

Segmentation

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Classification



Cat

Segmentation

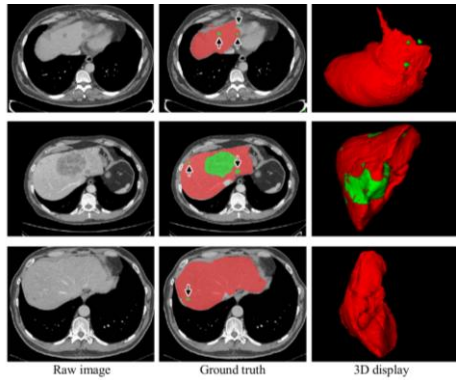


GRASS, CAT,
TREE, SKY

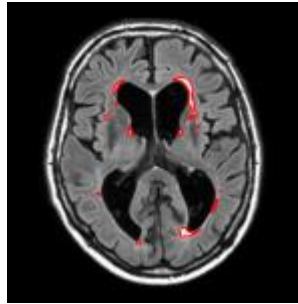
Detection



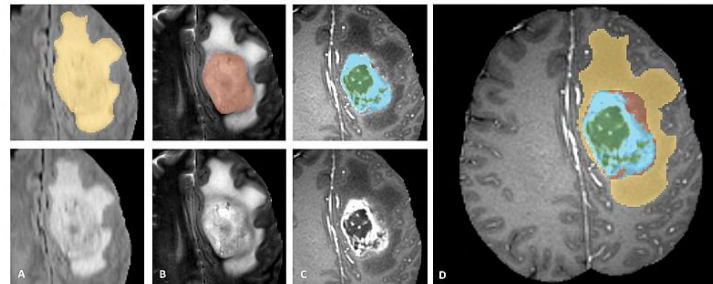
DOG, DOG, CAT



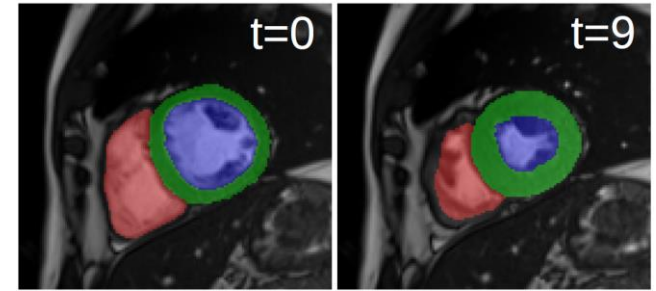
LiTS



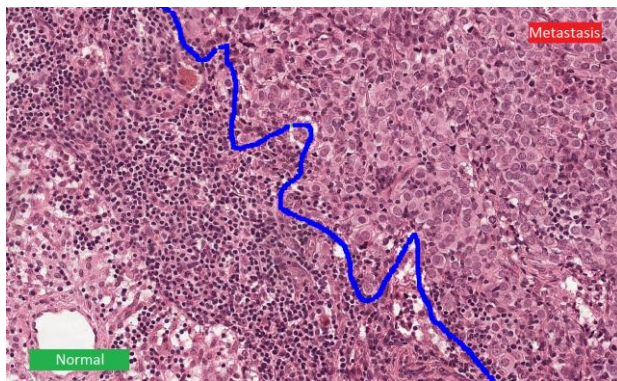
WMH



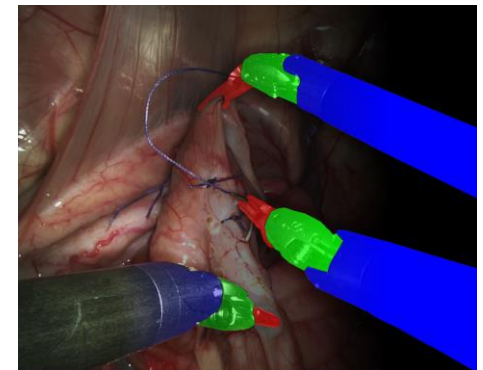
BraTS



ACDC



Camelyon



Endovis

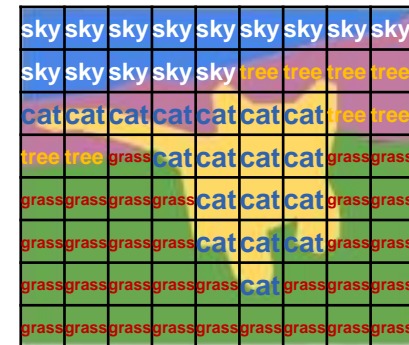
Classification



Cat

