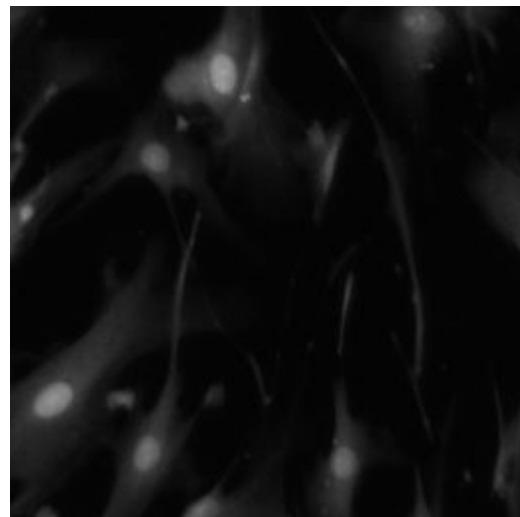
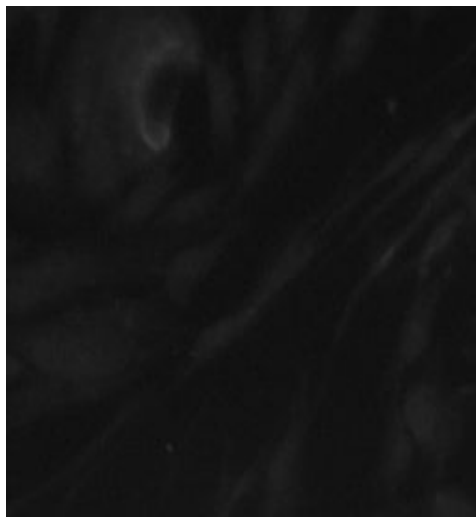
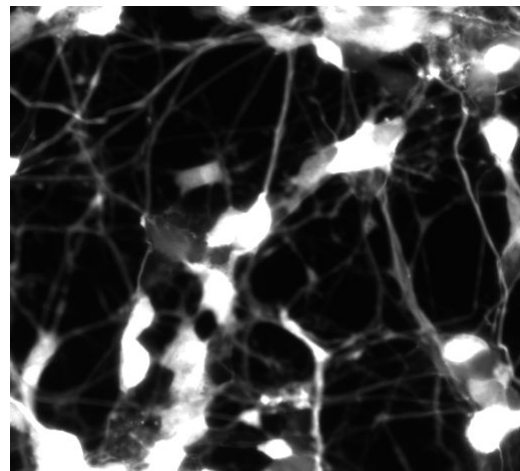
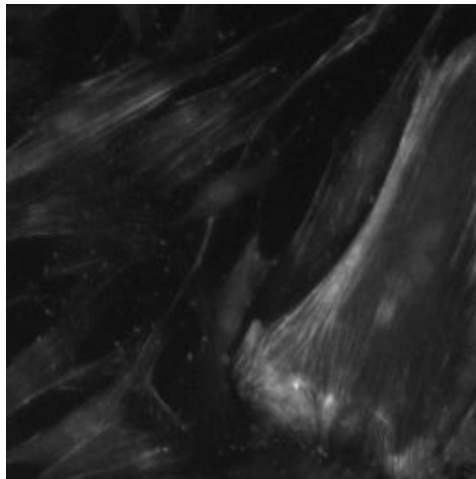


