PMLDL. X-Net

Anna Boronina

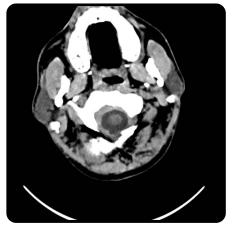
What is a brain stroke (ru: инсульт)?

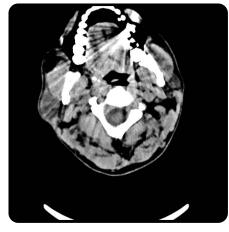
Brain stroke occurs when the blood supply to part(s) of one's brain is interrupted or reduced.

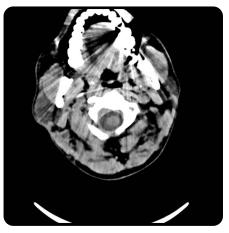


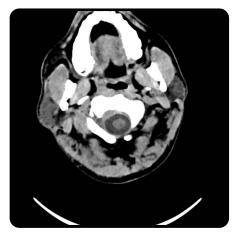
Problems

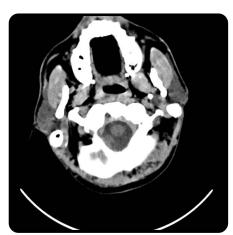
- sometimes even doctors cannot spot a brain stroke
- sometimes there is a sequence of MRI scans which should be processed together
- healthy VS stroke:

