

CADL2020

International Workshop on Computational Aspects of Deep Learning



Overview

# Introducing Region Pooling Learning

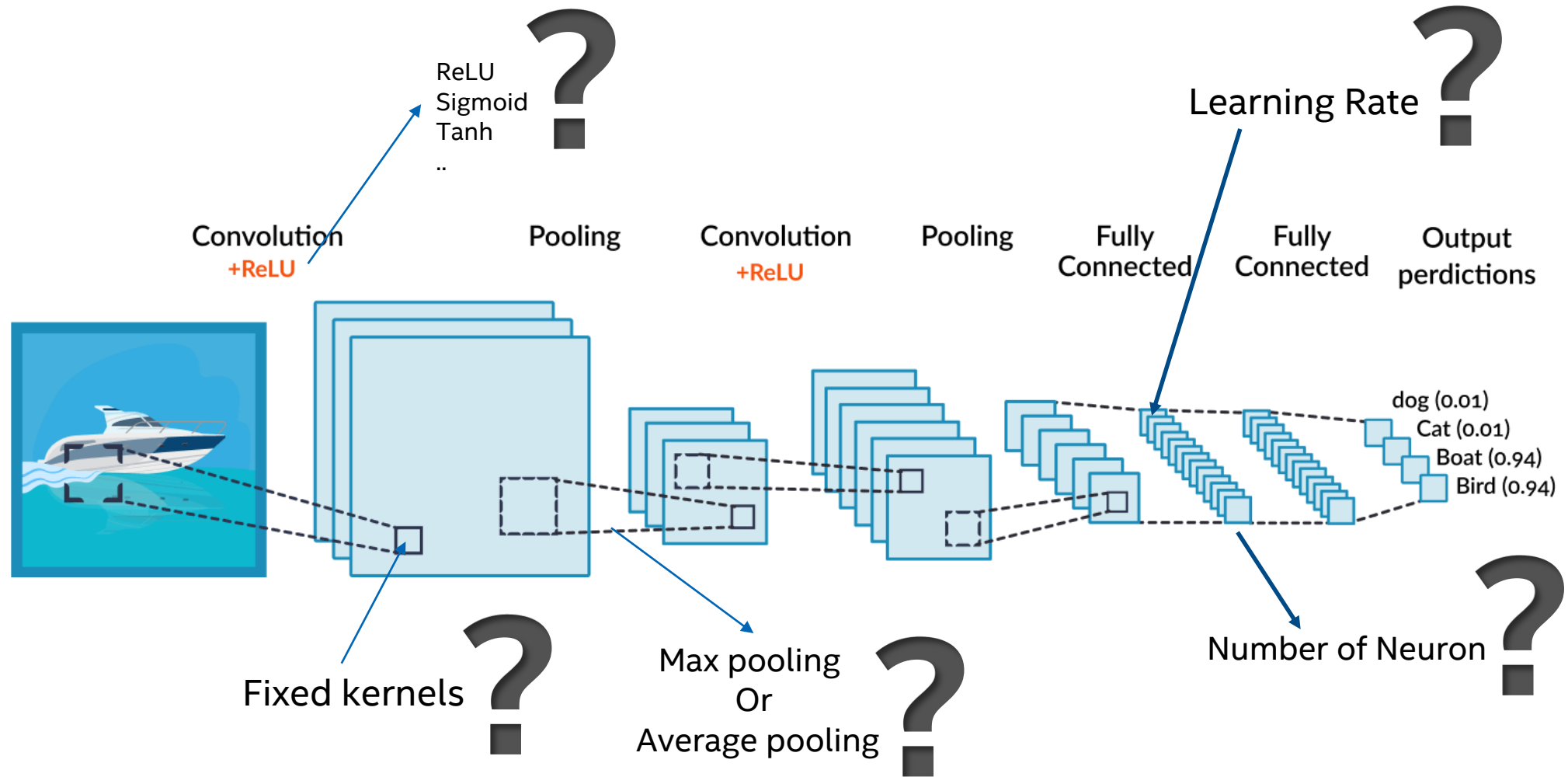
Presenter: Julio Zamora



Authors:

