



SegNet

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SegNet: A Deep Convolutional Encoder-Decoder Architecture for Image Segmentation

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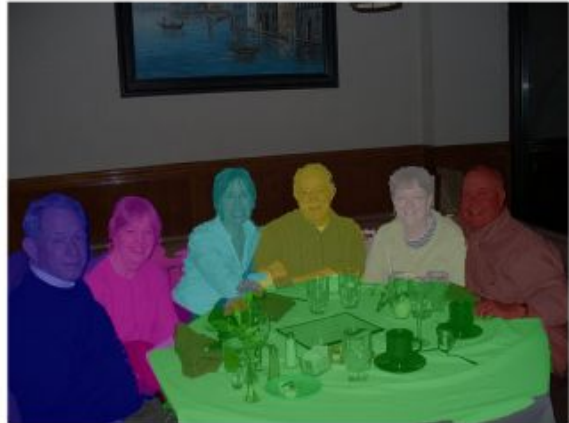
10 Oct 2016

What Semantic Segmentation is?

The process of assigning a label to every pixel in the image.

Semantic segmentation treats multiple objects of the same class as a single entity.

Example of various Scene Understanding tasks

 <p>Object Detection</p>	 <p>Semantic Segmentation</p>	 <p>Instance Segmentation</p>
<p>Tags: Person, Dining Table</p> <p>Image Classification</p>	<p>A group of people sitting at a table</p> <p>Image Captioning</p>	<p>Q: What were the people doing? A: Eating dinner</p> <p>Visual Question-Answering</p>

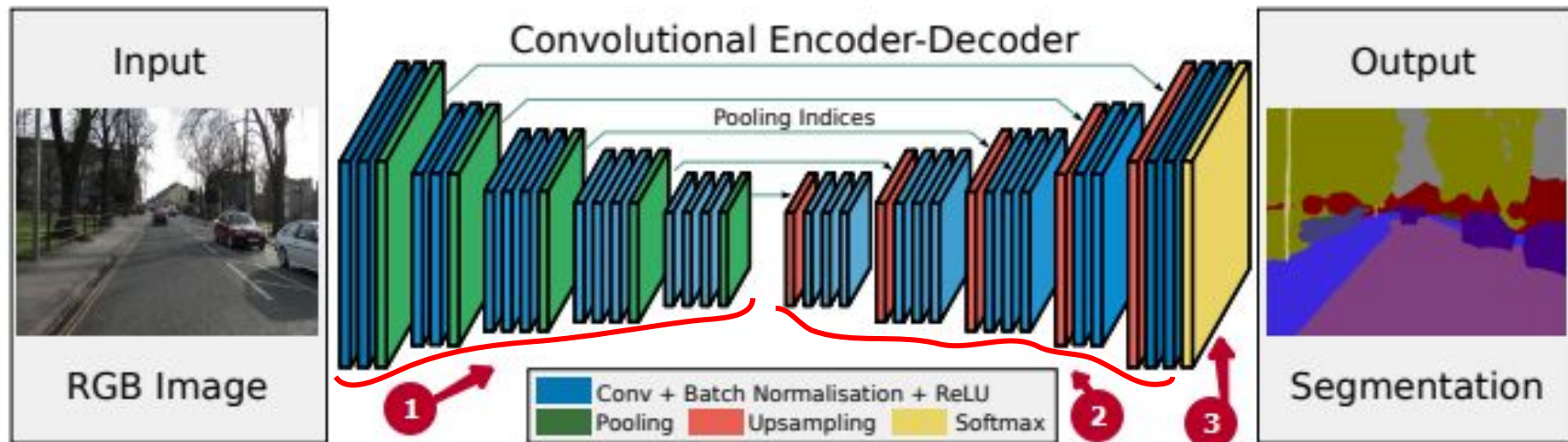
What SegNet is?

Novel and practical deep fully convolutional neural network architecture for semantic pixel-wise segmentation.

SegNet = (Encoder + Decoder) + Pixel-Wise Classification layer

It is primarily motivated by road scene understanding applications which require the ability to model appearance (road, building), shape (cars, pedestrians) and understand the spatial-relationship (context) between different classes such as road and side-walk.