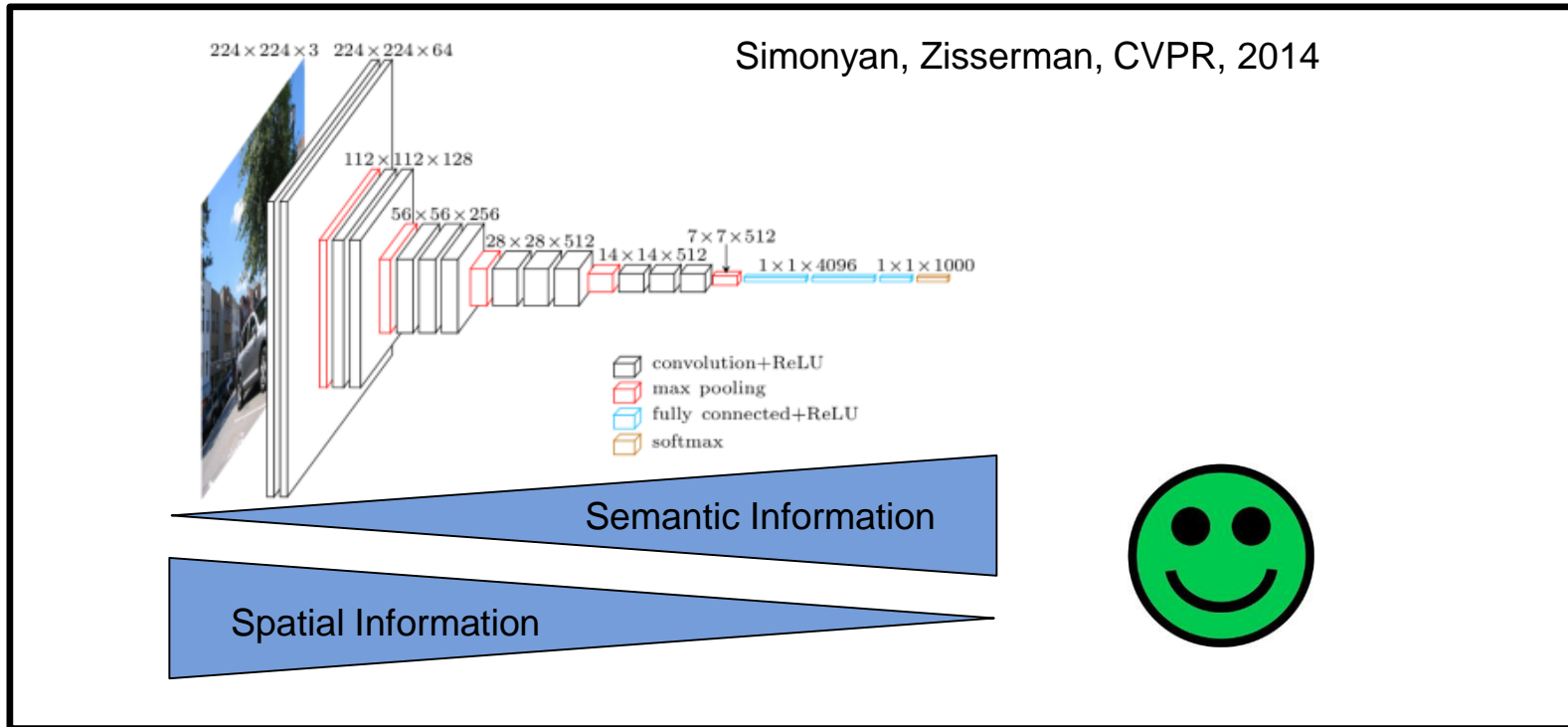
one label per image

Segmentation

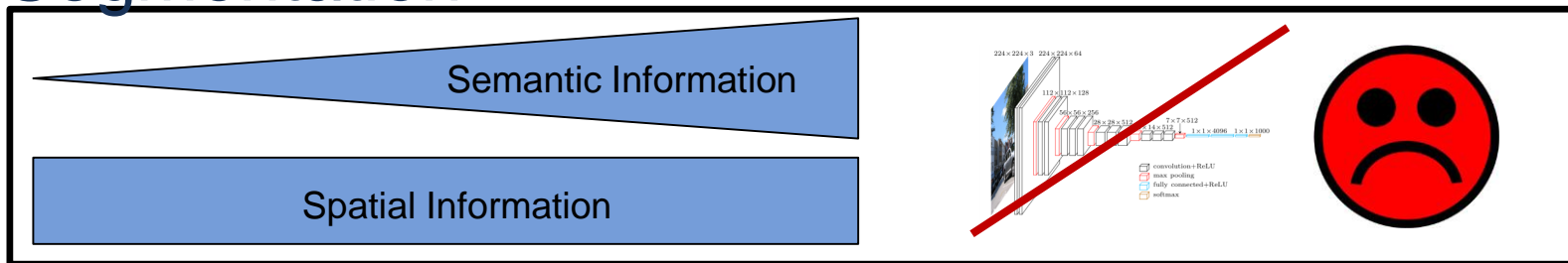


one label per pixel

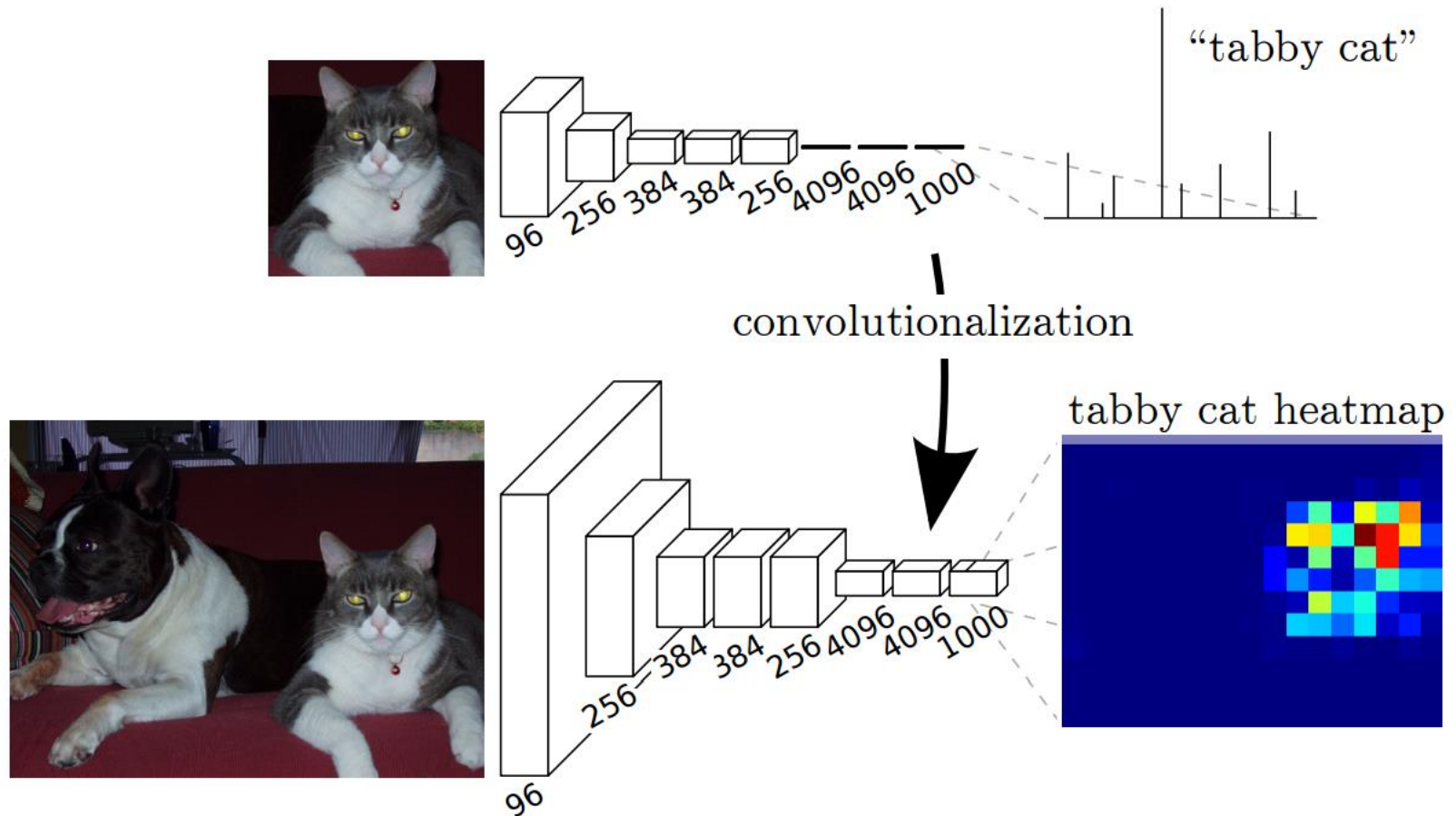
Classification



Segmentation

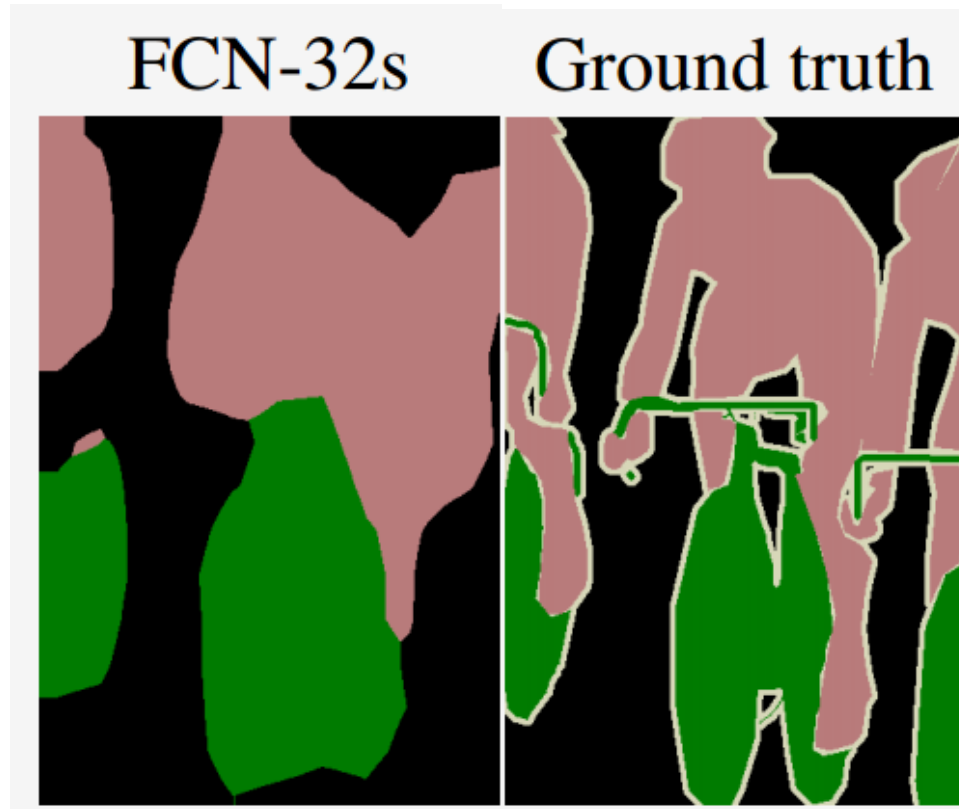


Fully Convolutional Networks for Semantic Segmentation



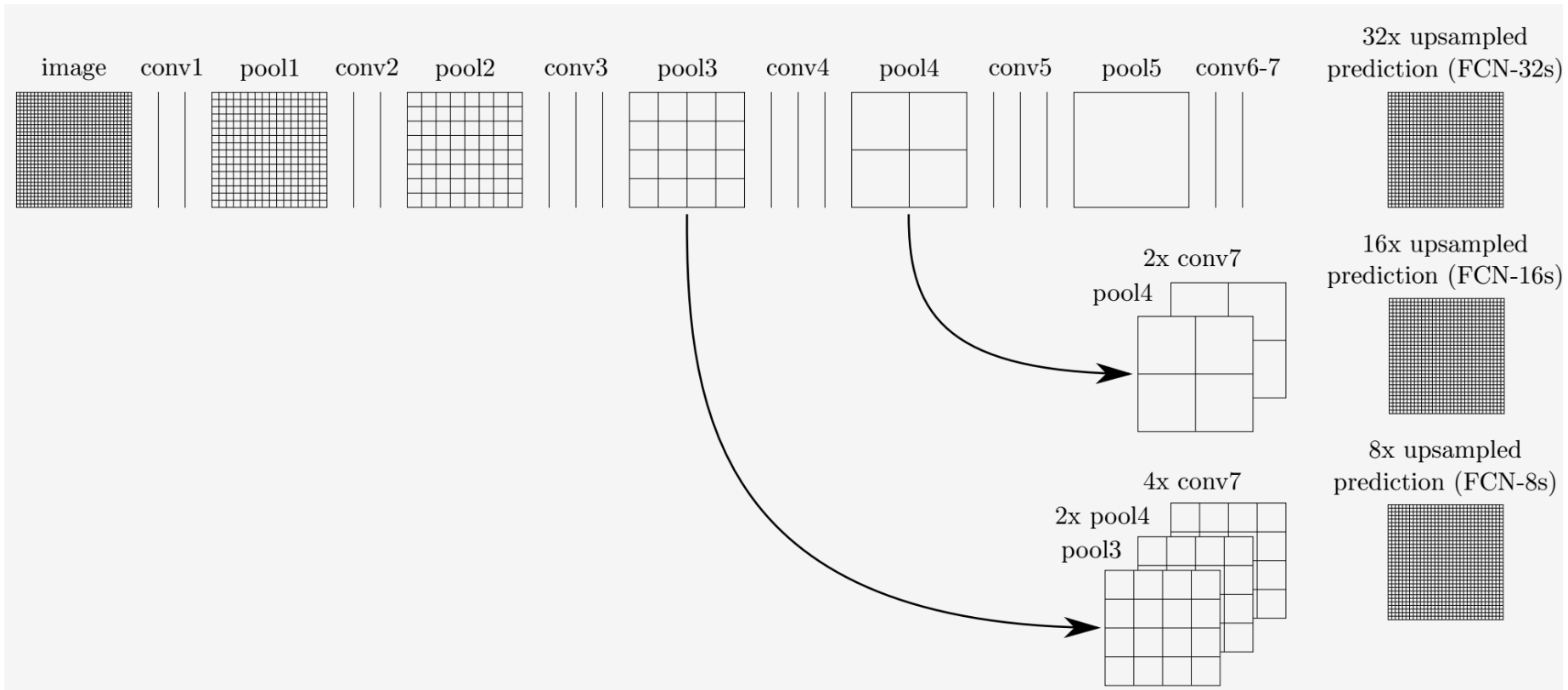
Long et al., CVPR, 2015

Fully Convolutional Networks for Semantic Segmentation



Long et al., CVPR, 2015

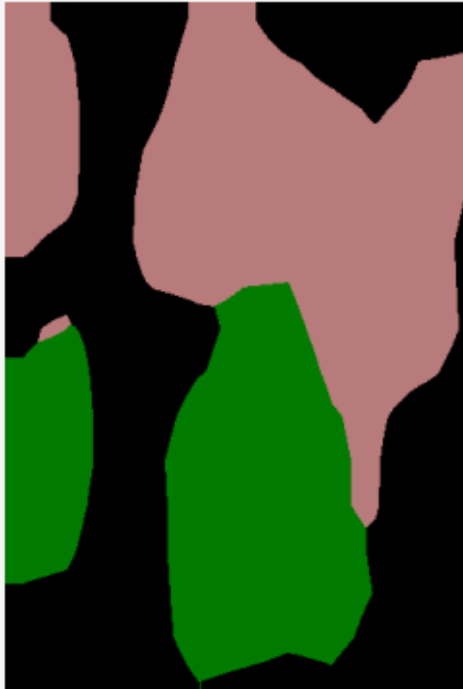
Fully Convolutional Networks for Semantic Segmentation