A U-Net like Solution to the 2018 Data Science Bowl Challenge

François Delarbre, Nathan Greffe and Pierre Keutgen

Problem

- Spot nuclei in images of cells
- Images are very heterogeneous
- Small dataset



Solution

Instance Segmentation are usually solved using either

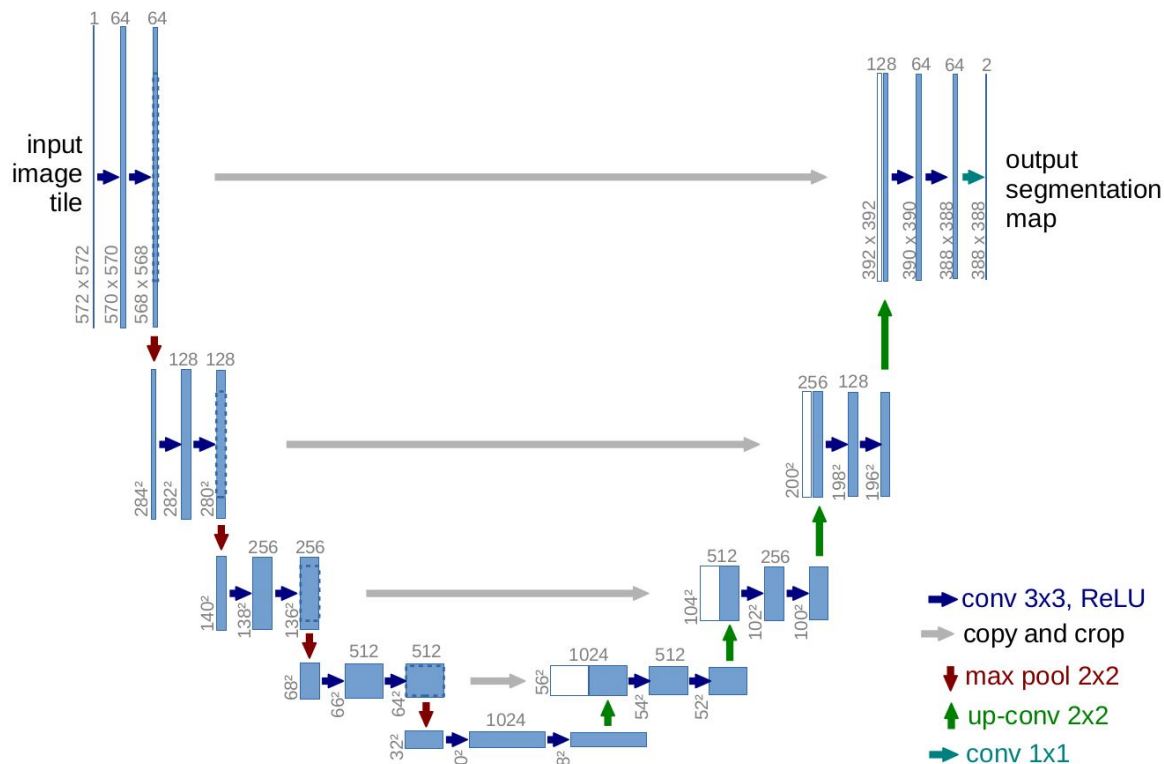
- Semantic segmentation network (U-Net) + post-processing (mainly watershed transform)
- Mask-RCNN based methods

Preprocessing

- Additional masks : Union, borders and centers
- Additional dataset from another competition
- Heavy data augmentation

Semantic Segmentation (1) : U-NET

- Downsampling path
- Upsampling path
- Skip connections between both
- Can be adapted for many architectures (ResNet, Xception, MobileNetV2, ...) and encoder weights can be pretrained (e.g. on ImageNet)
- Paper introduced many other ideas