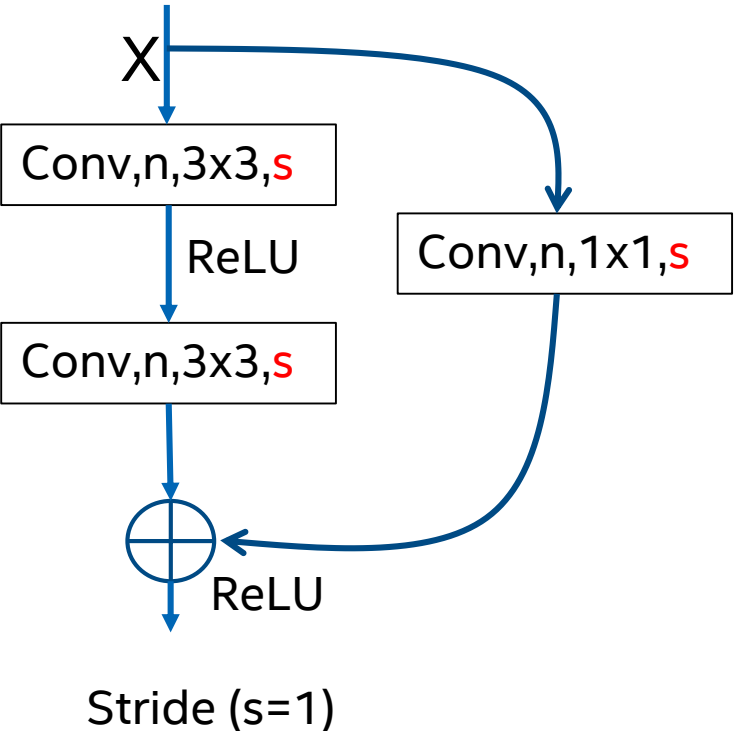
Adan Cruz, Julio Zamora , Omesh Tickoo

# Hyper parameters of a NN model

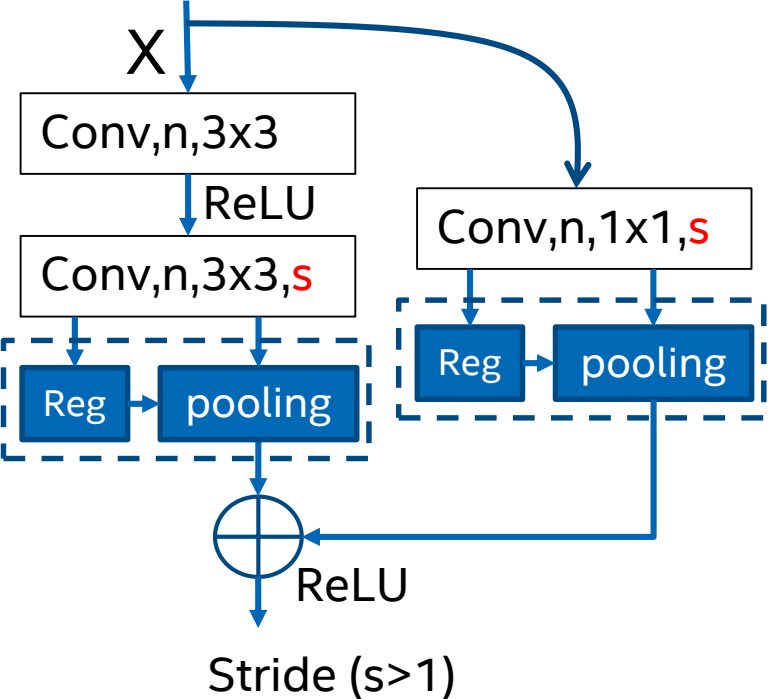


