PMLDL. X-Net

Anna Boronina

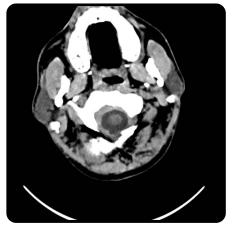
What is a brain stroke (ru: инсульт)?

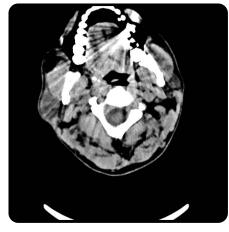
Brain stroke occurs when the blood supply to part(s) of one's brain is interrupted or reduced.

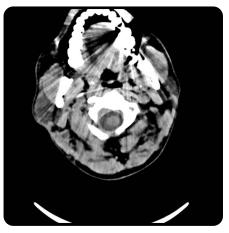


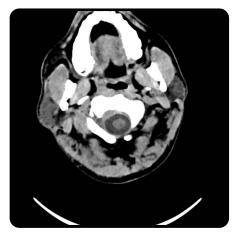
Problems

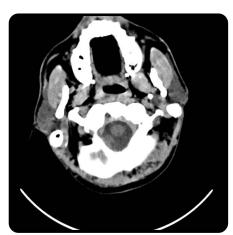
- sometimes even doctors cannot spot a brain stroke
- sometimes there is a sequence of MRI scans which should be processed together
- healthy VS stroke:

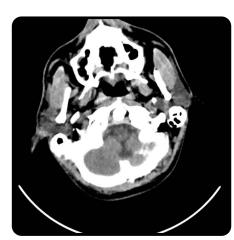




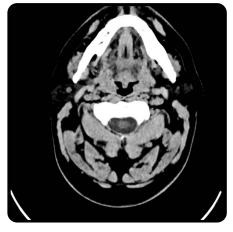


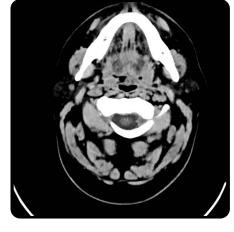




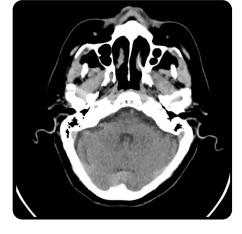














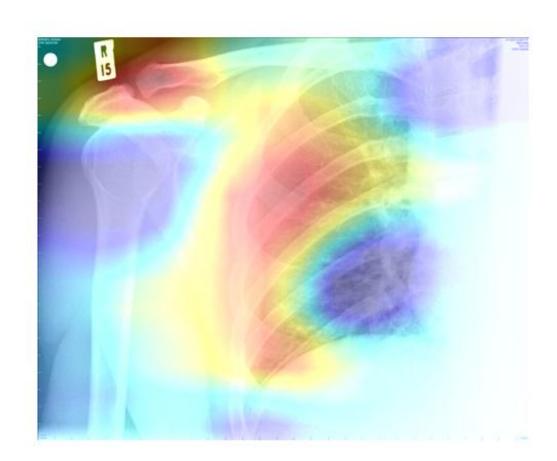


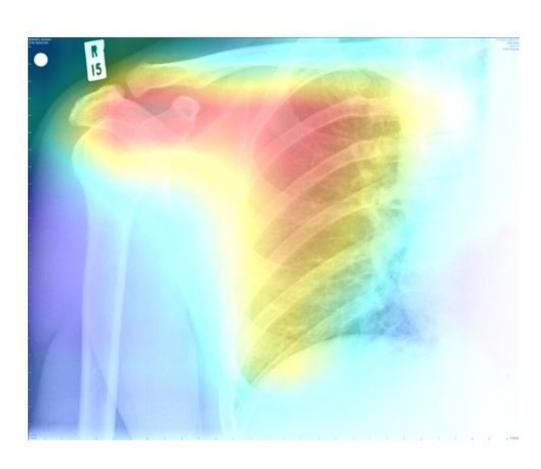


DL applications

- use Deep Learning model as a complementary tool
- use methods that explain model's decisions
 - GradCam
 - GradCam++
 - LIME
 - **.**..







X-Net Paper: Qi et. al



<u>International Conference on Medical Image Computing and Computer-Assisted Intervention</u>

MICCAI 2019: <u>Medical Image Computing and Computer Assisted Intervention – MICCAI 2019</u> pp 247-255 | <u>Cite as</u>

X-Net: Brain Stroke Lesion Segmentation Based on Depthwise Separable Convolution and Long-Range Dependencies

Authors Authors and affiliations

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Conference paper

First Online: 10 October 2019



Part of the Lecture Notes in Computer Science book series (LNCS, volume 11766)