Long et al., CVPR, 2015

Fully Convolutional Networks for Semantic Segmentation

FCN-32s



FCN-16s



FCN-8s

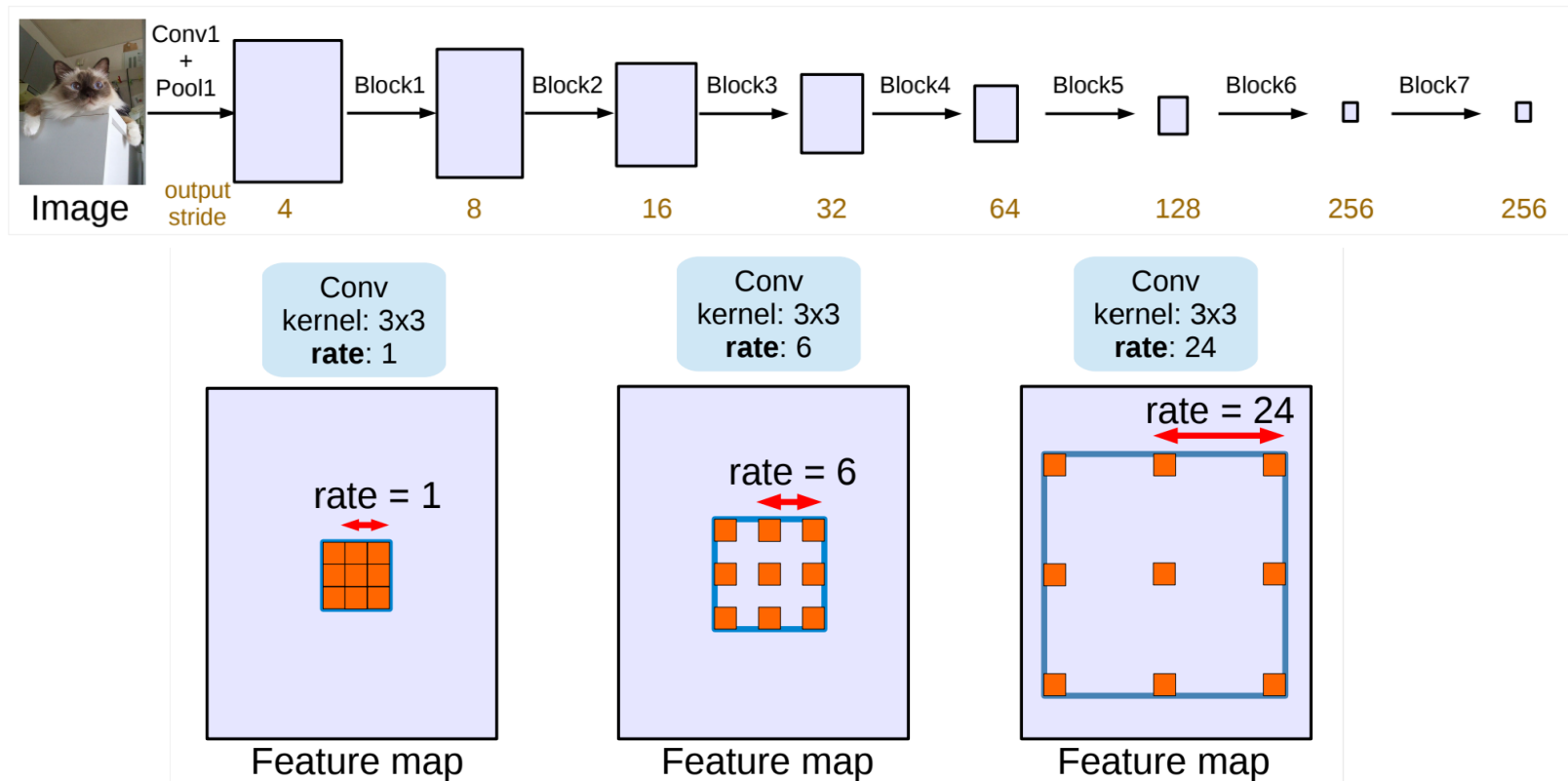


Ground truth



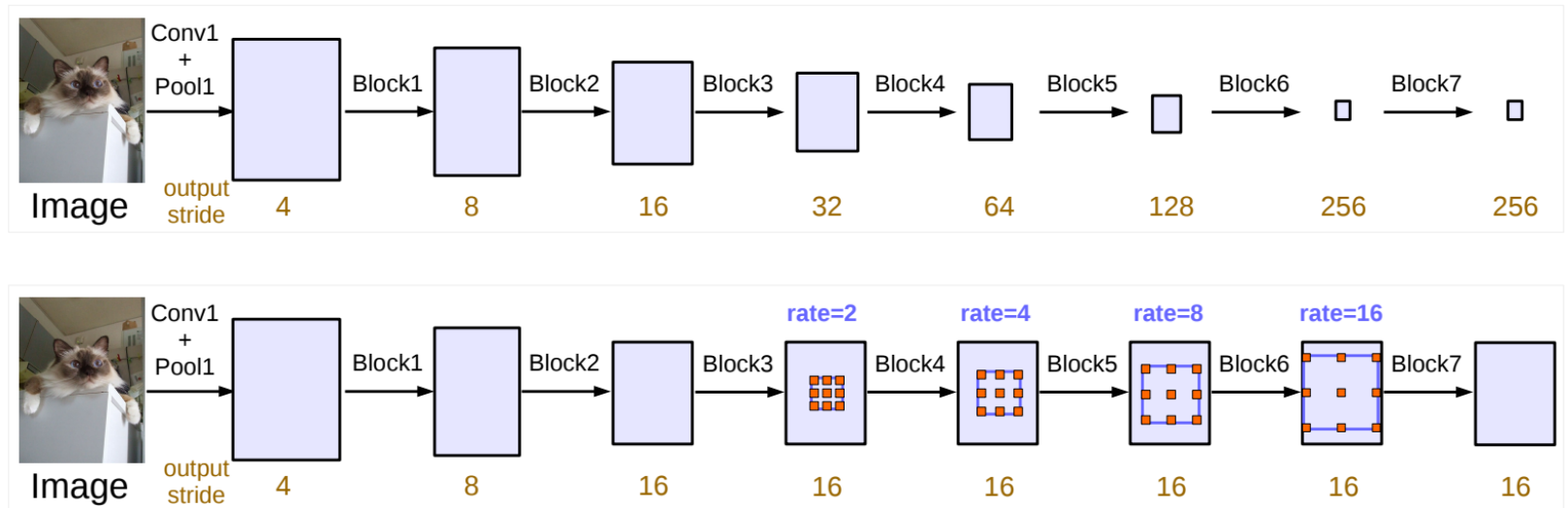
Long et al., CVPR, 2015

DeepLabv3



Chen et al., arXiv, 2017

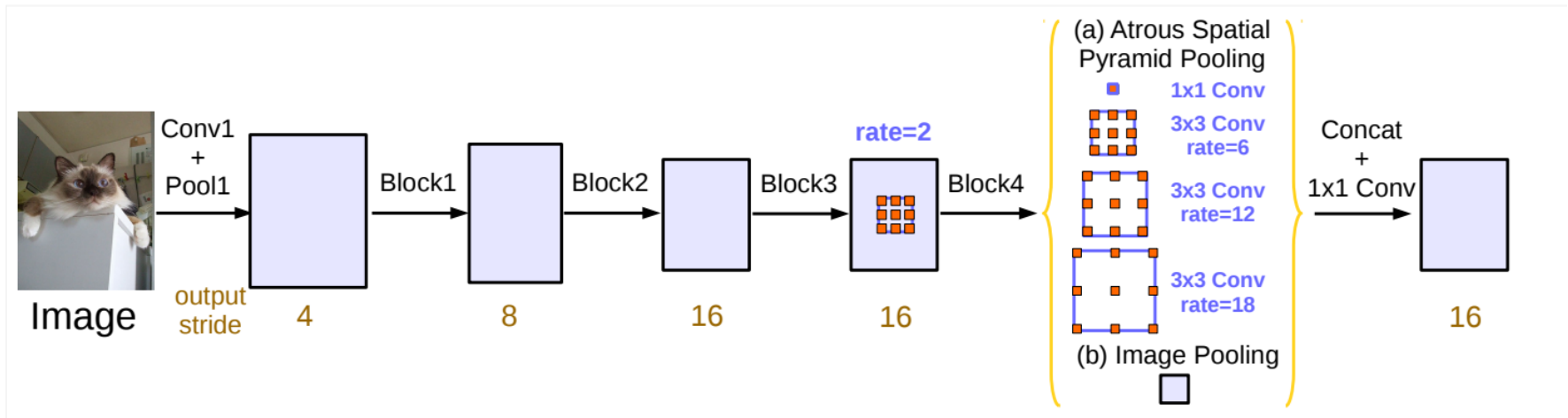
DeepLabv3



Chen et al., arXiv, 2017

DeepLabv3

Pretrained ResNet

