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Joint Non-rigid Motion and Image MRI SENSE Reconstruction

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Synopsis

In this work a method of the reconstruction of magnetic resonant image with a sence of non-rigid motion is presented. A non-rigid deformation model is introduced in the reconstruction problem in order to obtain, a motion-free image and 2) a set of non-rigid transformations that describe the motion of the structures being imaged. The method is able to reconstruct the motion information without reconstructing the images in additional motion states avoiding extra computational costs and resulting in reconstruction problem. Preliminary results for 2D cardiac cine MRI are shown.

Introduction

The presence of motion during MRI examinations is a major source of a stand to be degradation specially in modalities that require a long acquisition time or when imaging moving organs. Previously proposed approaches to correct involve to retrospectively bin the k-space data acquired in different motion states and to reconstruct the corresponding images [1]. Motion can then be estimated with certain image registration method and, once the deformation is known, a motion corrected reconstruction can be performed [2, 3]. The different motion states involves high computational costs and, in undersampled acquisitions, higher net acceleration factors making their reconstruction reproblem gets multiplied by the number of motion states considered, resulting in worse sed problem.

In this work, we present a method for the direct reconstruction of a motion from a set of spatial transformations without the need of additional images. The method departs from the basic assumption that the motion can be uescribed with rewer parameters than the deformed images themselves, resulting in a better posed reconstruction problem. It is a generalization of the method proposed in [4], where only rigid motion is considered.

Preliminary results for 2D, breath-hold cardiac cir : MR are shown, where instead of reconstructing an image for each cardiac phase, a single image : obtained plus a set of deformations describing the motion of the heart.

Theory and Methods

A. Method formulation

The SENSE reconstruction reconstruction problem can be generalized in the second reconstruction reconstruction problem can be generalized in the second reconstruction reconstruction problem can be generalized in the second reconstruction reconstruction problem can be generalized in the second reconstruction reconstruction problem can be generalized in the second reconstruction reconstruction problem can be generalized in the second reconstruction reconstruction problem can be generalized in the second reconstruction reconstru

$$(\hat{\mathbf{m}}, \hat{\boldsymbol{\Theta}}) \coloneqq \arg\min_{\mathbf{m}, \boldsymbol{\Theta}} \|A\mathcal{F}\mathbf{S}\mathbf{T}_{\boldsymbol{\Theta}}\mathbf{m} - \mathbf{y}\|_2^2$$

where y denotes the acquired k-space data, m the motion-free image of the set of parameters Θ , \mathcal{F} the discrete Fourier transformation and \mathbf{A} the same \mathbf{G} matrix.

The joint problem can be addressed by iterative solving two subproble ::

1) Image reconstruction:

Considering the deformation \mathbf{T}_{Θ} known, we solve

$$\hat{\mathbf{m}} = i - \lim_{\mathbf{m}} \|\mathbf{A}\mathcal{F}\mathbf{S}\mathbf{T}_{\Theta} - \mathbf{y}\|_2^2$$

The problem is equivalent to the one formulated in [2], where the open or $G = A \mathcal{F} S \mathcal{T}$ is defined, and can be solved by means of the conjugate gradient algorithm applied to the system of linear equations given by

$$\mathbf{G}^H \mathbf{G} \mathbf{m} = \mathbf{G}^H \mathbf{v}$$

2) Motion compensation:

Considering the image \$\mathbf{m}\$ to be known, the deformation fields are up ated by plving

$$\hat{\mathbf{\Theta}} = \epsilon$$
 $\mathbf{n} - \mathbf{y} \|_2^2$

In this case, the resulting problem is non-linear and can be solved by means of the non-linear conjugate gradient algorithm.

B. Experimental setup.

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A regular 2D breath-hold cine MRI was performed in one healthy volunteer with a 32 coil torso antenna in a 3T Philips Achieva TX scanner. Main acquisition parameters include $TR/TE/\alpha = 3.1ms/1.57ms/60^{\circ}$, $FOV=320x320mm^2$, pixel size= $2x2mm^2$, 30 cardiac phases. Fully sampled k-space data was retrospectively undersampling simulating a net acceleration factor of 16.

Results and discussion

Figure 1 shows the results obtained with the proposed method and with a compressed sensing (CS) reconstruction [5] for an acceleration factor of 16. Fully sampled image is shown as a reference. CS reconstruction presents residual undersampling artifacts and blurred edges in the endocardio. It is worth noticing that, with the proposed method, only one static image is reconstructe of cardiac phases. This aspect explains the capacity of the method to revent unlersampling artifacts.

In Figure 2 the left ventricle was manually segmented in the motion-fr image 1 the r ulting mask was propagated to the rest of the cardiac cycle. Diastolic and systolic phases are shown with a manual segmentation in the fully superposed. Excellent agreement between the propagated and manual segmentation can be shown.

Finally, in figure 3 the slice volumes of the left ventricle through the car cycle obta. I from the propagated mask and from manual segmentation are plotted, showing good agreement.

Conclusions

In this work preliminary results of a MRI reconstruction method in the presented as a proof of concept. The feasibility of the joint motion and a static image reconstruction approach has been tested and the applicability of the deformations fields to obtain left ventricle volume curves from a single segmentation along the cardiac cycle.

This method can be potentially applied to higher dimensional problems, such a 3D coro where the reconstruction of whole volumes in different cardio-respiratory states a estimate computational cost.

Acknowledgements

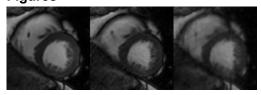
This work has been partially funded by MINECO and Junta de Castill ts TEC2014-57428-R and VA069U16 respectively

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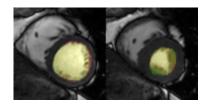
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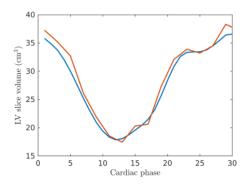
Figures

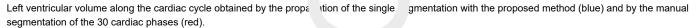


Reconstructed images with the proposed method (center), CS reconstruction (ht) for a acceleration factor of 16. Fully sampled image is show as reference



Segmentation of the left ventricle in diastole (left) and systole (right). Propagated mask from the proposed reconstruction (red) and manual segmentation on the fully sampled images (green) and overlapped areas (yellow).





Proc. Intl. Soc. Mag. Reson. Med. 25 (2017)

