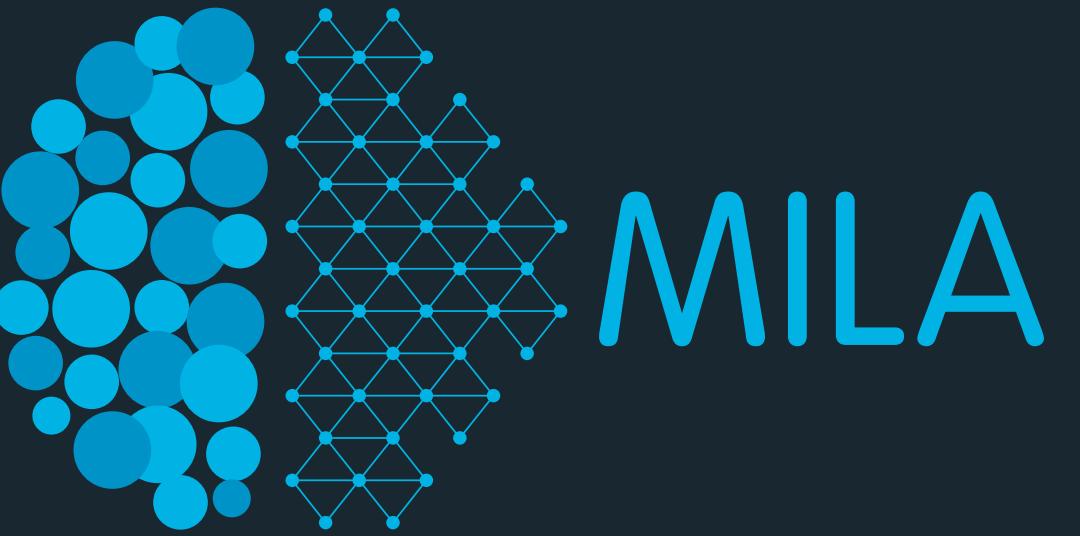


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d'apprentissage
de Montréal



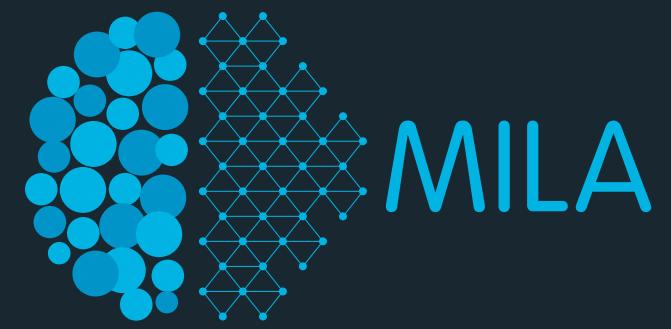
Fully Convolutional DenseNets for Semantic Segmentation

Simon Jégou, Michal Drozdzal, David Vazquez, Adriana Romero, Yoshua Bengio

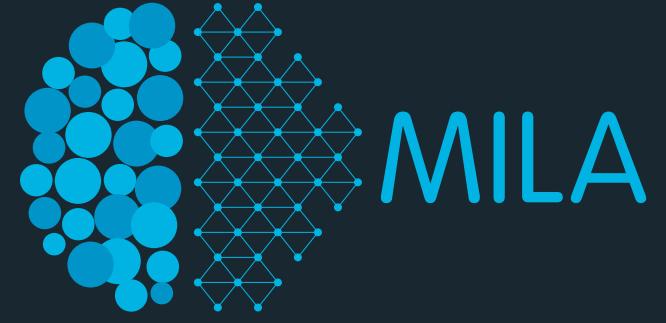
Outline

- Classical architectures and their extensions to FCN
- FC-DenseNets
- Results
- Wrap up

DL and Computer Vision

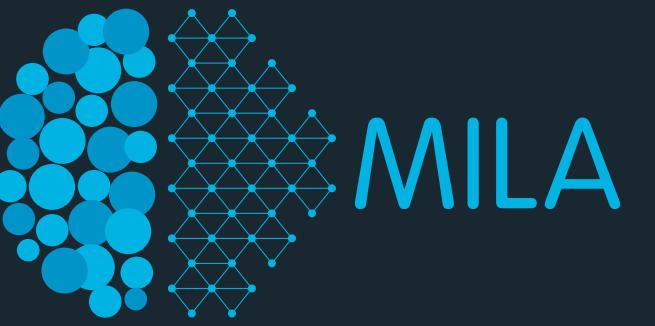


DL and Computer Vision



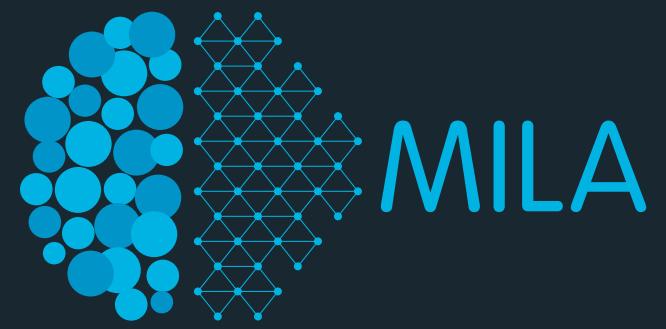
(Zhou et al., 2015)

DL and Computer Vision

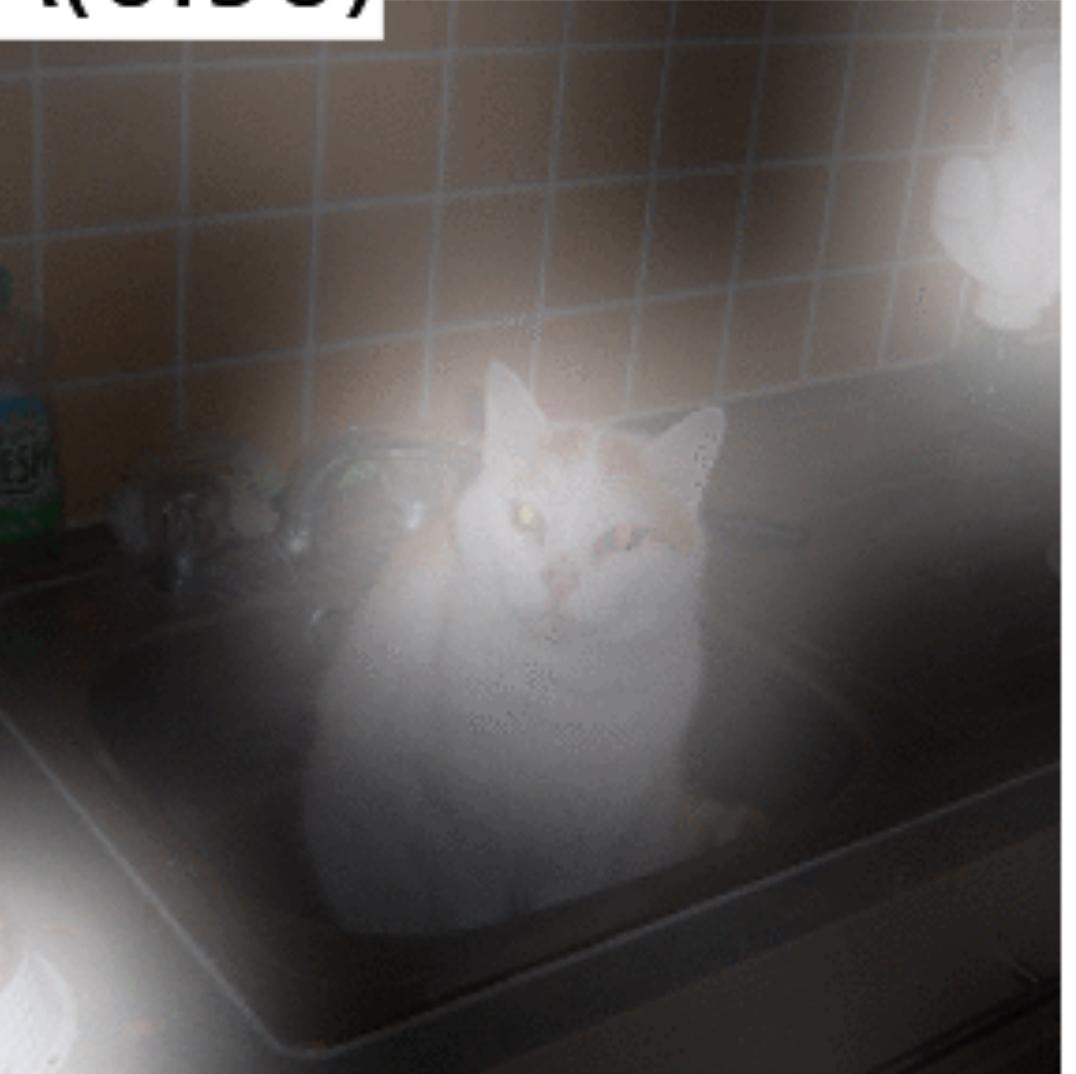


(Xu *et al.*, 2015)

DL and Computer Vision

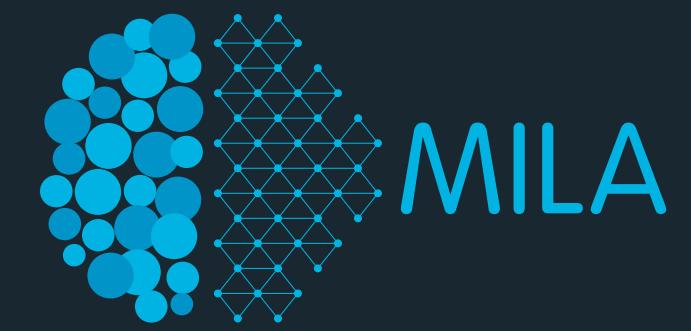


A(0.96)



(Xu *et al.*, 2015)

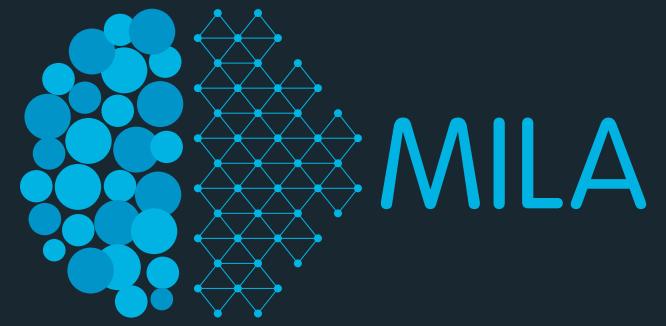
DL and Computer Vision



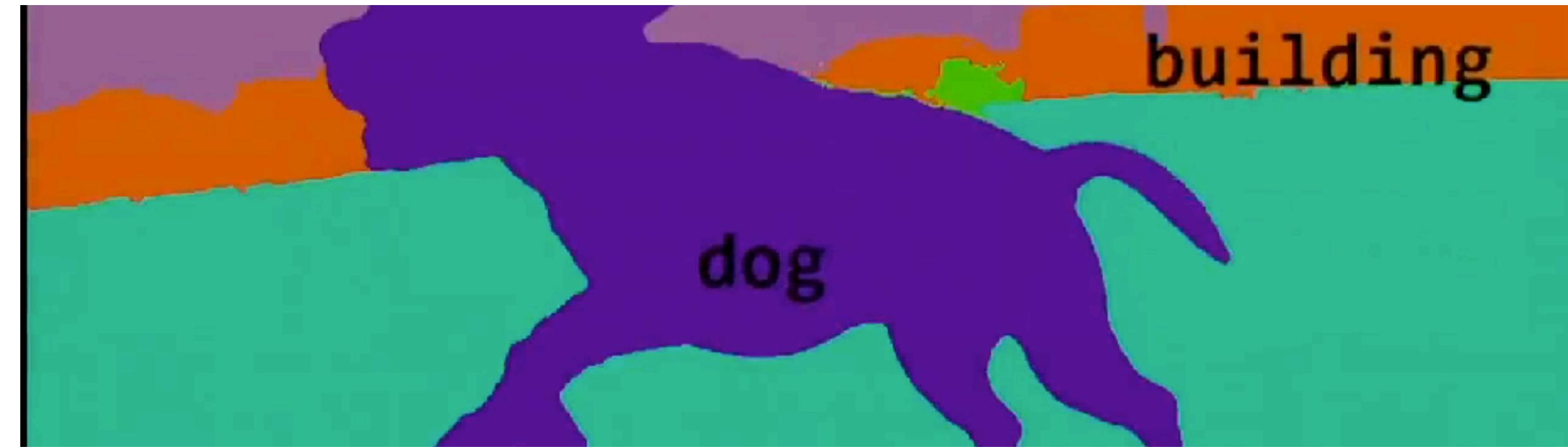
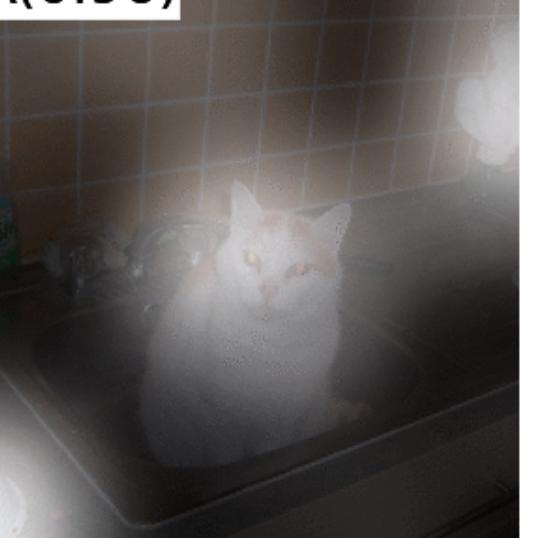
A(0.96)