SegNet architecture



1 - encoder network

2 - decoder network

3 - pixel-wise classification layer

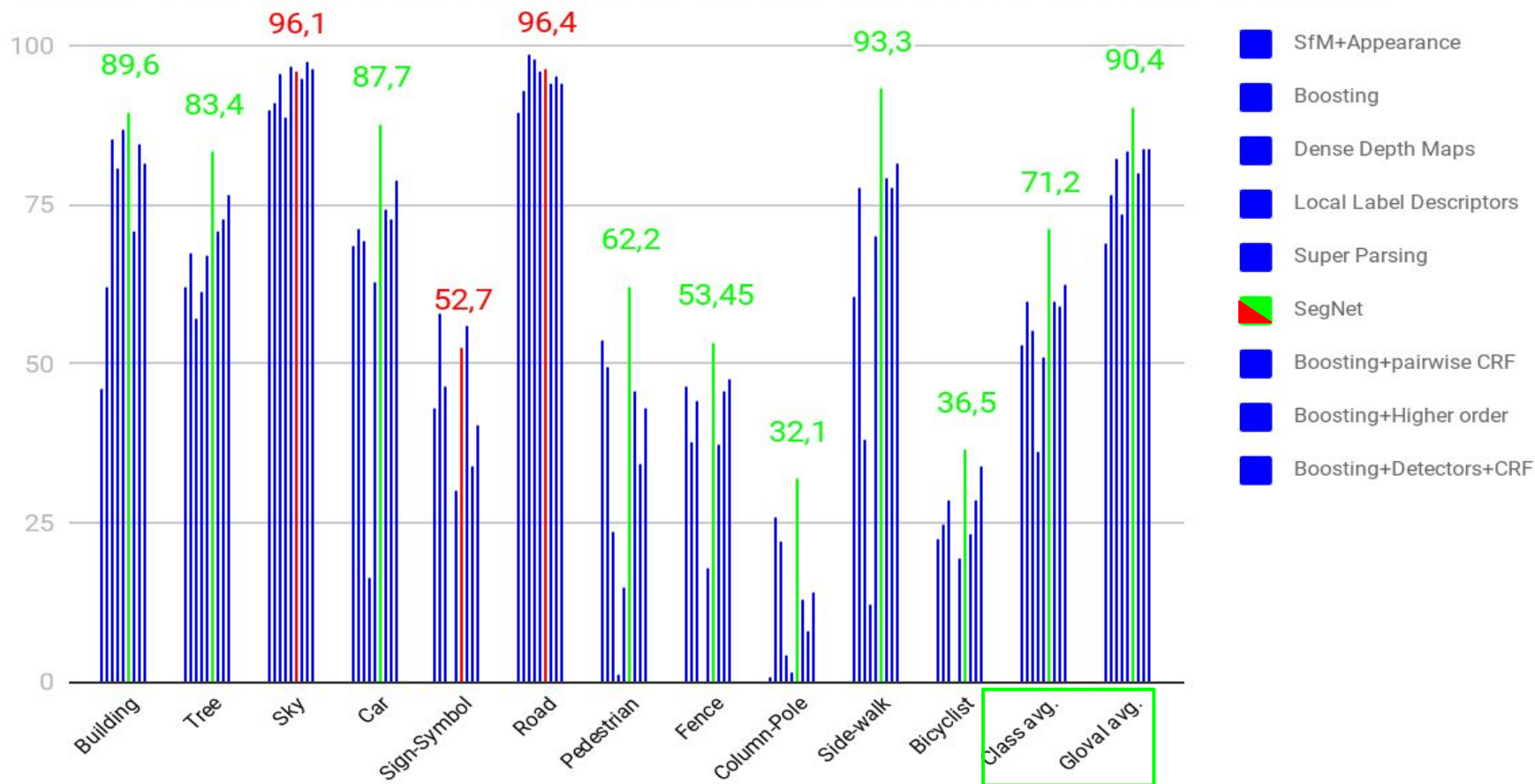
Main motivations behind SegNet

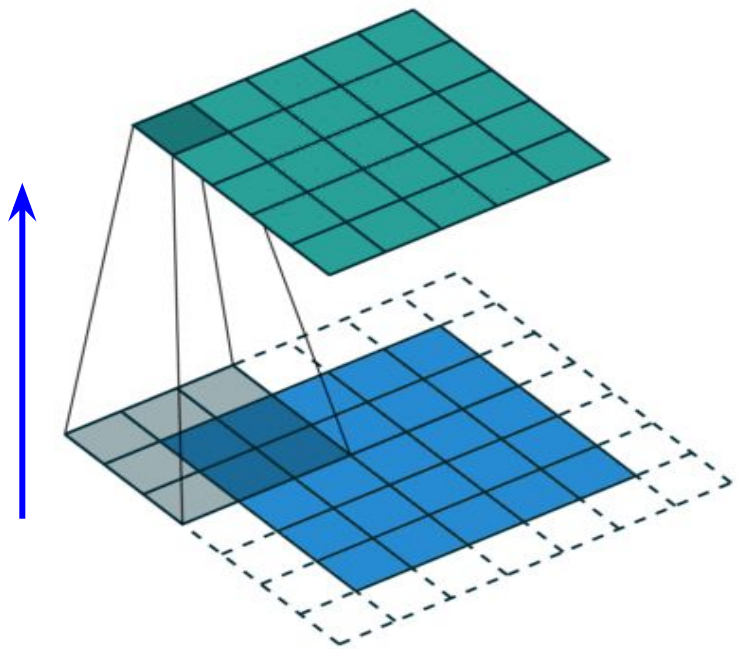
- Retain boundary information in the extracted image representation
- Efficient in terms of both memory and computation time
- Able to train end-to-end using efficient weight update technique

SegNet predictions on road scenes and indoor scenes

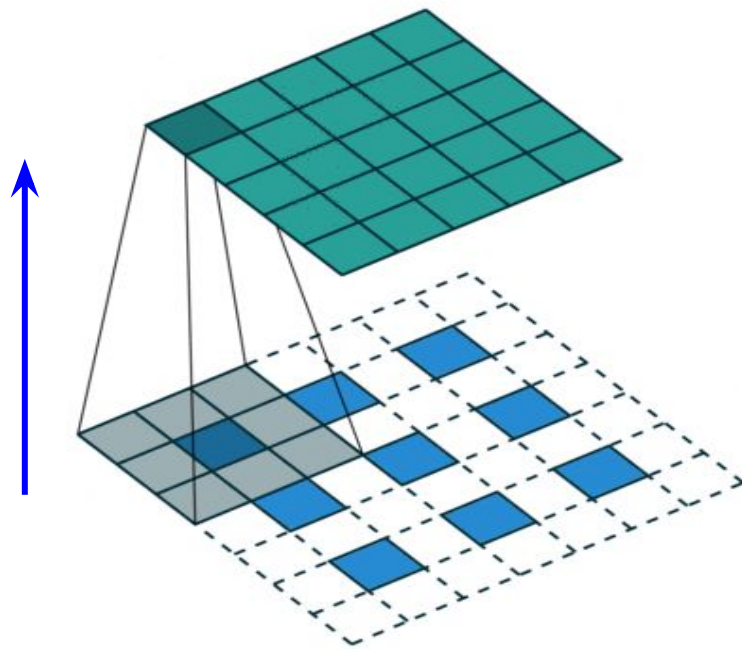


Quantitative comparisons of SegNet with traditional methods on the CamVid 11 road class segmentation problem



