Cardiac Segmentation With Strong Anatomical Guarantees

Nathan Painchaud^{1,2} Ph.D. candidate

¹Computer Science Department Université de Sherbrooke ²CREATIS INSA Lyon



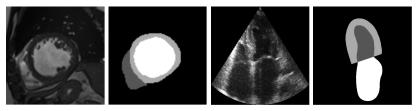
CREATIS

Imagining Imaging Symposium, December 9-10, 2020

Cardiac Imaging Automatic Cardiac Segmentation

Cardiac Imaging

2 modalities & 3 acquisition protocols \rightarrow diverse images and cardiac shapes



(a) Short-axis MRI

(b) Long-axis echocardiography (US)

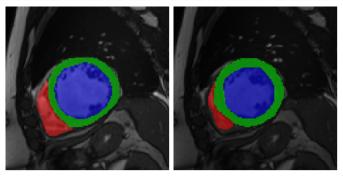
Examples of images from different modalities, with their annotations.

Cardiac Imaging Automatic Cardiac Segmentation

Cardiac Imaging: Cardiac Function

Objective

Determine the images' underlying **cardiac anatomy** to analyze the **cardiac function**.



(a) Heart at end diastole

(b) Heart at end systole

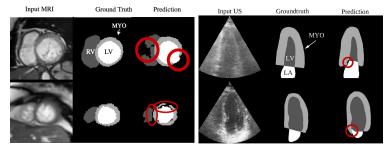
Cardiac Imaging Automatic Cardiac Segmentation

Automatic Cardiac Segmentation

Segmentation using deep neural networks:

Inter-observer variability
(Dice coeff., Hausdorff dist.)

🔀 Anatomical guarantees



(a) MRI

(b) US

[Painchaud et al., IEEE TMI, 2020]

Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Overview

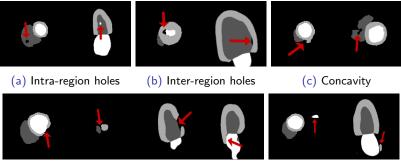
Post-processing of anatomically invalid images

- Before processing the images, we:
 - define anatomical criteria;
 - 2 learn a model of cardiac shapes.
- While processing images, we:
 - segment the images;
 - evaluate the anatomy of the segmented shapes;
 - Orrect invalid segmentations.

Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Anatomical Criteria

12 to 16 handcrafted metrics, i.e. implemented by code

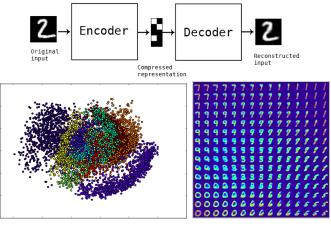


(d) Connectivity between LV and background (e) Disconnectivity

[Painchaud et al., IEEE TMI, 2020]

Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Autoencoders



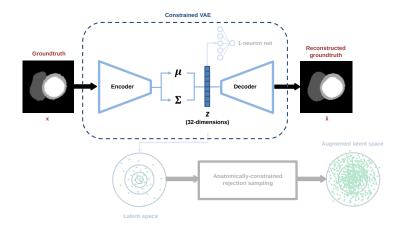
(a) 2D encoding of digits 0-9 (b) Decoded 2D samples

Source: https://blog.keras.io/building-autoencoders-in-keras.html

Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Cardiac Shape Modeling

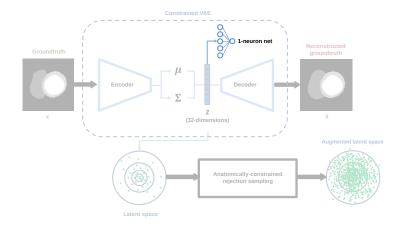
Variational autoencoder + topological contraint



Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Cardiac Shape Modeling

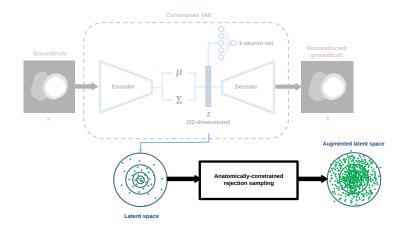
Variational autoencoder + topological constraint



Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Data Augmentation through Sampling

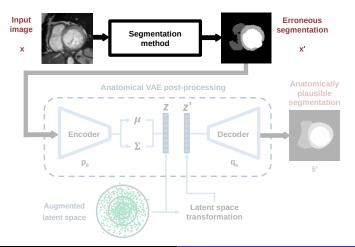
Sample the latent space posterior to increase density.



Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Anatomical Post-processing

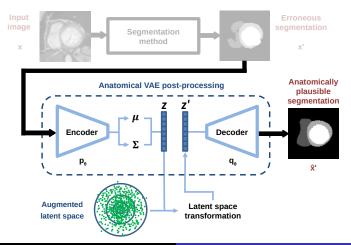
Cardiac segmentation with any chosen method



Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

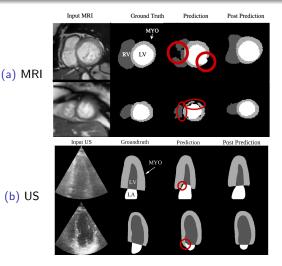
Anatomical Post-processing

Correction of invalid segmentations



Anatomical Criteria Cardiac Shape Modeling Data Augmentation through Sampling Anatomical Post-processing

Anatomical Post-processing: Results



[Painchaud et al., IEEE TMI, 2020]

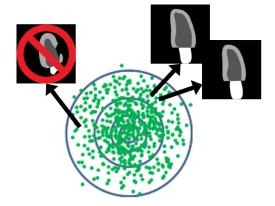
Nathan Painchaud

Deep Manifold Learning

Deep Manifold Learning

- ✓ Intra-sample anatomy
- □ Inter-sample similarity

Inter-image Consistency Interpretability



Deep Manifold Learning

Deep Manifold Learning

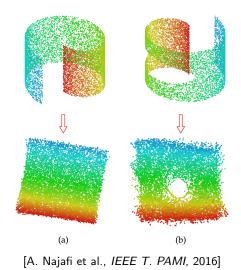
Attribute Manipulation

PCA Autoencoder z = (0.0, 0.0, 0.0)z = (-0.2., 0.0, 0.0)z = (-0.1, 0.0, 0.0)z = (0.1, 0.0, 0.0)z = (0.2, 0.0, 0.0)z = (0, 0, -0.4, 0.0)z = (0.0, -0.2, 0.0)z = (0.0, 0.0, 0.0)z = (0.0, 0.2, 0.0)z = (0.0, 0.4, 0.0)z = (0.0, 0.0, -0.6)z = (0.0, 0.0, -0.3)z = (0.0, 0.0, 0.0)z = (0.0, 0.0, 0.3)z = (0.0, 0.0, 0.6)

[Ladjal et al., arXiv, 2019]

Inter-image Consistency Interpretability Thank you for listening!

Manifold Linearization



Examples of linear 2D manifolds enclosed in 3D spaces

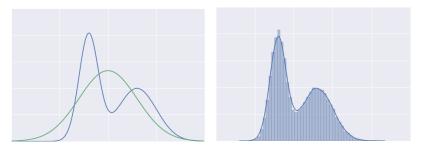
Nathan Painchaud Cardiac Segmentation w/ Guarantees

Rejection Sampling

Hard to sample target distribution (in blue).

We sample from a simpler proposed distribution (in green).

Sampling results.



Examples of rejection sampling with a 2D mixture of gaussians.