

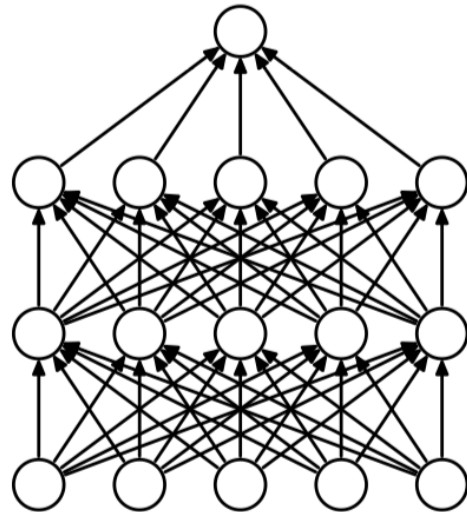
IN3063/INM702 - Programming and Mathematics for Artificial Intelligence

10.A - CNNs

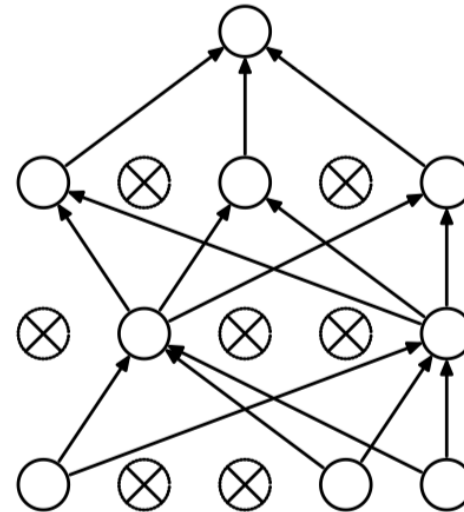
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Train / eval mode

- A Neural Network is by default in train mode
- When evaluating, the behavior of some components differ
- For example: Dropout



(a) Standard Neural Net



(b) After applying dropout.

Neural Networks should subclass nn.Module

- All weights of the module are registered as parameters
- Automatically sets them as differentiable
- Allows to create the computational graph

CLASS torch.nn.Module

Base class for all neural network modules.

Your models should also subclass this class.

Modules can also contain other Modules, allowing to nest them in a tree structure. You can assign the sub regular attributes:

```
import torch.nn as nn
import torch.nn.functional as F

class Model(nn.Module):
    def __init__(self):
        super(Model, self).__init__()
        self.conv1 = nn.Conv2d(1, 20, 5)
        self.conv2 = nn.Conv2d(20, 20, 5)

    def forward(self, x):
        x = F.relu(self.conv1(x))
        return F.relu(self.conv2(x))
```