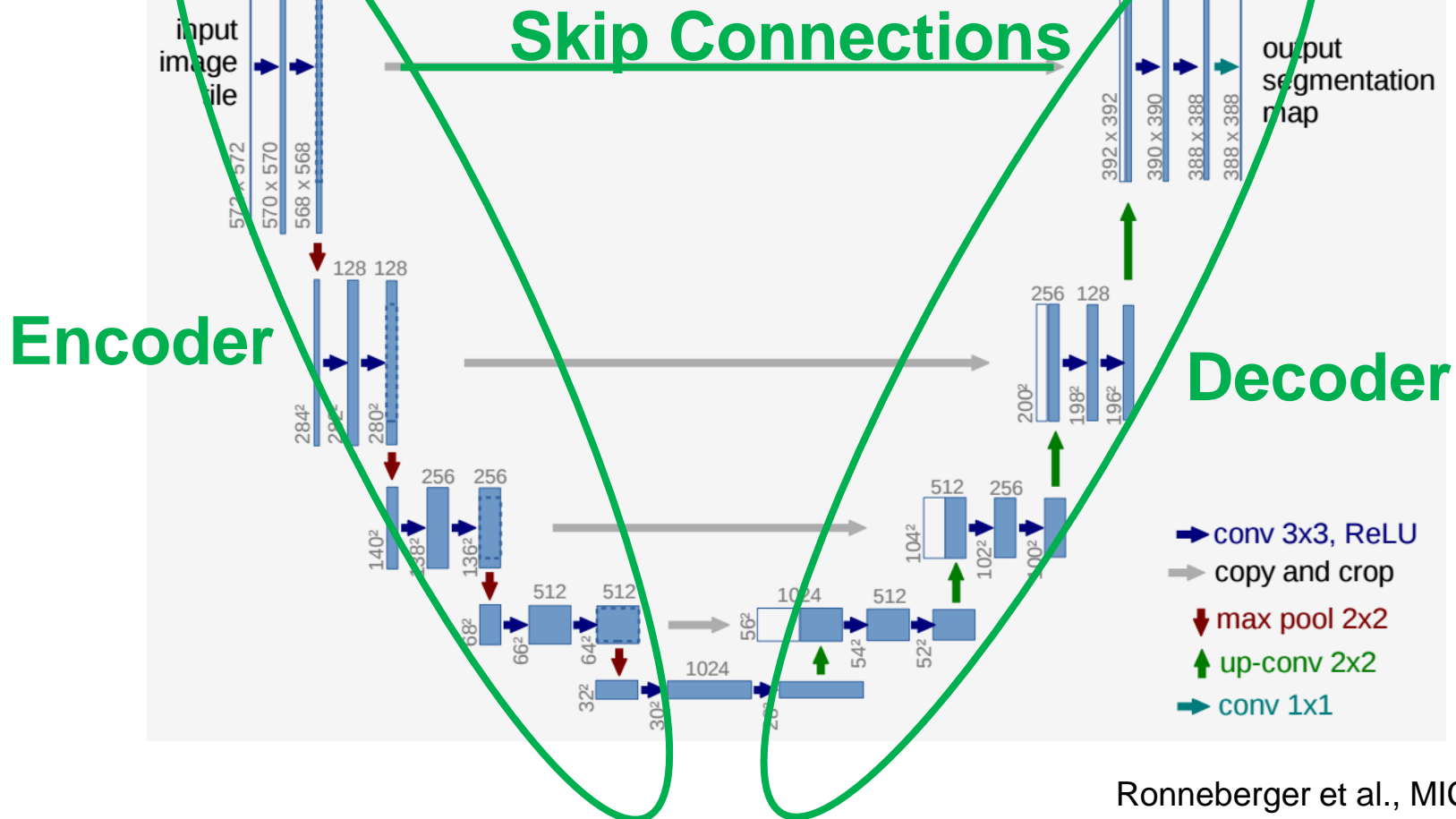


- + Effective utilization of pretrained network
- + Performs very well on semantically demanding tasks
- Output stride 16/8 -> no precise localization
- No pretraining possible for MIC

Chen et al., arXiv, 2017

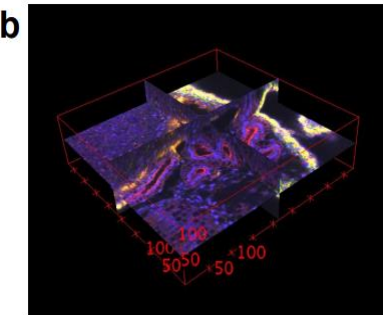
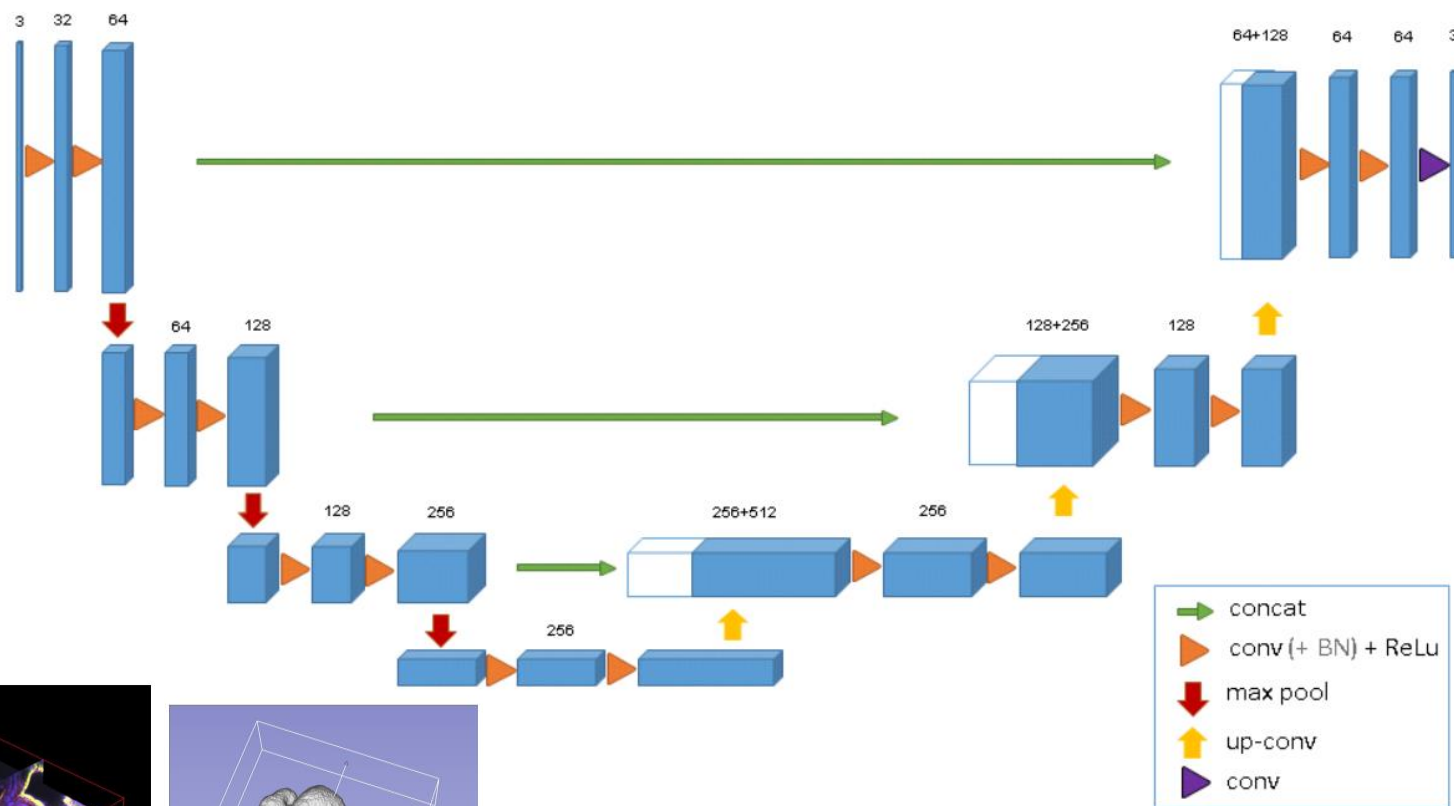
Encoder-Decoder: UNet

output stride 1!