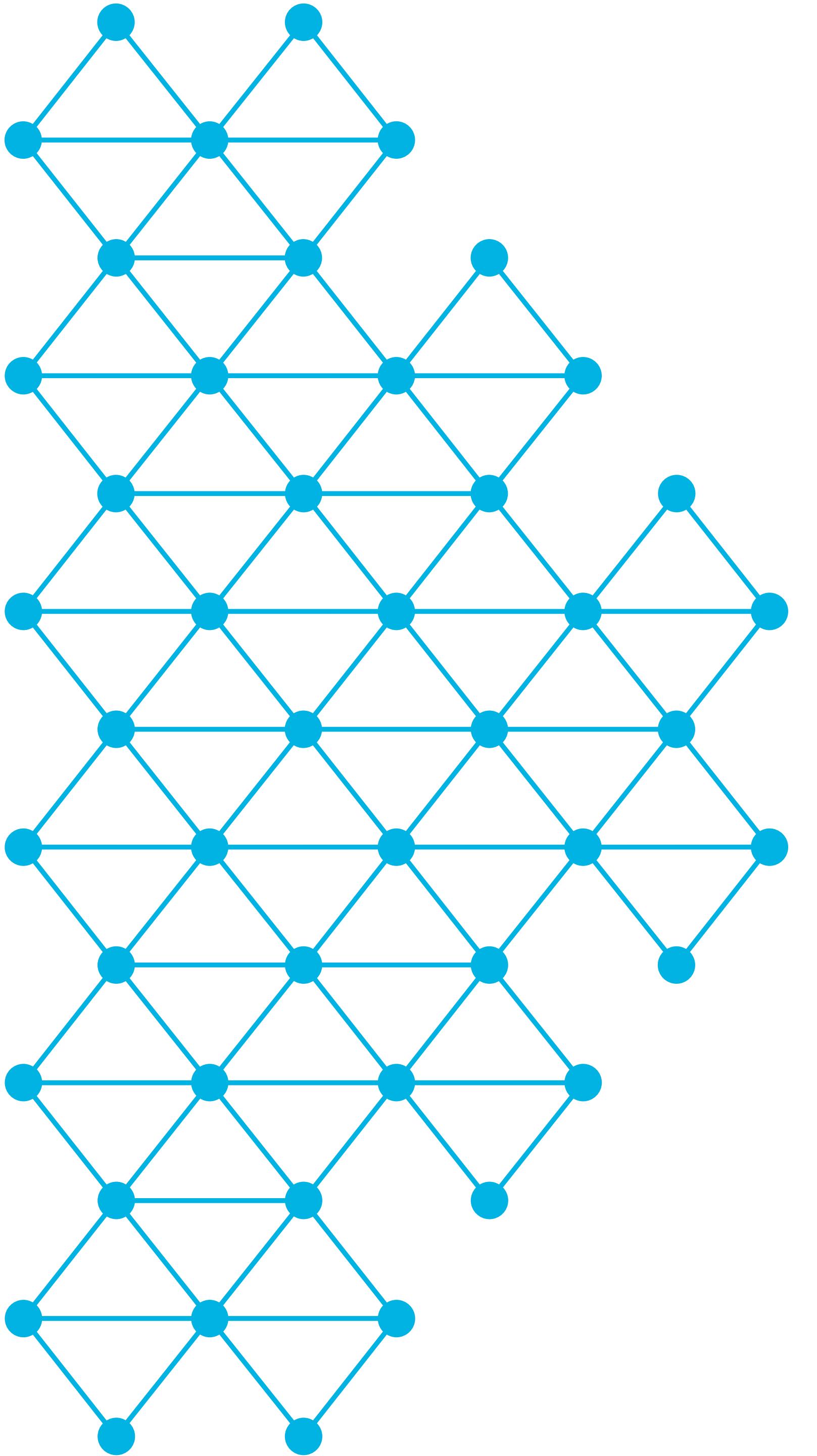
DL and Computer Vision



A(0.96)



(<https://www.youtube.com/watch?v=FroRjEejA30>, 2015)



**Classical architectures
and their extensions**

CNN - classification

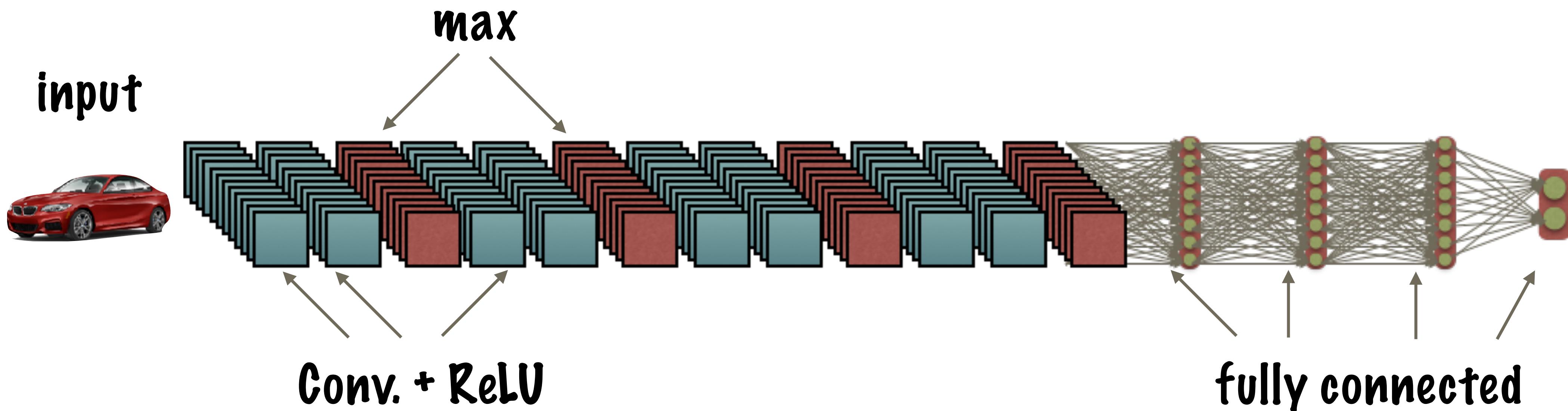
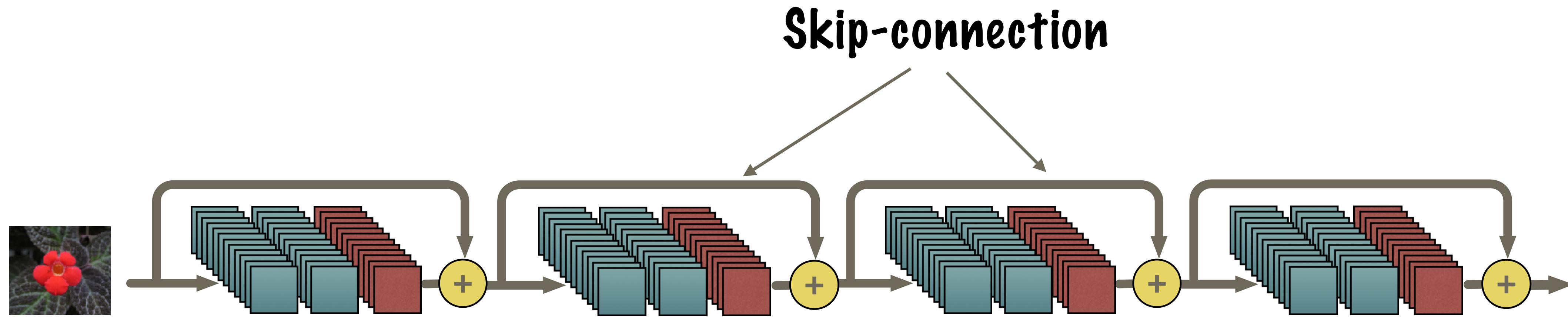


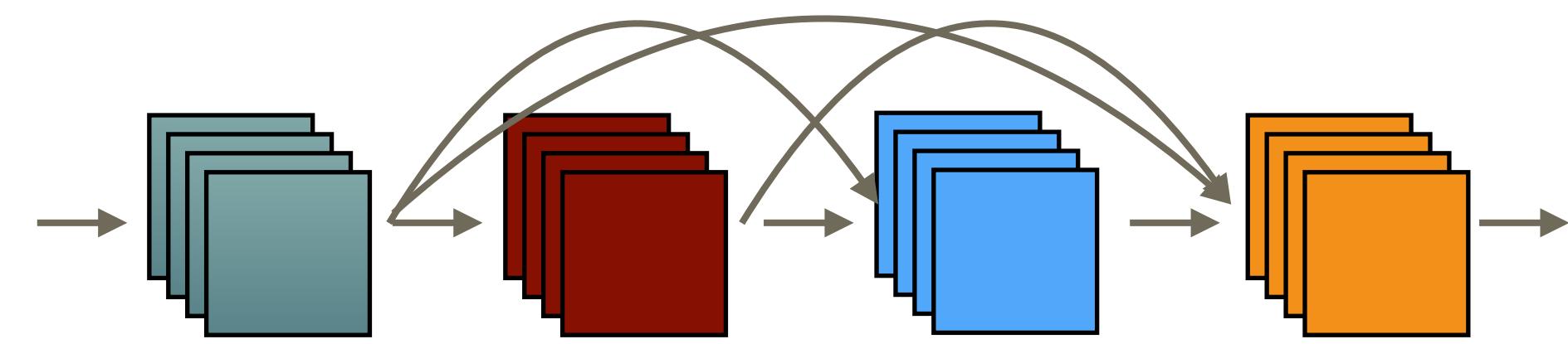
Figure courtesy of Kilian Weinberger

ResNet - classification



DenseNet - classification

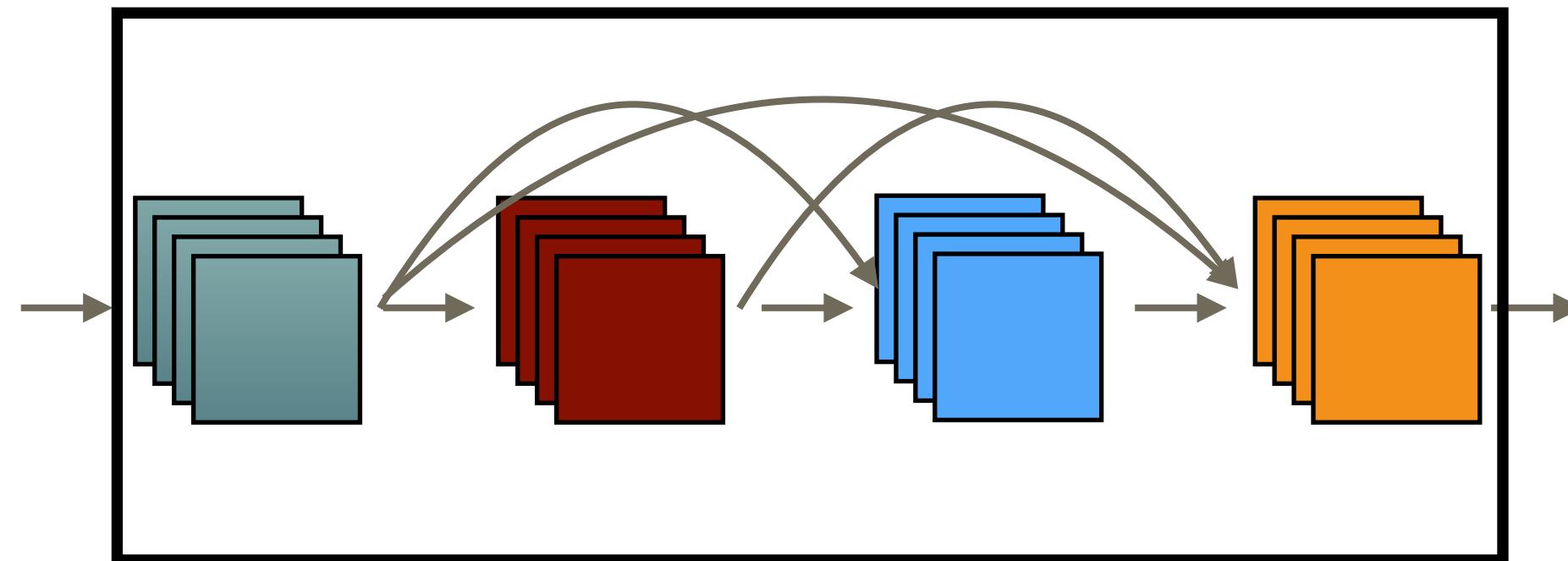
Connect every layer to every other layer of the same filter size (dense blocks).



DenseNet - classification

Connect every layer to every other layer of the same filter size (dense blocks).

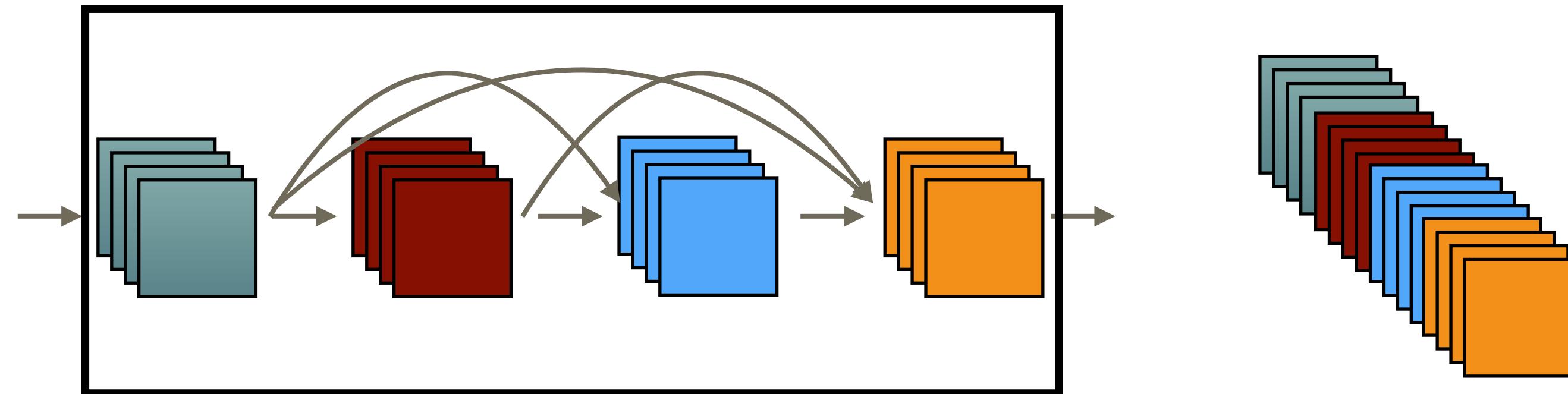
Dense Block (4 layers, growth rate 4)



DenseNet - classification

Connect every layer to every other layer of the same filter size (dense blocks).