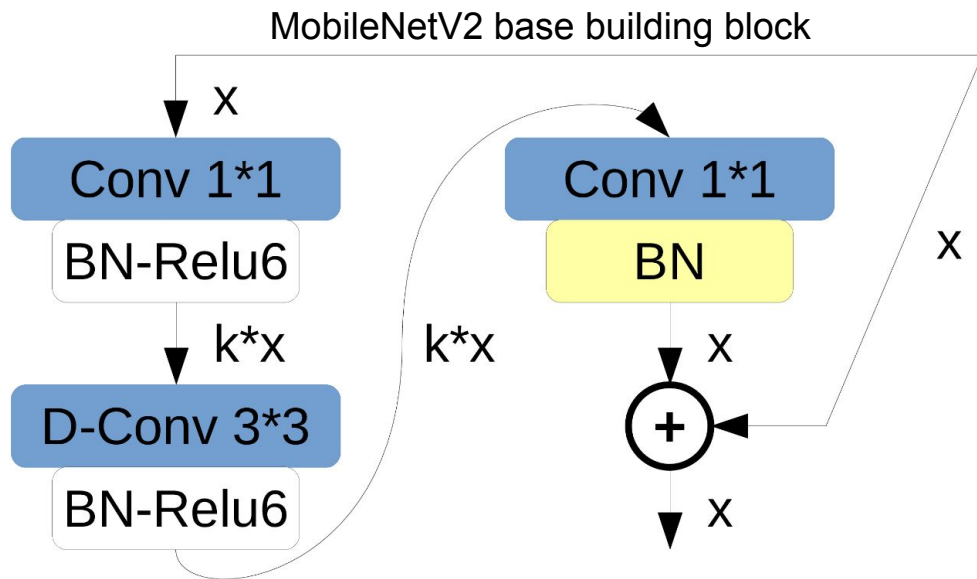


Semantic Segmentation (2)

After some experiments we choose to use:

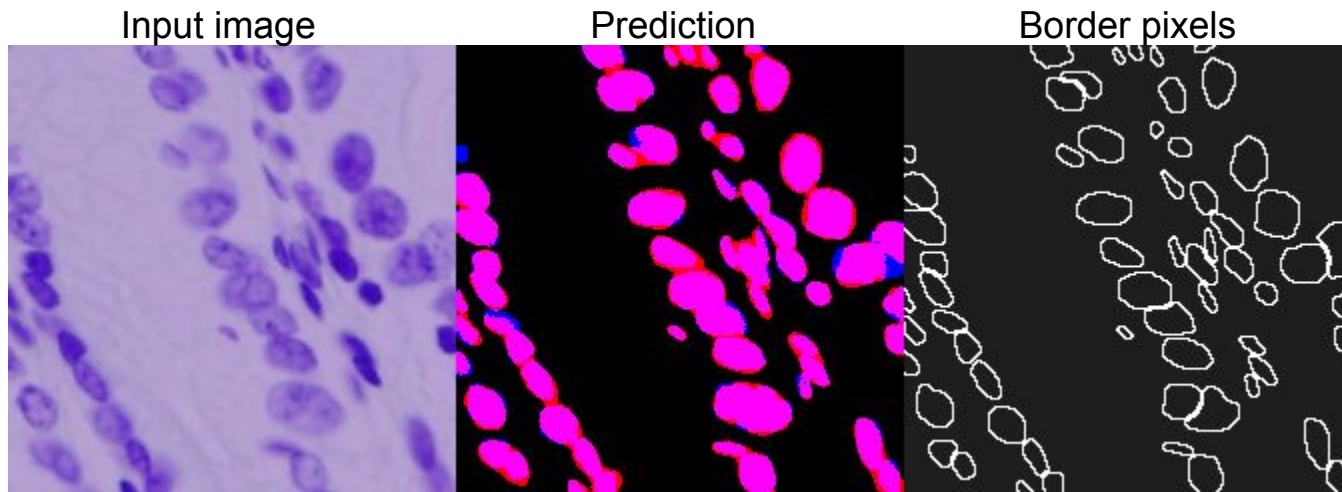
- Pretrained weights
- MobileNetV2-inspired Architecture (ResNet/ResNeXt/DenseNet were tested as well)
- Adam optimizer
- Loss is a composition of binary-cross-entropy and soft Dice Loss:

$$\text{softDC} = \frac{2 * \text{sum}(y_{\text{true}} * y_{\text{pred}}) + 1}{\text{sum}(y_{\text{true}} + y_{\text{pred}}) + 1}$$



Semantic Segmentation (3)