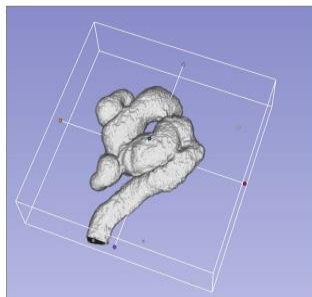


Ronneberger et al., MICCAI, 2015

UNet variants: UNet3D



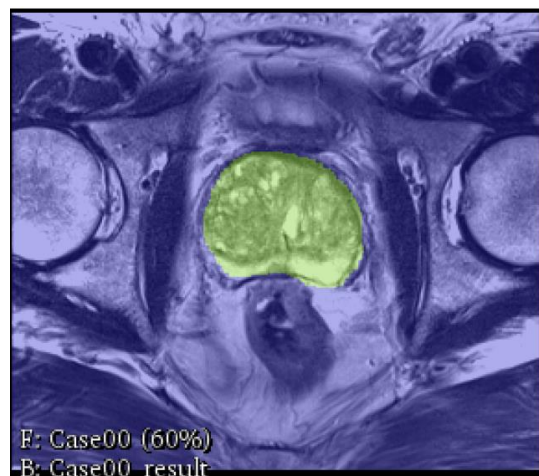
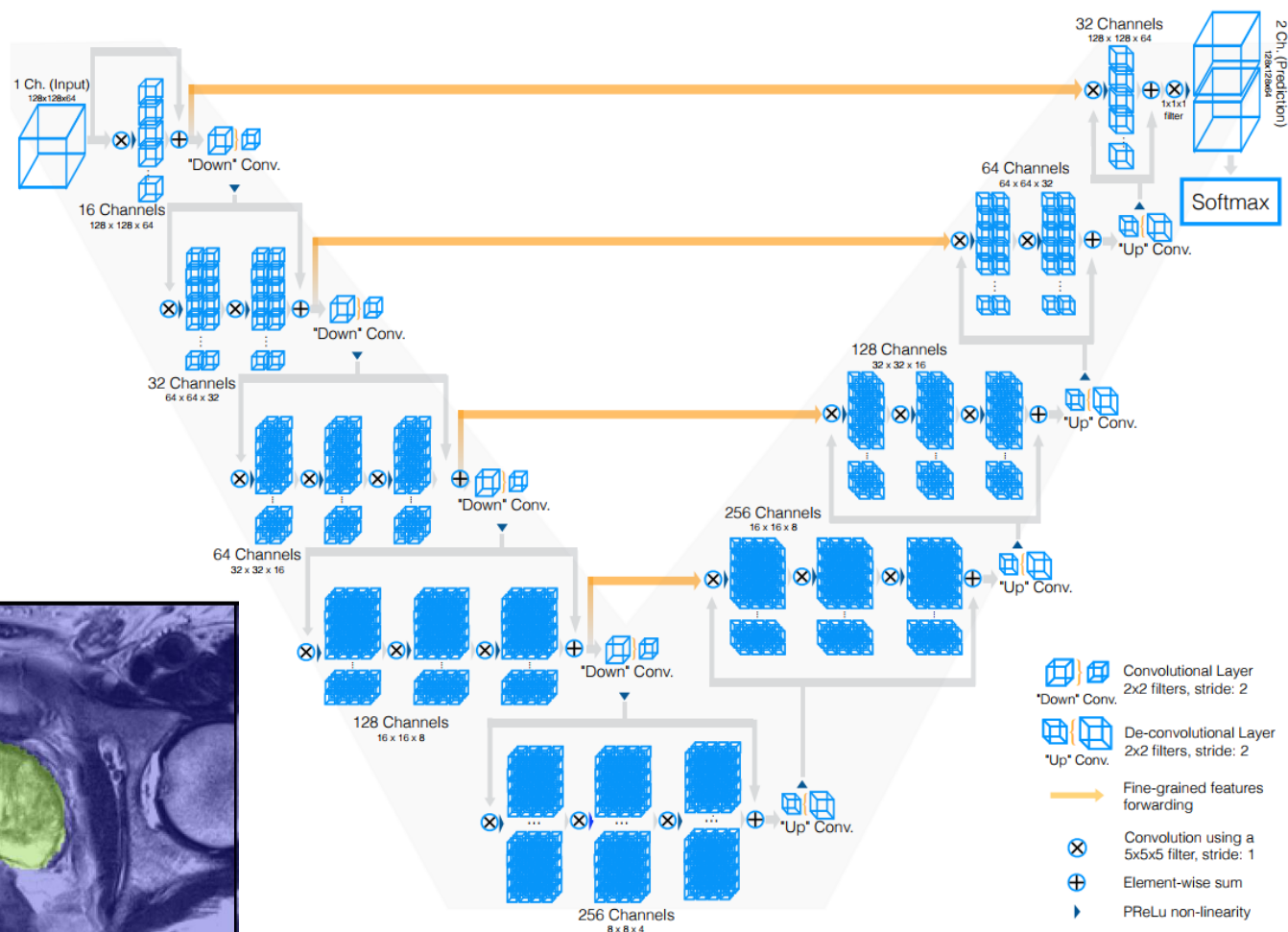
raw image



