

CNN segmentation: U-net

Ronneberger: U-net, MICCAI 2015

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2020

Previous segmentation strategies

- ▶ Sliding window
- ▶ Predict a class of the central pixel
- ▶ Many patches per image → not many images needed
- ▶ Slow
- ▶ Patch size - localization accuracy vs. context

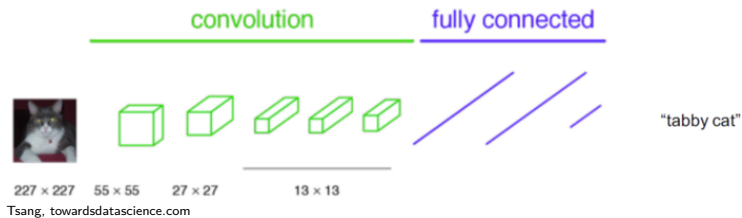
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Fully convolutional networks

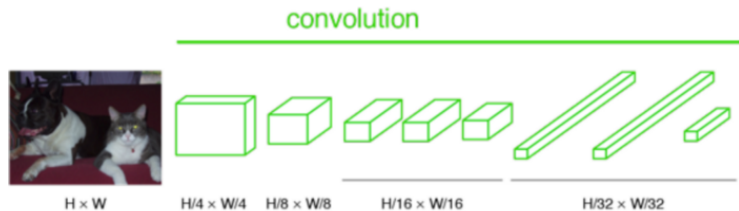
- ▶ no fully connected layers, only convolutional layers
- ▶ output - a (smaller) image