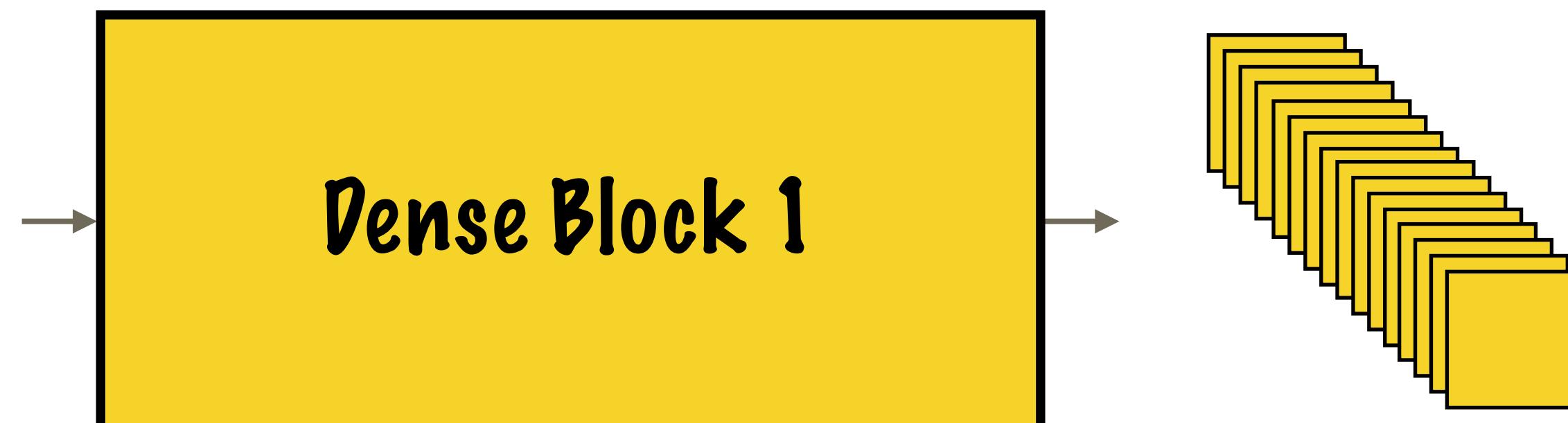
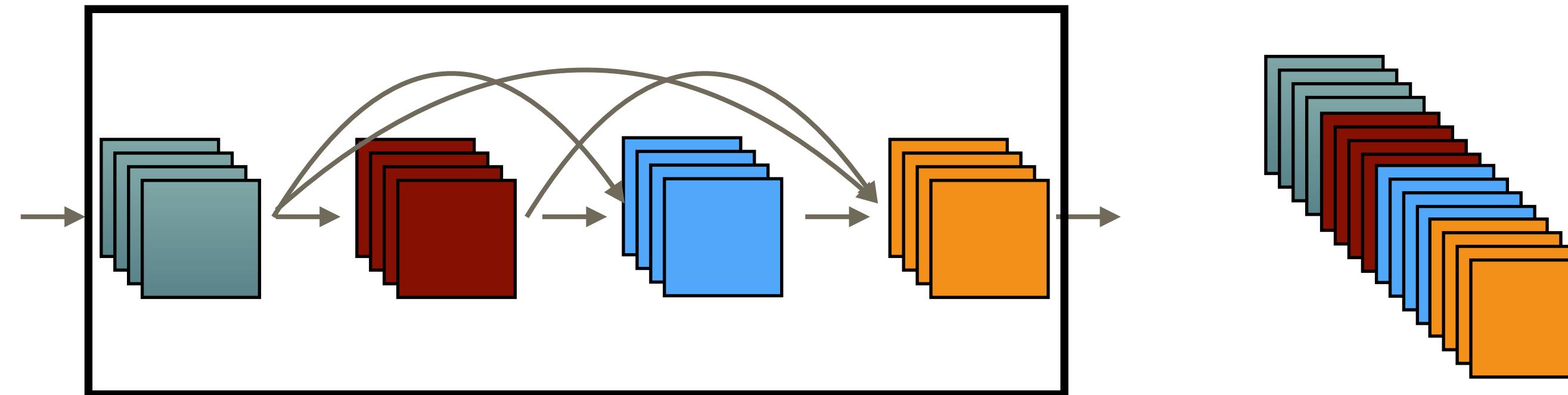
Dense Block (4 layers, growth rate 4)



DenseNet - classification

Connect every layer to every other layer of the same filter size (dense blocks).

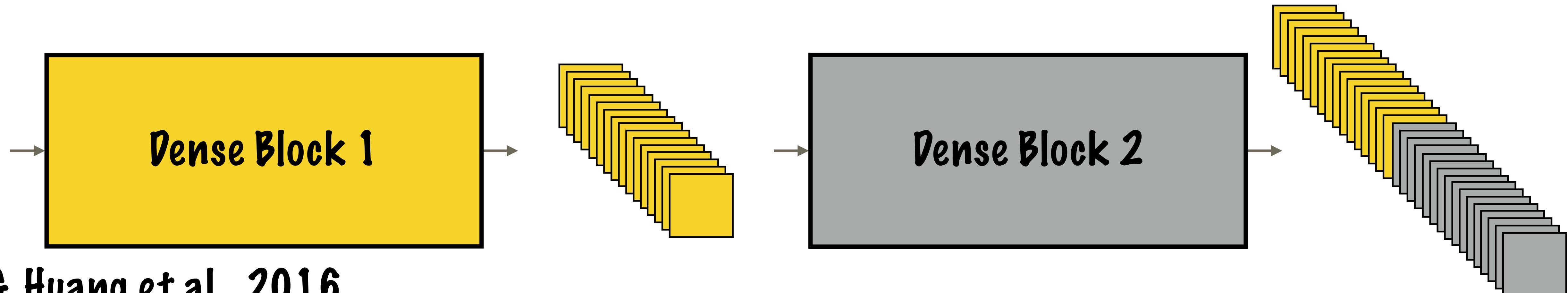
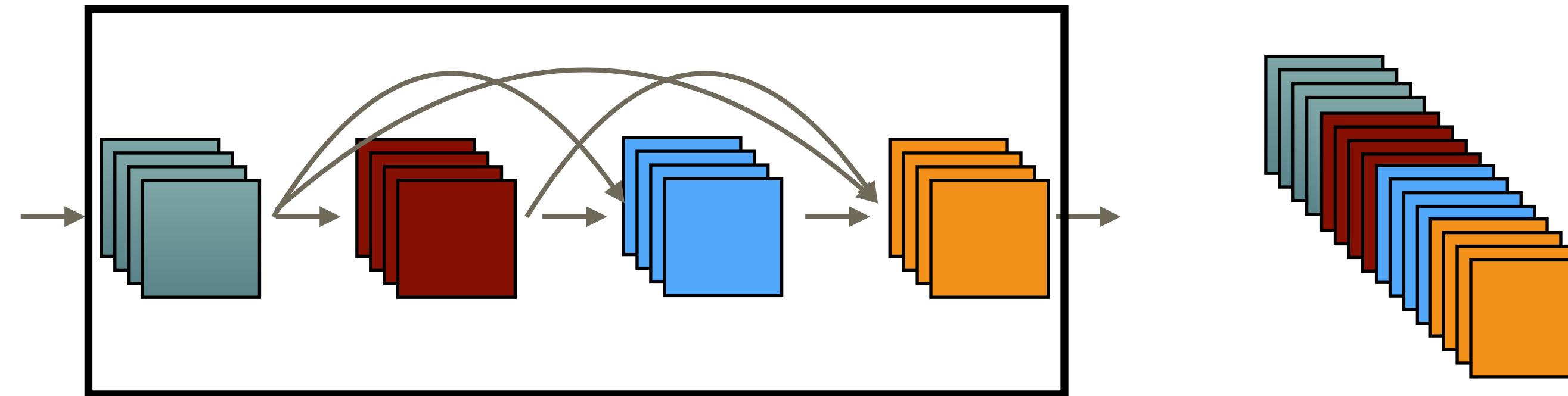
Dense Block (4 layers, growth rate 4)



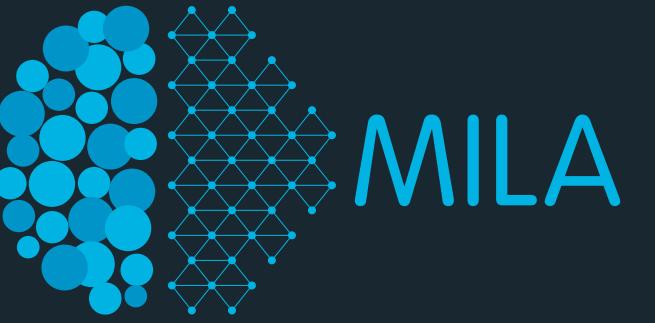
DenseNet - classification

Connect every layer to every other layer of the same filter size (dense blocks).

Dense Block (4 layers, growth rate 4)



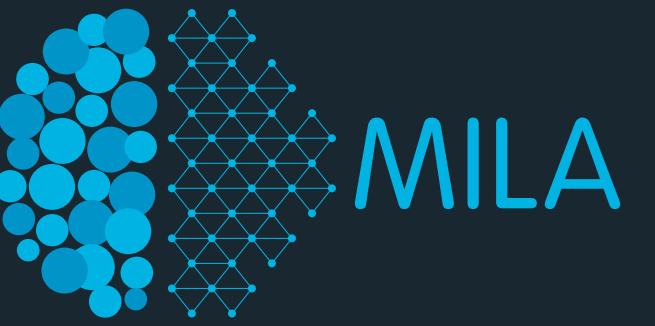
CNNs vs ResNets vs DenseNets



CNNs

$$x \downarrow F(x)$$

CNNs vs ResNets vs DenseNets

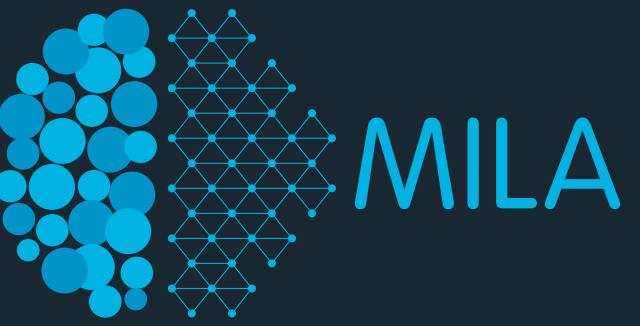


CNNs

$$x \downarrow F(x)$$

$$F(x) = \left(\begin{array}{c} \text{Conv.} \\ + \\ \text{ReLU} \end{array} \right) x n$$

CNNs vs ResNets vs DenseNets

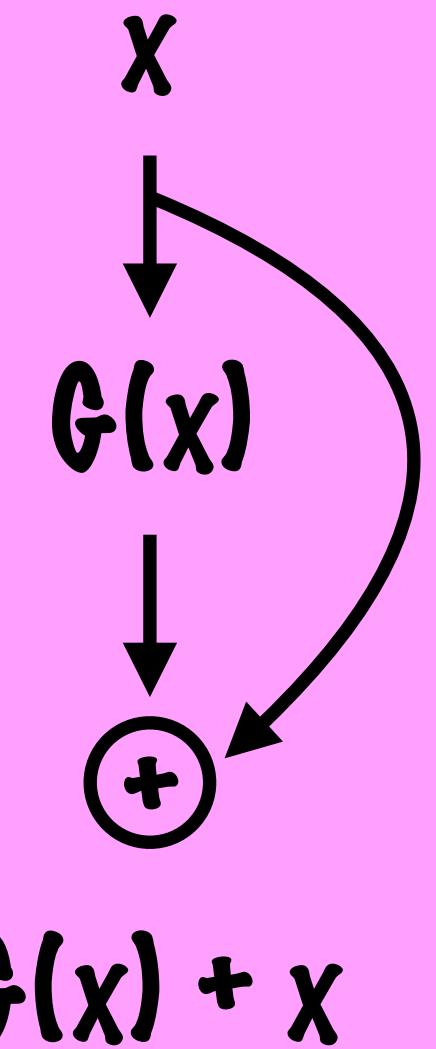


CNNs

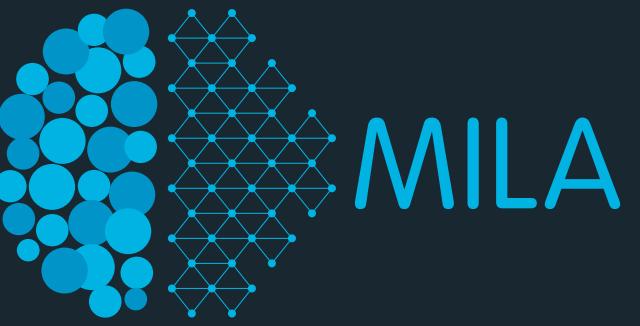
$$x \downarrow F(x)$$

$$F(x) = \left(\begin{array}{c} \text{Conv.} \\ + \\ \text{ReLU} \end{array} \right) x n$$

ResNets



CNNs vs ResNets vs DenseNets

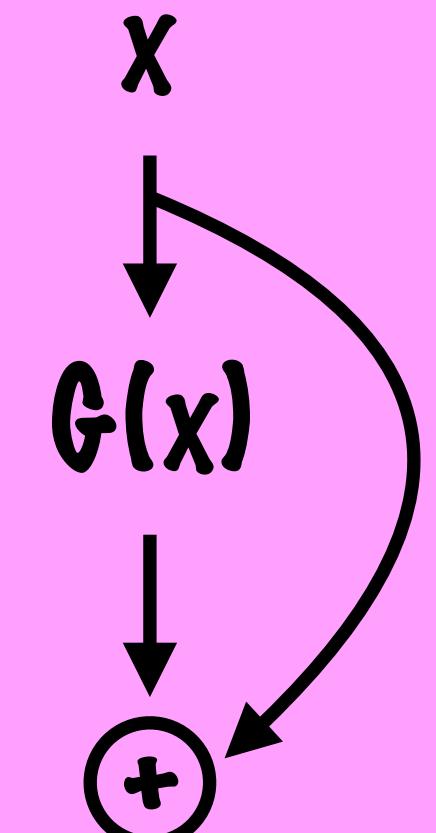


CNNs

$$x \downarrow \\ F(x)$$

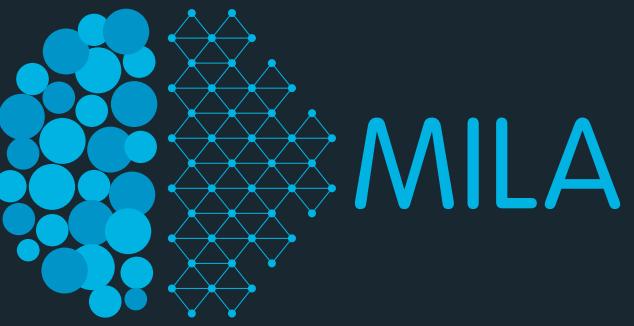
$$F(x) = \left(\begin{array}{c} \text{Conv.} \\ + \\ \text{ReLU} \end{array} \right) x n$$