Loss and Loss with logits

- Some loss functions already implement the activation of the last layer

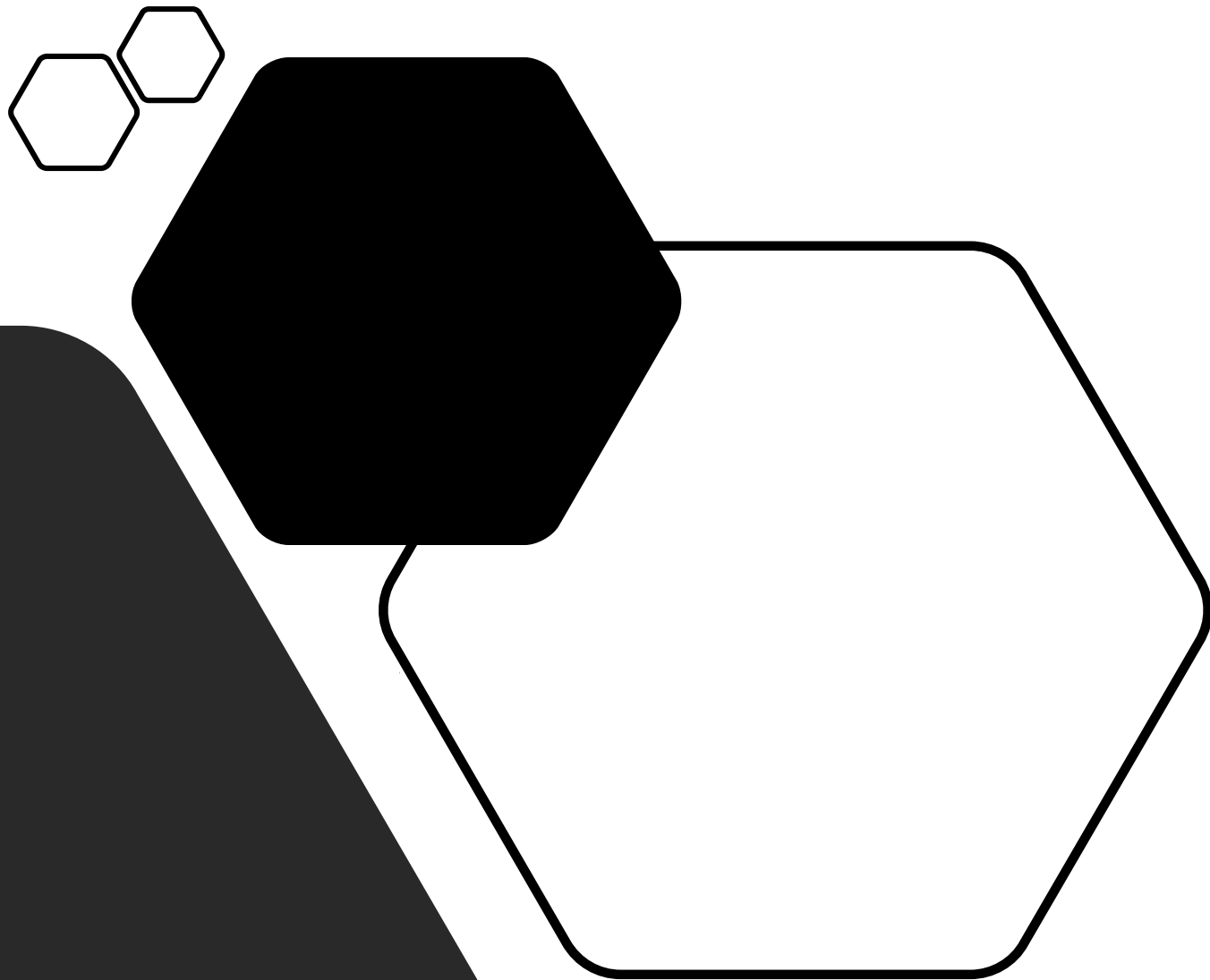
`nn.CrossEntropyLoss`

This criterion combines `nn.LogSoftmax()` and `nn.NLLLoss()` in one single class.