Previous attempts

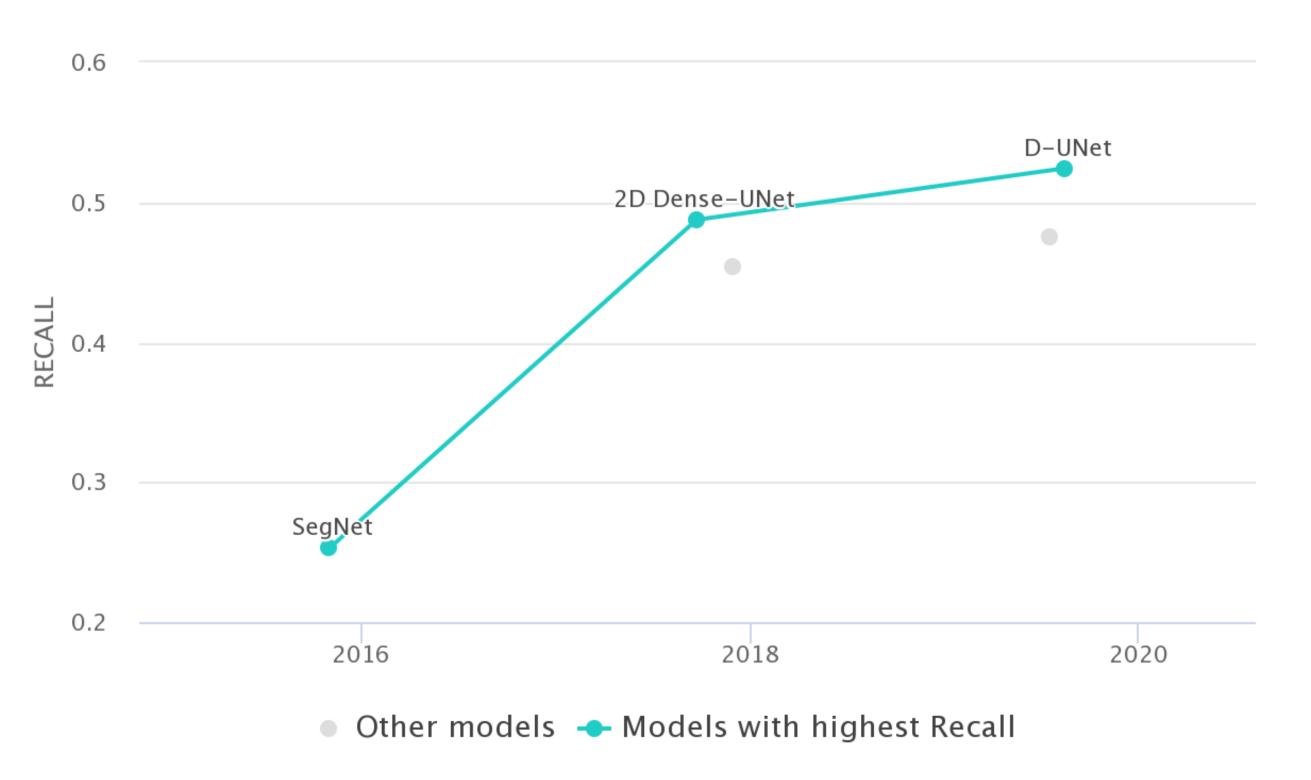
- SegNet symmetrical convolutional autoencoder
- UNet originally, non-residual convolutional network with "crop-and-concat"
- 2D Dense Unet residual dense network

Their problems according to Qi et. al

- heavy network parameters
- cannot capture long-range dependencies

Benchmark - ATLAS Dataset

Recall: How many relevant items are retrieved?

