition with artificial intelligence-assisted compressed sensing; FOV, field of view; TE, echo time; PI, parallel imaging; ACS, artificial intelligence-assisted compressed sensing.

the myocardial T1 and T2 relaxation times (T1 =1,200 ms and T2 =40 ms). Additionally, 1 R-R interval was skipped between each slice to allow for signal recovery between the inversion pulses.

Patient population

This study was conducted in accordance with the Declaration of Helsinki (as revised in 2013) and was approved by the institutional ethics board of Tongji Hospital, Tongji Medical College, Huazhong University of Science and Technology. Informed consent was obtained from all patients. From August to December 2021, 33 patients with clinical indications including cardiomyopathy, cardiac valve disease, myocarditis, and other cardiovascular diseases were prospectively recruited.

Image acquisition

All cardiac MR (CMR) scans were performed on a 3T scanner (uMR 790, United Imaging Healthcare, Shanghai, China) with a dedicated 24-channel cardiac coil. Before the scan, all patients were told to hold their breath after exhalation. The MS-PI protocol was performed before the other protocols, and the proposed MS-ACS and SS-ACS protocols were subsequently carried out. Nine short-axis

slices were acquired for each protocol to cover the entire left ventricle (LV) from the mitral valve to the LV apex. The scanning parameters for all protocols are listed in *Table 1*.

In addition, patients with heart failure were additionally distinguished to validate the feasibility of these protocols in patients that failed to perform the long breath-holding. Therefore, cardiac function was measured from images acquired from the continuous short-axial balanced steady-state free precession pulse (bSSFP) sequence (30) (echo time =1.32 ms, repetition time =2.84 ms, flip angle =55°, matrix =208×208, field of view =320×320). Patients with an LV ejection fraction (LVEF) <50% (31) and the clinical indication of heart failure were considered and included in the heart failure cohort.

Image analysis

The myocardium and blood pool regions of interest (ROIs) in the MS-PI, MS-ACS, and SS-ACS were contoured on a workstation by 2 CMR technicians (X.Y. and Y.L.). The myocardial signal intensity (SI) was measured at the septum. Blood signals were measured at the center of the blood pool in the left ventricle away from the papillary muscles and chordae tendineae. SI was obtained using the mean signal of the ROI, and noise was obtained by the standard deviation (SD) of the myocardium or blood pool.

 Table 1 Scanning parameters of the CMR imaging techniques

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For patients with myocardial edema, the edematous region areas were evaluated using commercial software cvi42, v. 5.3 (Circle Cardiovascular Imaging, Calgary, AB, Canada) according to the so-called 2SD method (32). The LV endocardium and epicardium were delineated at all slices, and the remote normal myocardium ROI was selected to calculate the mean and SD values. Next, myocardium with a signal intensity >2 SD were automatically detected as an edematous region. The total LV mass and edema myocardium mass were calculated, and the extent of edema was defined by the ratio of edema mass and LV mass (%LVM).

In addition, the SNR, contrast ratio (CR), and contrastto-noise ratio (CNR) were calculated to compare the images of the different protocols, and were defined as follows:

$$SNR_{myocardium} = \frac{SI_{myocardium}}{SD_{myocardium}}$$
[2]

$$CR_{blood, myocardium} = \frac{SI_{myocardium}}{SI_{blood}}$$
[3]

$$CNR_{blood, myocardium} = \frac{SI_{myocardium} - SI_{blood}}{SD_{blood}}$$
[4]

$$CR_{edema, myocardium} = \frac{SI_{edema}}{SI_{myocardium}}$$
[5]

$$CNR_{edema, myocardium} = \frac{SI_{edema} - SI_{myocardium}}{SD_{blood}}$$
[6]

In the qualitative evaluation, the images were blindly reviewed by 3 radiologists (L.H., L.R., and Z.Y.), all of whom had more than 5 years of experience in cardiovascular imaging. The overall image quality, visual quality of blood nulling, free wall of the right ventricle (RV), free wall of the LV, and the interventricular septum were evaluated using a 5-point ordinal Likert scale (33) as follows: 1, nondiagnostic; 2, poor; 3, fair; 4, good; and 5, excellent. In addition, motion-related artifacts and signal dropout in the LV free wall were evaluated in the heart failure cohort.

