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

Juan Miguel Valverde, Artem Shatillo, Riccardo de Feo, Jussi Tohka

## Why hemisphere segmentation?



The ratio $\frac{\text { contralateral hemisphere volume }}{\text { ipsilateral hemisphere volume }}$
is an important biomarker for acute stroke.

Region of interest


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

## MedicDeepLabv3+ (convolutional neural network)



## Experiments

## 1. Comparison with eight other methods

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

Convolutional Neural Networks
2. Brain midline volume


Brain extraction

## 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 | mase + 6 | $\begin{gathered} \text { Neariceeplanasu+ } \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & \text { 罵 } \\ & \hline \end{aligned}$ | MedicDeepLabv3+ | $\mathbf{0 . 9 5 2} \pm 0.04$ | $\mathbf{1 . 8 5 6} \pm \mathbf{0 . 9 1}$ |  |  |
|  | VoxResNet | $0.951 \pm 0.04$ | $2.042 \pm 1.02$ |  |  |
|  | HighRes3DNet | $0.949 \pm 0.04$ | $1.858 \pm 1.04$ |  |  |
|  | V-Net | $0.948 \pm 0.04$ | $1.920 \pm 1.05$ |  |  |
|  | UNet (2D) | $0.947 \pm 0.05$ | $3.477 \pm 1.20$ |  |  |
|  | 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 | $0.781 \pm 0.10$ | $3.628 \pm 0.46$ |  |  |
| 寻 | MedicDeepLabv3+ | $0.944 \pm 0.04$ | $\mathbf{2 . 0 6 4} \pm \mathbf{1 . 8 5}$ |  |  |
|  | VoxResNet | $0.944 \pm 0.04$ | $2.265 \pm 1.86$ |  |  |
|  | HighRes3DNet | $0.942 \pm 0.04$ | $2.205 \pm 1.86$ |  |  |
|  | V-Net | $0.940 \pm 0.04$ | $2.218 \pm 1.86$ |  |  |
|  | UNet (2D) | $0.941 \pm 0.05$ | $3.689 \pm 1.64$ |  |  |
|  | DeepLabv3+ | $0.921 \pm 0.04$ | $2.411 \pm 1.80$ |  |  |

## 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+ | $\underline{0.008}$ | $[-0.013,0.035]$ |
| 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.
$>3 \mathrm{D}$ convolutional neural networks achieved better segmentations than 2D.

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## GENOMMED charles river



UNIVERSITY OF EASTERN FINLAND

A.I.VIRTANEN

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