

Brain Tumor Segmentation With Deep Neural Networks

Axel Davy¹, Mohammad Havaei², David Warde-Farley³, Antoine Biard⁴, Lam Tran⁵, Pierre-Marc Jodoin², Aaron Courville³, Hugo Larochelle², Chris Pal^{3,6,*} and Yoshua Bengio³

Motivations

- Provide a fast and well-performing segmentation system for brain tumor segmentation
- Compare a Deep Convolutional Neural Network (DCNN) approach to other systems

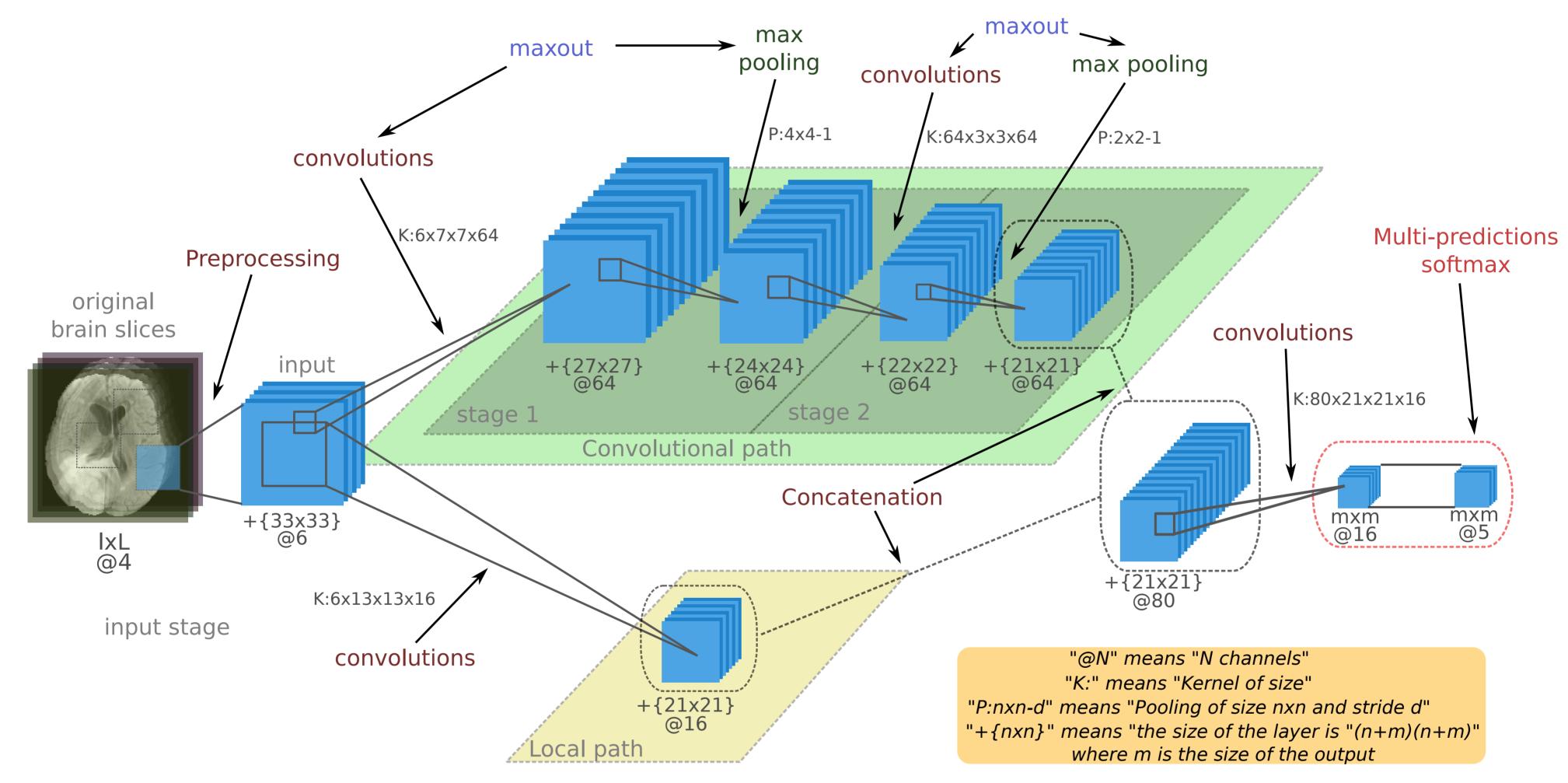
Why Use Deep Convolutional Neural Networks?