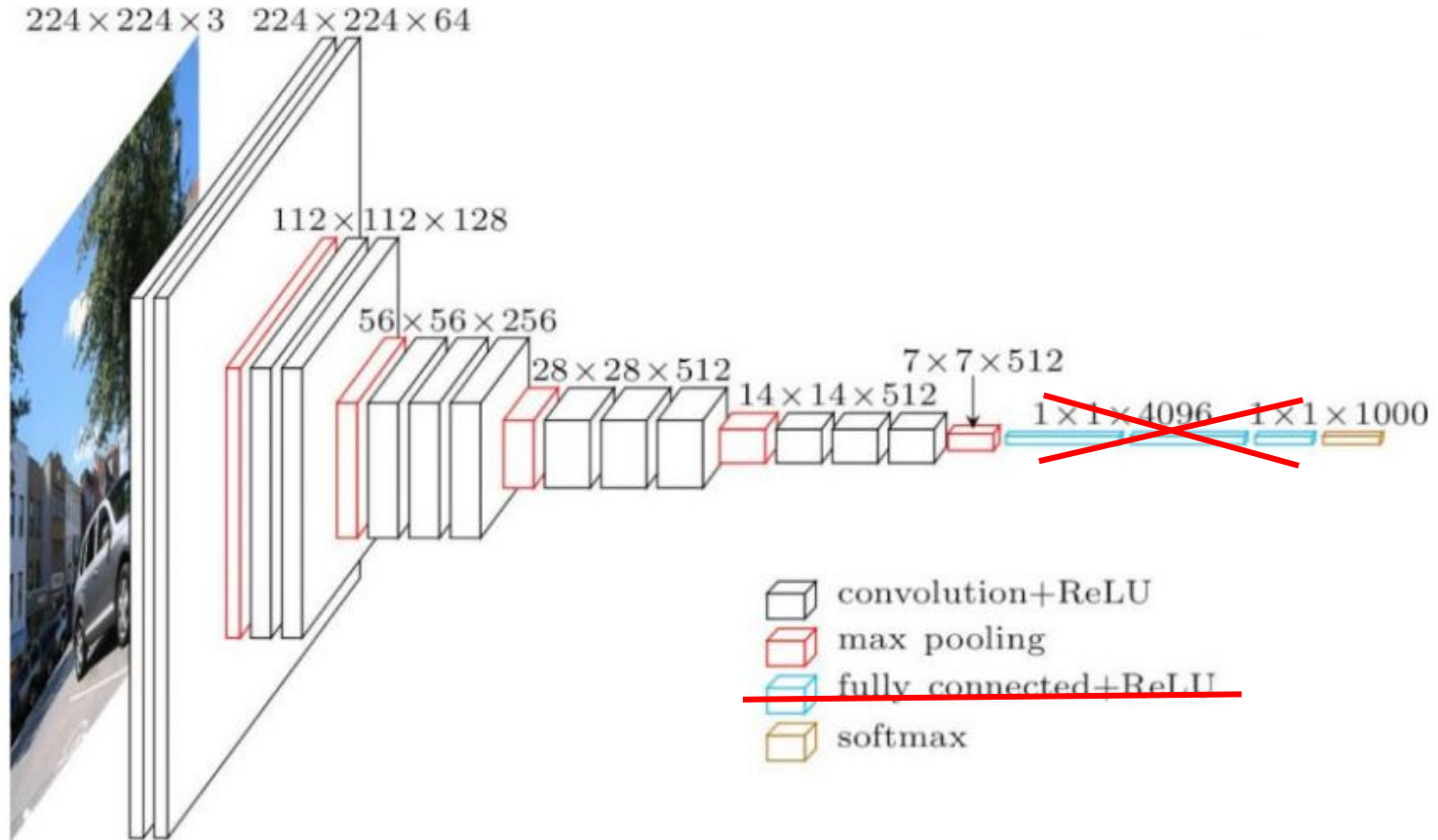


Encoding

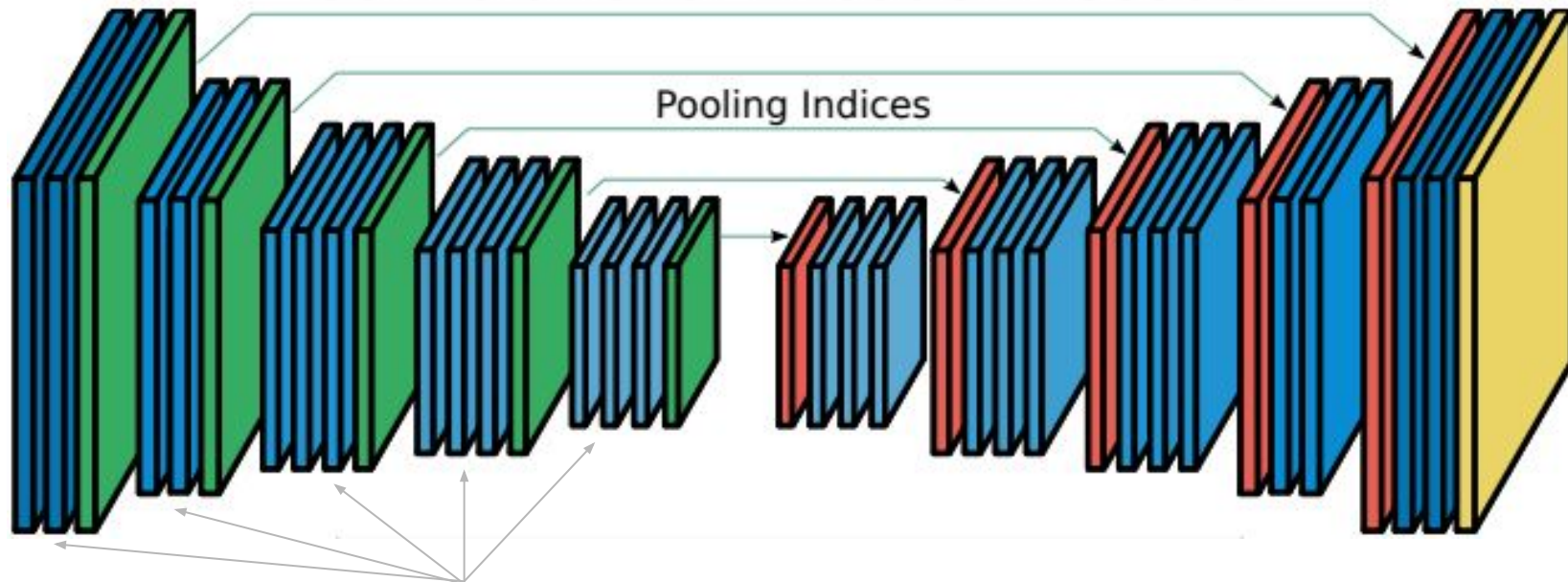


Decoding

VGG16 architecture



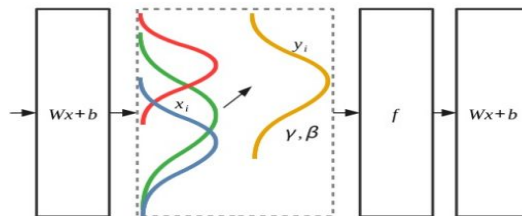
Convolutional Encoder-Decoder



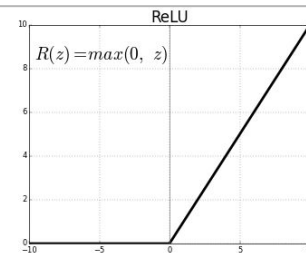
1x1	1x0	1x1	0	0
0x0	1x1	1x0	1	0
0x1	0x0	1x1	1	1
0	0	1	1	0
0	1	1	0	0