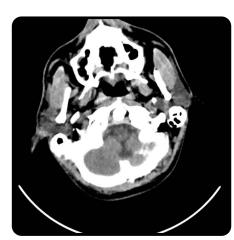




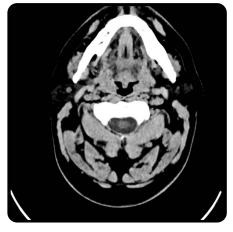


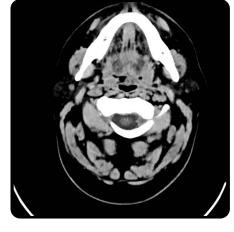




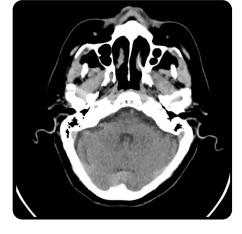














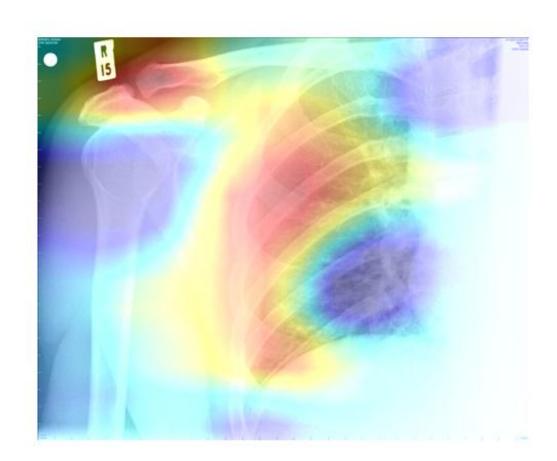


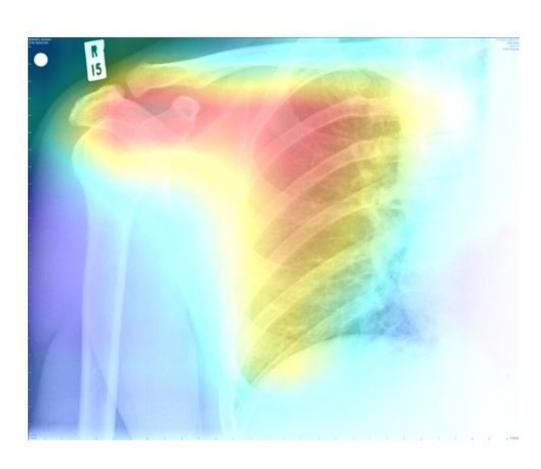


DL applications

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







Benchmark - ATLAS dataset

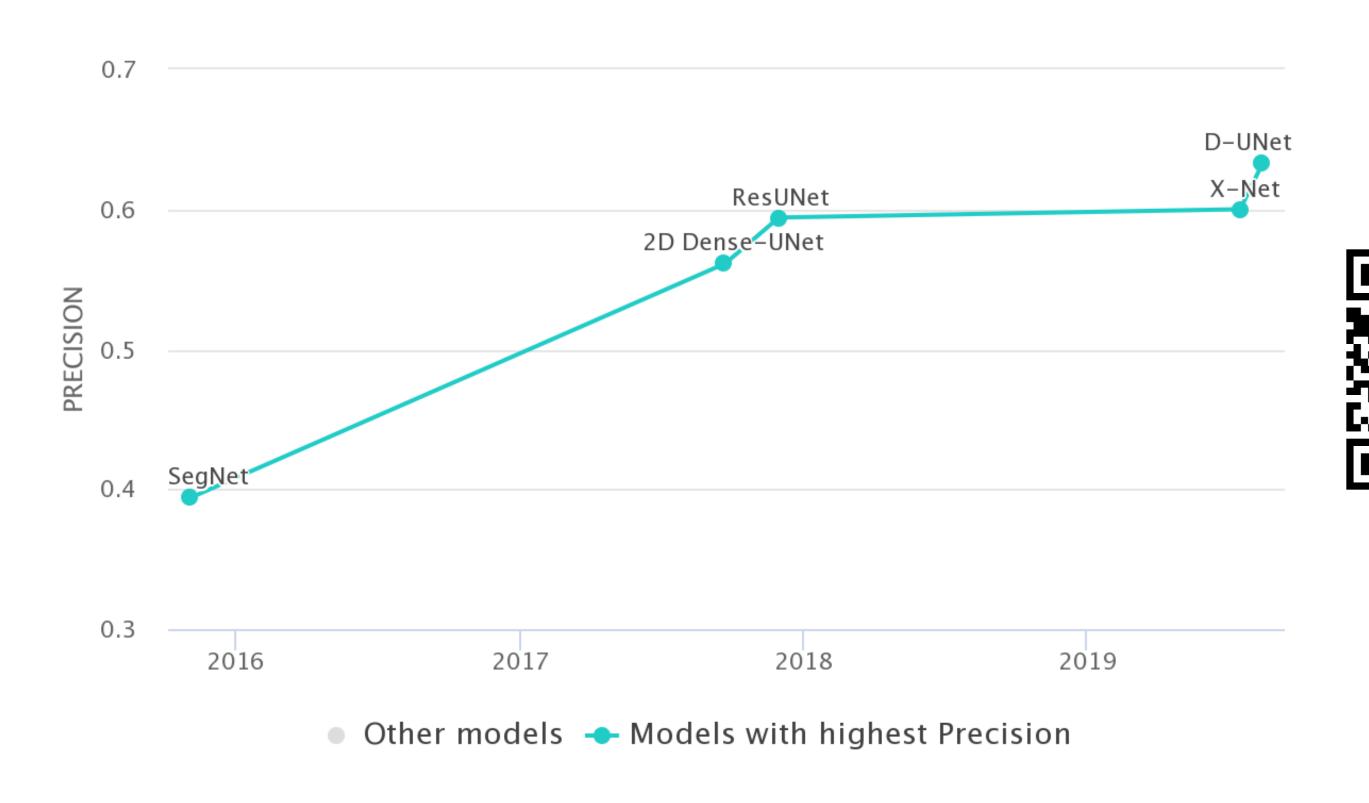
Recall: How many relevant items are retrieved?