`nn.NLLLoss`

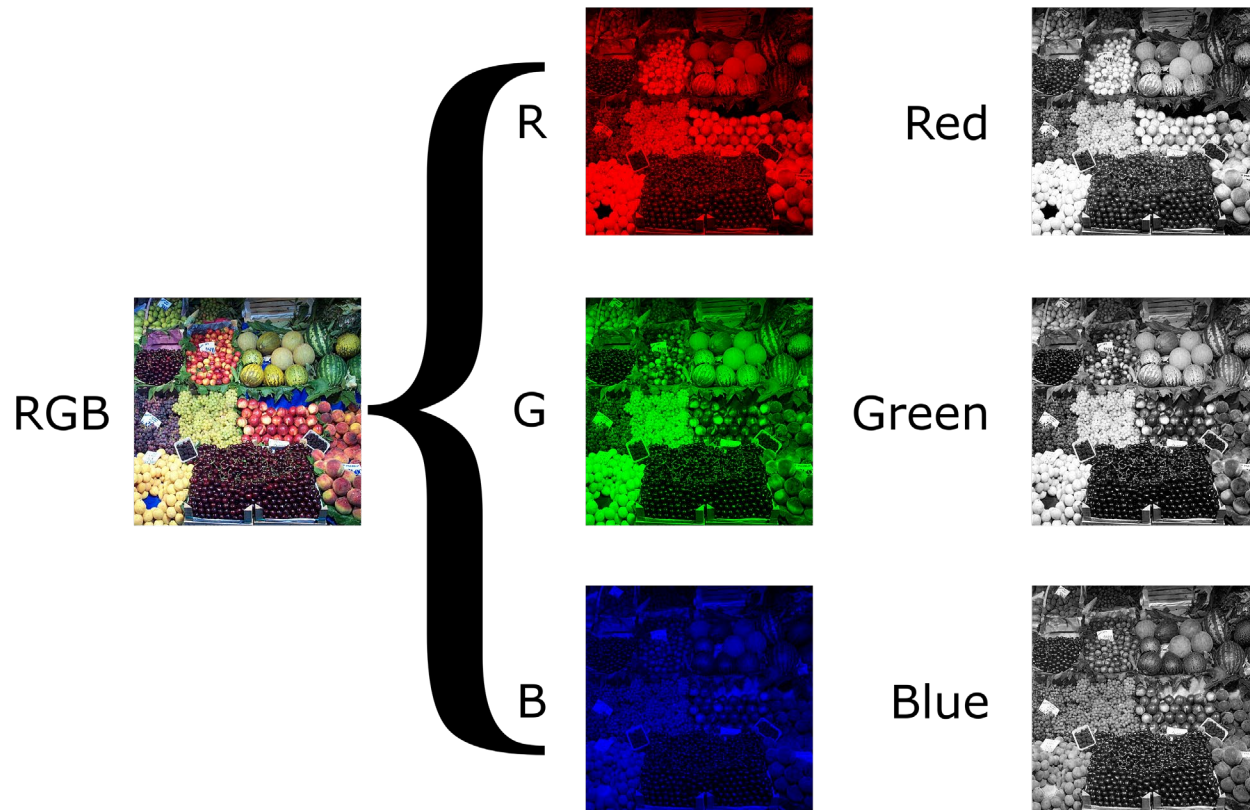
The negative log likelihood loss.

Learning on Images



Properties of Images

- 2-Dimensional structure (3 if you count the channel)