3x3conv

4		

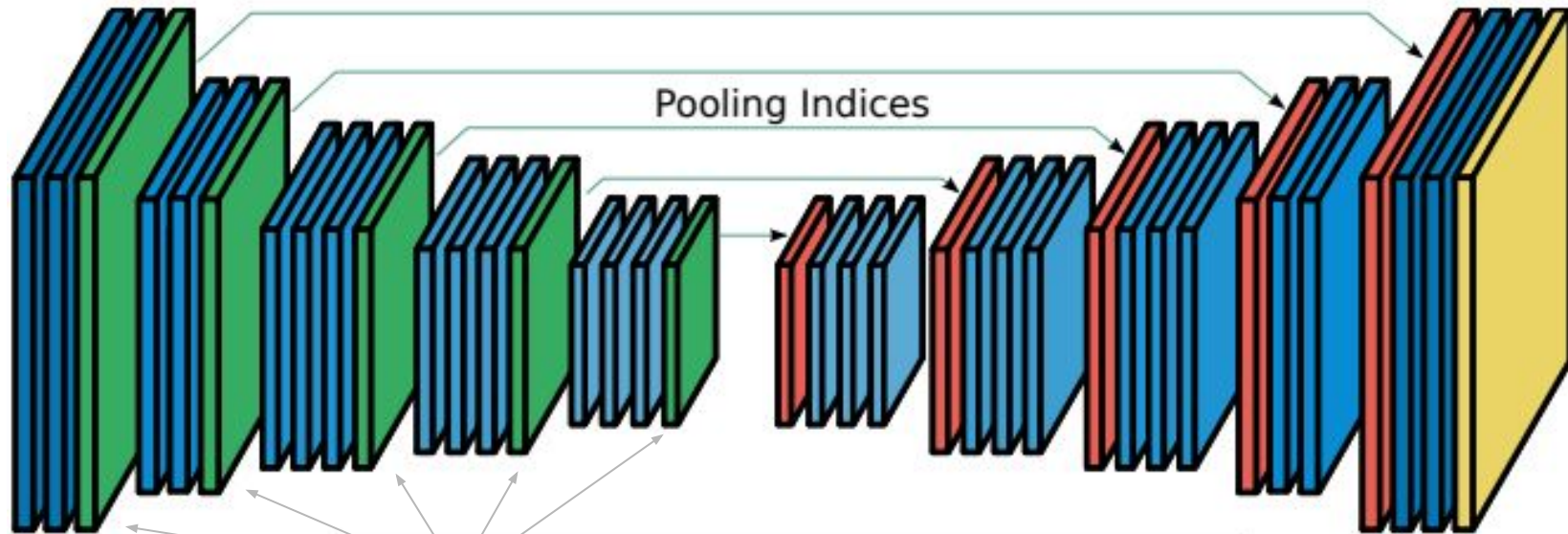


Batch Normalization



ReLU

Convolutional Encoder-Decoder

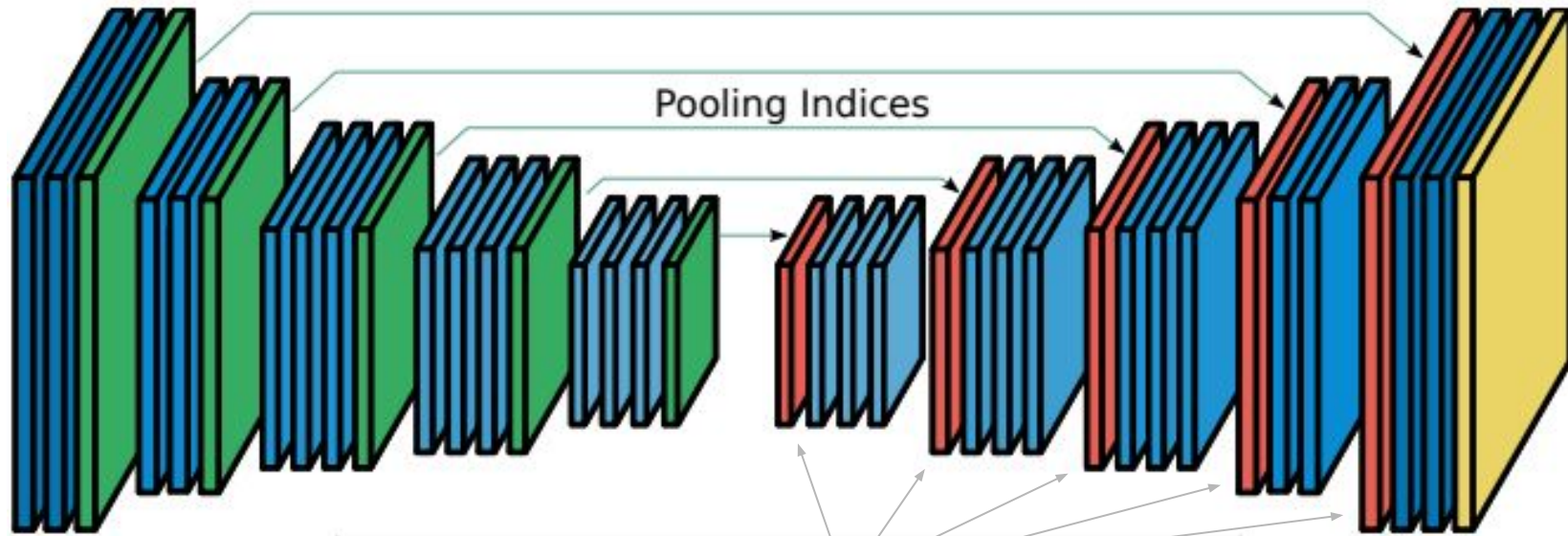


1	3	2	9
7	4	1	5
8	5	2	3
4	2	1	4

7	9
8	

2x2 Max-pooling