Statistical analysis

Statistical analysis was performed using SPSS (version 23.0; IBM Corp., Armonk, NY, USA). Continuous data are described as the mean \pm SD or median and interquartile range, and categorical data are presented as the frequency and percentage. Normality was evaluated using the Kolmogorov-Smirnov test, and homogeneity of variance was assessed using Levene test. Before analysis, the ordinal

Likert scores from the 3 radiologists and quantitative parameters calculated by 2 observers were averaged. Analysis of variance or the Friedman test was used to assess the differences between the ordinal Likert scores and quantitative parameters among MS-PI, MS-ACS, and SS-ACS. The differences in qualitative scores and quantitative values between the 3 protocols pairwise were respectively evaluated by paired t-tests or paired Wilcoxon signedrank test. Kendall W was used to assess the interobserver agreement of the scoring data. The intraclass correlation coefficient (ICC) was used to assess the interobserver agreement between $SNR_{myocardium}$, $CR_{blood,myocardium}$, CNR_{blood,myocardium}, CR_{edema,myocardium}, and CNR_{edema,myocardium}. The agreement between the edema extents obtained from MS-PI, MS-ACS, and SS-ACS was compared using Bland-Altman analysis. P<0.05 was considered statistically significant.

Results

All 33 patients successfully underwent the scans with both the conventional approach and the 2 proposed protocols. Among these, 9 patients (8 males 89%; aged 45.7 ± 14.0 years) were classified as patients with heart failure (LVEF $23.7\%\pm8.9\%$) and added into the heart failure cohort, including 8 (89%) cases of dilated cardiomyopathy and 1 (11%) case of hypertension heart disease. Moreover, 11 (33%) patients had myocardial edema, including myocardial infarction (n=5), hypertrophic cardiomyopathy (n=4), and myocarditis (n=2). The characteristics of the included patients are listed in *Table 2*.

The acquisition time per slice of MS-ACS was shorter than that of MS-PI ($6.1\pm1.0 vs. 9.7\pm1.7 s$; P<0.001). In addition, the entire acquisition time of SS-ACS was only about 17 s in a single breath-hold, while the other approaches were longer than 55 s with 9 breath-holds. In the 9 patients with heart failure, the breath-holding time for 1 slice was significantly shortened using MS-ACS compared to MS-PI ($5.5\pm1.0 vs. 8.8\pm1.5 s$; P<0.001), and the total acquisition time for 9 slices was shortened by 36% ($50.0\pm9.1 vs. 79.6\pm13.5 s$; P<0.001).

The quantitative and qualitative results are listed in *Table 3*. The $SNR_{myocardium}$, $CR_{blood,myocardium}$, and $CNR_{blood,myocardium}$ of SS-ACS were significantly higher than those of MS-PI (all P values <0.001) and MS-ACS (all P values <0.05). In addition, the $SNR_{myocardium}$ and $CNR_{blood,myocardium}$ of MS-ACS were statistically significant higher than those of MS-PI (both P values <0.001), whereas the $CR_{blood,myocardium}$ was not (P=0.143). The heart

Table 2 The patients' clinical characteristics

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Clinical characteristics	Value	
Sex, male	26 (79%)	
Age, years	40.5±14.8	
Height, cm	169.7±7.3	
Weight, kg	73.8±15.4	
BSA, m ²	1.83±0.22	
Heart rate, bpm	69.6±13.9	
Clinical diagnosis		
Pulmonary hypertension	1	
Cardiomyopathy	15	
Hypertensive heart disease	2	
Myocardial infarction	5	
Valvular heart diseases	2	
Myocarditis	2	
Other diseases	6	

The results are shown as the number (%) or mean \pm SD, as specified. Continuous data presented as mean \pm SD. BSA, body surface area; SD, standard deviation.