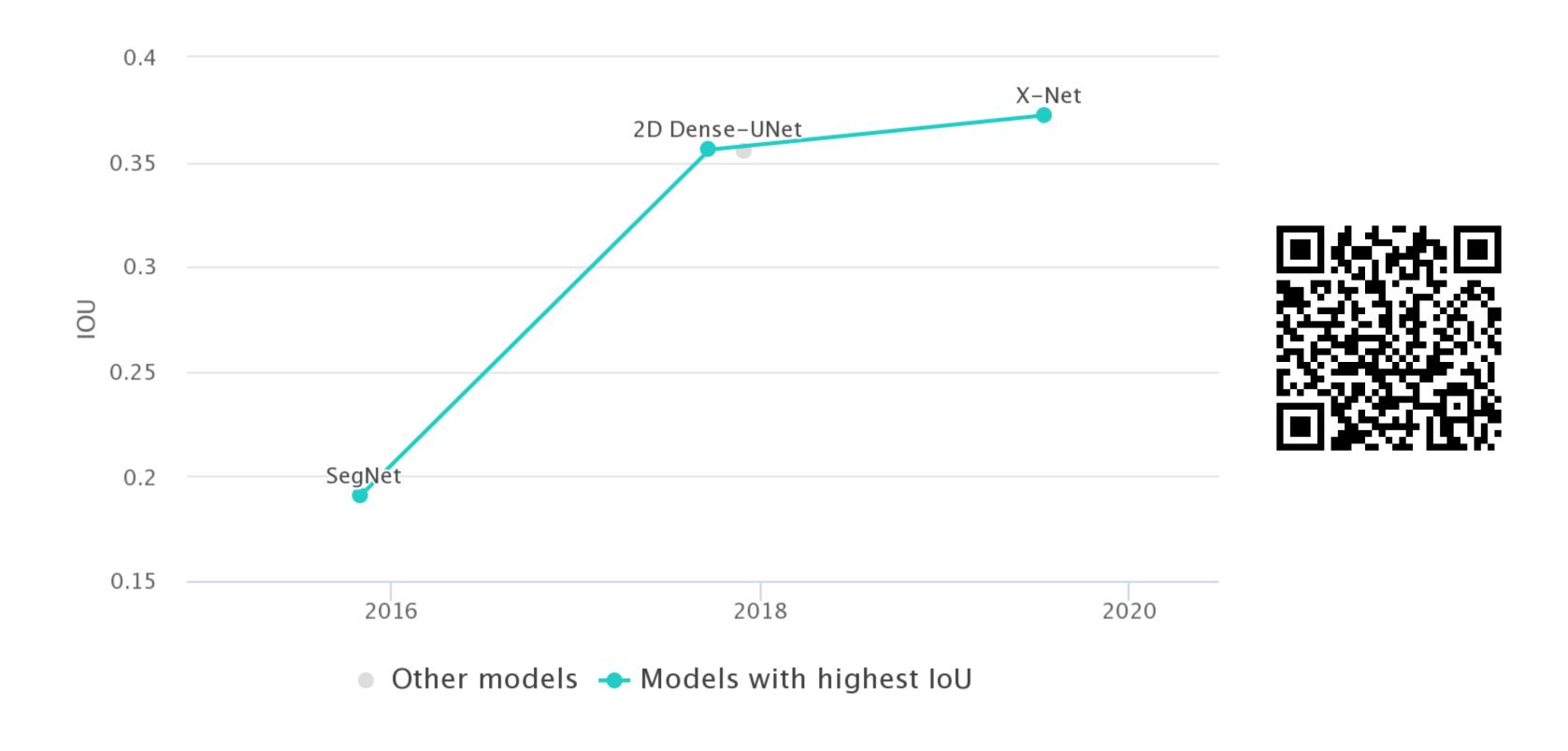
Benchmark - ATLAS dataset

Precision: How many retrieved items are relevant?



Benchmark - ATLAS dataset

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



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

Kehan Qi, Hao Yang, Cheng Li, Zaiyi Liu, Meiyun Wang, Qiegen Liu, Shanshan Wang

Conference paper

First Online: 10 October 2019



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



Previous attempts

CHECK THIS INFO

- SegNet symmetrical convolutional autoencoder
- UNet residual convolutional network
- 2D Dense Unet residual dense network

