

# Automatic cerebral hemisphere segmentation in rat MRI with lesions via attention-based convolutional neural networks

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# Why hemisphere segmentation?



Region of interest

Source: https://neuroscience-graphicdesign.com/2017/08/01/post-1-rat-brain-gallery/

The ratio <u>contralateral hemisphere volume</u> ipsilateral hemisphere volume

is an important **biomarker** for acute stroke.

### MedicDeepLabv3+ (convolutional neural network)



# Experiments

### **1.** Comparison with eight other methods

VoxResNet, HighRes3DNet, V-Net, UNet, DeepLabv3+, Demon, RATS, RBET

**Convolutional Neural Networks** 

Brain extraction

### 2. Brain midline volume



### 3. Hemispheric ratio

Are the hemispheric ratios in the ground truth significantly different from the automatic segmentations?

- Effect size (Cohen's d)
- Confidence interval

### Results

#### 1. Comparison

	Approach	Dice	HD	Image + GT MedicDeep
Brain	MedicDeepLabv3+	$\textbf{0.952} \pm \textbf{0.04}$	$1.856\pm0.91$	No to A
	VoxResNet	$0.951 \pm 0.04$	$2.042 \pm 1.02$	A.C.
	HighRes3DNet	$0.949 \pm 0.04$	$1.858 \pm 1.04$	
	V-Net	$0.948 \pm 0.04$	$1.920 \pm 1.05$	
	UNet <b>(2D)</b>	$0.947 \pm 0.05$	$3.477 \pm 1.20$	
	DeepLabv3+	$0.936 \pm 0.04$	$2.149 \pm 1.02$	
	Demon (2D)	$0.934 \pm 0.04$	$3.621 \pm 1.17$	
	RATS	$0.913 \pm 0.01$	$2.221 \pm 0.51$	
	RBET	$0.781 \pm 0.10$	$3.628 \pm 0.46$	M 16
CH	MedicDeepLabv3+	$0.944 \pm 0.04$	$\textbf{2.064} \pm \textbf{1.85}$	
	VoxResNet	$0.944 \pm 0.04$	$2.265 \pm 1.86$	
	HighRes3DNet	$0.942 \pm 0.04$	$2.205 \pm 1.86$	
	V-Net	$0.940\pm0.04$	$2.218 \pm 1.86$	
	UNet <b>(2D)</b>	$0.941 \pm 0.05$	$3.689 \pm 1.64$	and -
	DeepLabv3+	$0.921 \pm 0.04$	$2.411 \pm 1.80$	B

### 2. Brain midline volume

- MedicDeepLabv3+ outperformed the baseline DeepLabv3+.
- UNet provided slightly higher (0.02) Dice coefficients.

### 3. Hemispheric ratio

Approach	Cohen's d	Confidence Interval	
MedicDeepLabv3+	0.008	<u>[-0.013, 0.035]</u>	
VoxResNet	-0.042	[-0.060, -0.025]	
HighRes3DNet	-0.102	[-0.125, -0.080]	
V-Net	0.003	[-0.042, 0.022]	
UNet	-0.038	[-0.054, -0.021]	
DeepLabv3+	0.050	[-0.008, 0.099]	

Small d Zero-centered C.I.

### **Discussion / Conclusion**

> Our method provided **excellent** and **more accurate** segmentations than the other methods.

- ➤ Our method takes one second to segment 3D volumes.
- > 3D convolutional neural networks achieved better segmentations than 2D.

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