Table 3 Quantitative analysis and qualitative scoring data of all subjects

In the heart failure cohort, the images from a total of 4 (44%) patients showed obvious artifacts under MS-PI, while no

	0 ,			
Quantitative parameters/qualitative scores	MS-PI	MS-ACS	SS-ACS	P value [#]
Total acquisition time, s	87.5±16.3	55.2±9.2*	17.0±3.1* ^{/&}	<0.001
SNR _{myocardium}	65.2±28.1	80.7±30.6*	109.9±37.5* ^{/&}	<0.001
CR _{blood,myocardium}	14.7±5.8	15.6±5.0	18.0±5.7* ^{/&}	0.043
CNR _{blood,myocardium}	62.8±27.9	77.9±30.2 *	95.7±32.8* ^{/&}	<0.001
CR _{edema,myocardium} [†]	1.40±0.28	1.37±0.27	1.33±0.23	0.833
CNR _{edema,myocardium} *	20.0±17.5	27.3±24.2	29.7±29.7	0.624
Overall image quality	3.9±0.6	4.1±0.4*	4.0±0.4*	0.011
Blood nulling	4.6±0.6	4.7±0.5	4.8±0.5*	<0.001
RV wall visibility	3.8±0.5	3.9±0.5	3.8±0.6	0.399
LV wall visibility	3.9±0.6	4.4±0.5*	4.3±0.5*	<0.001
Septum visibility	4.4±0.6	4.5±0.5	4.6±0.5*	0.034

Data are presented as the mean ± SD.[†], CR_{edema,myocardium} was counted across 11 patients with myocardial edema; [‡], CNR_{edema,myocardium} was counted across 11 patients with myocardial edema; [#], P value was calculated by analysis of variance or the Friedman test among the 3 DB-T2W protocols; *, P<0.05 of the MS-ACS or SS-ACS methods, as compared to the MS-PI method; [&], P<0.05 between the MS-ACS and SS-ACS methods. MS-PI, multi-shot acquisition with parallel imaging; MS-ACS, multi-shot acquisition with artificial intelligence-assisted compressed sensing; SS-ACS, single-shot acquisition with artificial intelligence-assisted compressed sensing; SNR, signal-to-noise ratio; CNR, contrast-to-noise ratio; CR, contrast ratio; RV, right ventricle; LV, left ventricle.

Table 4 Interreader agreement of all patients							
Qualitative scores/quantitative parameters	MS-PI	MS-ACS	SS-ACS				
Overall image quality	0.735	0.767	0.614				
Blood nulling	0.718	0.715	0.729				
RV wall visibility	0.639	0.637	0.708				
LV wall visibility	0.668	0.762	0.659				
Septum visibility	0.669	0.693	0.738				
SNR _{myocardium}	0.773	0.771	0.763				
CR _{blood,myocardium}	0.924	0.812	0.946				
CNR _{blood,myocardium}	0.807	0.783	0.781				
CR _{edema,myocardium} [†]	0.927	0.960	0.961				
CNR _{edema,myocardium} ‡	0.816	0.833	0.883				

[†], CR_{edema,myocardium} was counted across 11 patients with myocardial edema; [‡], CNR_{edema,myocardium} was counted across 11 patients with myocardial edema. MS-PI, multi-shot acquisition with parallel imaging; MS-ACS, multi-shot acquisition with artificial intelligence-assisted compressed sensing; SS-ACS, single-shot acquisition with artificial intelligence-assisted compressed sensing; SNR, signal-to-noise ratio; CNR, contrast-to-noise ratio; RV, right ventricle; LV, left ventricle.

artifact was seen on the MS-ACS images. *Figure 2* illustrates an example of artifact comparison from the heart failure cohort in which an obvious artifact was observed on the MS-PI image, whereas no artifact was seen on the MS-ACS image.