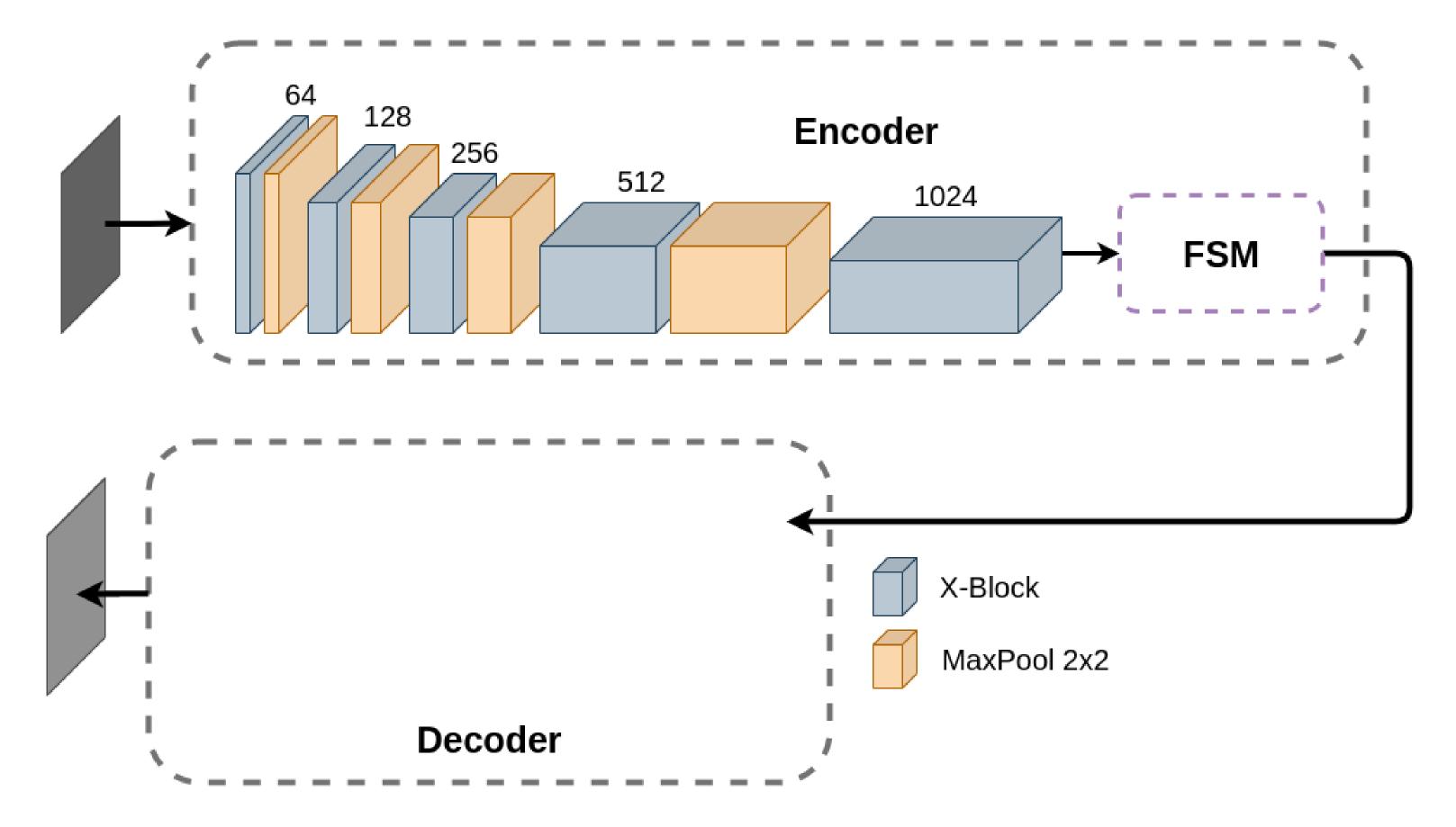
Their problems according to Qi et. al

- heavy network parameters
- cannot capture long-range dependencies

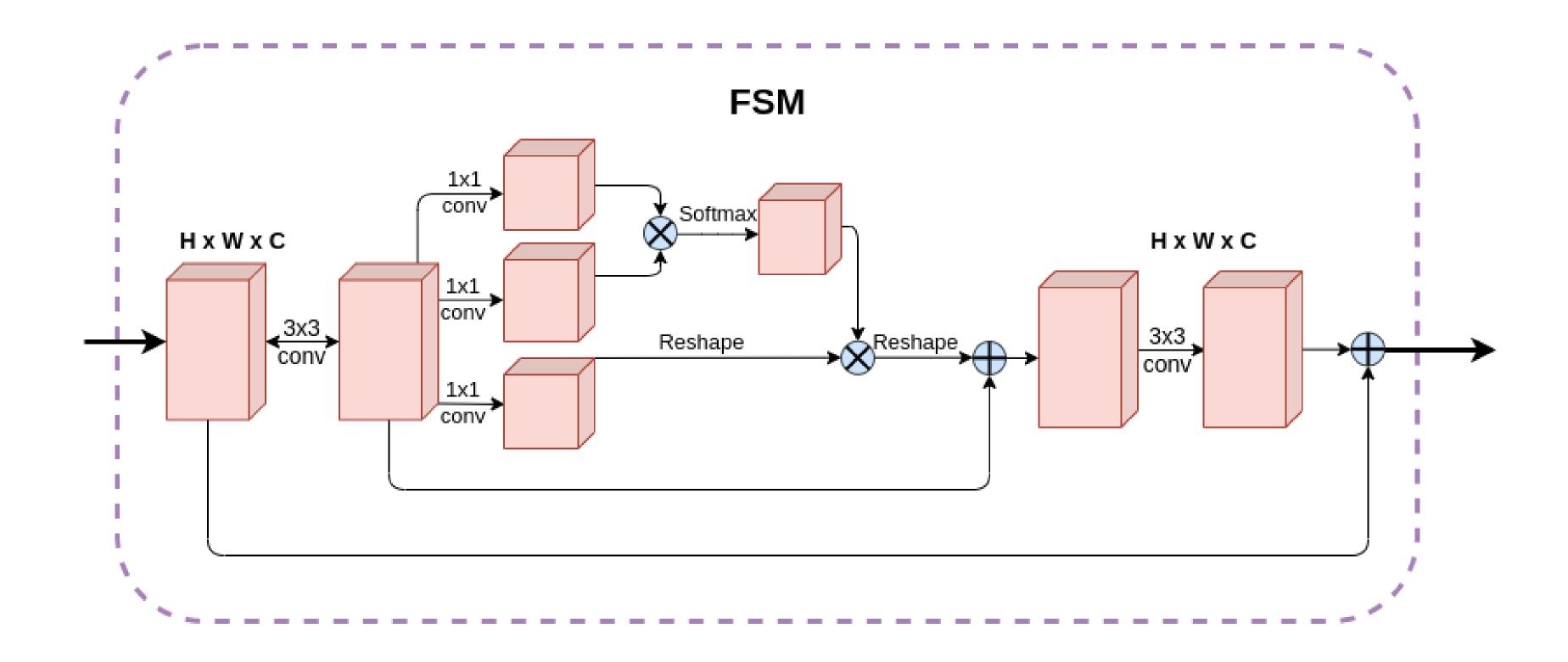
Solution proposed by Qi et. al