ResNets



$$G(x) = \left(\begin{array}{c} \text{BN} \\ + \\ \text{ReLU} \\ + \\ \text{Conv.} \end{array} \right) x n$$

CNNs vs ResNets vs DenseNets

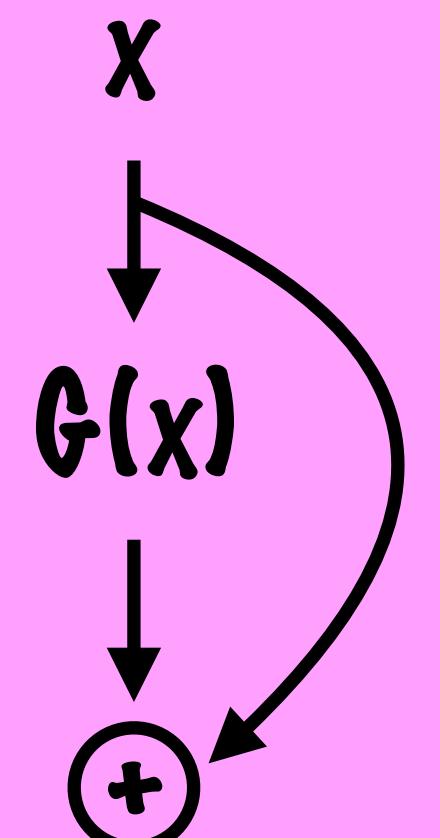


CNNs

$$x \downarrow F(x)$$

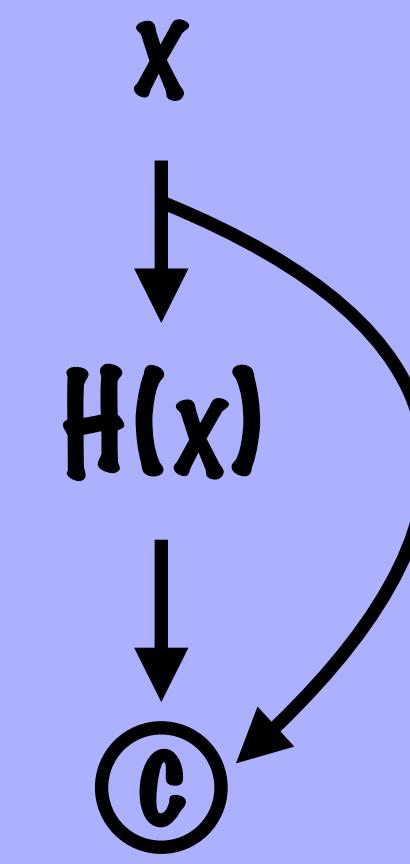
$$F(x) = \left(\begin{array}{c} \text{Conv.} \\ + \\ \text{ReLU} \end{array} \right) x n$$

ResNets



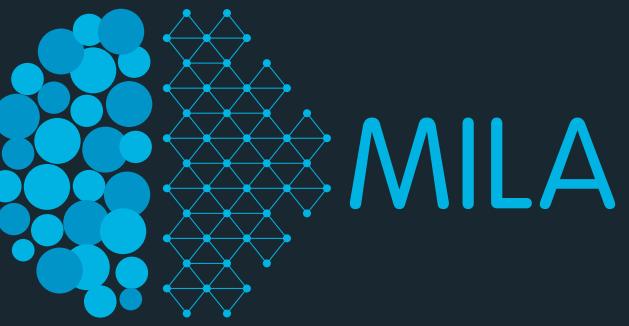
$$G(x) = \left(\begin{array}{c} \text{BN} \\ + \\ \text{ReLU} \\ + \\ \text{Conv.} \end{array} \right) x n$$

DenseNets



$$[H(x), x]$$

CNNs vs ResNets vs DenseNets

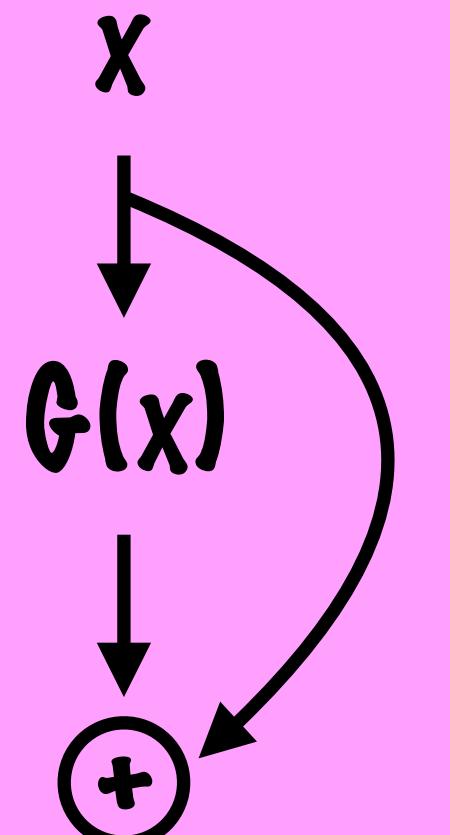


CNNs

$$x \downarrow F(x)$$

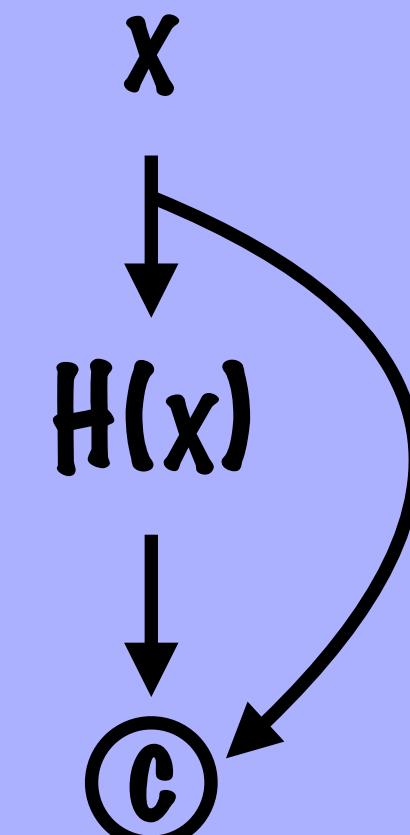
$$F(x) = \left(\begin{array}{c} \text{Conv.} \\ + \\ \text{ReLU} \end{array} \right) x n$$

ResNets



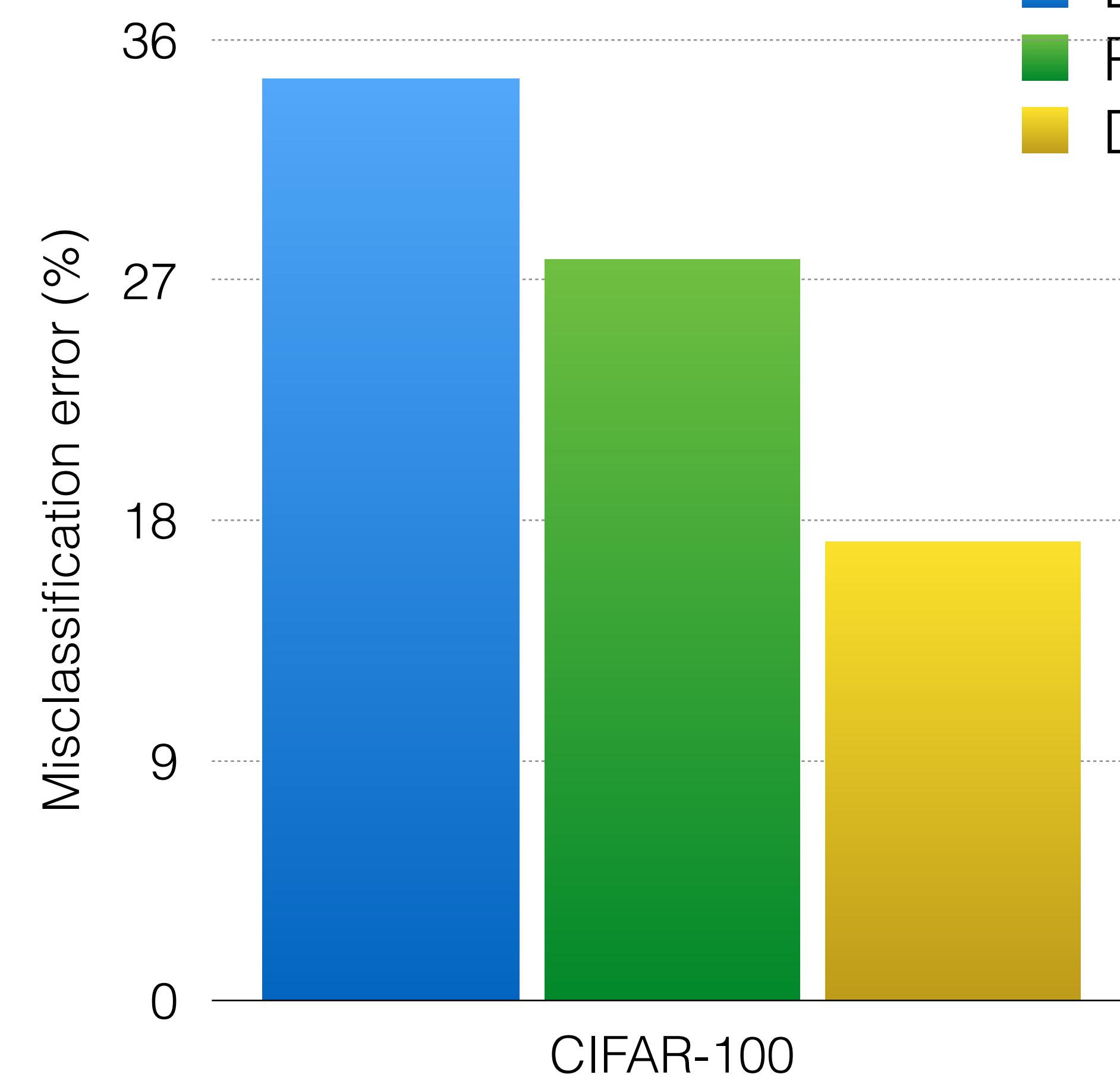
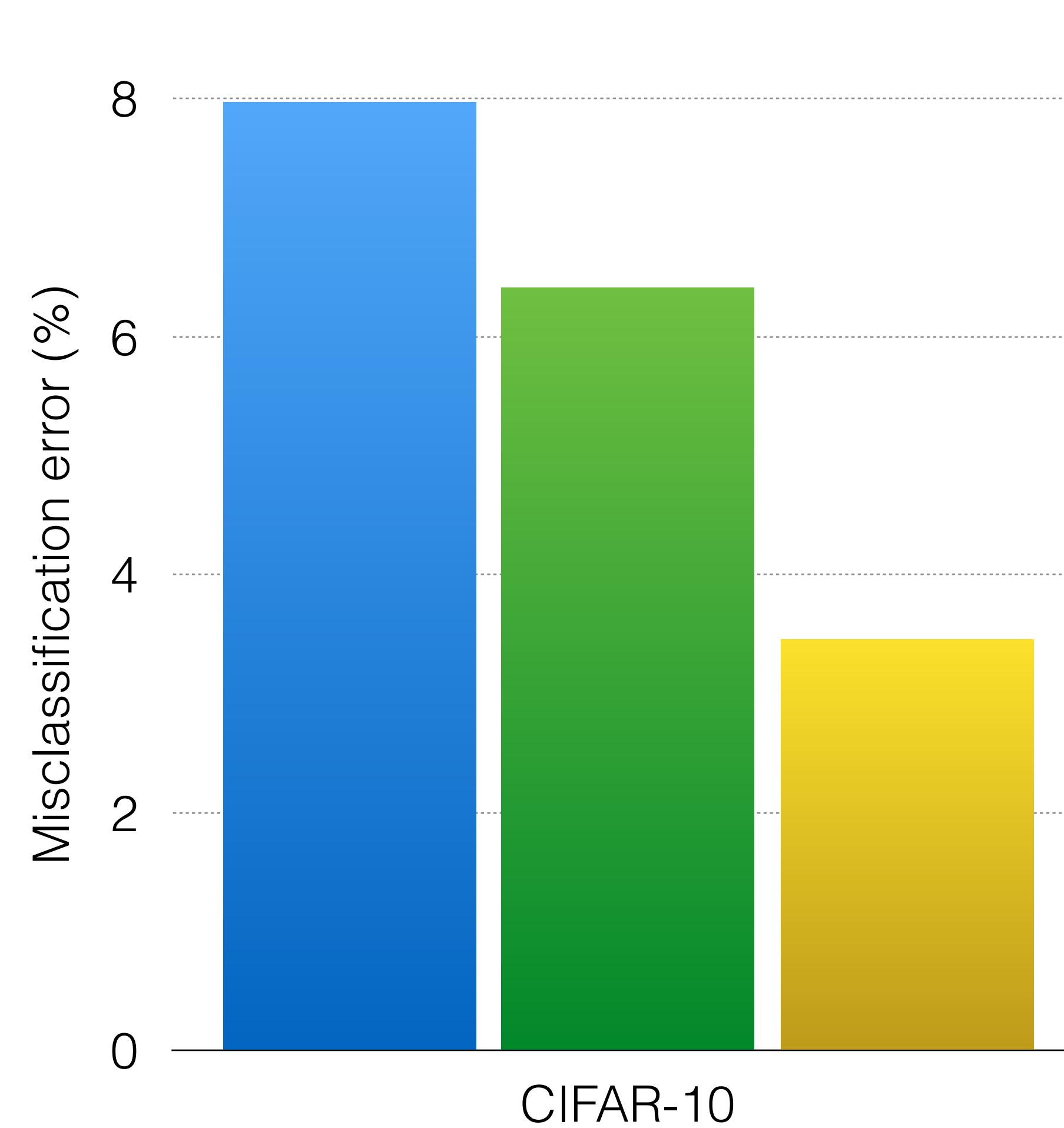
$$G(x) = \left(\begin{array}{c} \text{BN} \\ + \\ \text{ReLU} \\ + \\ \text{Conv.} \end{array} \right) x n$$