A comparison of the MS-PI and SS-ACS images is shown in *Figure 3*, in which multiple breath-holds were performed during the MS acquisition and a single breathhold was performed in the SS acquisition. Fewer motionrelated artifacts in the myocardium and a better dark blood effect were observed in the SS-ACS images, especially at slices adjacent to the apex.

Figure 4 displays the edema comparison of MS-PI and SS-ACS in a patient in whom edema was shown with a similar location and size under both sequences. The Bland-Altman plots of edema extent among the different techniques are shown in *Figure 5*. The edema extents of SS-ACS was in good agreement with that of MS-PI (bias 0.35%; 95% confidence interval: –9.7% to 10.4%), while those of SS-ACS and MS-PI exhibited no significant difference (P=0.859). Furthermore, the edema extent of MS-ACS showed good agreement with that of MS-PI (bias –0.03%, 95% confidence interval: –6.8% to 6.8%), while those of MS-ACS and MS-PI displayed no significant difference (P=0.979).

Discussion

In this study, an ACS technique was implemented to

accelerate DB-T2W myocardial imaging. Two acquisition modes of ACS, MS-ACS and SS-ACS, were used to compare to conventional DB-T2W imaging with the PI method. The breath-holding time of MS-ACS method was 4 heartbeats shorter than that of MS-PI method for each slice, and 36 heartbeats of scan time could be reduced with the acquisition of 9 slices. The total scan time of SS-ACS was reduced to 17 s and the acquisition could be implemented within a single breath-hold, while the total scan time of the MS acquisition was longer than 55 s for 9 slices, and multiple breath-holds were needed during the acquisition. Meanwhile, the image quality of DB-T2W imaging with ACS was comparable or even superior to that using the PI method.

Fast spin echo-based DB-T2W imaging is routinely used for cardiac morphological evaluation but is severely impaired by signal loss due to the long echo train length (4). Alternatively, T2-prepared SSFP approaches have developed rapidly without the limitation of signal loss due to the fast acquisition of gradient echo (34,35). However, the signal of SSFP is proportional to $\sqrt{(T2/T1)}$ (36), which reduces the signal change of different tissue. Moreover, the SSFP is sensitive to field inhomogeneity (37), which may lead to a banding artifact, especially at higher magnetic fields. To some extent, the fast spin echo-based methods are preferable for DB-T2W imaging because they are insensitive to field inhomogeneity and have higher SNRs and CNRs (5).



Figure 2 The whole heart image comparison from a patient (female; 59 years old; LVEF 29%). The images were acquired using MS-PI and MS-ACS; although both required multiple breath-holds, MS-ACS used a shorter breath-holding duration. The yellow arrows indicate the obvious artifacts. MS-ACS yielded fewer motion artifacts than did MS-PI. MS-PI, multi-shot acquisition with parallel imaging; MS-ACS, multi-shot acquisition with artificial intelligence-assisted compressed sensing; LVEF, left ventricle ejection fraction.

In addition, the signal loss caused by the long echo train length of the fast spin echo could be largely eliminated with the integration of the ACS technique for its super-fast acquisition.

The ACS technique in this application uses the advantages of AI and CS reconstruction. To guarantee generalizability, the network was trained by millions of fully sampled bits of data, including various image indications, scan conditions, body habiti, and anatomies from phantoms and volunteers in a clinical setting (24). In addition, to guarantee the reliability of the reconstruction, the output of the network was regarded as a reference image of the additional regularization term in the CS reconstruction (38). Previous studies reported that the ACS in DB-T2W imaging significantly shortened the breath-holding time of the patients and improved the image quality by reducing motion artifacts for patients (26-28). In this study, it took 7 heartbeats per slice using the MS-ACS approach and 11 heartbeats per slice using MS-PI protocol, which improved patient cooperation and thus reduced respiratory motion artifacts and avoided myocardial signal loss. In the heart failure cohort, the MS-ACS also took 7 heartbeats per slice, which shortened the acquisition time.

