

WB-XIC, Lab5:

# Konwolucyjne sieci neuronowe w praktyce: ResNet & DenseNet

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# Cel:

1. Stride
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3. Batch Normalization
4. ResNet
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# Stride: Conv2d parameter

<https://poloclub.github.io/cnn-explainer>

## Understanding Hyperparameters

Input Size:



Padding:



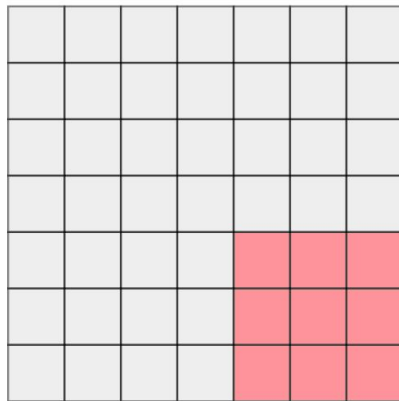
Kernel Size:



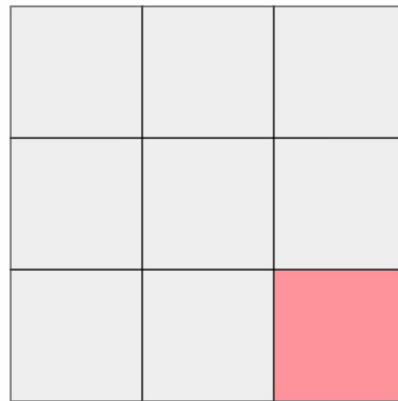
Stride:



Input (7, 7)  
After-padding (7, 7)



Output (3, 3)



*Hover over the matrices to change kernel position.*

Google Colab (powtórka)

# Batch Normalization

<https://pytorch.org/docs/stable/generated/torch.nn.BatchNorm2d.html>

<https://arxiv.org/abs/1502.03167>

Docs > torch.nn > BatchNorm2d



## BATCHNORM2D

```
CLASS torch.nn.BatchNorm2d(num_features, eps=1e-05, momentum=0.1, affine=True,  
    track_running_stats=True, device=None, dtype=None) [SOURCE]
```

Applies Batch Normalization over a 4D input (a mini-batch of 2D inputs with additional channel dimension) as described in the paper [Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift](#).

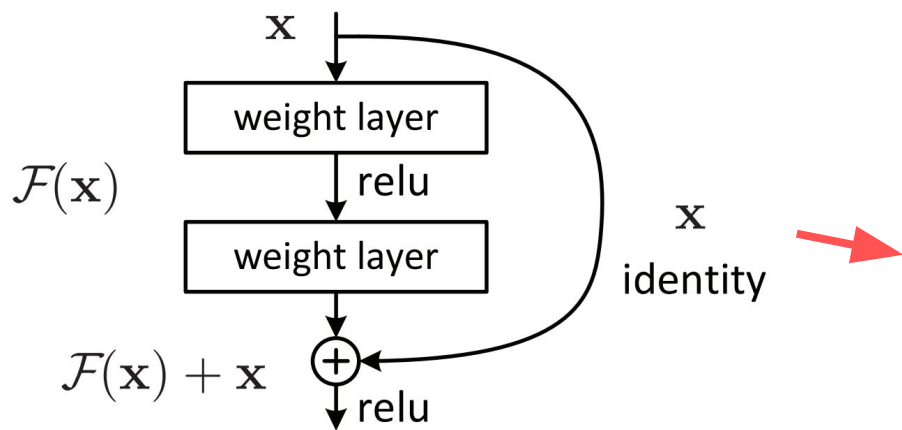
$$y = \frac{x - \mathbb{E}[x]}{\sqrt{\text{Var}[x] + \epsilon}} * \gamma + \beta$$

The mean and standard-deviation are calculated per-dimension over the mini-batches and  $\gamma$  and  $\beta$  are learnable parameter vectors of size C (where C is the input size). By default, the elements of  $\gamma$  are set to 1 and the elements of  $\beta$  are set to 0. The standard-deviation is calculated via the biased estimator, equivalent to `torch.var(input, unbiased=False)`.

# ResNet

<https://arxiv.org/abs/1512.03385>

<https://github.com/kuangliu/pytorch-cifar/blob/master/models/resnet.py>



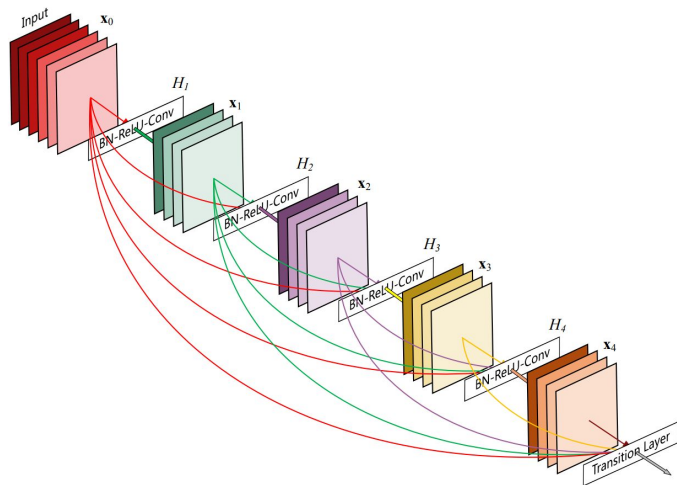
```
def forward(self, x):  
    out = F.relu(self.bn1(self.conv1(x)))  
    out = self.bn2(self.conv2(out))  
    out += self.shortcut(x)  
    out = F.relu(out)  
    return out
```

Figure 2. Residual learning: a building block.

# DenseNet

<https://arxiv.org/abs/1608.06993>

<https://github.com/kuangliu/pytorch-cifar/blob/master/models/densenet.py>



```
def forward(self, x):  
    out = self.conv1(F.relu(self.bn1(x)))  
    out = self.conv2(F.relu(self.bn2(out)))  
    out = torch.cat([out,x], 1)  
    return out
```

**Figure 1:** A 5-layer dense block with a growth rate of  $k = 4$ . Each layer takes all preceding feature-maps as input.

Pytania, wnioski, zadanie domowe