DenseNets

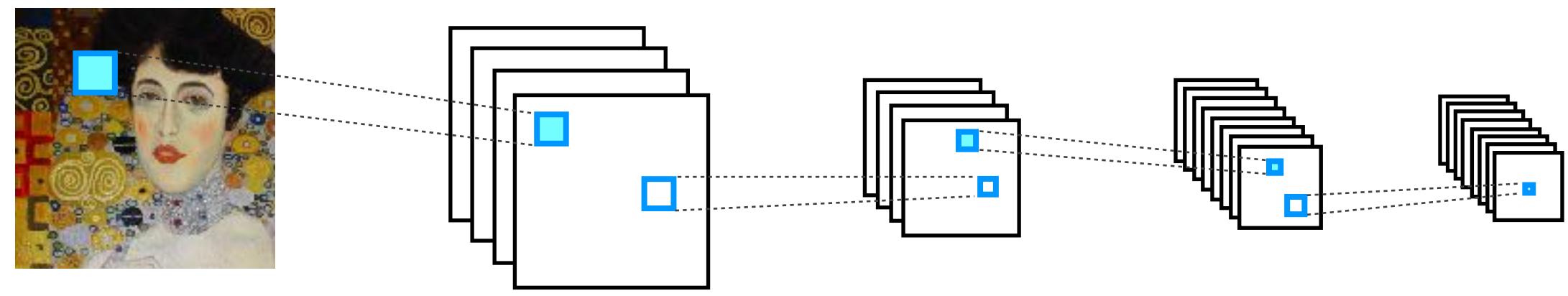
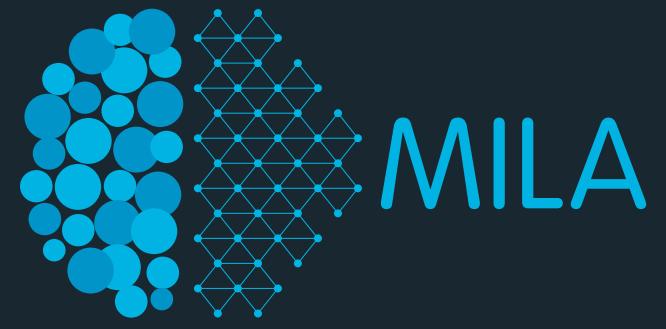


$$H(x) = \left(\begin{array}{c} \text{BN} \\ + \\ \text{ReLU} \\ + \\ \text{Conv.} \\ + \\ \text{Dropout} \end{array} \right)$$

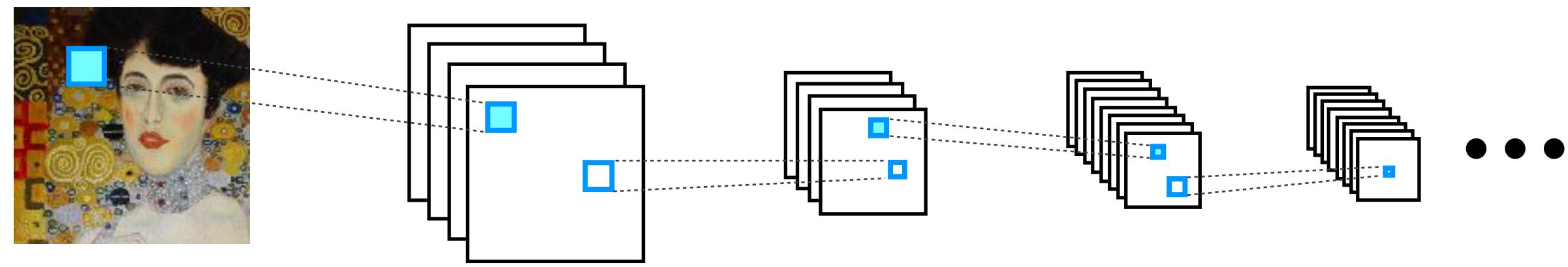
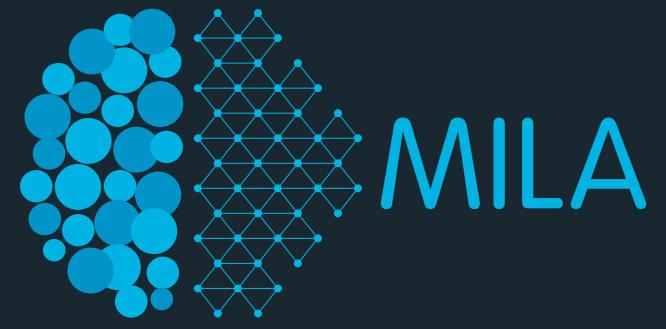
Classification results



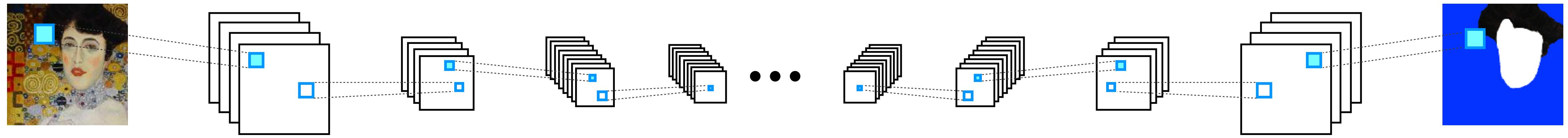
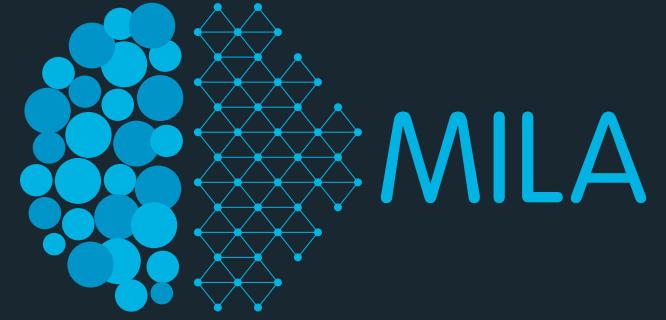
FCN - semantic segmentation



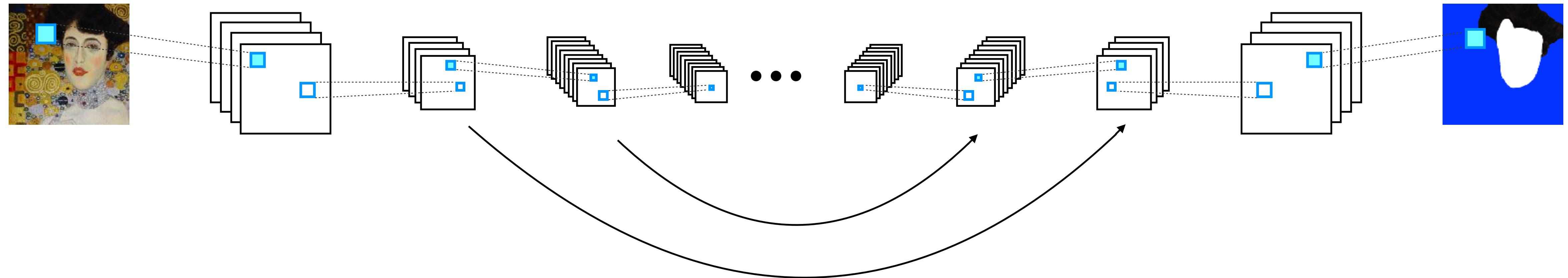
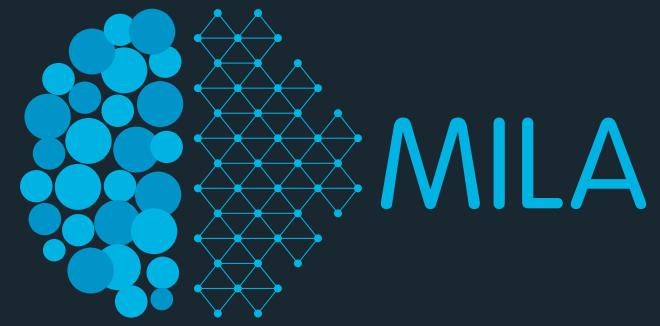
FCN - semantic segmentation



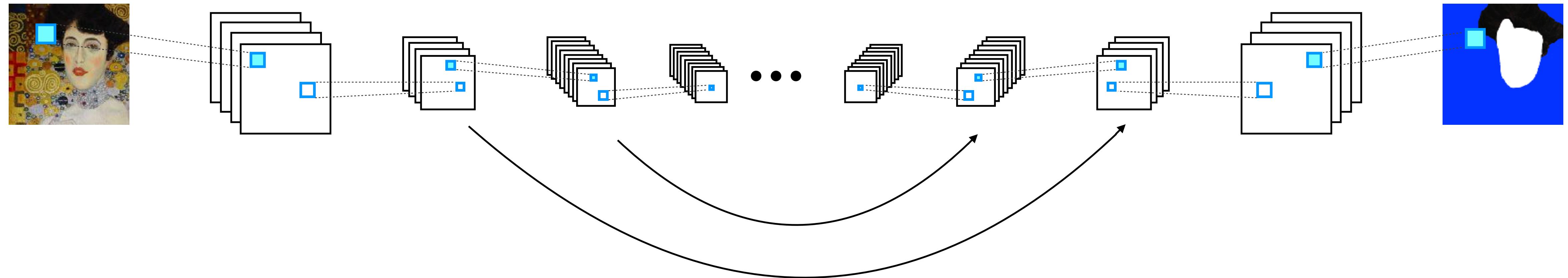
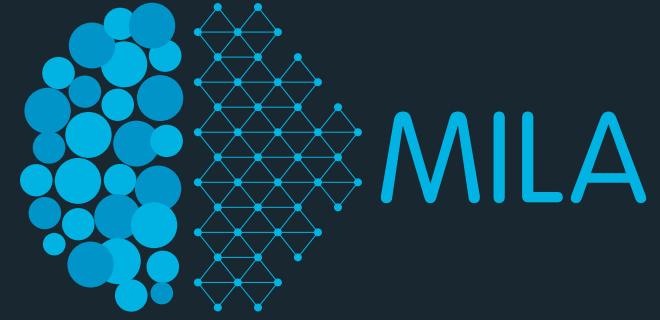
FCN - semantic segmentation



FCN - semantic segmentation

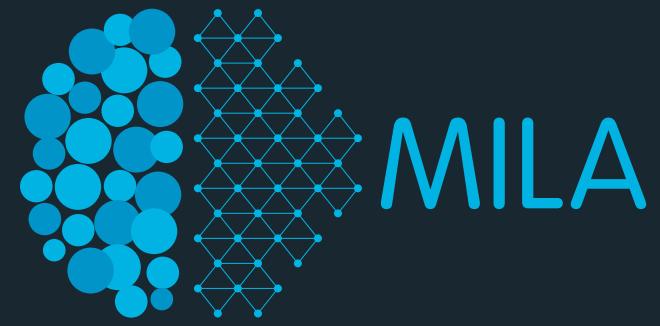


FCN - semantic segmentation

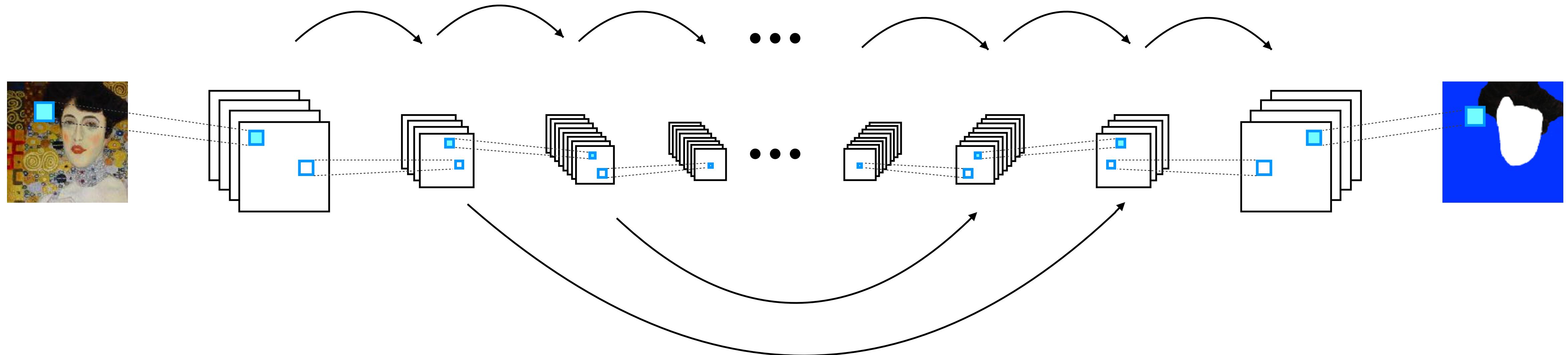


- * **Sum** (J. Long et al., 2015)
- * **Concat** (O. Ronneberger et al., 2015)
- * **Index tracking** (V. Badrinarayanan et al., 2015)