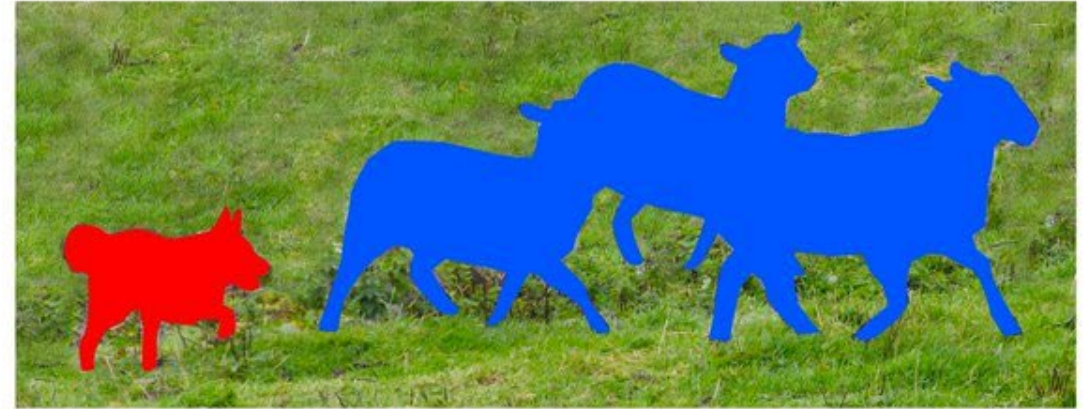
Local features



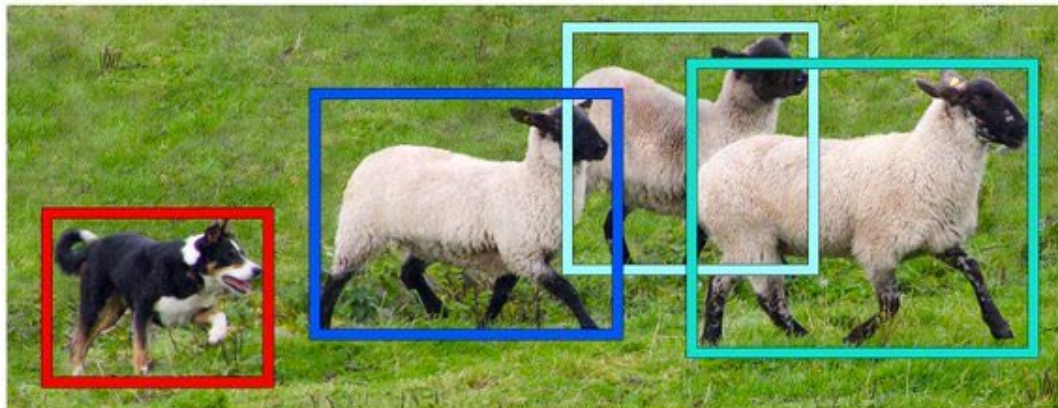
Different Deep Learning Problems



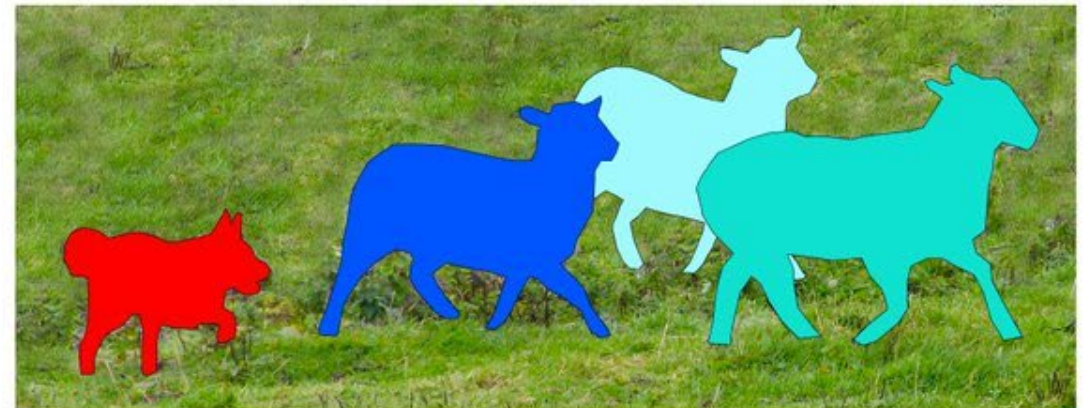
Image Recognition



Semantic Segmentation



Object Detection



Instance Segmentation