Convolutional Encoder-Decoder



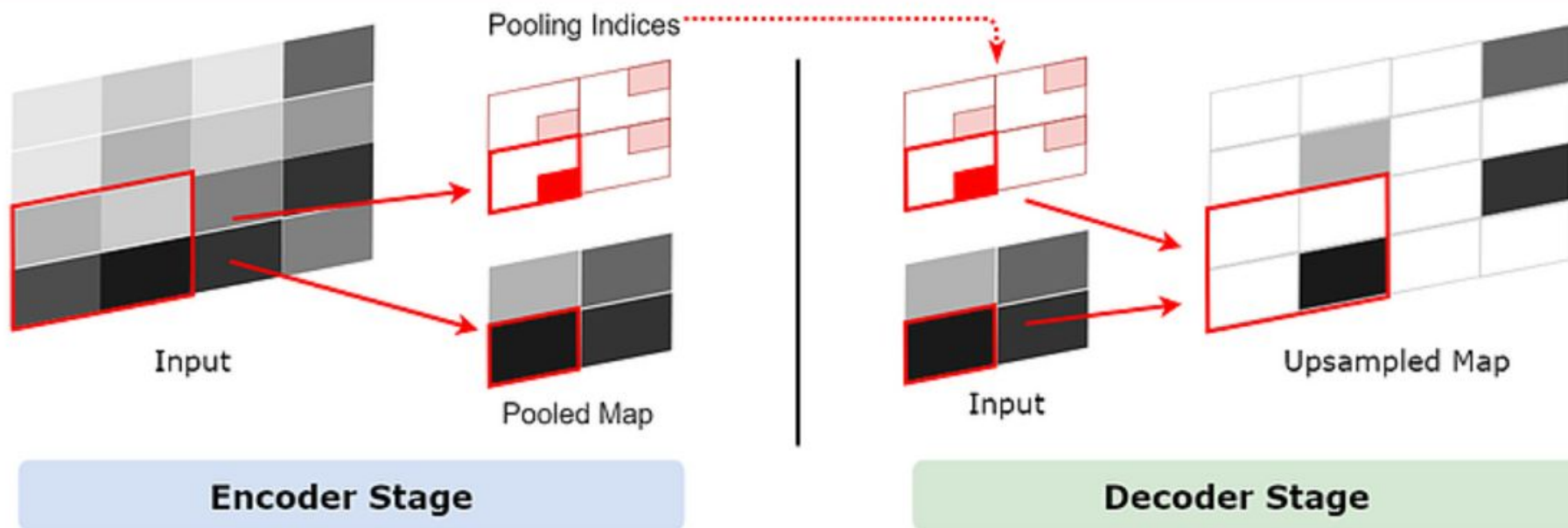
a	0	0	0
0	0	b	0
0	0	0	d
c	0	0	0

a	b
c	d

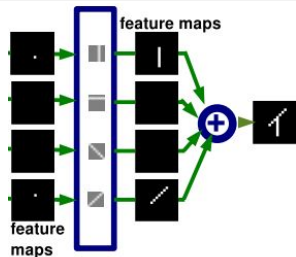
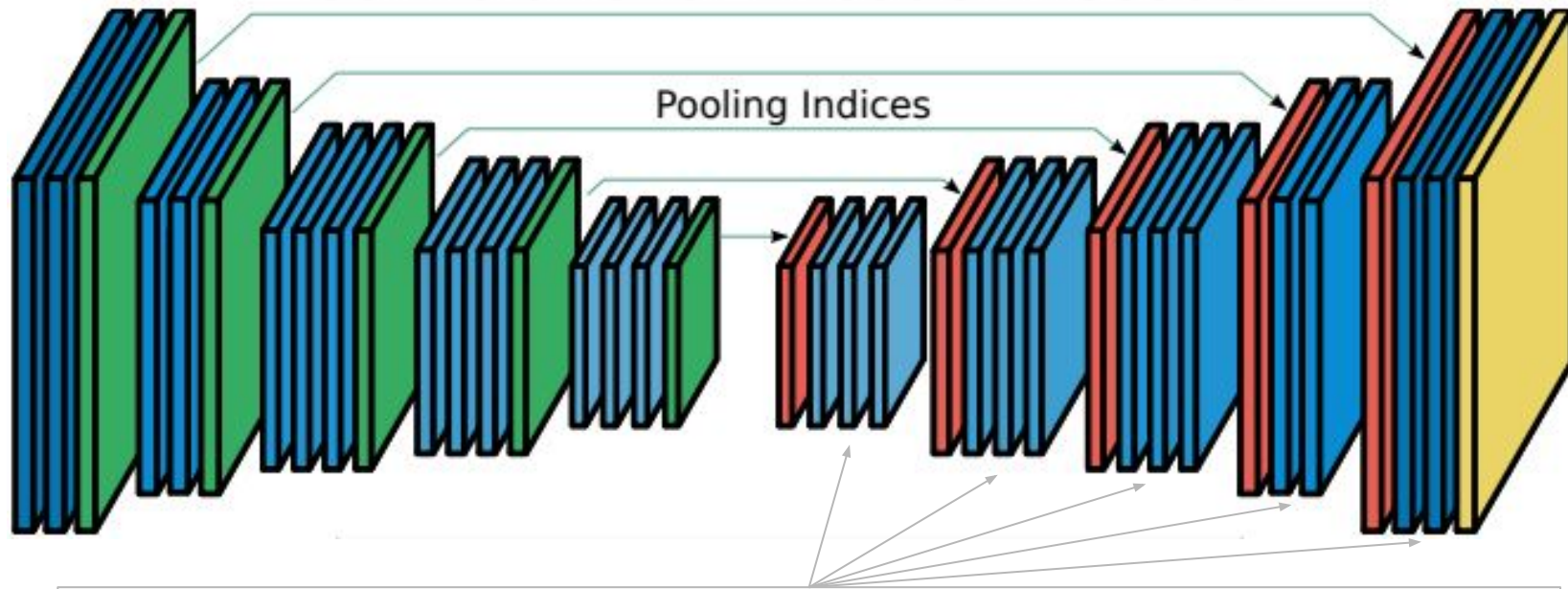
Max-pooling
Indices