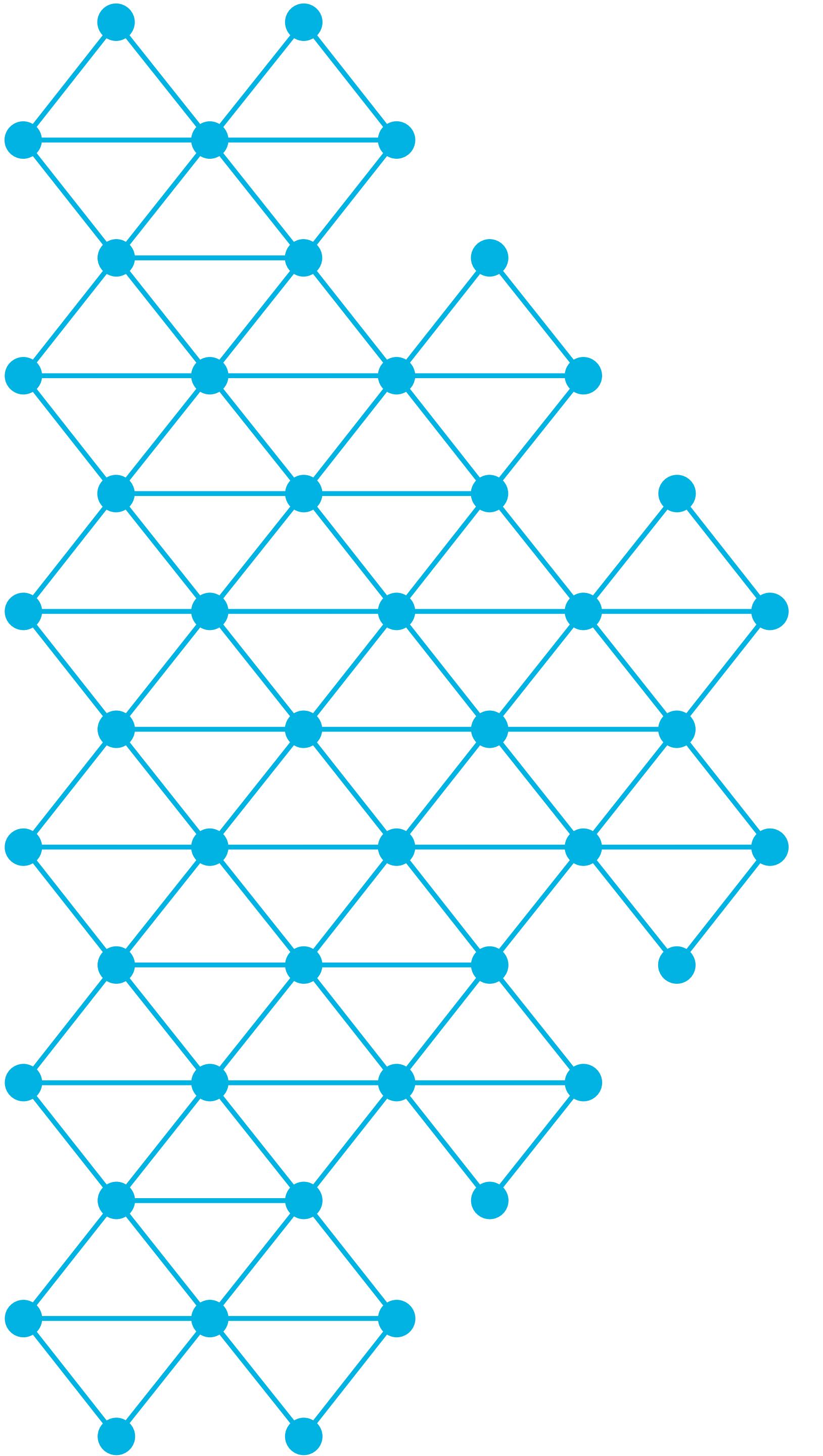
FCN - semantic segmentation



* ResUnet (M. Drozdzal et al., 2016)

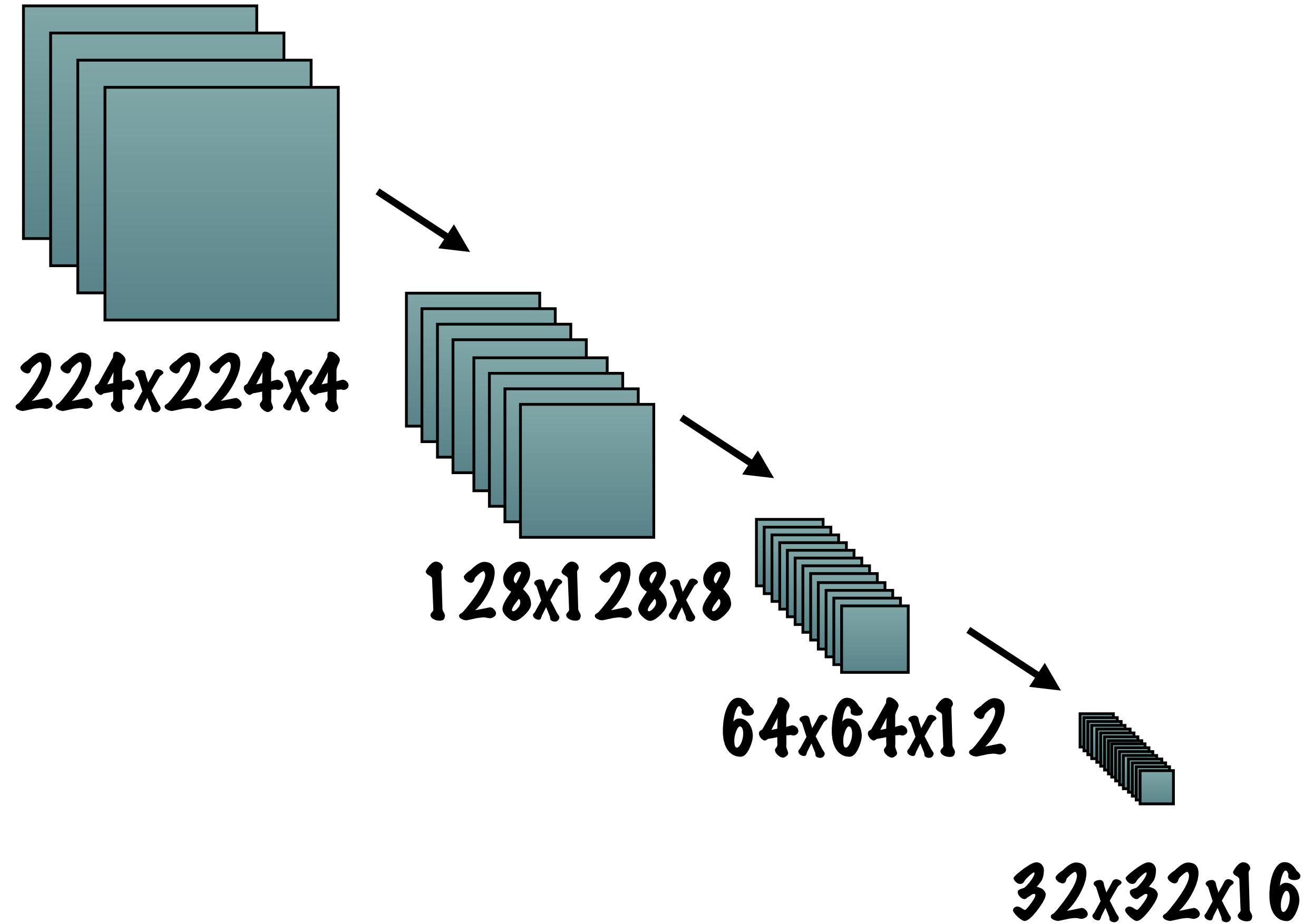
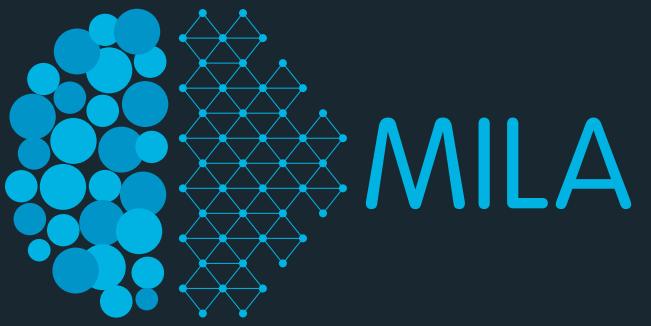


- * Sum (J. Long et al., 2015)
- * Concat (O. Ronneberger et al., 2015)
- * Index tracking (V. Badrinarayanan et al., 2015)

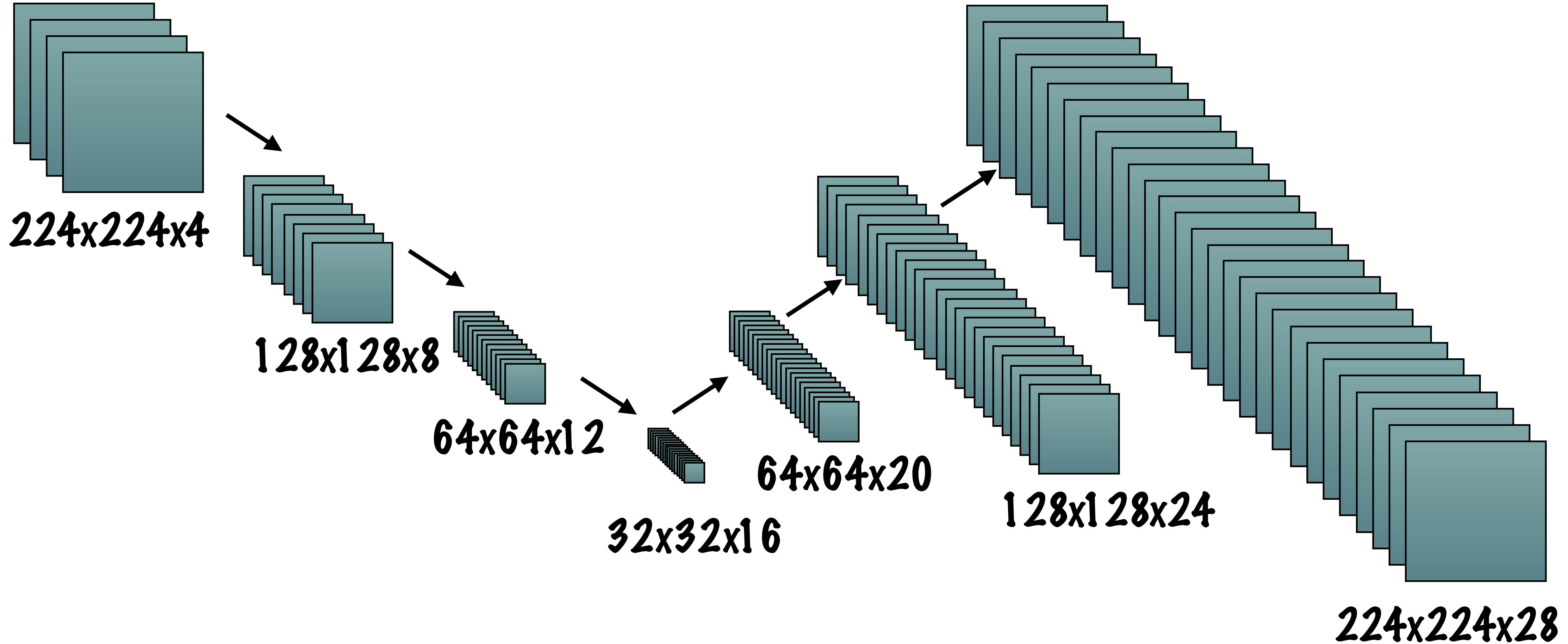


FC-DenseNets

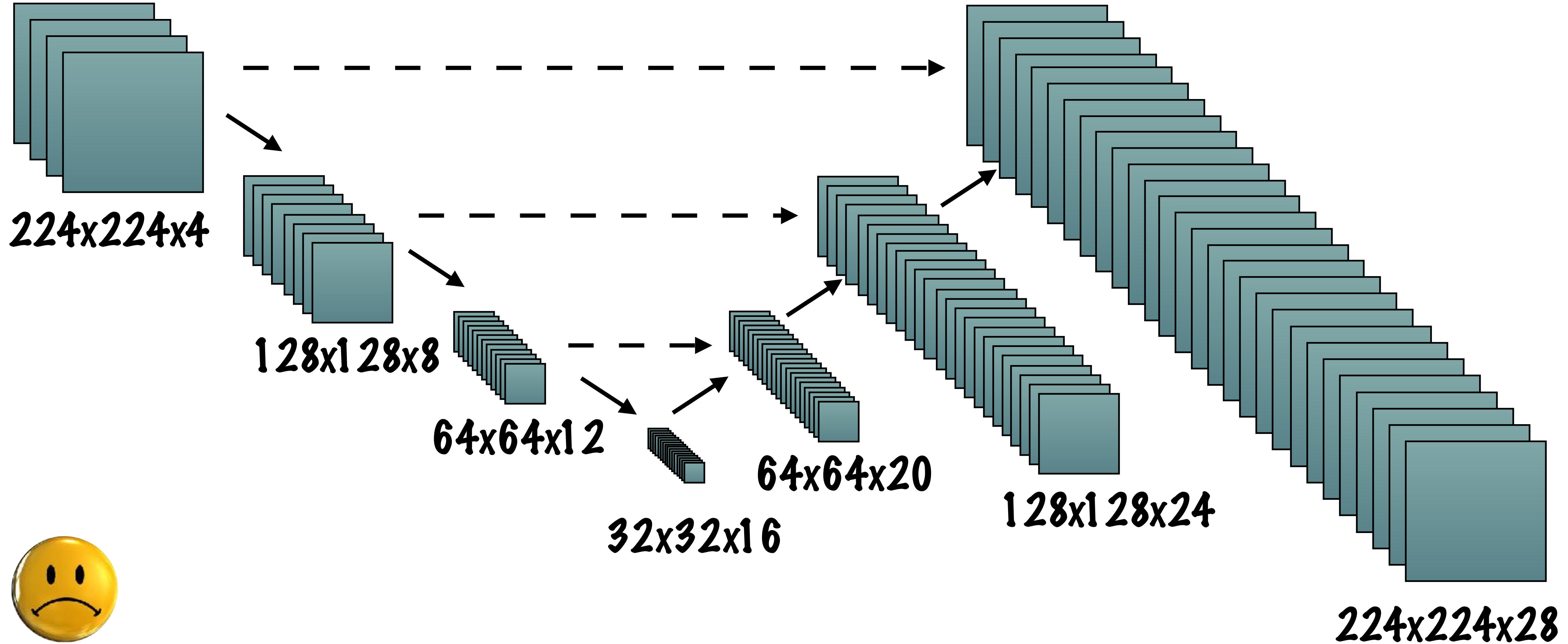
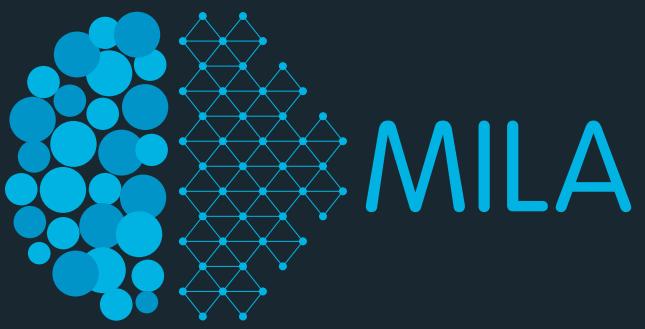
Naive extension of DenseNets