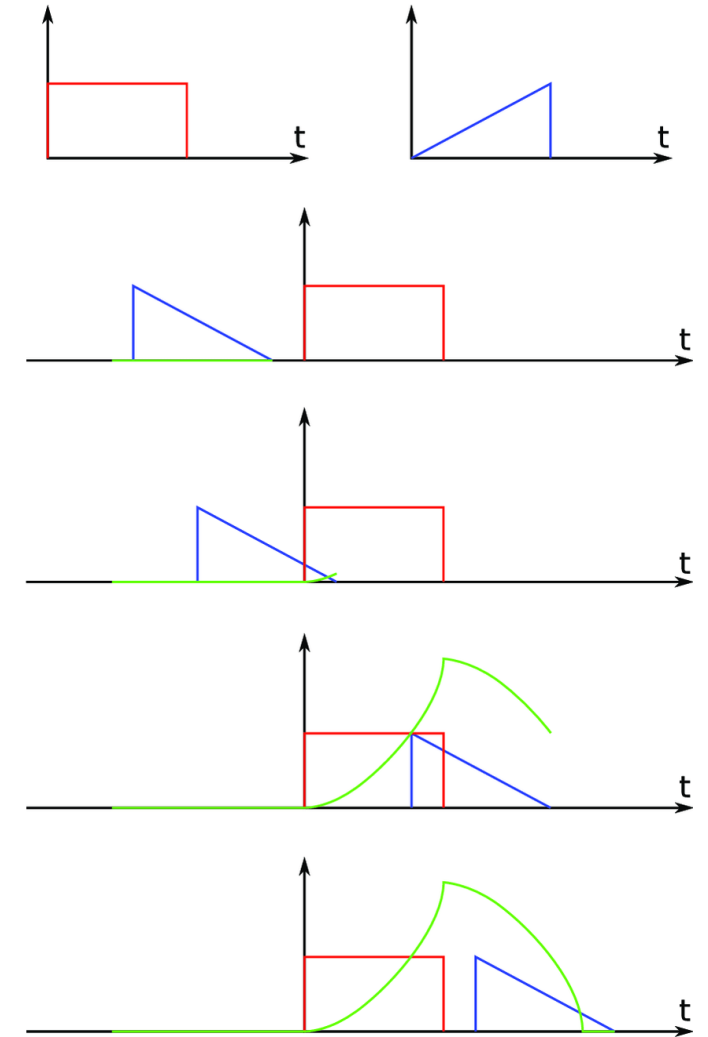
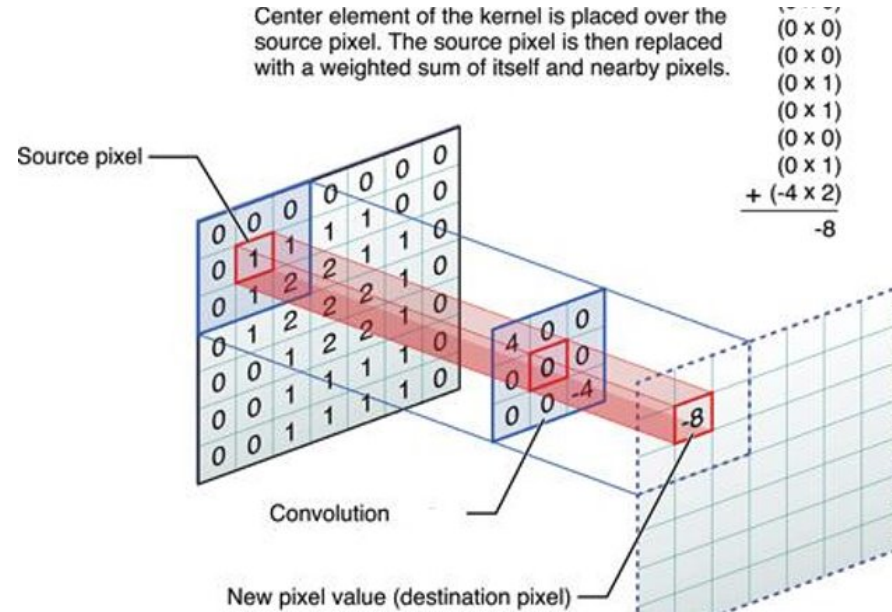
How to solve this broad range of problem?

- Classifying pixel by pixel?
- Classifying the whole image?
- Local features can occur at different locations of the image