Upsampling

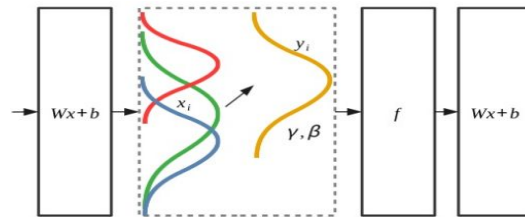
Using pooling indices for upsampling



Convolutional Encoder-Decoder

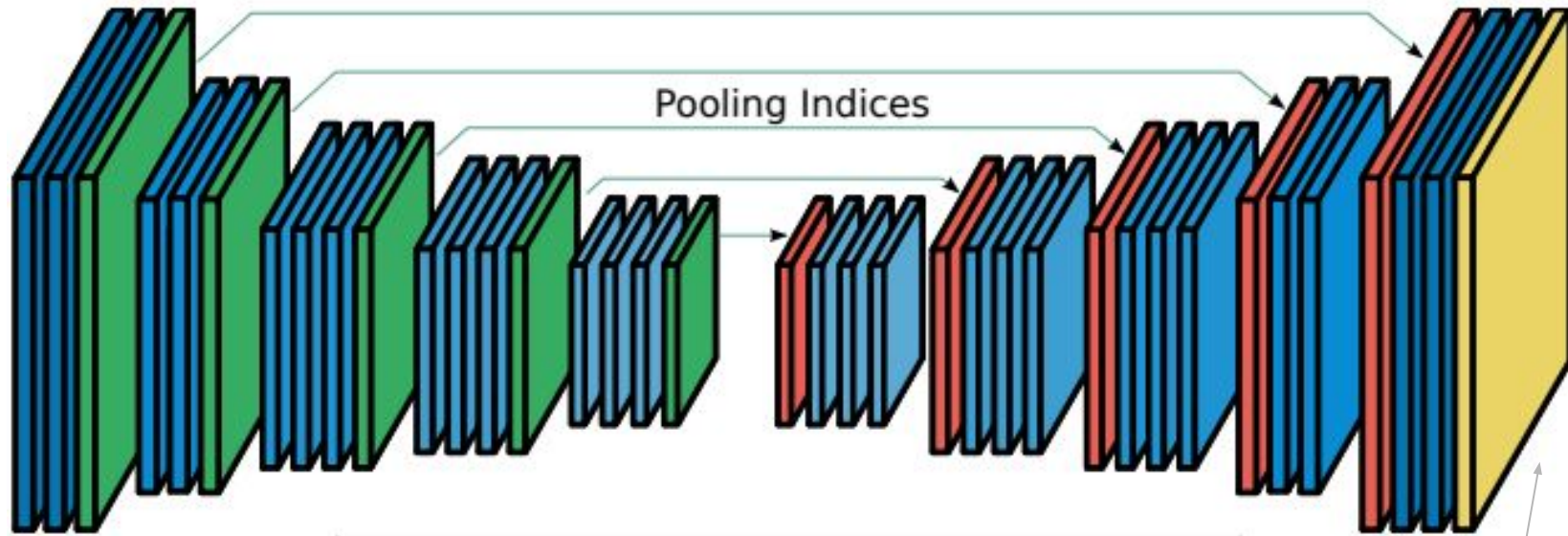


Decoder Filter Bank



Batch Normalization

Convolutional Encoder-Decoder



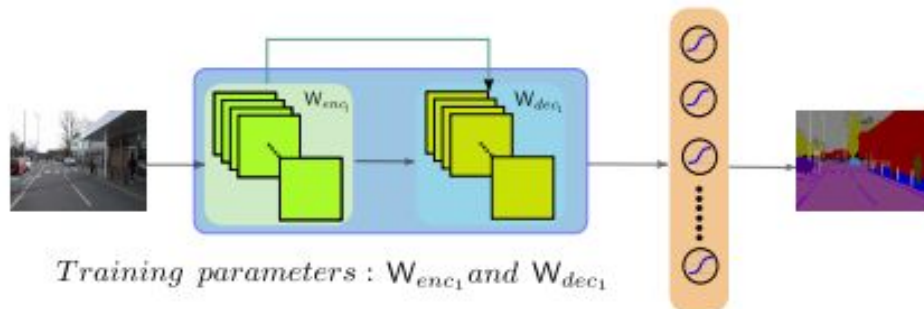
Pooling Indices

Softmax classifier

The diagram shows a vertical stack of five circles, each containing a sine wave. The circles are connected by a vertical line. An arrow points from the stack to the right. Below the stack is the formula for the Softmax classifier:

$$y_k = \frac{\exp(a_k)}{\sum_i \exp(a_i)}$$

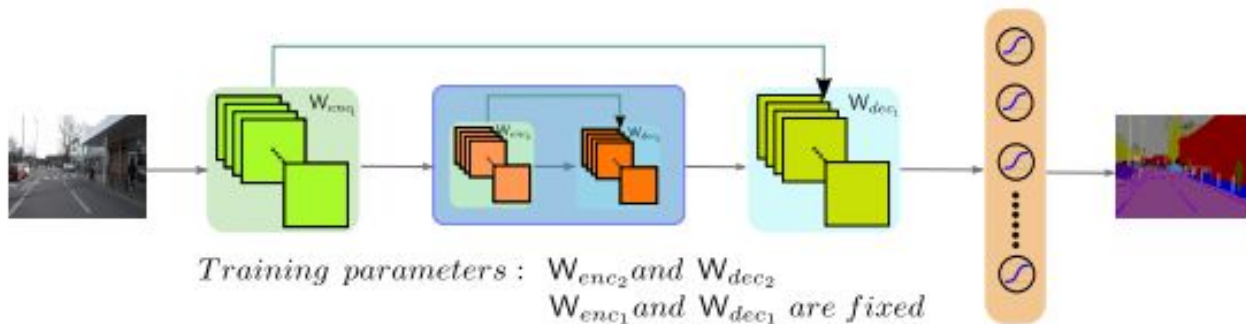
Training the SegNet



Training parameters : W_{enc1} and W_{dec1}

(a)

$$y_k = \frac{\exp(a_k)}{\sum_i \exp(a_i)}$$



Training parameters : W_{enc2} and W_{dec2}
 W_{enc1} and W_{dec1} are fixed

(b)