dense segmentation

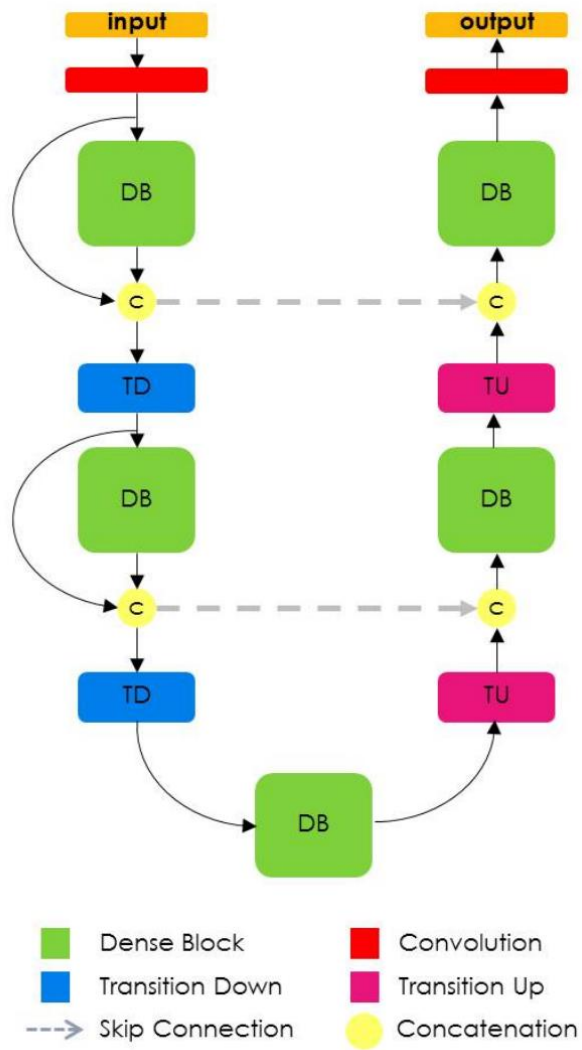
Cicek et al., MICCAI, 2016

UNet variants: V-Net

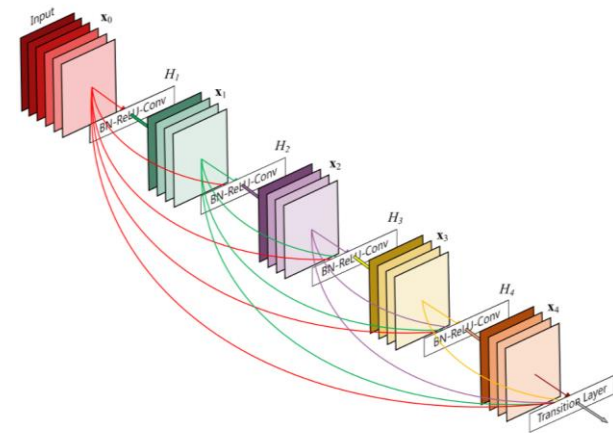


Milletari et al., 3DV, 2016

UNet variants: Dense UNet



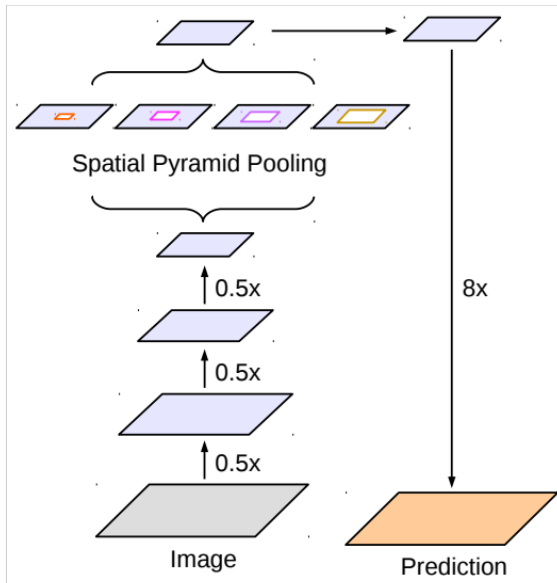
Jegou et al., CVPRW, 2017



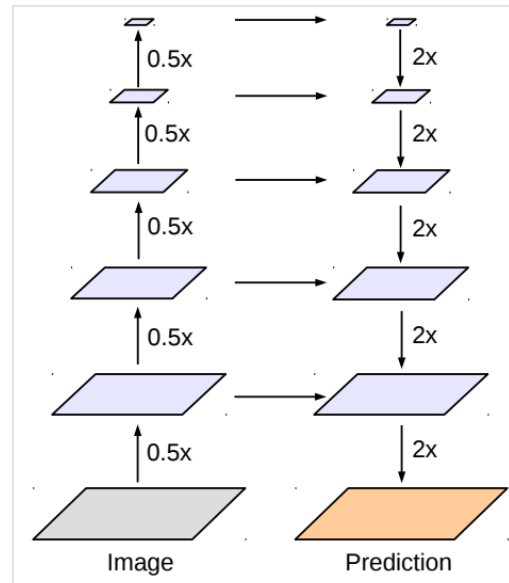
Huang et al., CVPR, 2017



DeepLabv3+



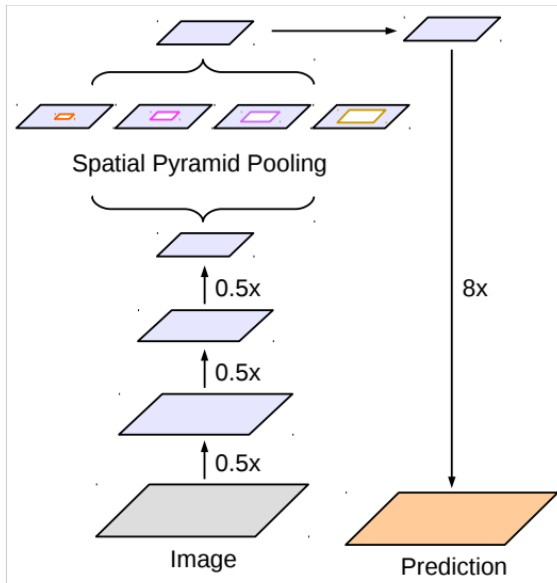
DeepLabv3



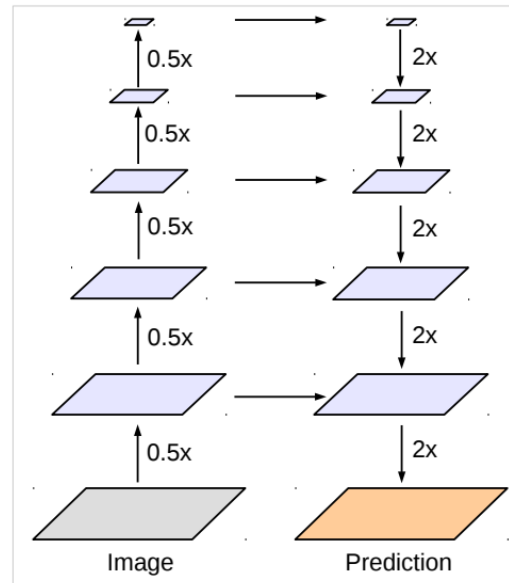
Encoder-Decoder

Chen et al., arXiv, 2018

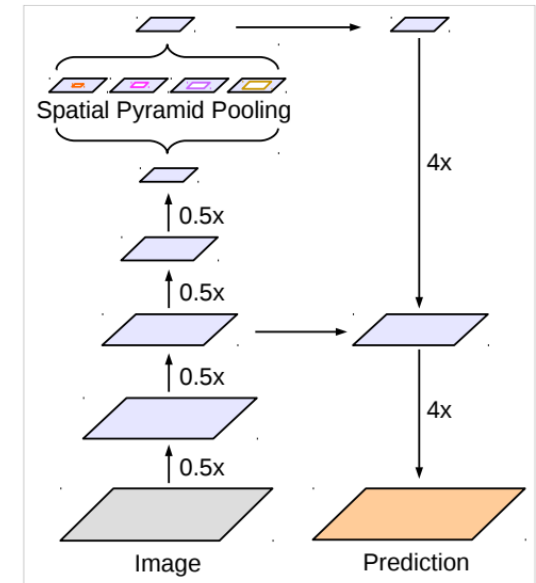
DeepLabv3+



DeepLabv3



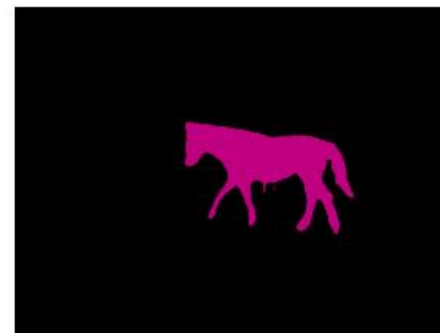
Encoder-Decoder



DeepLabv3+

Chen et al., arXiv, 2018

DeepLabv3+



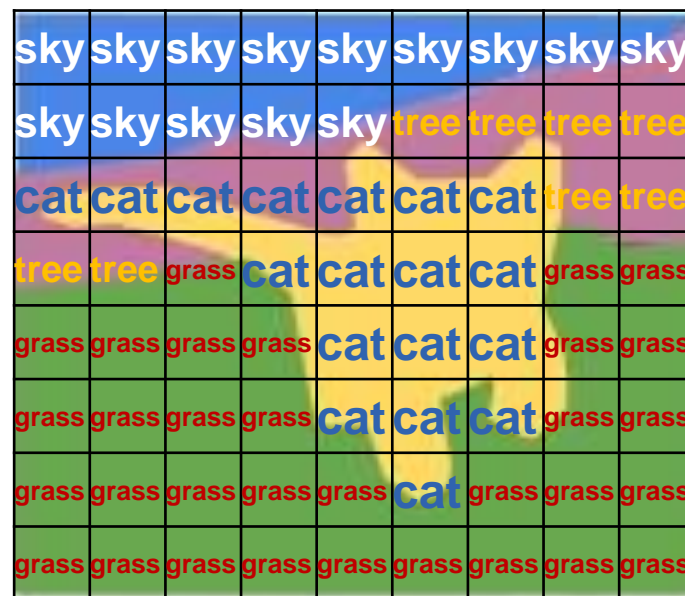
(a) Image

(b) w/ BU

(c) w/ Decoder

Chen et al., arXiv, 2018

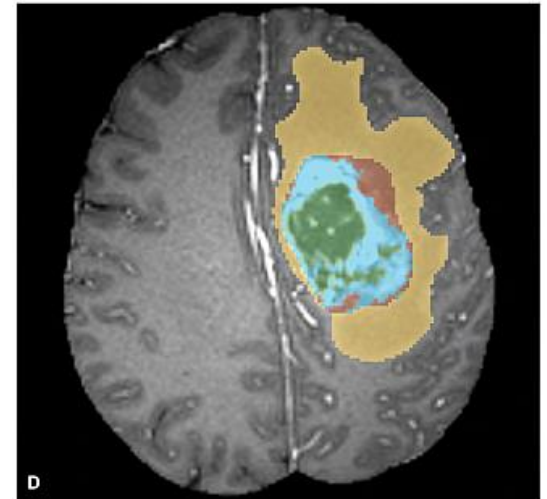
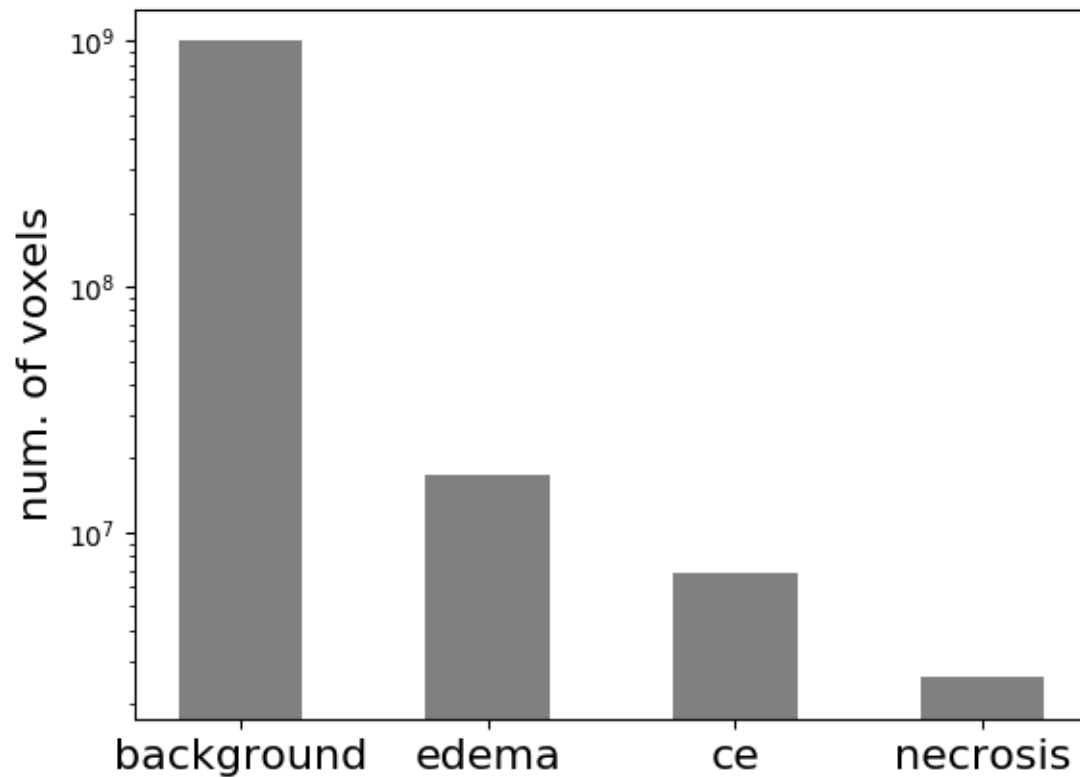
Medical Image Segmentation – Loss functions



Categorical Crossentropy

$$L_{CE}(p, g) = - \frac{1}{N} \sum_{i \in N} \sum_{k \in K} g_{i,k} \log p_{i,k}$$

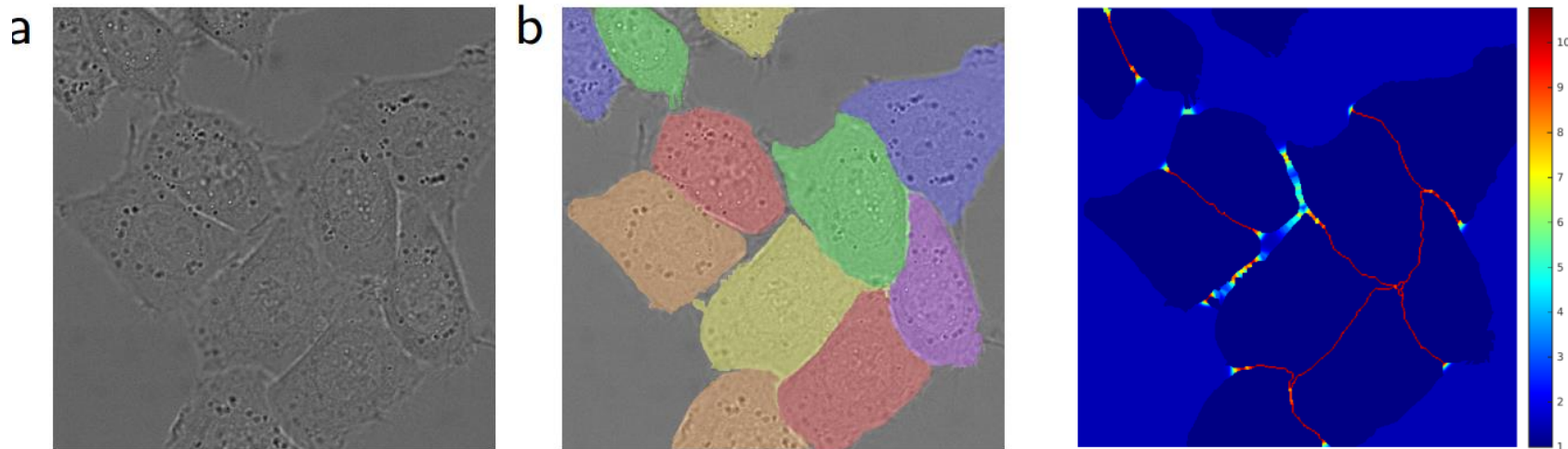
Medical Image Segmentation – Loss functions



Medical Image Segmentation – Loss functions

Categorical Crossentropy (weighted)