# Region Pooling Learning

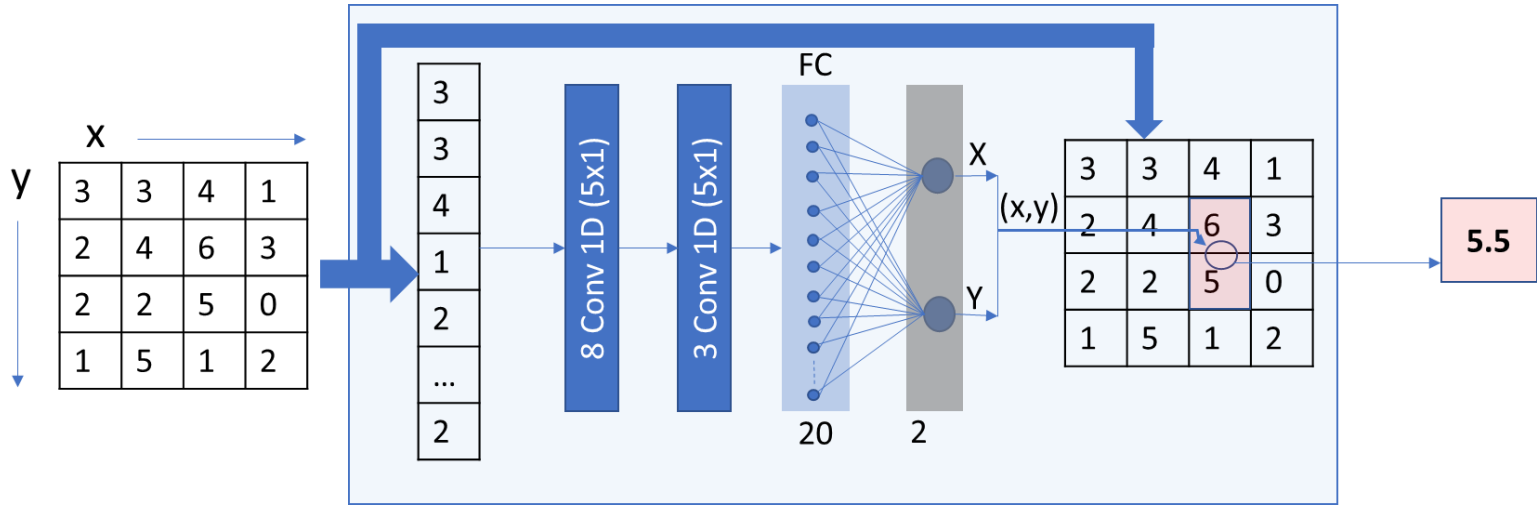
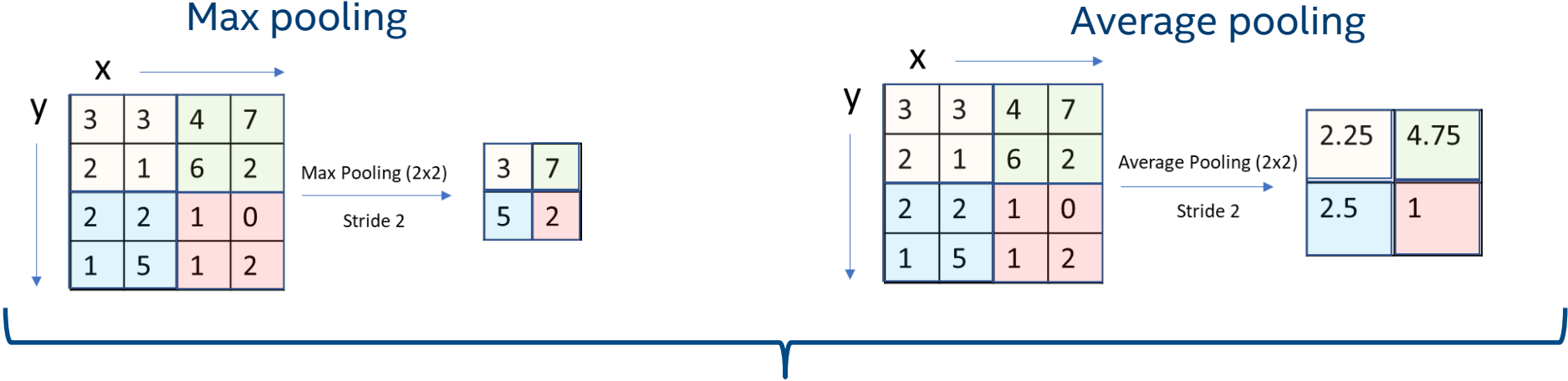
## Resnet Block



