

# Multiple resolution residual network for automatic thoracic organs-at-risk segmentation from CT

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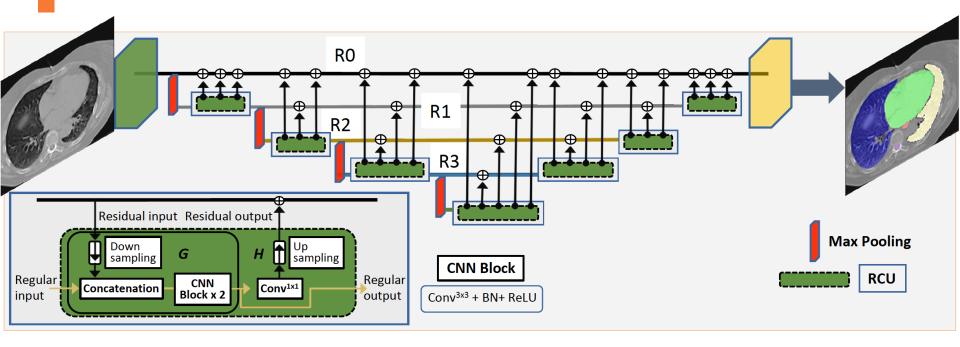
## **Motivation**

- Radiotherapy treatment planning requires highly accurate segmentations for precise tumor targeting while reducing unnecessary dose to critical normal organs<sup>1</sup>
- Clinical treatments use manual delineations done by physicians<sup>2</sup>
  - Time consuming
  - Highly variable between same and different physicians
- Current methods for automatic segmentation of thoracic OARs (e.g. U-Net and FCN architectures) still pose a challenge for narrow, thin structures located in the mediastinum (with little soft-tissue contrast) such as the esophagus
  - Loss of resolution in the deeper convolutional layers
- 1. Thomas Rockwell Mackie et al. Image guidance for precise conformal radiotherapy. International Journal of Radiation Oncology, Biology, Physics, 56(1):89-105, 2003

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2. Jinzhong Yang et al. A statistical modeling approach for evaluating auto-segmentation methods for image-guided radiotherapy. Comput Med Imag Graph, 36(6):492-500, 2012

# Multiple Resolution Residual Network (MRRN)



The MRRN simultaneously combines information from multiple feature streams computed at different image resolution levels through residual connections



# **Experiments**

- Datasets
  - CT scans of 241 internal patients with LA-NSCLC
    - Training: N = 206
    - Validation: N = 35
  - 60 CT scans from the 2017 AAPM Thoracic Auto-Segmentation Challenge<sup>3</sup>
    - Testing set 1: N = 48 (training + offline testing)
    - Testing set 2: N = 12 (online testing)
- Implementation
  - Training in 2D with 21441 images, validation with 2104 images
  - Image size: 256x256, after cropping and resizing
- 3. J Yang, Veeraraghavan H, Armato S.G., K Farahani, J.S Kirby, J Kalpathy-Kramer, W van Elmpt, A Dekker, X Han, X Feng, P Aljabbar, B Oliviera, B van der Heyden, L Zamdborg, D Lam, M Gooding, and G.C. Sharp. Autosegmentation for thoracic radiation treatment planning: A grand challenge at AAPM 2017

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# Results

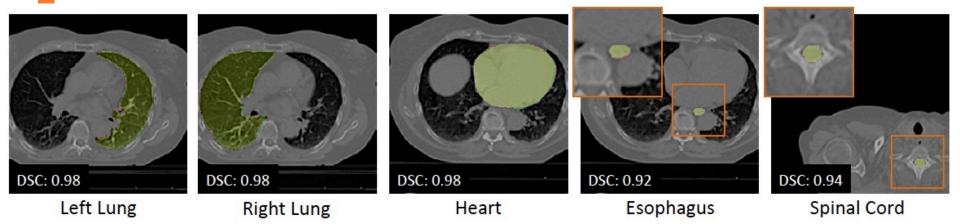
- Median DSC and IQR achieved for testing set 1
  - 0.97 (IQR:0.97-0.98) for the left and right lungs
  - 0.93 (IQR: 0.93-0.95) for the heart
  - 0.78 (IQR: 0.76-0.80) for the esophagus
  - 0.88 (IQR: 0.86-0.89) for the spinal cord

### DSC achieved for thoracic OARs in the AAPM online testing set (testing set 2)

Method	2D/3D	Left Lung	Right Lung	Heart	Esophagus	Spinal Cord
MRRN	2D	0.96 ± 0.01	0.96 ± 0.02	$0.93 \pm 0.03$	0.77 ± 0.04	0.87 ± 0.017
Elekta	2.5D/3D	$0.97 \pm 0.02$	$0.97 \pm 0.02$	$0.93 \pm 0.02$	$0.72 \pm 0.10$	$0.88 \pm 0.037$
UVa	3D	$0.98 \pm 0.01$	$0.97 \pm 0.02$	$0.92 \pm 0.03$	$0.64 \pm 0.20$	$0.89 \pm 0.042$
Mirada	2D	$0.98 \pm 0.02$	$0.97 \pm 0.02$	0.91 ± 0.02	0.71 ± 0.12	0.87 ± 0.110



# Results



Example segmentations for the analyzed organs. Green mask = expert delineation, red mask = algorithm-generated segmentation, yellow mask = combined segmentation