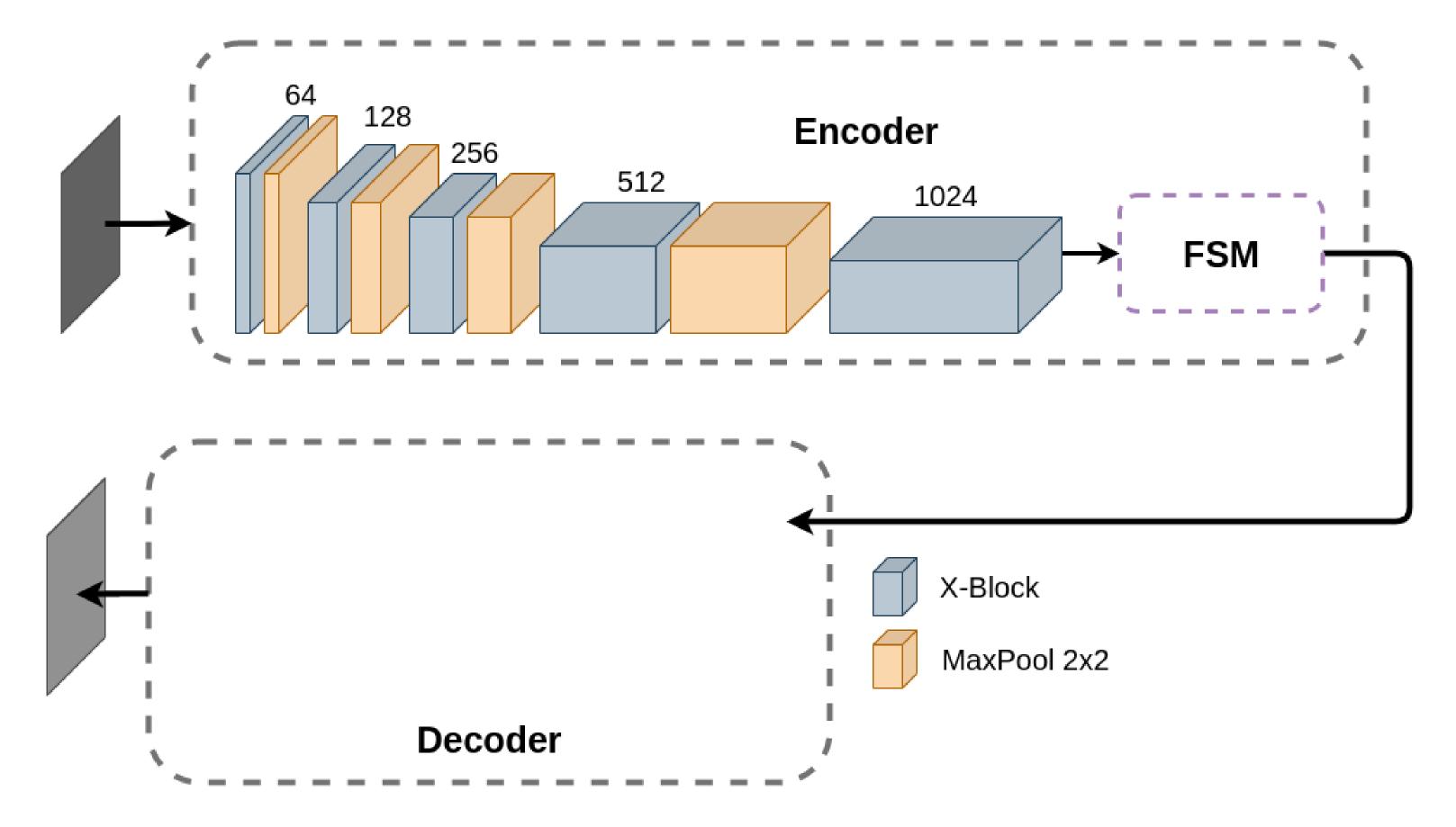




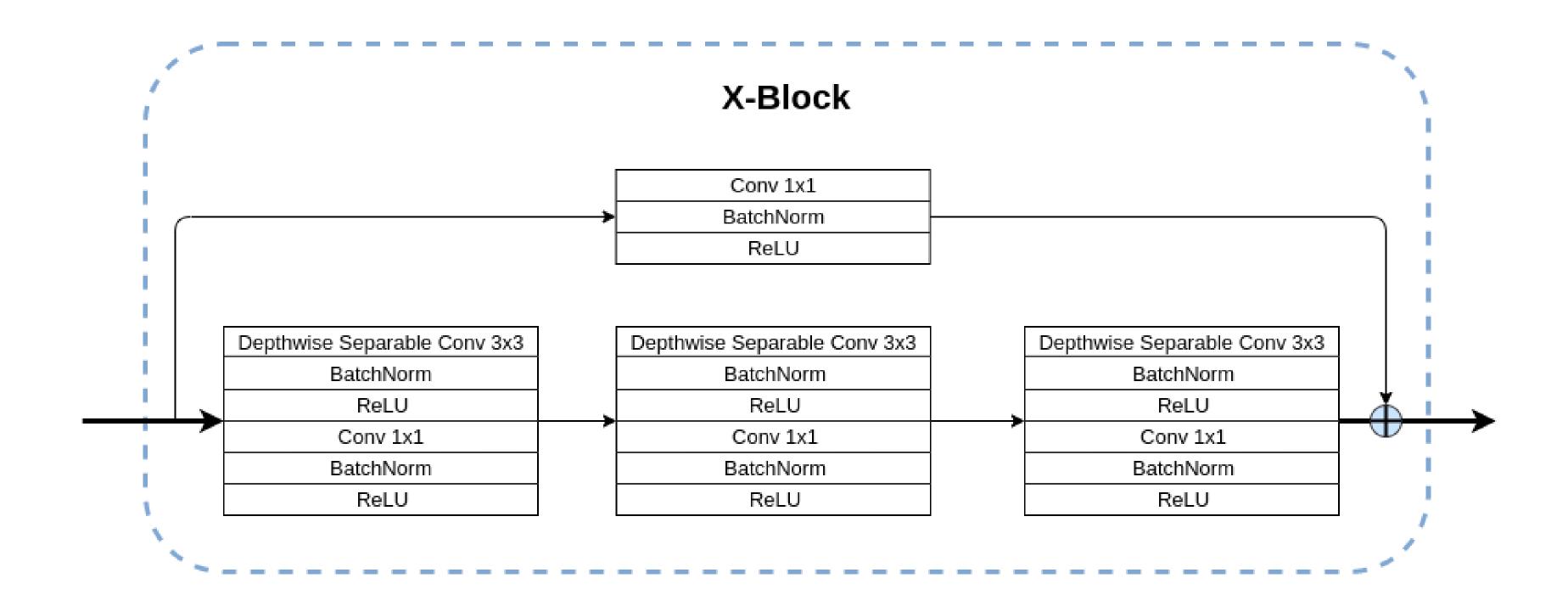
Solution proposed by Qi et. al

- reduce number of parameters
 - separable convolution
 - it is efficient
 - gives better segmentation results
 - basic upsampling
- Feature Similarity Module (FSM) to capture long-range dependencies
- skip-and-concat to solve vanishing gradient problem and overfitting

Architecture. Encoder



Encoder. XBlock



Encoder. X-Block. Separable Convolution

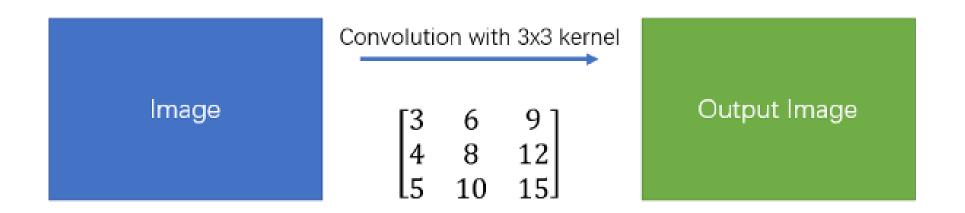
Separating a 3x3 kernel spatially

$$\begin{bmatrix} 3 & 6 & 9 \\ 4 & 8 & 12 \\ 5 & 10 & 15 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ 5 \end{bmatrix} \times \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}$$