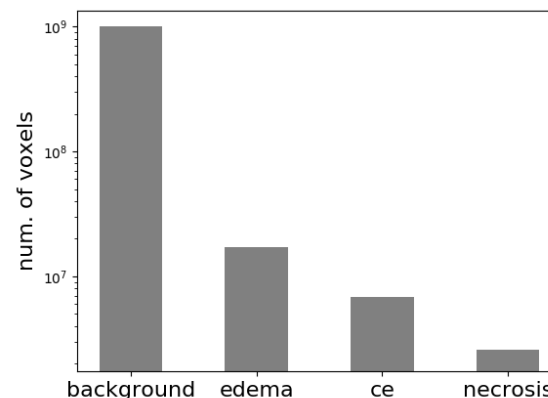
$$L_{WCE}(p, g) = -\frac{1}{N} \sum_{i \in N} \sum_{k \in K} w_i g_{i,k} \log p_{i,k}$$



Ronneberger et al., MICCAI, 2015

Medical Image Segmentation – Loss functions

$$DICE(A, B) = \frac{2 |A \cap B|}{|A| + |B|}$$



Dice Loss

$$L_{DICE}(p, g) = - \frac{2}{|K|} \sum_{k \in K} \frac{\sum_{i \in N} p_{i,k} g_{i,k}}{\sum_{i \in N} p_{i,k} + \sum_{i \in N} g_{i,k}}$$

Milletari et al., 3DV, 2016

Sudre et al., DLMIA/ML-CDS (MICCAI), 2017

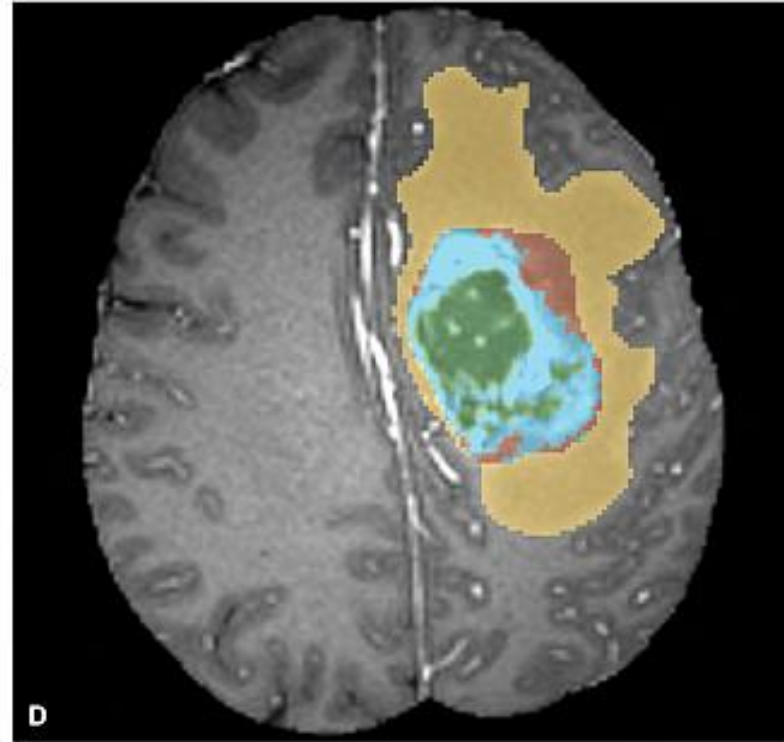
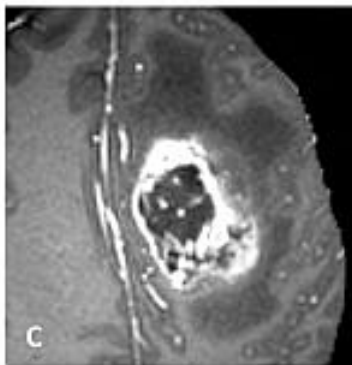
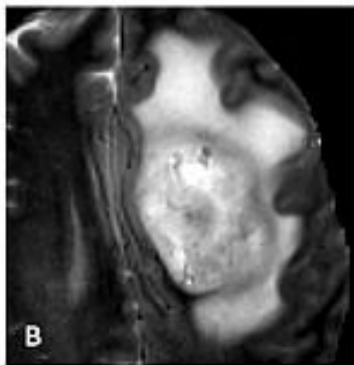
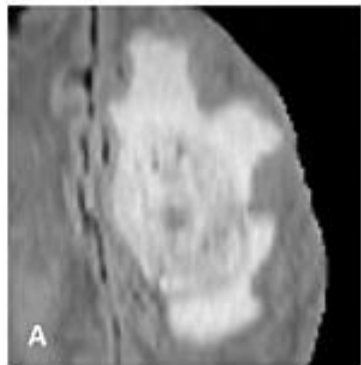
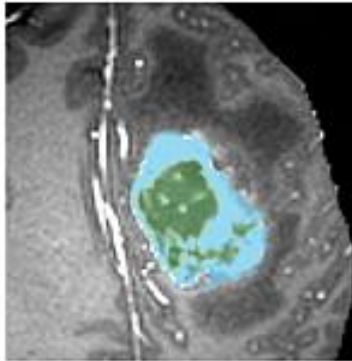
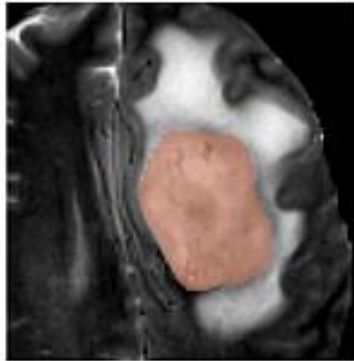
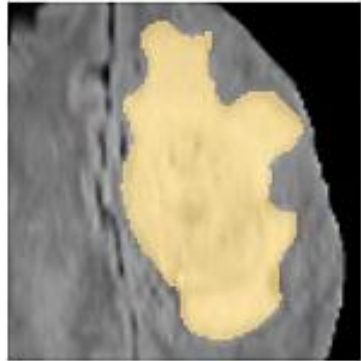
Drozdal et al., DLMIA/LABELS (MICCAI), 2016