- Weight more importantly pixels from borders of cells (idea used in U-Net)



Purple: TP Dark: TN
Blue: FN Red: FP

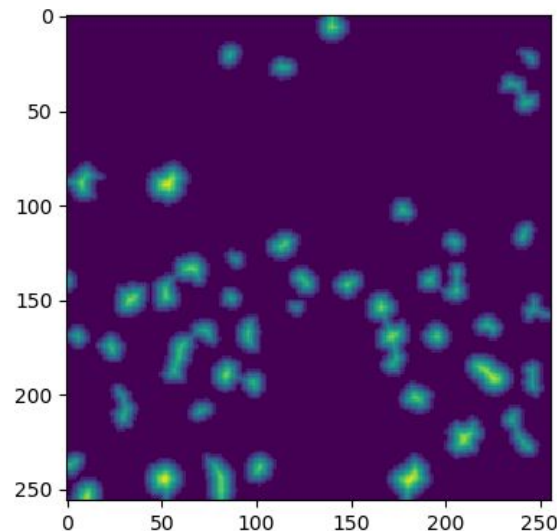
Post Processing

Watershed transform: critical part is to place the markers

1) Markers as local minima of the distance transform on “is a cell mask”

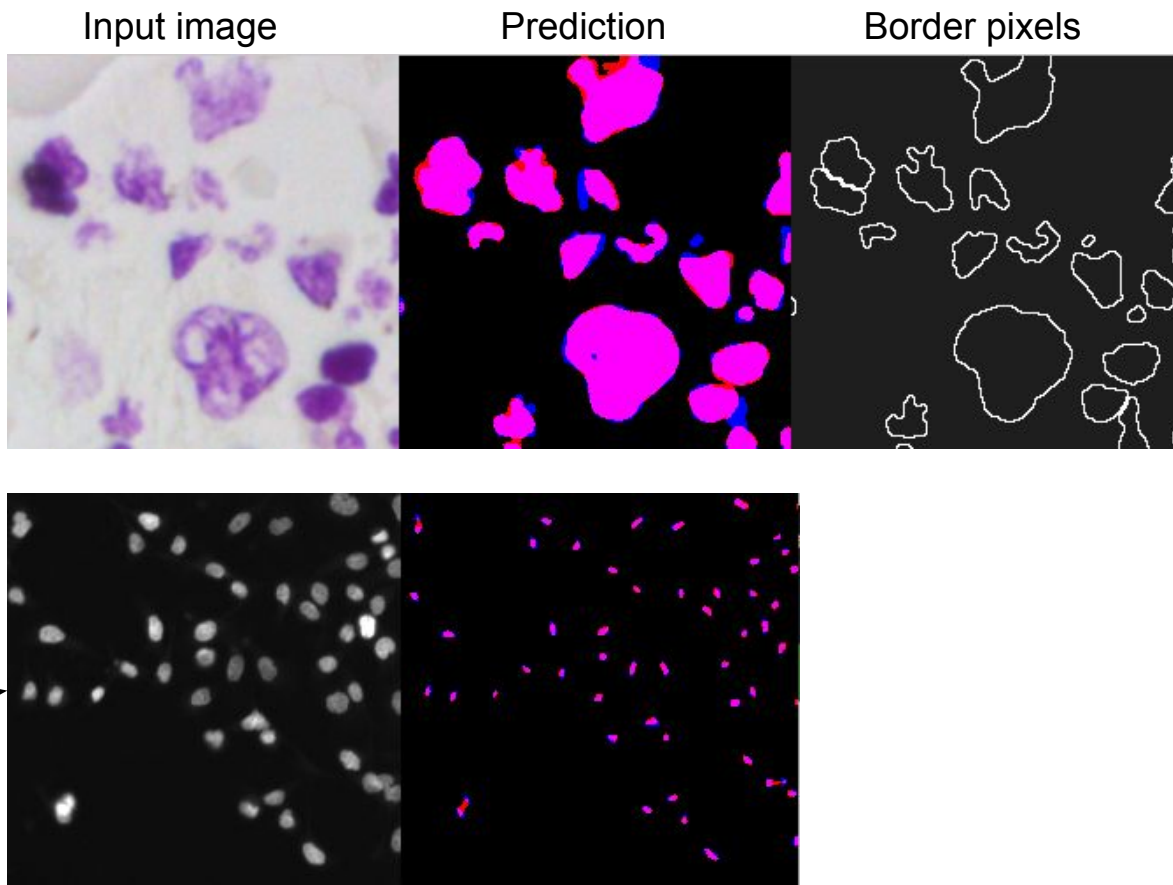
2) 1) helped by the predicted borders

3) Predicted centers as markers



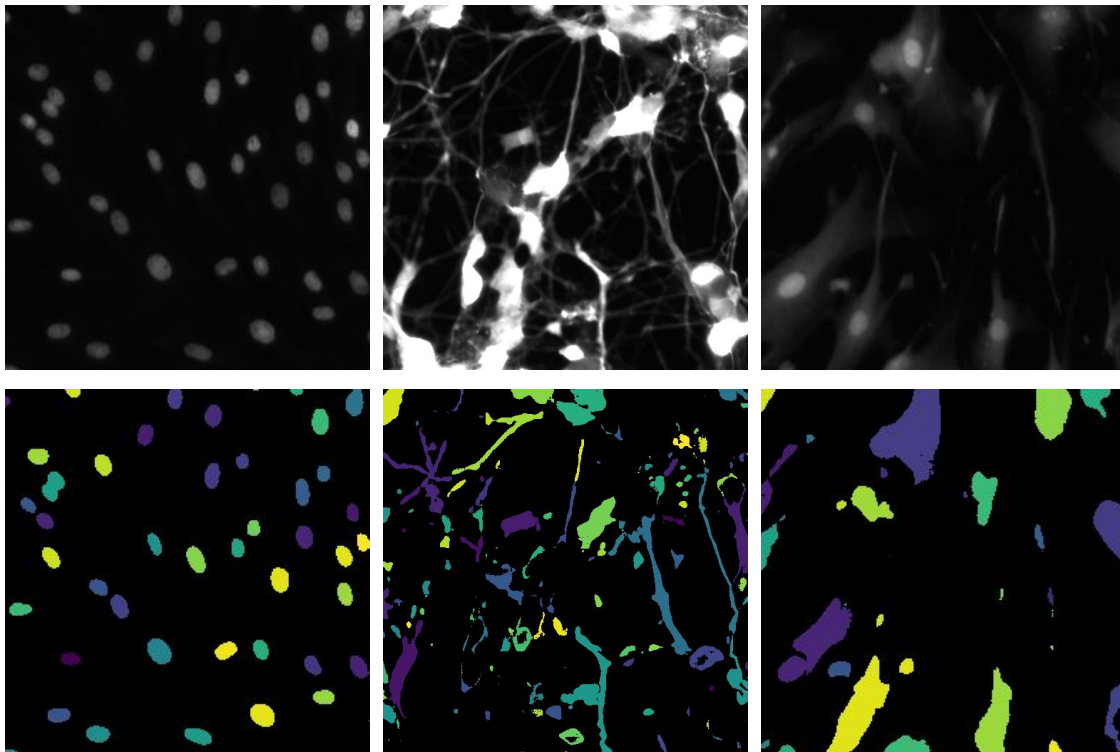
Final solution

- U-Net + emphasis on borders for semantic segmentation mask →
- Watershed with the predicted centers as markers
- U-Net with same architecture for centers →



Scores with our final solution

- The score is computed following the competition formula
- On our validation set: 0.279
- On our test set: 0.226



To go further

- More post-processing
- Deep-watershed transform
- Larger dataset
- Deeper networks