Encoder. X-Block. Separable Convolution

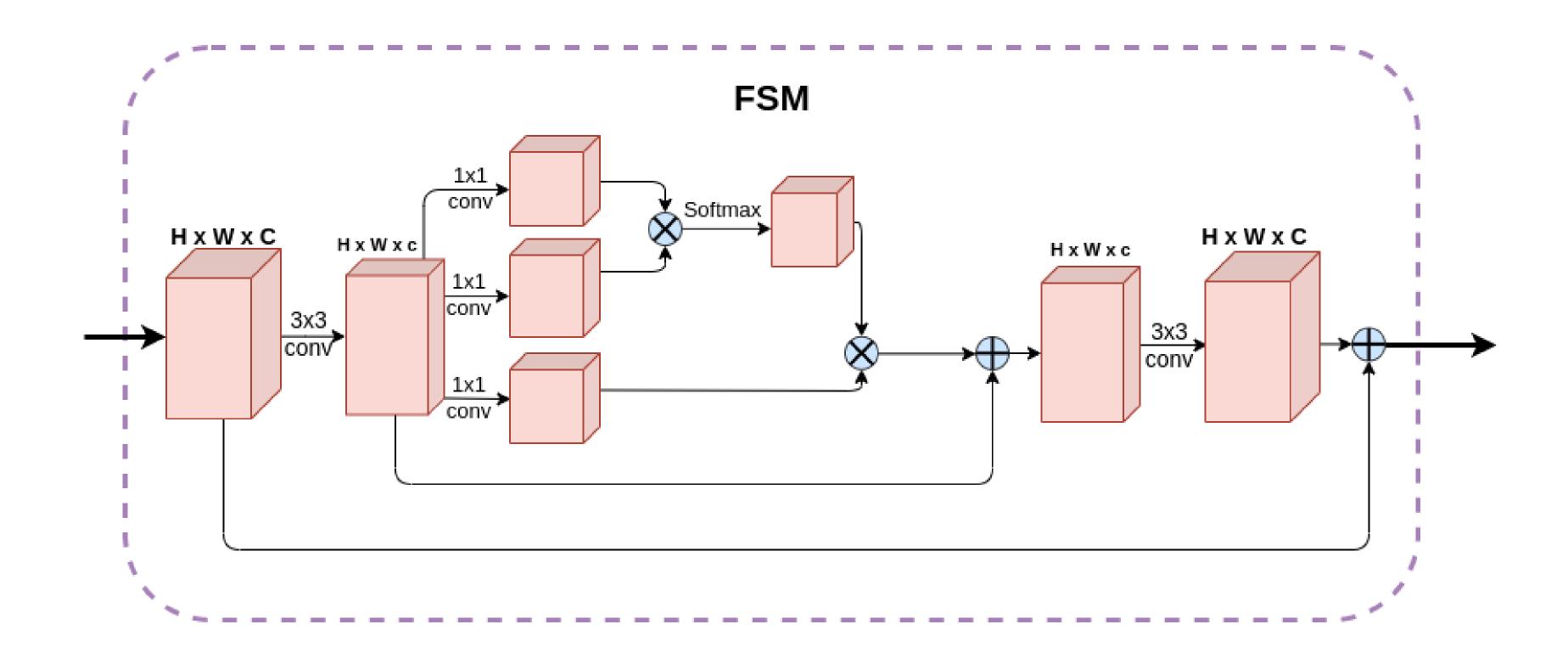
Separable Convolution can enhance efficiency and decrease overfitting without significantly reducing effectiveness