Medical Image Segmentation – Example: BraTS

Whole Tumor

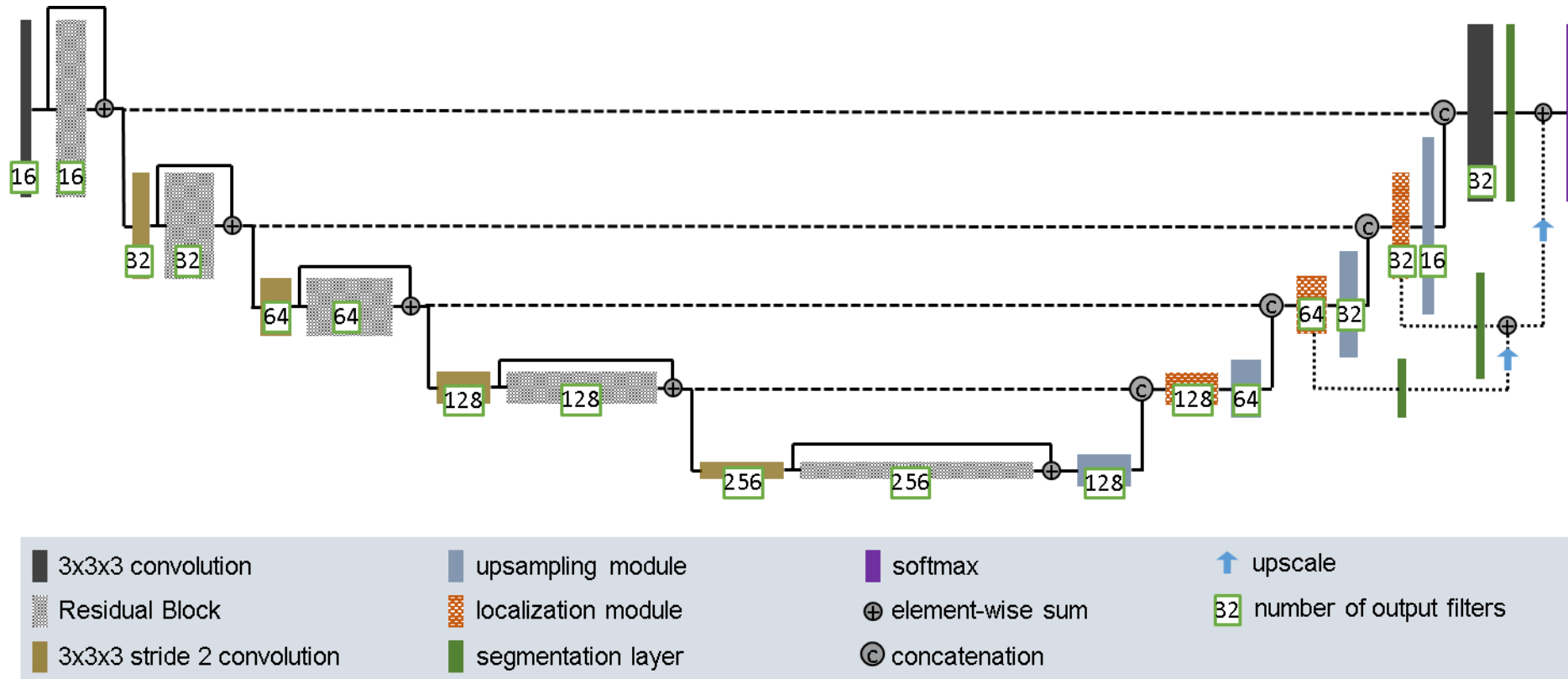
Tumor Core

Necrosis and
Enhancing Tumor



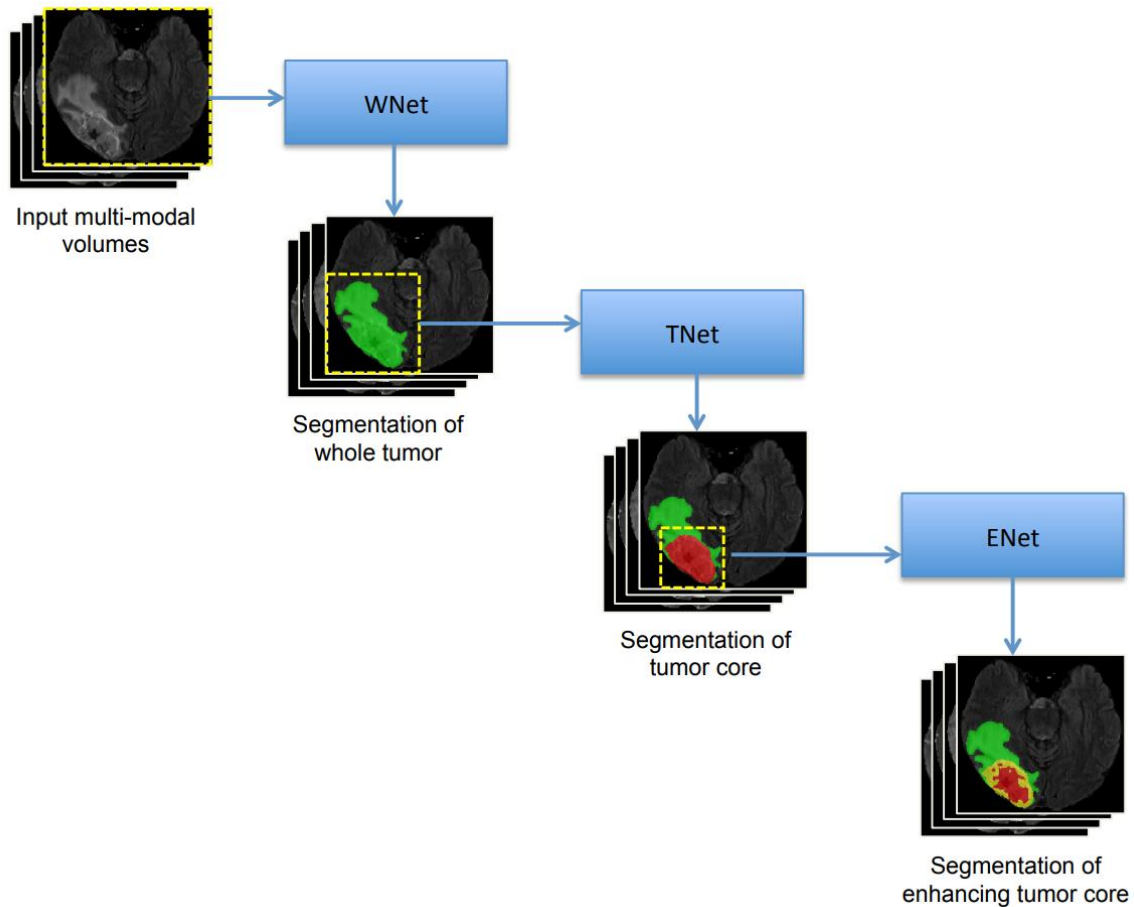
www.braintumorsegmentation.org

BraTS 2017 3rd Place (you can get a long way with a well trained UNet)



- Train on large patches (128x128x128)
- DICE loss
- A lot of data augmentation

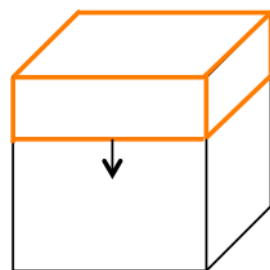
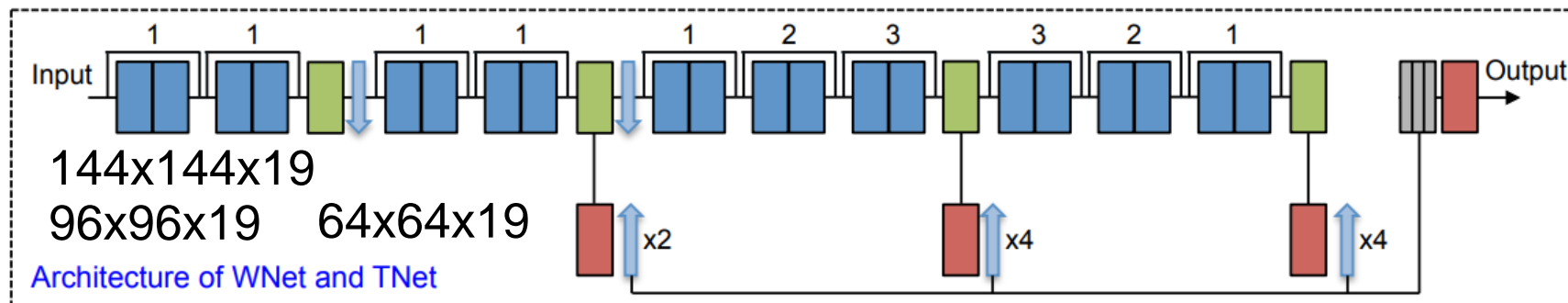
BraTS 2017 2nd Place



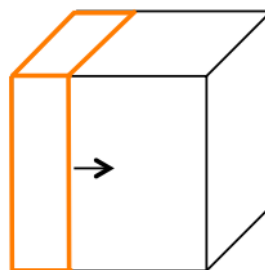
Cascade: simulate annotation procedure

Wang et al., MICCAI-BRATS, 2017

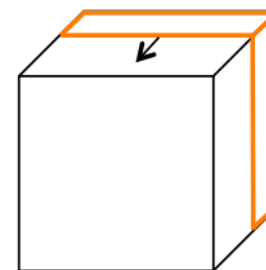
BraTS 2017 2nd Place



axial



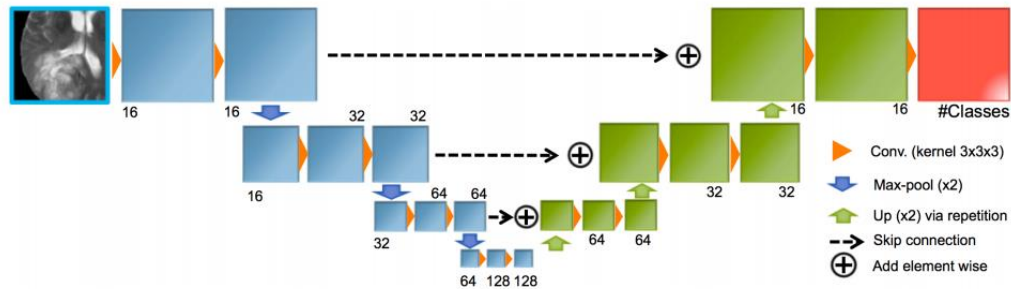
sagittal



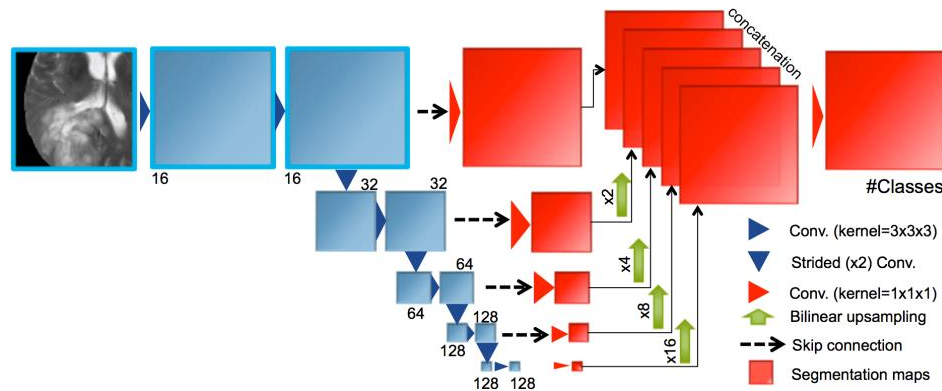
coronal

Wang et al., MICCAI-BRATS, 2017

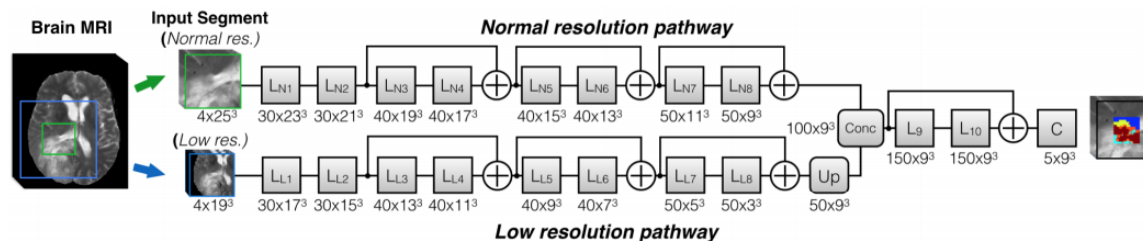
BraTS 2017 1st Place



UNet (3D)



FCN (3D)

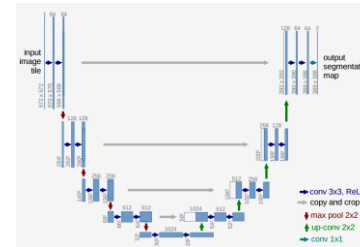


DeepMedic

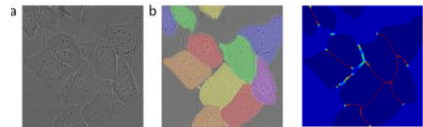
Kamnitsas et al., MICCAI-BRATS, 2017

Segmentation in Medical Image Computing: Short Summary

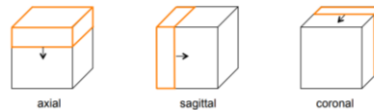
- Use Encoder-Decoder with OS1



- Think about your loss function $L_{DICE}(p, g) = - \frac{2}{|K|} \sum_{k \in K} \frac{\sum_{i \in N} p_{i,k} g_{i,k}}{\sum_{i \in N} p_{i,k} + \sum_{i \in N} g_{i,k}}$



- Properly aggregate semantic information



- Data Augmentation



- Ensembling

Thank you!