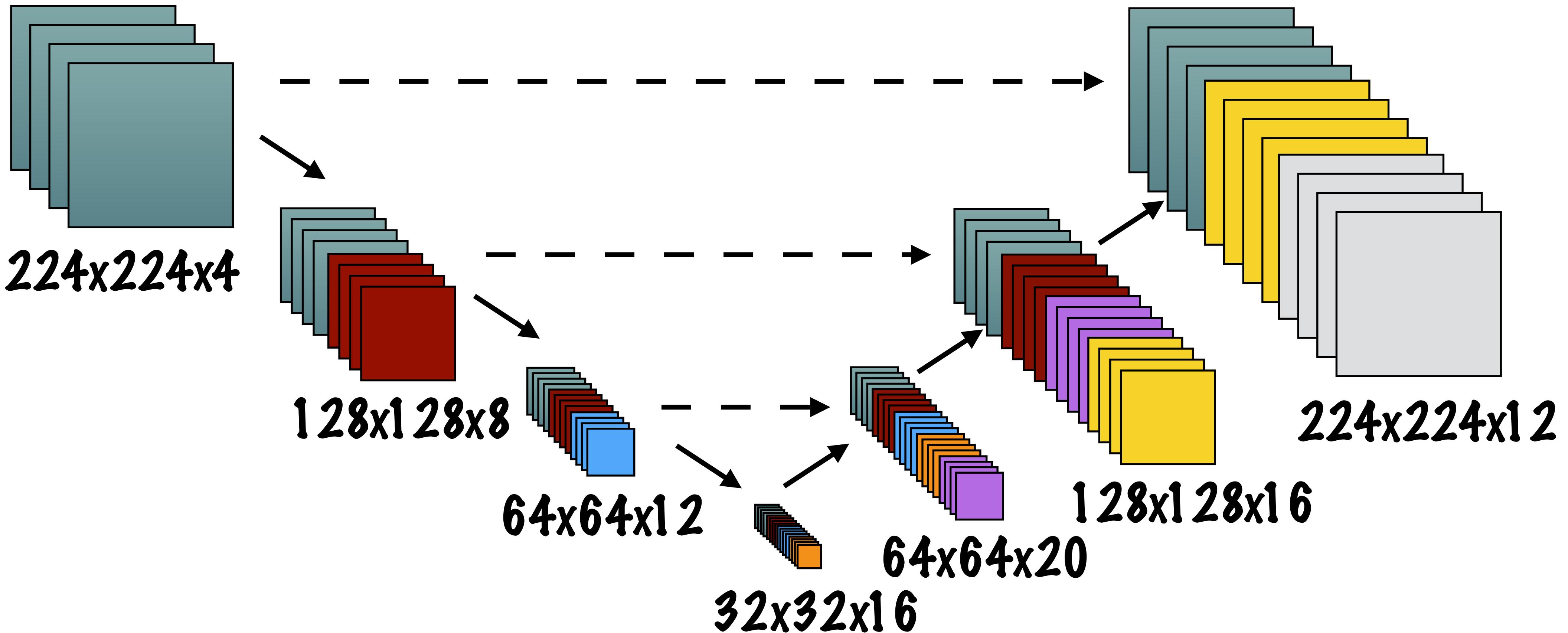
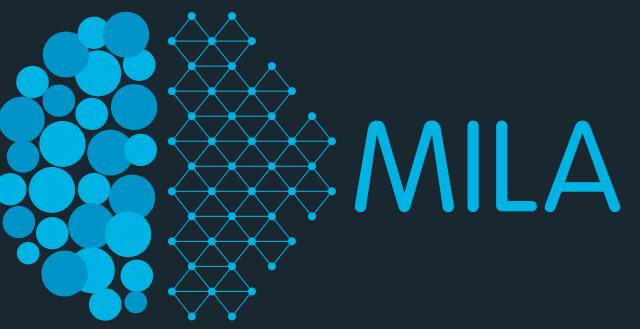
Naive extension of DenseNets



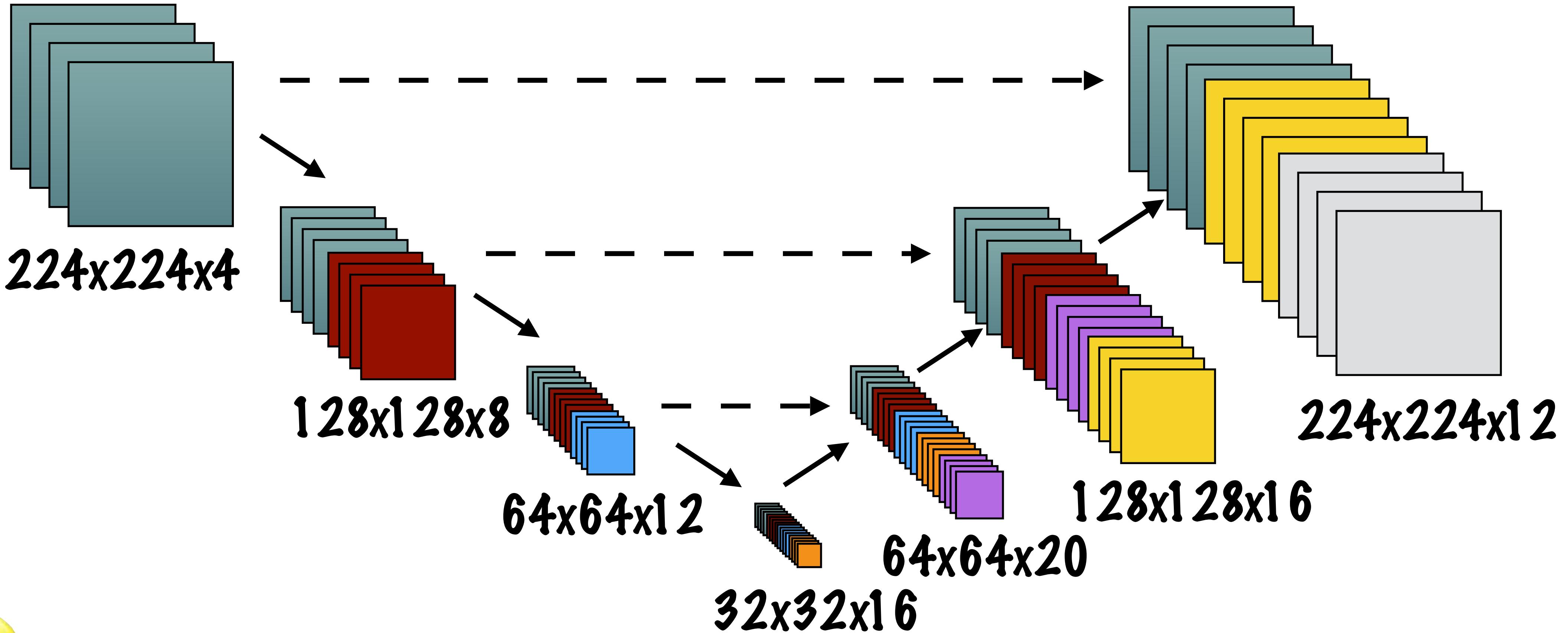
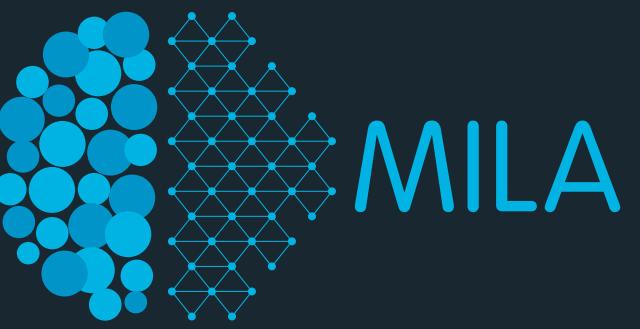
Naive extension of DenseNets