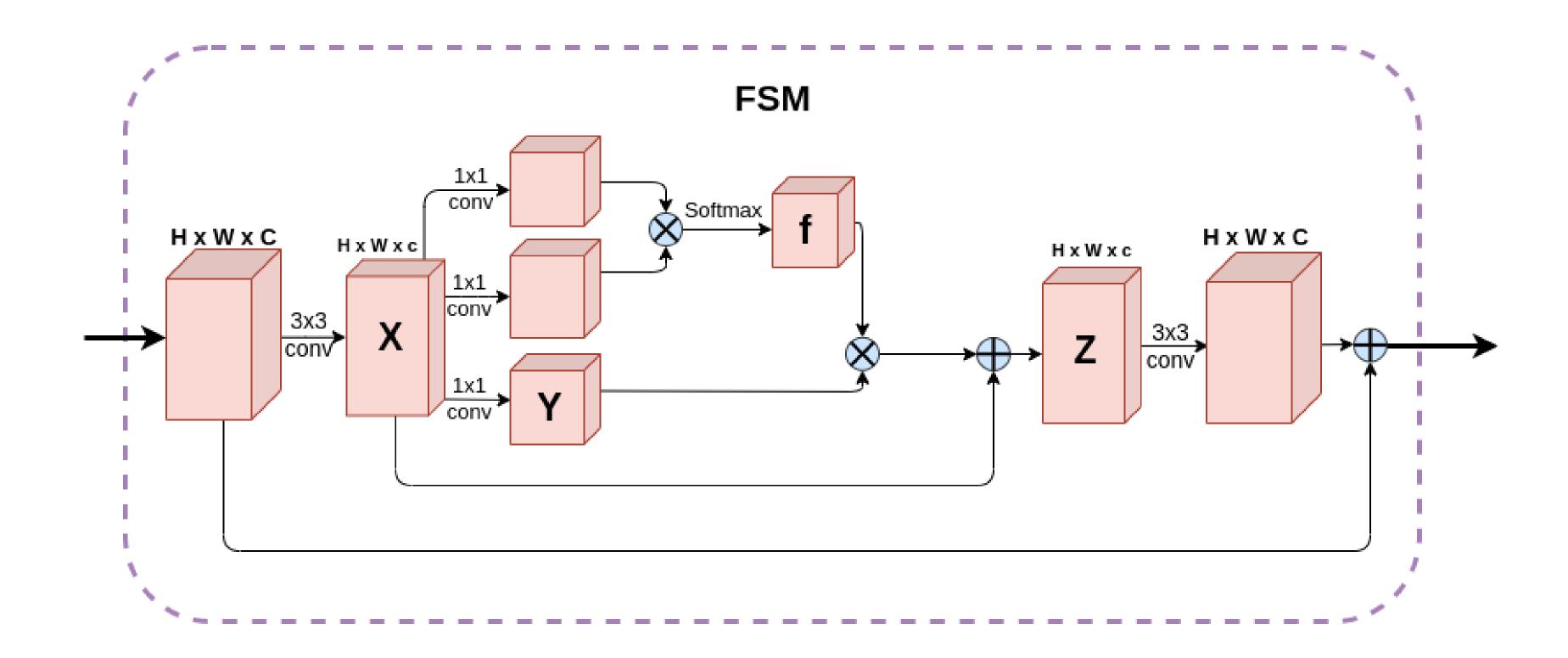
Simple Convolution

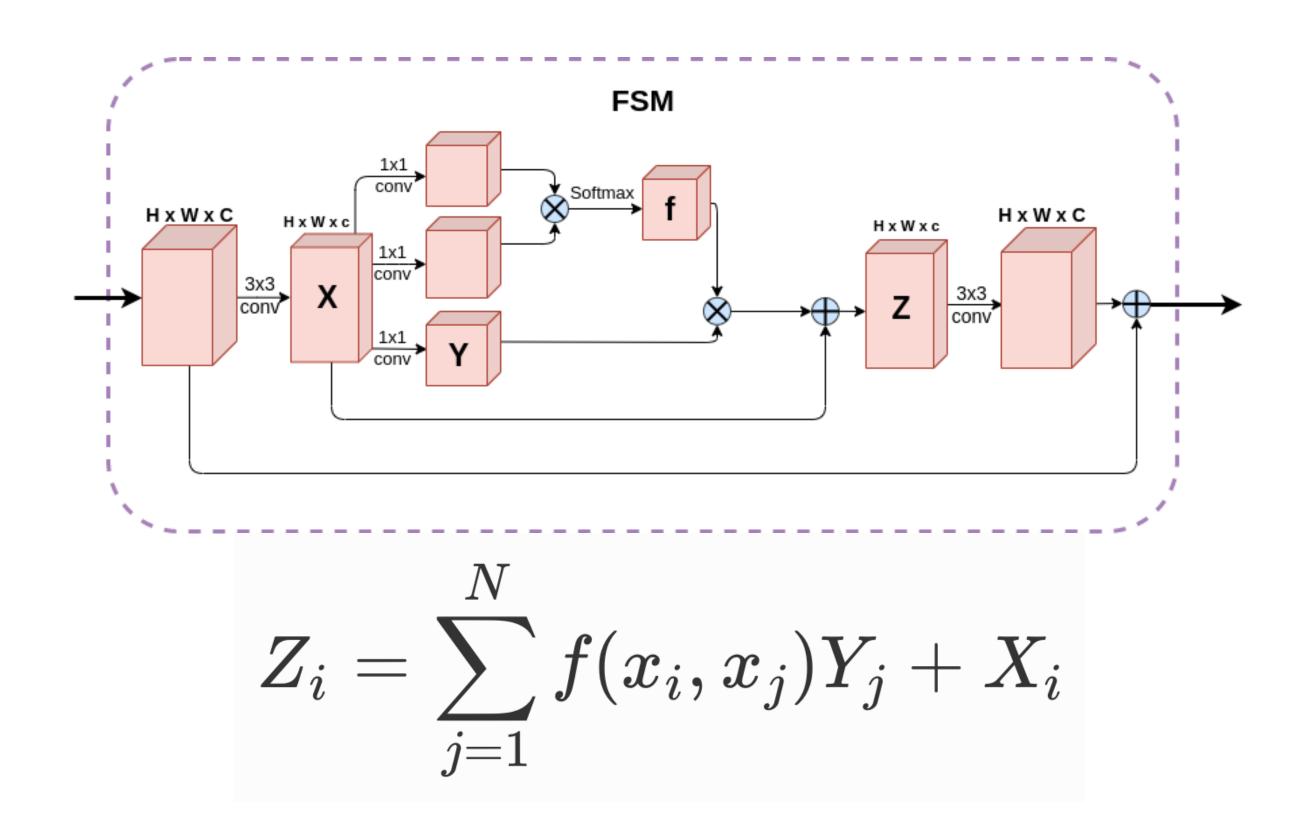


Spatial Separable Convolution

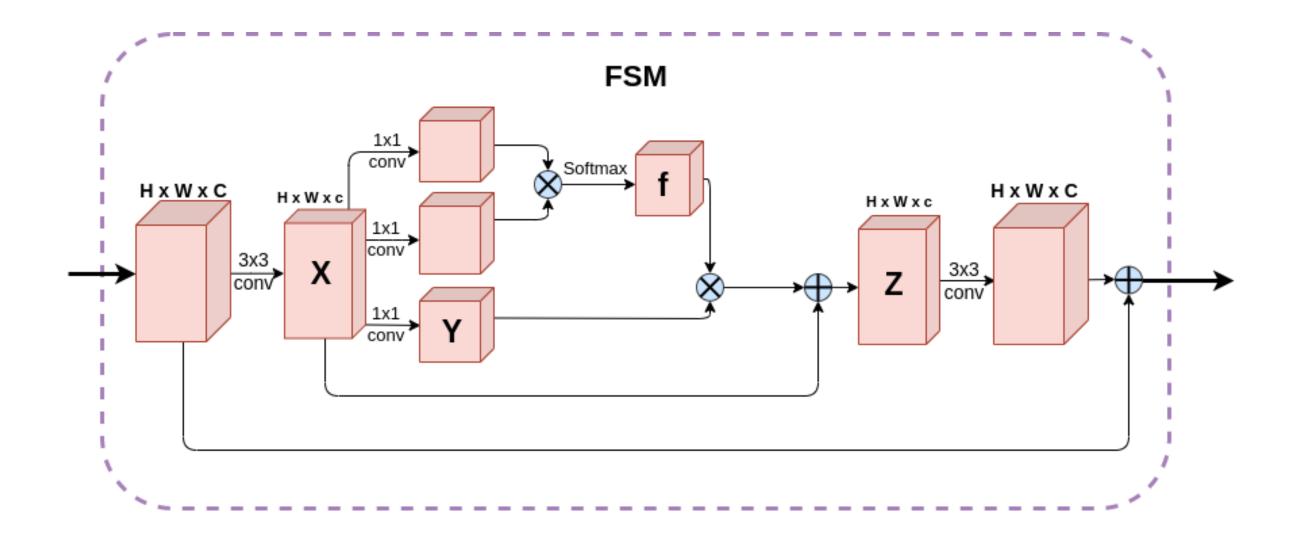




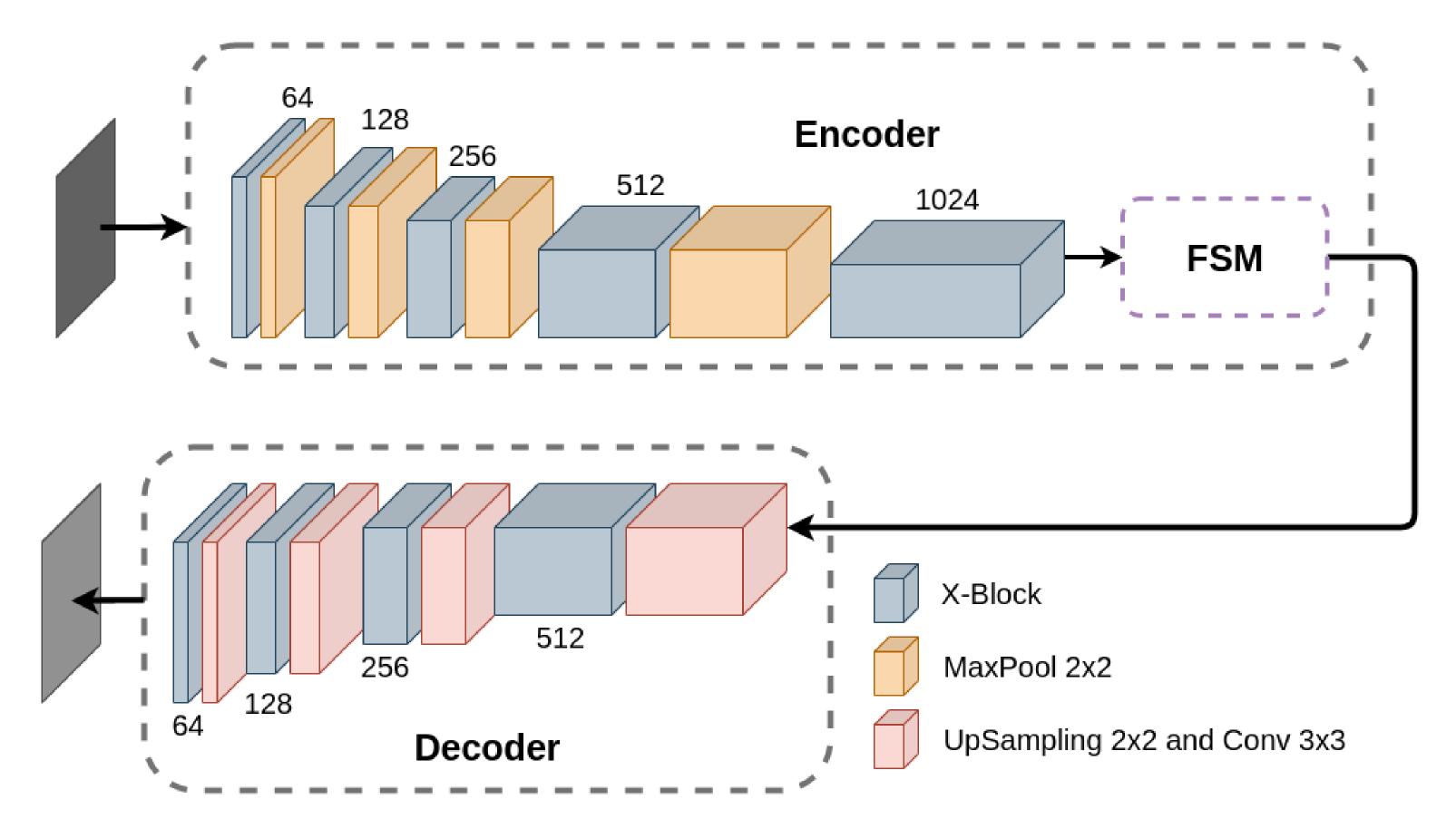




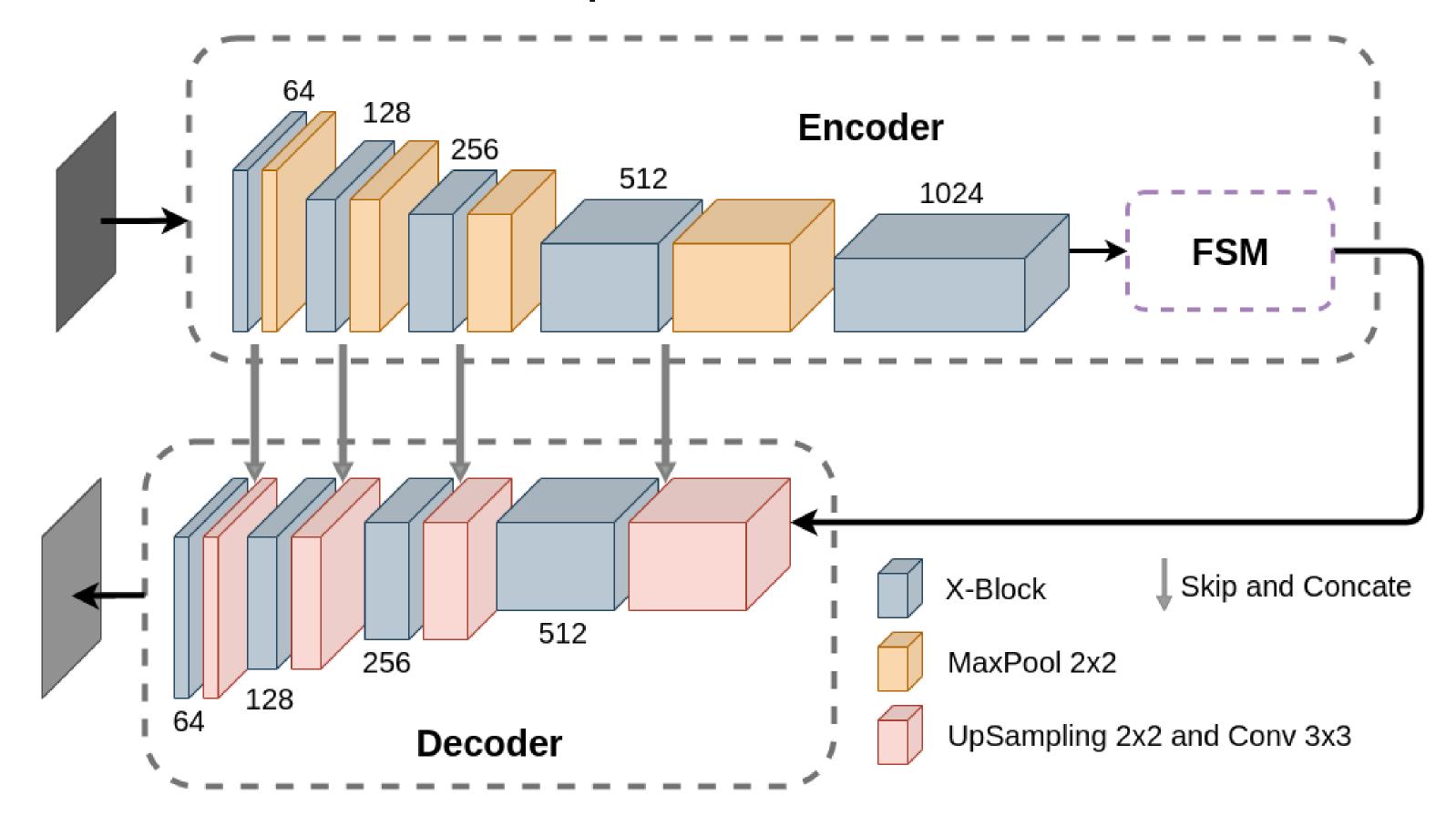