Classification network

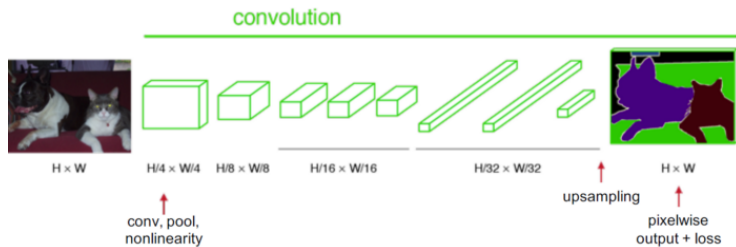


Sik-Ho

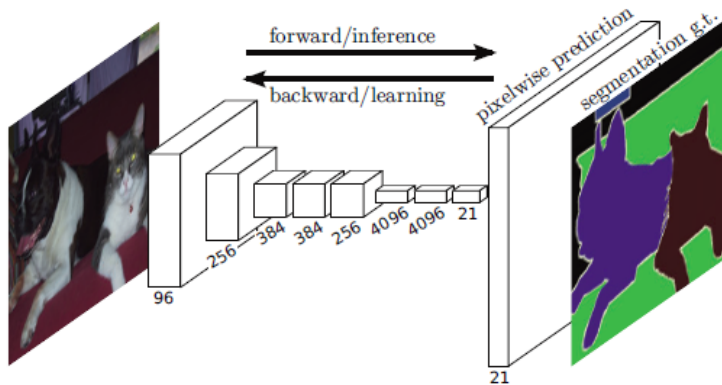
Fully convolutional network



Segmentation

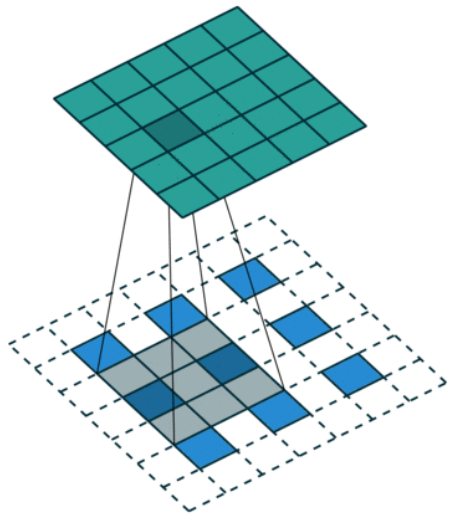


Fully convolutional structure

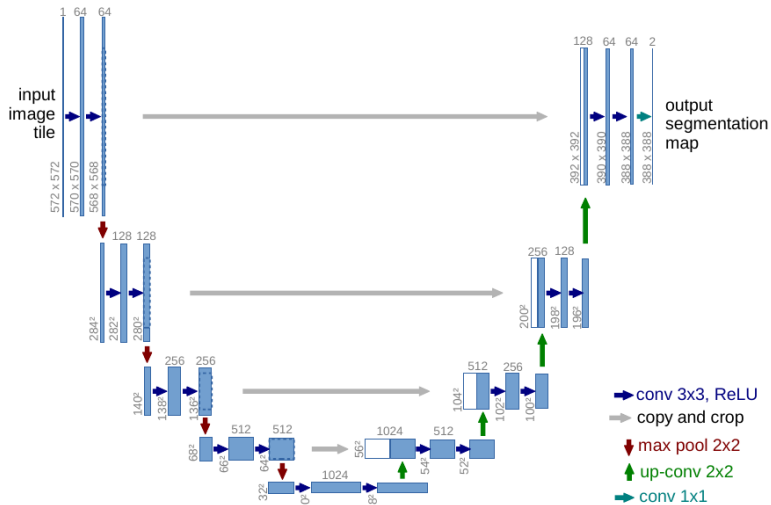


Upconvolution

Transposed convolution, (Deconvolution)



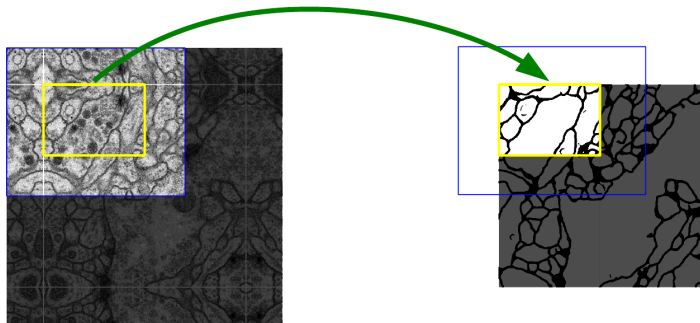
U-net structure



U-net notes

- ▶ no fully connected layers
- ▶ upsampling to get to the original resolution
- ▶ skip connections
- ▶ small kernels (3×3)
- ▶ only fully valid convolutions, no padding \rightarrow cropping
- ▶ 2×2 max-pooling \rightarrow fixed-size input, even size at all levels
- ▶ number of channels increase
- ▶ softmax in the final level
- ▶ “downsampling” block can be replaced
- ▶ extensive augmentation (elastic deformation)
- ▶ random weight initialization with $\sigma = \sqrt{2/N}$
- ▶ dropout

Tiling strategy



Loss function

► Cross-entropy

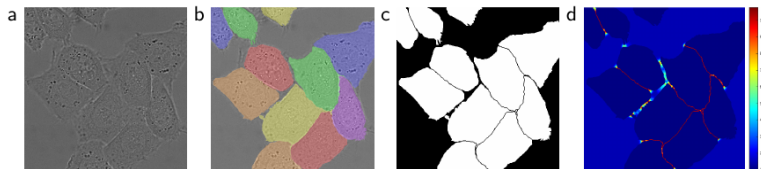
$$E = \sum_{\mathbf{x} \in \Omega} w(\mathbf{x}) \log(p_{\ell(\mathbf{x})}(\mathbf{x}))$$

The separation border is computed using morphological operations. The weight map is then computed as

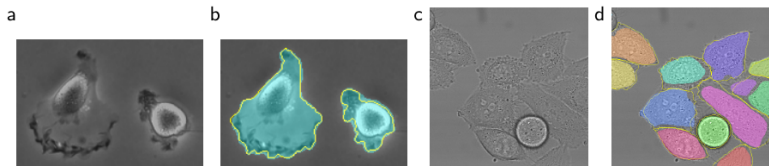
$$w(\mathbf{x}) = w_c(\mathbf{x}) + w_0 \cdot \exp\left(-\frac{(d_1(\mathbf{x}) + d_2(\mathbf{x}))^2}{2\sigma^2}\right) \quad (2)$$

where $w_c : \Omega \rightarrow \mathbb{R}$ is the weight map to balance the class frequencies, $d_1 : \Omega \rightarrow \mathbb{R}$ denotes the distance to the border of the nearest cell and $d_2 : \Omega \rightarrow \mathbb{R}$ the distance to the border of the second nearest cell. In our experiments we set $w_0 = 10$ and $\sigma \approx 5$ pixels.

Border weight example



Example segmentation



Quantitative results

IoU, cell tracking challenge

Name	PhC-U373	DIC-HeLa
IMCB-SG (2014)	0.2669	0.2935
KTH-SE (2014)	0.7953	0.4607
HOUS-US (2014)	0.5323	-
second-best 2015	0.83	0.46
u-net (2015)	0.9203	0.7756