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

Architecture. Encoder

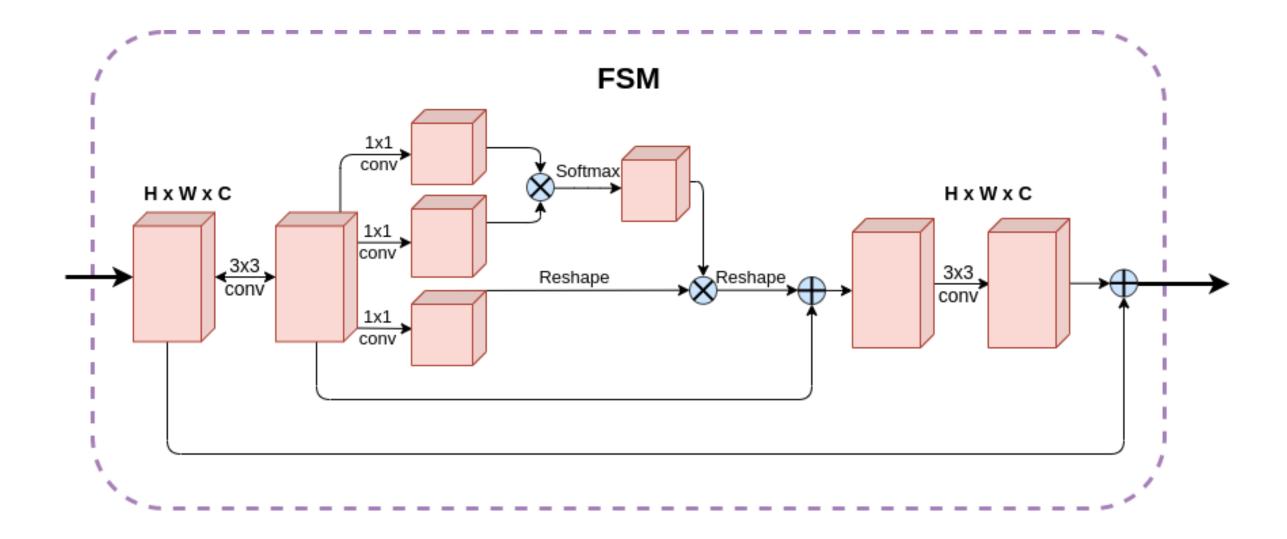


Architecture. Encoder. FSM



Architecture. Encoder. FSM

- 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. Encoder. FSM. Separable Convolution