- 1. FSM is a non-local operation for capturing long-range dependencies by learning a relationship between any two positions of a feature map
- 2. FSM has a reduced number of training parameters (thanks to separable convolution)
- 3. Can be plugged into any model



Architecture. Decoder



Final Architecture. Skip-and-concate



A bit more information

Optimizer: Adam, lr=0.001

Loss Function: Dice + Cross Entropy losses

Batch Size: 8



Benchmark - ATLAS dataset

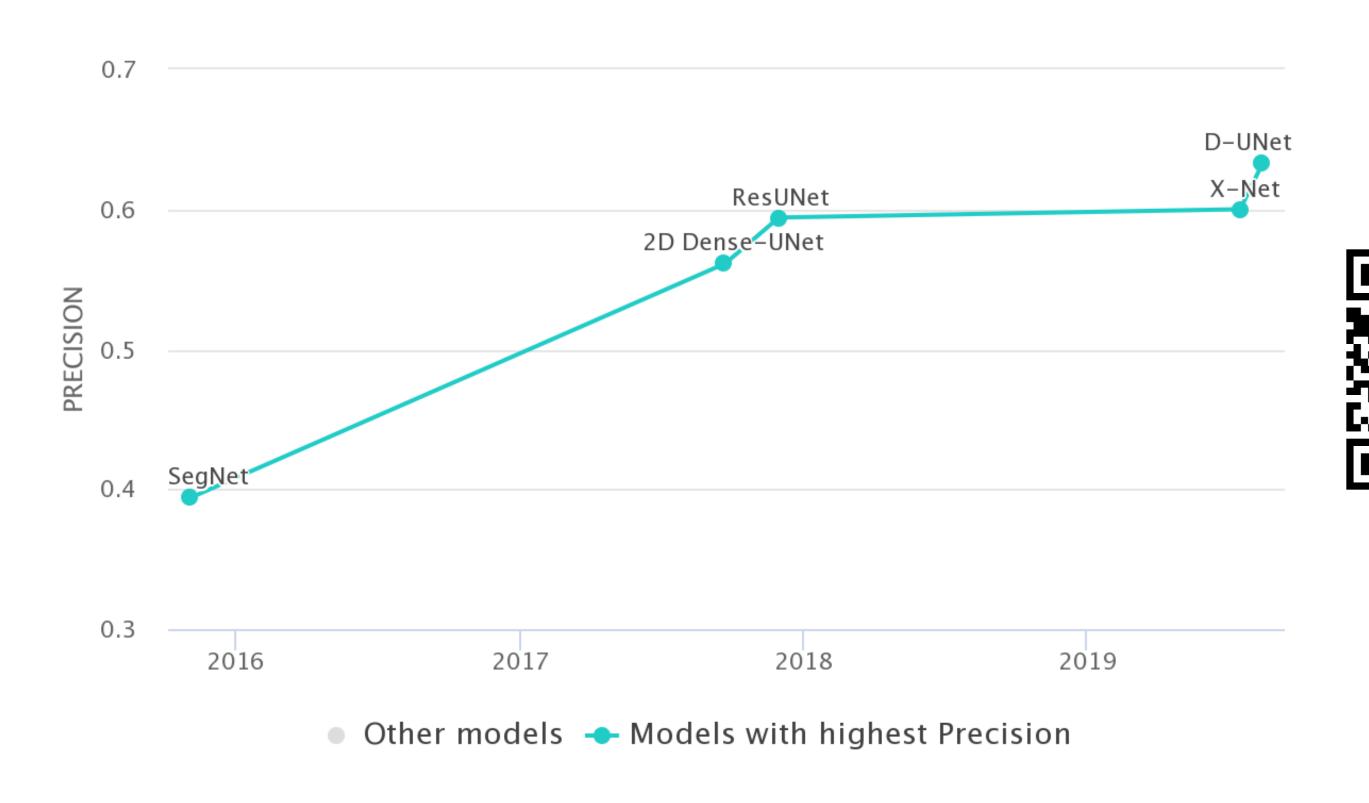
Recall: How many relevant items are retrieved?