## ResNet Block & Region Pooling

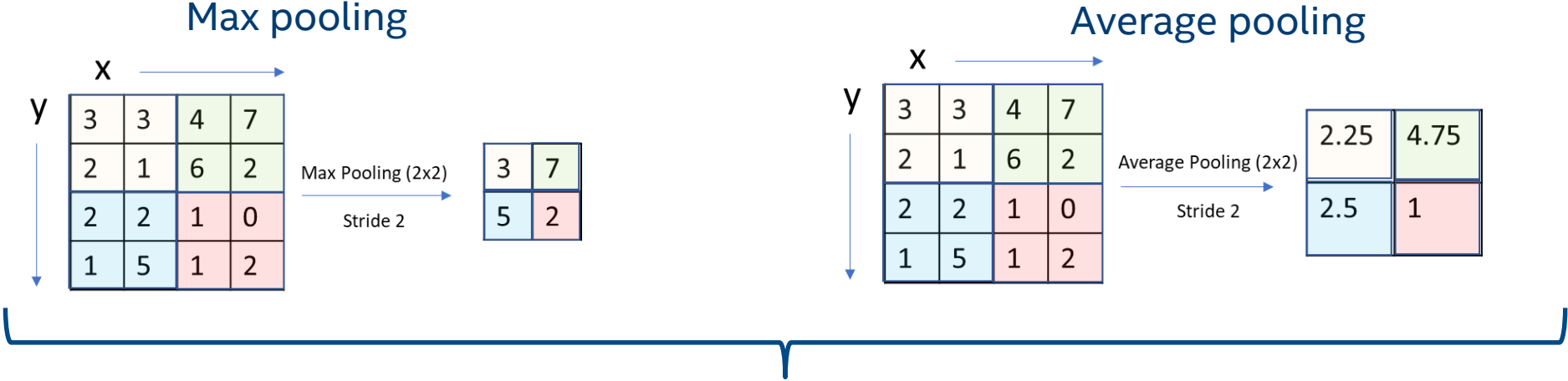


# Region Pooling Learning

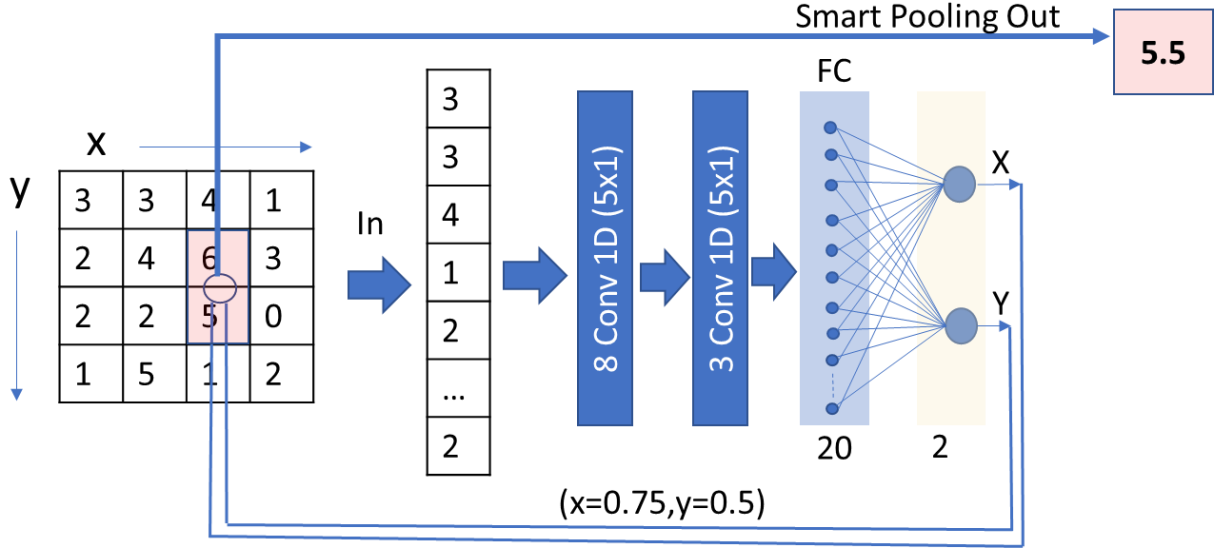


# Region Pooling Learning

Classical:

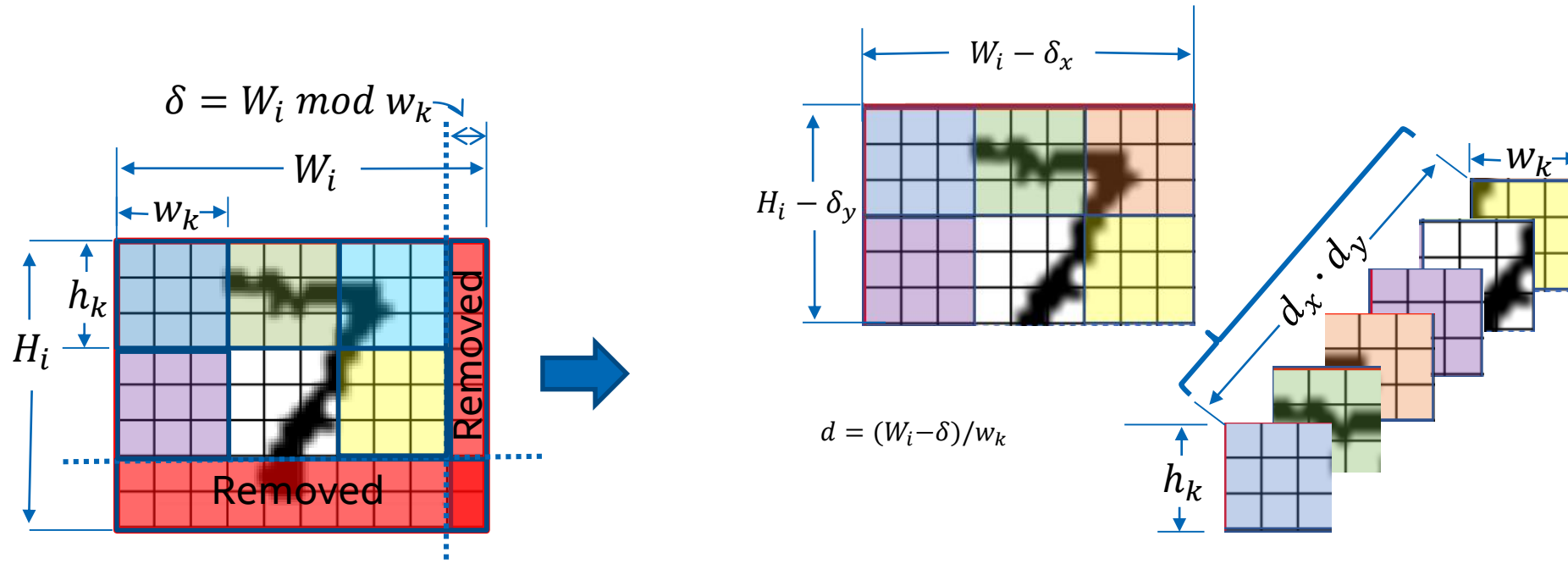


Adaptive:



# Parallelization

Pooling is a process that can be easily parallelized



# Smart Pooling Results

## Smart Pooling Results

The results using this technology in comparison with academic state-of-the-art neural nets shows competitive performance for image recognition datasets, with much smaller parameter size.

- 163X memory compression with better accuracy
- ~3% improvement changing a single layer
- Changing a single layer produces an accuracy improvement for ImageNet

Table 2. Results for different ResNet topologies reported Vs ResNet18\* with added smart pooling

Neural Network	Depth	#Parameters	Accuracy%
ResNet18	18	11M	93.02
ResNet50	50	25.6M	93.62
ResNet100	100	44.5M	93.75
<b>ResNet18*</b>	<b>18</b>	<b>0.27M</b>	<b>94.08</b>

Table 4. Accuracy results for ResNet18\* with smart pooling Vs reported ResNet topologies for ImageNet

Neural Network	#Param	Top1%	Top5%
ResNet18	11M	69.76	89.08
ResNet50	25.6M	76.15	92.87
ResNet101	44.5M	77.37	93.56
<b>ResNet18*</b>	<b>11M</b>	<b>71.33</b>	<b>90.07</b>

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