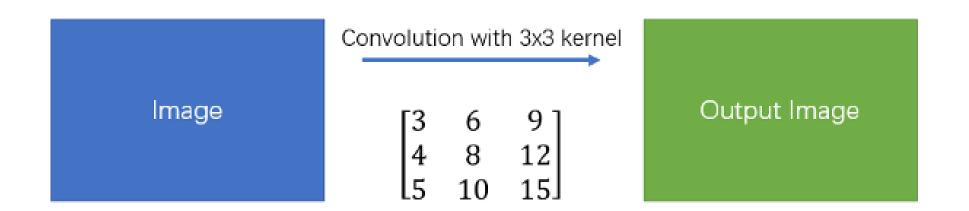
Separating a 3x3 kernel spatially

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

Architecture. Encoder. FSM. Separable Convolution

Separable Convolution can enhance efficiency without significantly reducing effectiveness

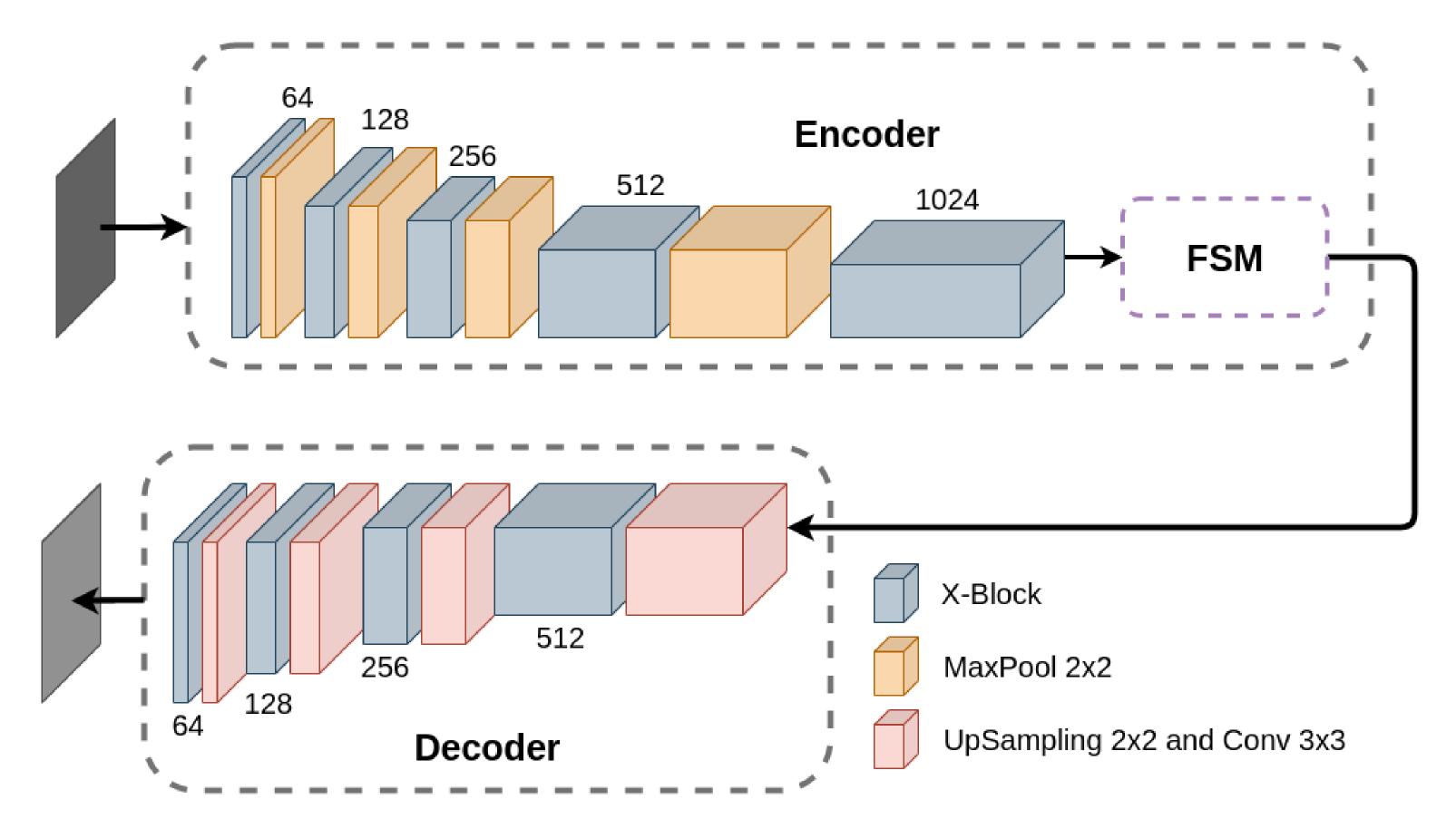
Simple Convolution



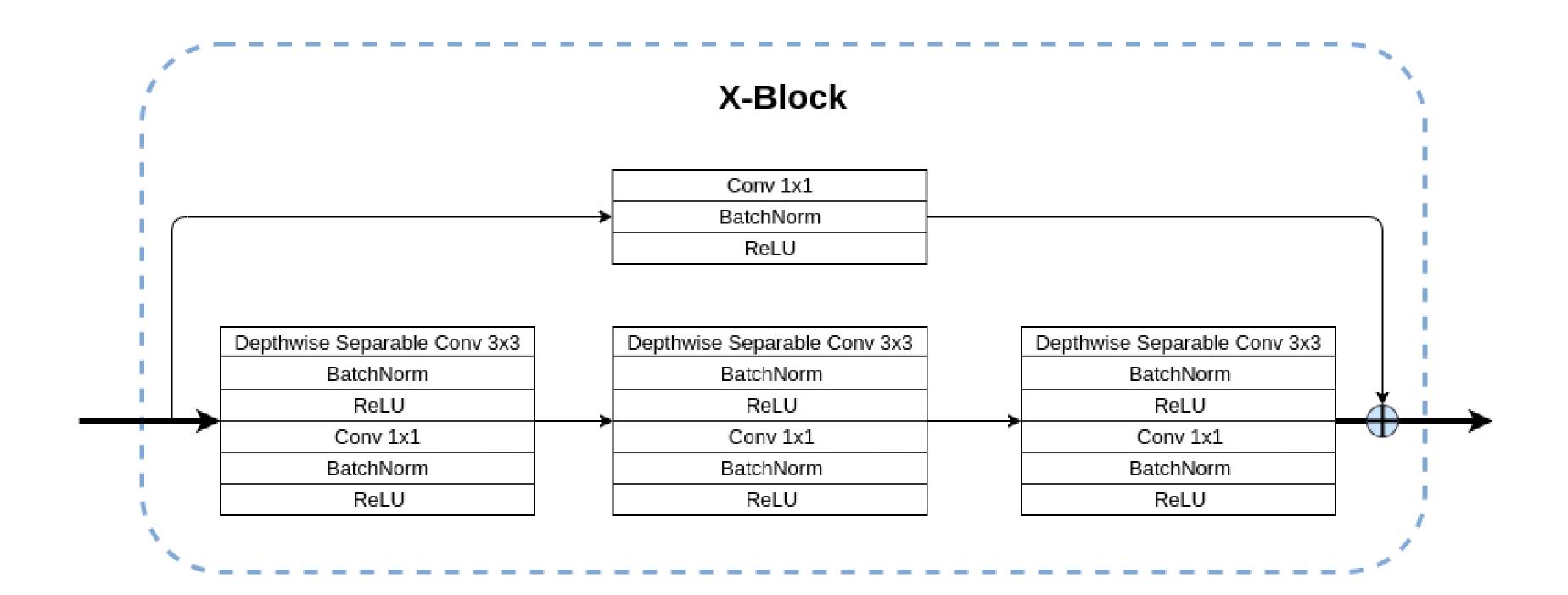
Spatial Separable Convolution



Architecture. Decoder



Architecture. Decoder. XBlock



Architecture. Skip-and-concate