- Able to learn feature representations for new domains and modalities
- Tend to behave particularly well when a lot of data is available
- Have demonstrated state of the art performance for other computer vision problems

Framework

- We remove intensity outliers of the MRI images (FLAIR, T1, T1C, T2)
- We apply N4ITK on T1 and T1c
- We normalize the data per brain and per channel
- We compute two channels (T1c T1), and a channel containing T1c mirror(T1c), the mirror being applied on a line basis)
- The Neural Network is applied to give a label probability map
- We remove some false positives by removing tumor pieces with unlikely shape

Our Network



4: École Polytechnique, Palaiseau, France

5: University of Rochester, New York, USA

6: École Polytechnique de Montréal, Canada

Filters

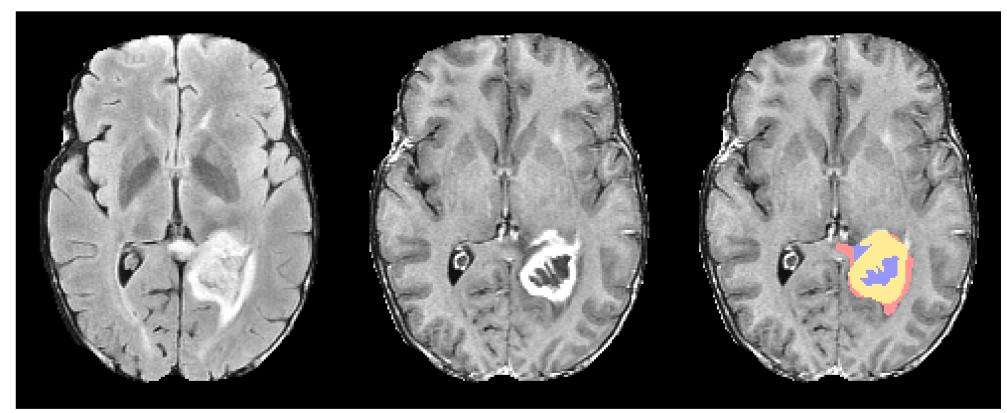
对现代名称 经现金的复数 医医眼动物 **包括设备的数据条据的通用数据**

Extract of some filters learned by the network at the first layer of the 'convolutional path'. From top to bottom, the images are the filters on the different input channels: FLAIR, T1, T1c, T2, T1c - T1, T1c - mirror(T1c). The filters are grouped by two: The Maxout Convolutional layer does the max over the two filters applied to the input.

- 1: École normale supérieure, Paris, France
- 2: Université de Sherbrooke, Sherbrooke, Canada

3: Université de Montréal, Montréal, Canada

Results on BRATS 2013 data



Example of predictions on brain HG_0310 slice 77

Name	Dice score			Positive Predictive Value			Sensitivity		
	Complete	Core	Enhancing	Complete	Core	Enhancing	Complete	Core	Enhancing
Train HG/LG	0.79	0.68	0.57	0.81	0.75	0.54	0.79	0.67	0.63
Leaderboad	0.72	0.63	0.56	0.69	0.64	0.50	0.82	0.68	0.68
Challenge	0.85	0.74	0.68	0.85	0.74	0.62	0.85	0.78	0.77

Results on the 2013 datasets, with a model trained with the 2013 data.

*Contact Information: Chris Pal Email: christopher.pal@polymtl.ca