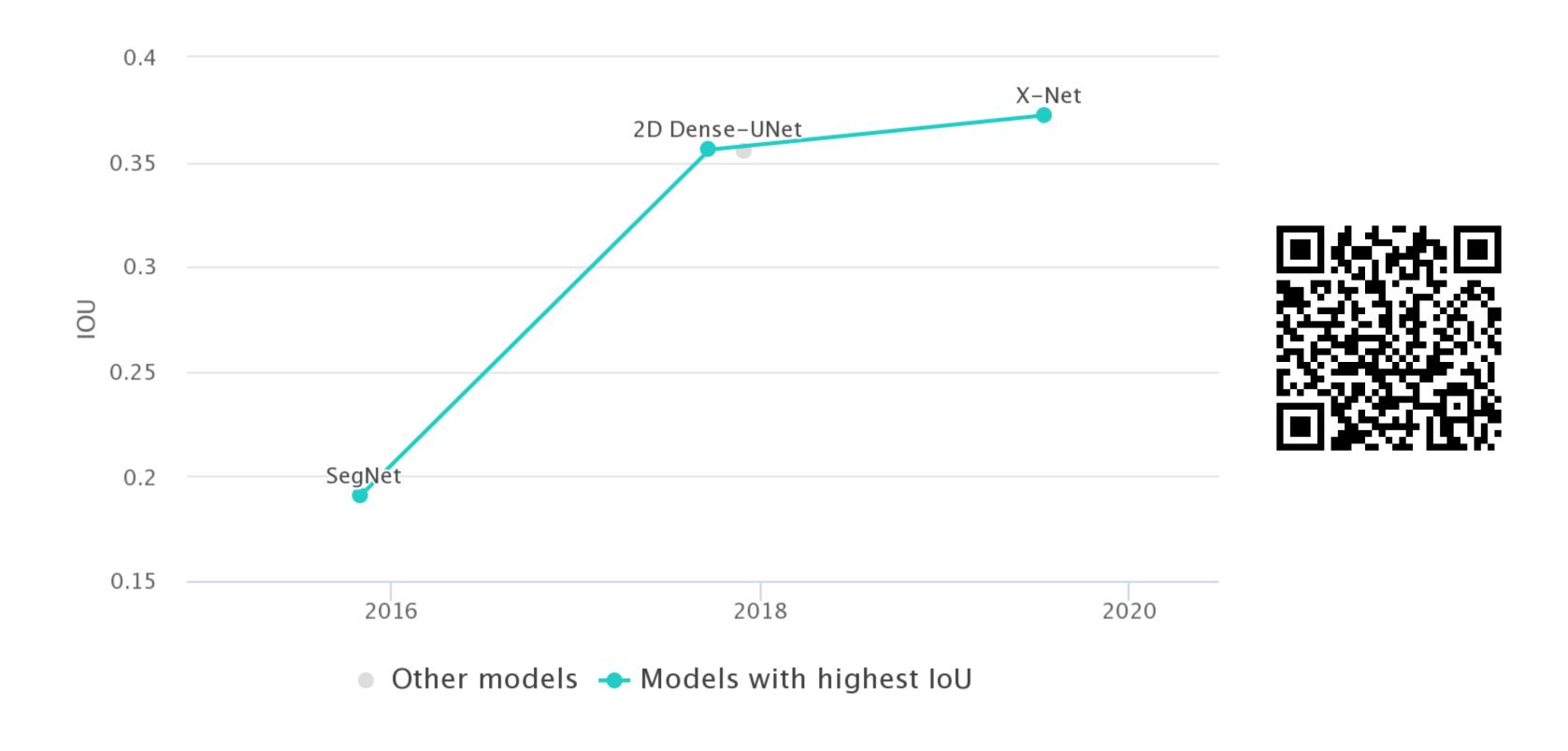
Benchmark - ATLAS dataset

Precision: How many retrieved items are relevant?



Benchmark - ATLAS dataset

IoU: What is the % of overlap between ground-truth segment and predicted segment?



The authors' results

Method	IoU	# parameters
2D Dense-UNet	0.35	50.0M
U-Net	0.34	34.5M
SegNet	0.35	29.5M
X-Net	0.37	15.1M