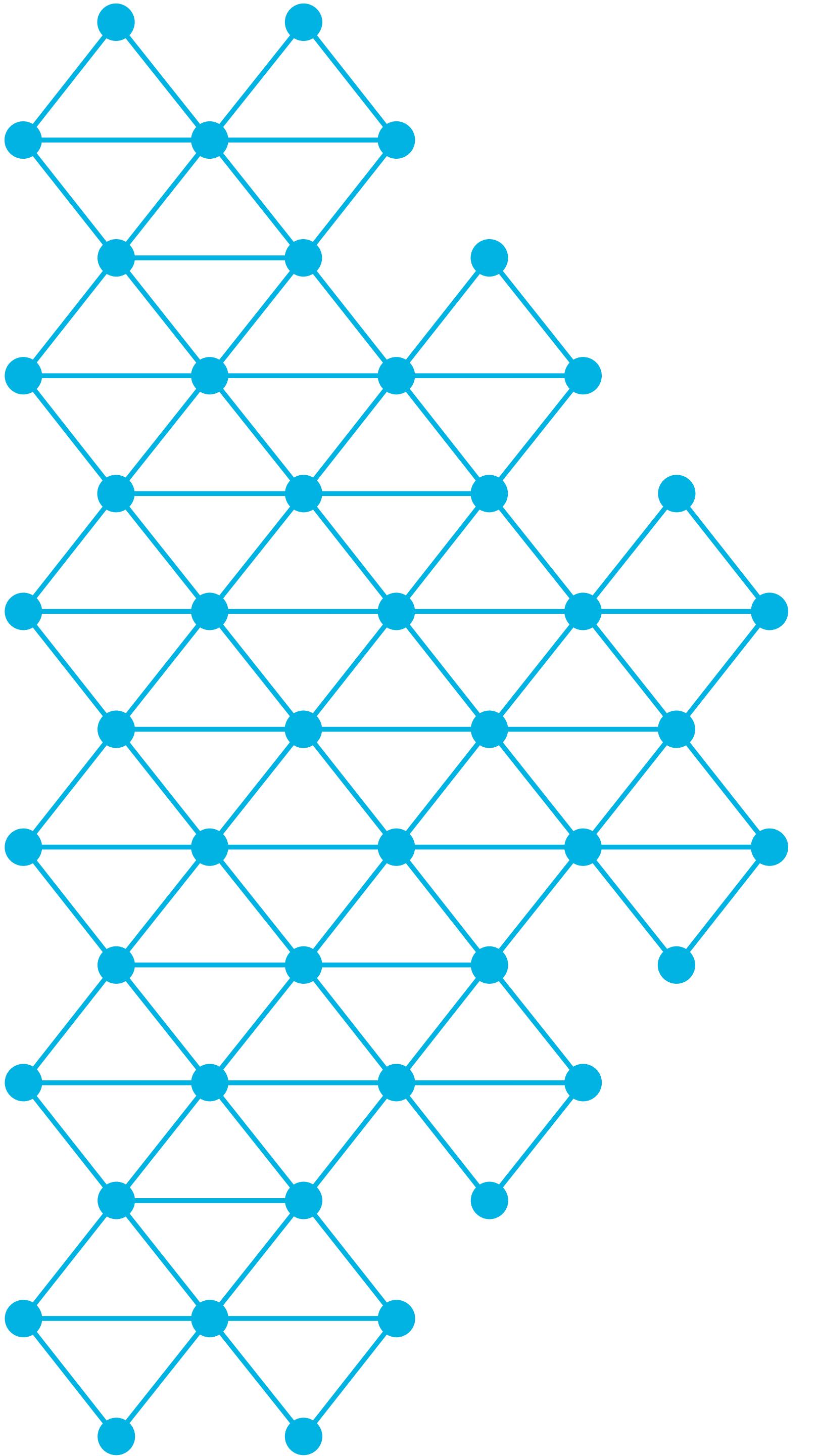


Fully Convolutional DenseNets



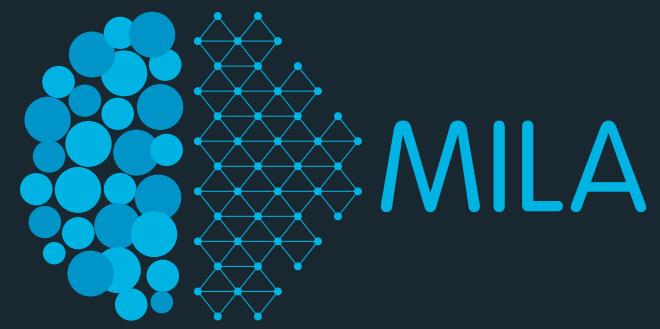
Fully Convolutional DenseNets





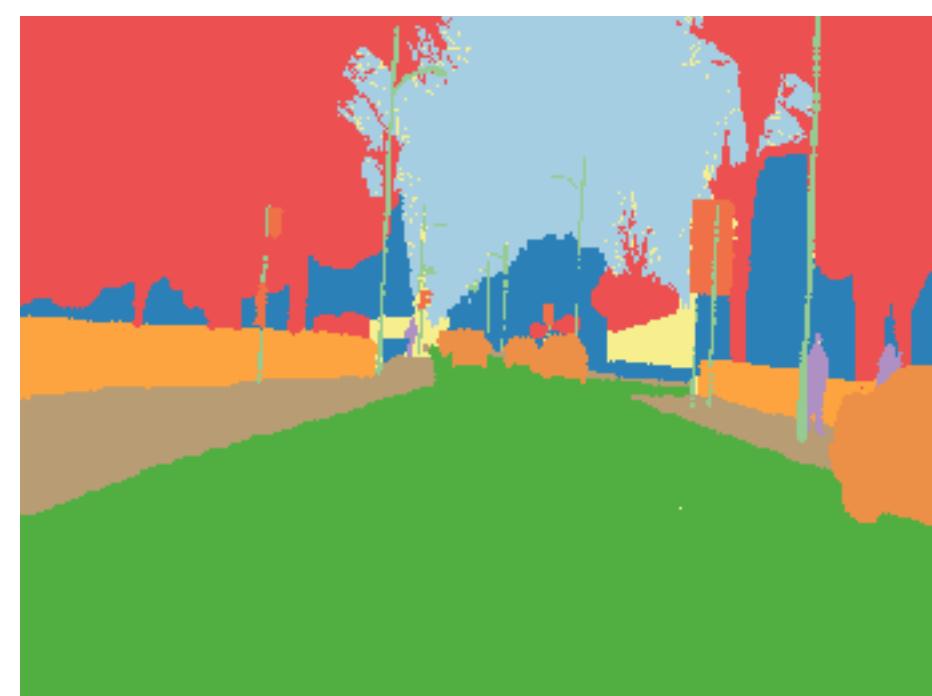
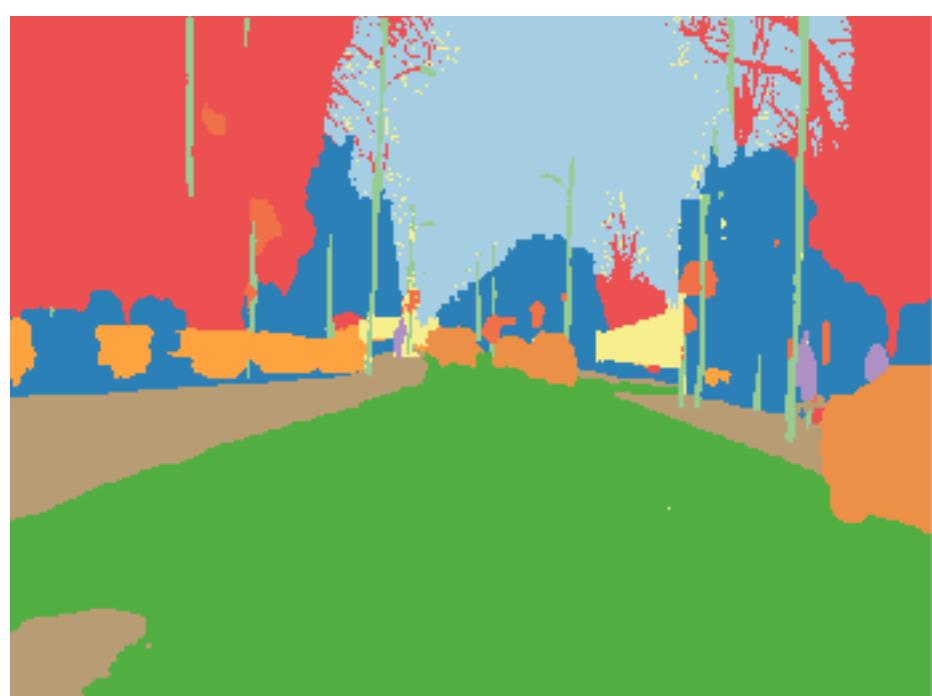
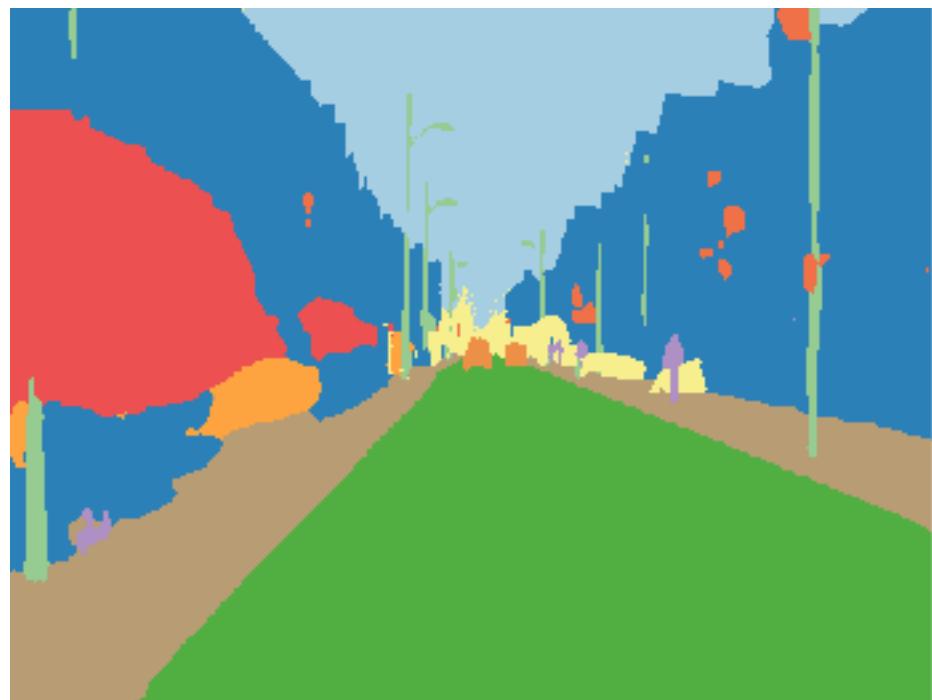
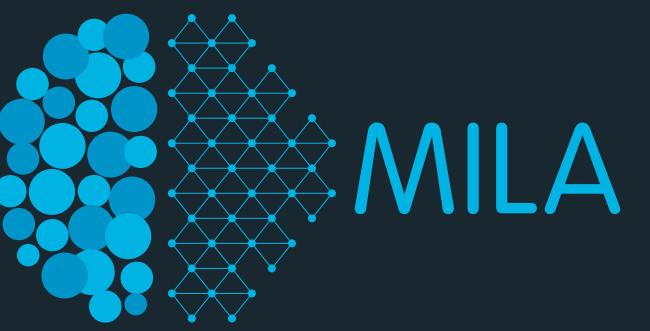
Results

Quantitative results - CamVid

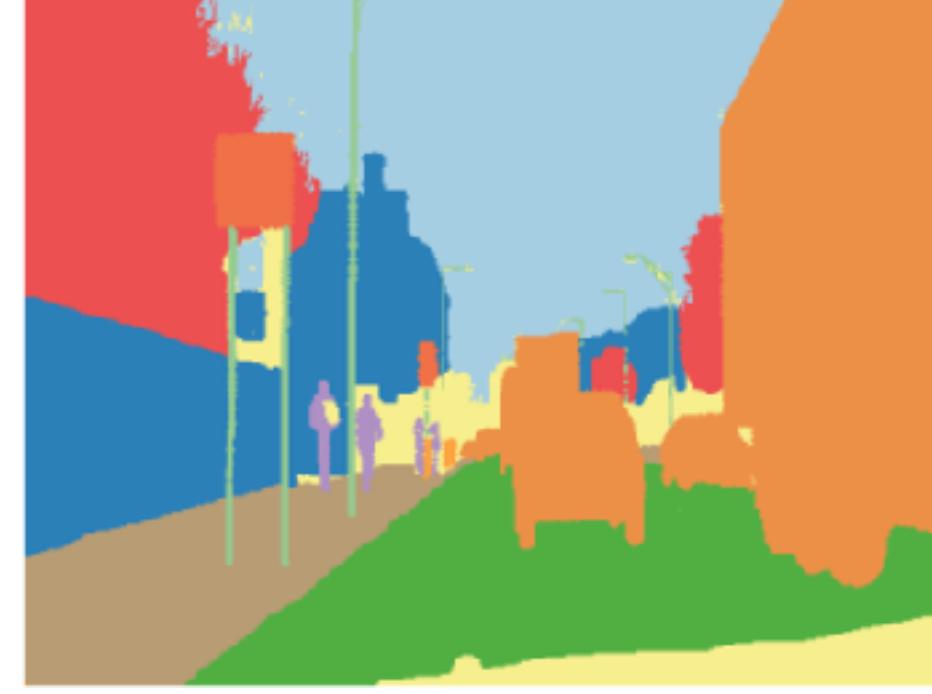
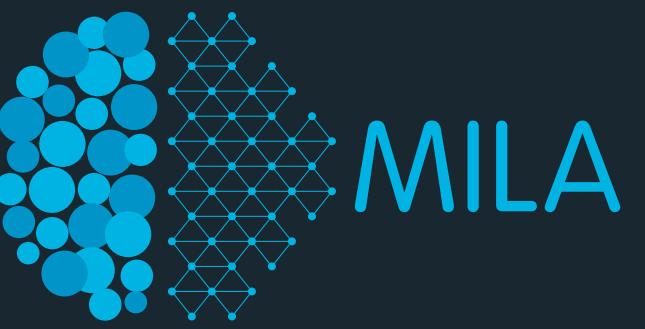


Model	Pretrained	# parameters (M)	Building	Tree	Sky	Car	Sign	Road	Pedestrian	Fence	Pole	Sidewalk	Cyclist	Mean IoU	Global accuracy
SegNet [1]	✓	29.5	68.7	52.0	87.0	58.5	13.4	86.2	25.3	17.9	16.0	60.5	24.8	46.4	62.5
Bayesian SegNet [14]	✓	29.5							n/a					63.1	86.9
DeconvNet [20]	✓	252							n/a					48.9	85.9
Visin et al. [35]	✓	32.3							n/a					58.8	88.7
FCN8 [19]	✗	134.5	77.8	71.0	88.7	76.1	32.7	91.2	41.7	24.4	19.9	72.7	31.0	57.0	88.0
DeepLab-LFOV [5]	✓	37.3	81.5	74.6	89.0	82.2	42.3	92.2	48.4	27.2	14.3	75.4	50.1	61.6	—
Dilation8 [36]	✓	140.8	82.6	76.2	89.0	84.0	46.9	92.2	56.3	35.8	23.4	75.3	55.5	65.3	79.0
Dilation8 + FSO [16]	✓	140.8	84.0	77.2	91.3	85.6	49.9	92.5	59.1	37.6	16.9	76.0	57.2	66.1	88.3
Classic Upsampling	✗	20	73.5	72.2	92.4	66.2	26.9	90.0	37.7	22.7	30.8	69.6	25.1	55.2	86.8
FC-DenseNet56 (k=12)	✗	1.5	77.6	72.0	92.4	73.2	31.8	92.8	37.9	26.2	32.6	79.9	31.1	58.9	88.9
FC-DenseNet67 (k=16)	✗	3.5	80.2	75.4	93.0	78.2	40.9	94.7	58.4	30.7	38.4	81.9	52.1	65.8	90.8
FC-DenseNet103 (k=16)	✗	9.4	83.0	77.3	93.0	77.3	43.9	94.5	59.6	37.1	37.8	82.2	50.5	66.9	91.5

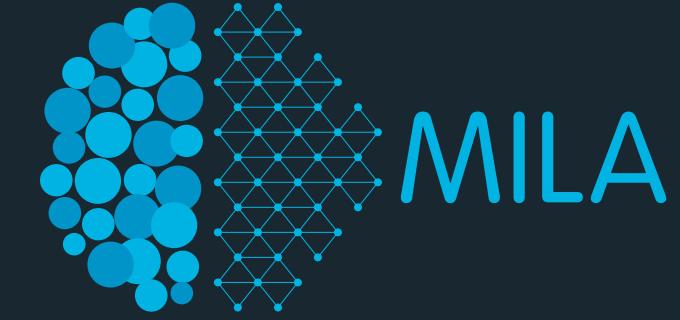
Qualitative results - good



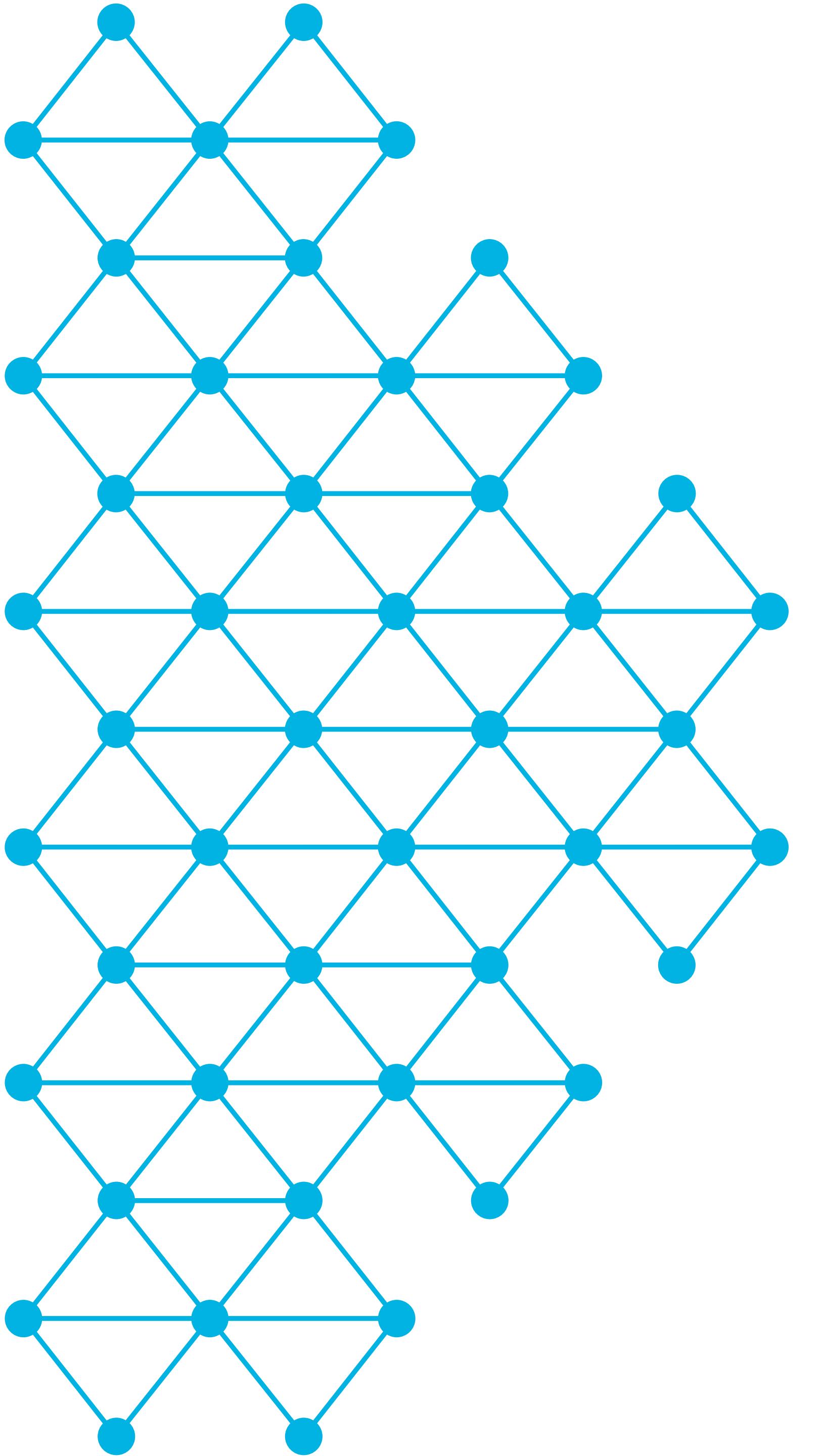
Qualitative results - fail



Quantitative results - Gatech



Model	Acc.
<i>2D models (no time)</i>	
2D-V2V-from scratch [33]	55.7
FC-DenseNet103	79.4
<i>3D models (incorporate time)</i>	
3D-V2V-from scratch [33]	66.7
3D-V2V-pretrained [33]	76.0



Wrap up

Wrap up

- We presented an extension of DenseNets for semantic segmentation.
- FC-DenseNets encourage deep supervision and feature reuse, while mitigating the feature explosion.
- FC-DenseNets as an ensemble of variable depth networks.

Fully Convolutional DenseNets for Semantic Segmentation

Simon Jégou, Michal Drozdzal, David Vazquez, Adriana Romero, Yoshua Bengio

Thank you!