Additionally, the application of ACS makes it possible to acquire whole-heart images during a single breath hold in DB-T2W imaging by cooperating with an SS acquisition, which shortens the scan time considerably, simplifies the scanning workflow, and lessens the breath-holding burden of patients. Moreover, the SS-ACS solution could



Figure 3 The whole heart image comparison from another patient (male; 60 years old; LVEF: 40%). The images were acquired using MS-PI, which required multiple breath-holds, and single-shot acquisition with ACS reconstruction (SS-ACS), which required just a single breath-hold. The yellow arrows indicate the obvious artifacts. SS-ACS largely reduced the motion artifacts. MS-PI, multi-shot acquisition with parallel imaging; SS-ACS, single-shot acquisition with artificial intelligence-assisted compressed sensing; LVEF, left ventricle ejection fraction.



MS-PI

SS-ACS

Figure 4 A patient case (male; 35 years old; LVEF: 59%) for edema imaging comparison of MS-PI (left) and SS-ACS (right). Edema (green arrows) had a similar location and size in imaging by both sequences. MS-PI, multi-shot acquisition with parallel imaging; SS-ACS, single shot acquisition with artificial intelligence-assisted compressed sensing; LVEF, left ventricle ejection fraction.

substantially reduce the motion artifacts induced by variant R-R intervals through SS acquisition with 1 slice per cardiac cycle. In other words, respiratory-triggered free-breathing acquisition is an alternate method of obtaining DB-T2W imaging for patients who cannot tolerate the breath-hold.

Limitations

In this study, MS DB-T2W imaging with PI was considered a reference for technique comparison because it is hard to collect fully sampled cardiac imaging data as a gold standard. Moreover, the current study lacked technical comparison with other CS or AI methods. Although a total of 33 cardiovascular patients is sufficient to validate the proposed methods, a larger sample size of different



Figure 5 The Bland-Altman plots of edema extents obtained from the images of different techniques. MS-PI, multi-shot acquisition with parallel imaging; MS-ACS, multi-shot acquisition with artificial intelligence-assisted compressed sensing; SS-ACS, single-shot acquisition with artificial intelligence-assisted compressed sensing; LVM, left ventricle mass; SD, standard deviation.

cardiac diseases is warranted for the statistical comparisons of various cardiac diseases. In addition, validation across multiple sites is also warranted to eliminate the potential selection bias. Although the SS acquisition was applied to minimize the effects of cardiac motion due to the variant R-R intervals that occur in the MS acquisition, the image quality of the DB-T2W imaging using the ACS method could have been impaired by the serious arrhythmias or extremely high heart rates of the patients.

Conclusions

The ACS method for DB-T2W imaging significantly reduced the acquisition time of the MS scheme to 7 heartbeats per slice and of the SS scheme to 19 heartbeats for 9 slices. It further provided better image quality compared to the conventional PI method. Moreover, DB-T2W imaging was implemented to cover the whole heart with 9 slices in a single breath-hold via the application of SS acquisition combined with ACS, which could help to reduce the breath-holding burden of patients and improve diagnostic efficiency.

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Footnote

Conflicts of Interest: All authors have completed the ICMJE uniform disclosure form (available at https://qims. amegroups.com/article/view/10.21037/qims-22-607/coif). Shiyu Zhang and Shuheng Zhang are employed by United Imaging Healthcare. JX is employed by United Imaging Healthcare America, Inc. The other authors have no conflicts of interest to declare.

Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013) and was approved by the institutional ethics board of Tongji Hospital, Tongji Medical College, Huazhong University of Science and Technology. Informed consent was obtained from all of the patients.

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