$$y_k = \frac{\exp(a_k)}{\sum_i \exp(a_i)}$$

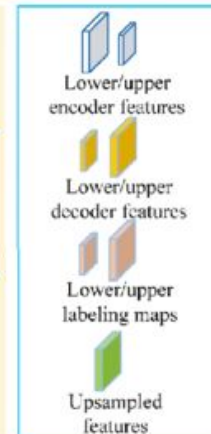
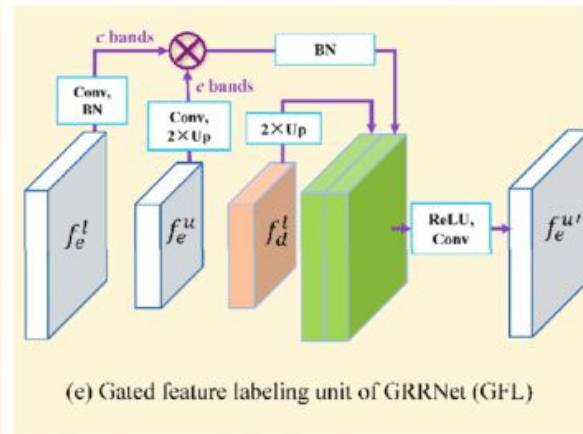
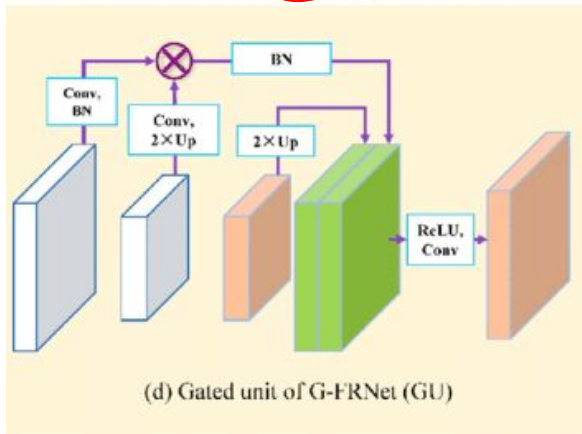
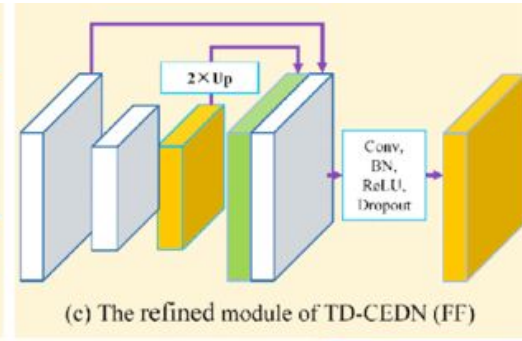
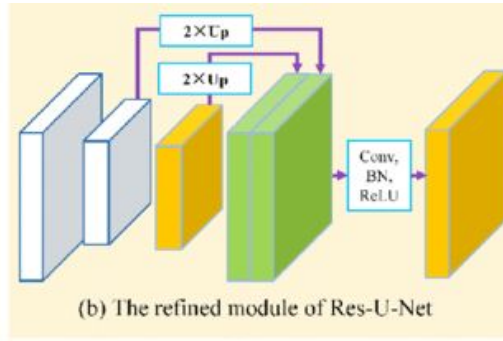
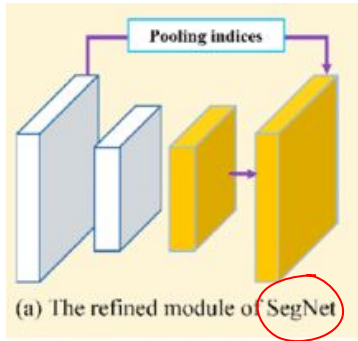
Comparison of decoder variants

Variant	Params (M)	Storage multiplier	Inference time (ms)
Fixed upsampling			
Bilinear-Interpolation	0.625	0	24.2
Upsampling using max-pooling indices			
SegNeg-Basic	1.425	1	52.6
SegNeg-Basic-Encoder Addition	1.425	64	53.0
SegNeg-Basic-SingleChannelDecoder	0.625	1	33.1
Learning to upsample			
FCN-Basic	0.65	11	24.2
FCN-Basic-NoAddition	0.65	n/a	23.8
FCN-Basic-NoDimReduction	1.625	64	44.8
FCN-Basic-NoAddition-NoDimReduction	1.625	0	43.9

Summary of different decoders analysis

- The **best performance** is achieved **when encoder feature maps are stored in full**
- **Compressed forms** of encoder feature maps can be stored and used for decoding to **meet memory constraints**
- **Larger decoders increase performance** for a given encoder network

Schematic representation of different architectures

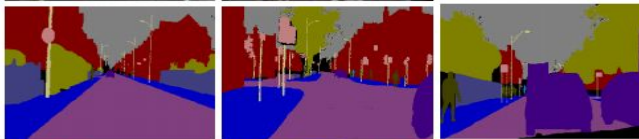


Road scene segmentation

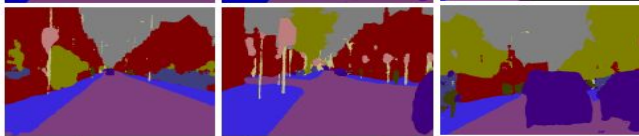
Test samples



Ground Truth



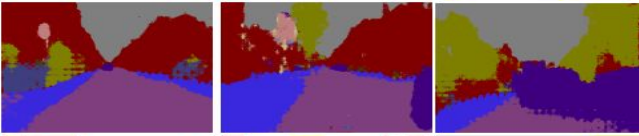
SegNet



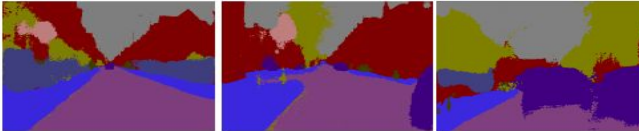
DeepLab-LargeFOV



FCN



DeconvNet



Indoor scene segmentation

Test samples



Ground Truth



SegNet



DeepLab-LargeFOV



FCN (learnt deconv)



DeconvNet



Conclusion

- SegNet is more efficient compared to other architectures since it only stores the max-pooling indices of the feature maps and uses them in its decoder network to achieve good performance
- On large and well known datasets SegNet performs competitively, achieving high scores for road scene understanding
- End-to-end learning of deep segmentation architectures is a harder challenge

References

- 1) <http://mi.eng.cam.ac.uk/projects/segnet/>
- 2) <https://www.cyberailab.com/home/segnet-an-image-segmentation-neural-network>
- 3) <http://www.robots.ox.ac.uk/~tvb/publications/2017/CRFMeetCNN4SemanticSegmentation.pdf>
- 4) <https://neurohive.io/en/popular-networks/vgg16/>

Questions?

