Towards joint segmentation and registration of the myocardium in CP-BOLD MRI at rest

Ilkay Oksuz1*, Rohan Dharmakumar2, Sotirios Tsaftaris3,1

From 19th Annual SCMR Scientific Sessions
Los Angeles, CA, USA. 27-30 January 2016

Background
A new contrast and stress-free approach for detecting myocardial ischemia has been recently demonstrated using Cardiac Phase-resolved Blood Oxygen-Level-Dependent (CP-BOLD) MRI. To detect the disease, CP-BOLD relies on the observation that myocardial signal intensity varies as a function of cardiac phase. These changes are not readily visible and post-processing is necessary; including myocardial segmentation and registration. Automated analysis approaches, which can obtain pixel-level determination of ischemia, are desirable since they may lead to improved accuracy in detection of disease. To achieve this, precise segmentation and non-linear registration of the myocardium among the frames (the cardiac phases) in the cine stack would be required. Unfortunately, at present due to BOLD contrast variations, classical approaches to segmentation and registration fail to reach desirable accuracy. Here we propose an algorithm that jointly finds suitable myocardial segmentation and elastically registers the heart.

Methods
Flow- and motion-compensated 2D short-axis CP-BOLD was acquired along the mid ventricle in 10 canines at baseline and under severe LAD stenosis. Imaging studies were performed on a 1.5T Espree (Siemens Healthcare). TR/TE

Figure 1 The figure at top shows the flowchart of the algorithm. First 9 × 9 patches are extracted for each pixel and concatenated with HOG and Gabor features of each patch. Dictionaries $D_B$ and $D_M$ are calculated for the initialization of registration process. The sparse representation of each pixel $p$ for background $X_B$ and myocardium $X_M$ are calculated and concatenated; $X_p = [X_B, X_M]$. In data term of the objective function of registration, the similarity metric in between two consecutive frames is defined as L1 norm of the concatenated supports: $S_1(p, t, t+1) = ||X_p^t - X_p^{t+1}||_1$. The approach iterates refinement step of the segmentation using the registration for the agreement of classification and then finding the $D$, then $X$. The fusion of labels in between the different frames is used for classification of the myocardium region.

© 2016 Oksuz et al. This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited. The Creative Commons Public Domain Dedication waiver (http://creativecommons.org/publicdomain/zero/1.0/) applies to the data made available in this article, unless otherwise stated.
6.2/3.1 ms; spatial resolution = 1.2 × 1.2 × 8 mm³, flip-angle = 70°, ~25 frames per cardiac cycle. The main principle of the proposed approach is to use an external initial database of pre-segmented images (and dictionaries for myocardium and background), to first come up with an initial segmentation (relying on classification when projecting on discriminatory dictionaries) of an input CP-BOLD patient stack. Then, images in the cardiac cycle (cine acquisition) are registered using the sparse coefficients of the projections to establish a new similarity metric. Afterwards, the obtained segmentation is refined; which gives the opportunity to refine the dictionaries (adapting them to the dataset under consideration). This refinement of segmentation and dictionaries is continued till convergence. The flowchart and detailed explanation of the algorithm is shown in figure 1.

Results
As seen in table 1, experiments under baseline and ischemia conditions show that the proposed approach improves the results thanks to the refinement with registration. Moreover, the proposed approach obtains superior accuracy when compared to state of the art methods for registration (DRAMMS and MIND).

Conclusions
The experiments clearly underline the need for a new representation in image registration and segmentation. Using the information from the same subject to train the dictionaries improves the accuracy of myocardium segmentation both in the cases of health and disease. Although further experiments are necessary to validate this approach, CP-BOLD can open the road to repeatable, truly non-invasive diagnosis of ischemic heart disease.

Dice overlap measures for the proposed method compared with segmentation without refinement and two state of the art methods for image registration DRAMMS (Ou et al., MIA, 2011) and MIND (Heinrich et al., MIA, 2012)

|                        | CP-BOLD BASELINE | CP-BOLD ISCHEMIA |
|------------------------|------------------|------------------|
| Proposed Approach      | 0.68 ± 0.08      | 0.63 ± 0.06      |
| Proposed without       | 0.66 ± 0.09      | 0.60 ± 0.13      |
| refinement DRAMMS      | 0.61 ± 0.07      | 0.54 ± 0.06      |
| MIND                   | 0.62 ± 0.07      | 0.53 ± 0.09      |

Authors’ details
1IMT Institute for Advanced Studies Lucca, Lucca, Italy. 2Cedars-Sinai Medical Center, Los Angeles, CA, USA. 3University of Edinburgh, Edinburgh, United Kingdom.

Published: 27 January 2016