Convolution

- Signal processing method
- In the case of images: 3D kernels

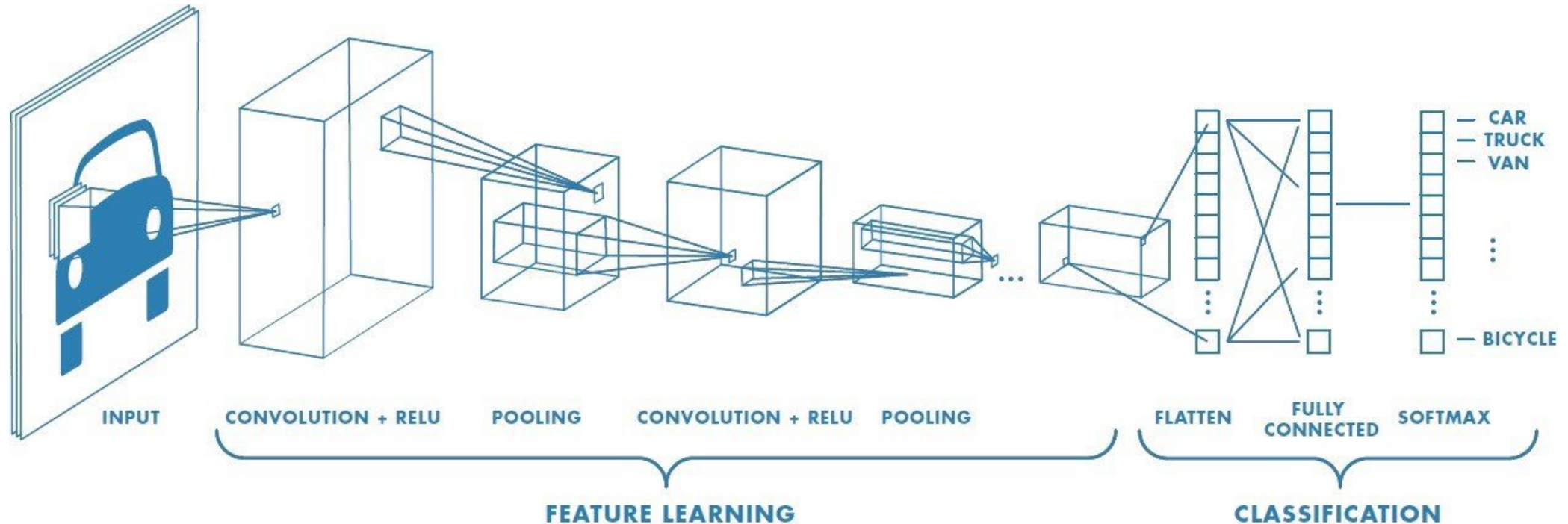


Convolutional Neural Networks



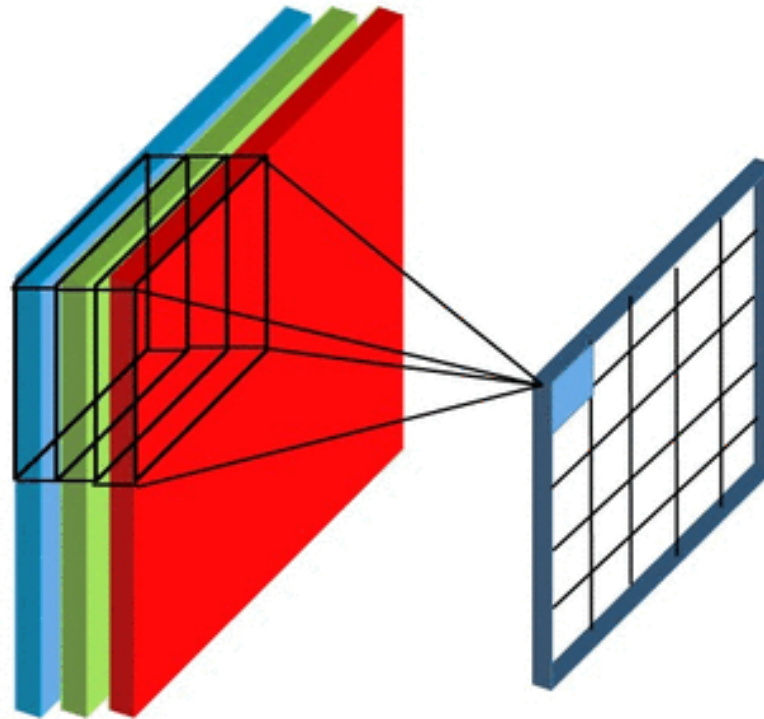
Convolutional Neural Networks

- The kernel is a list of parameter.
- Local feature learning



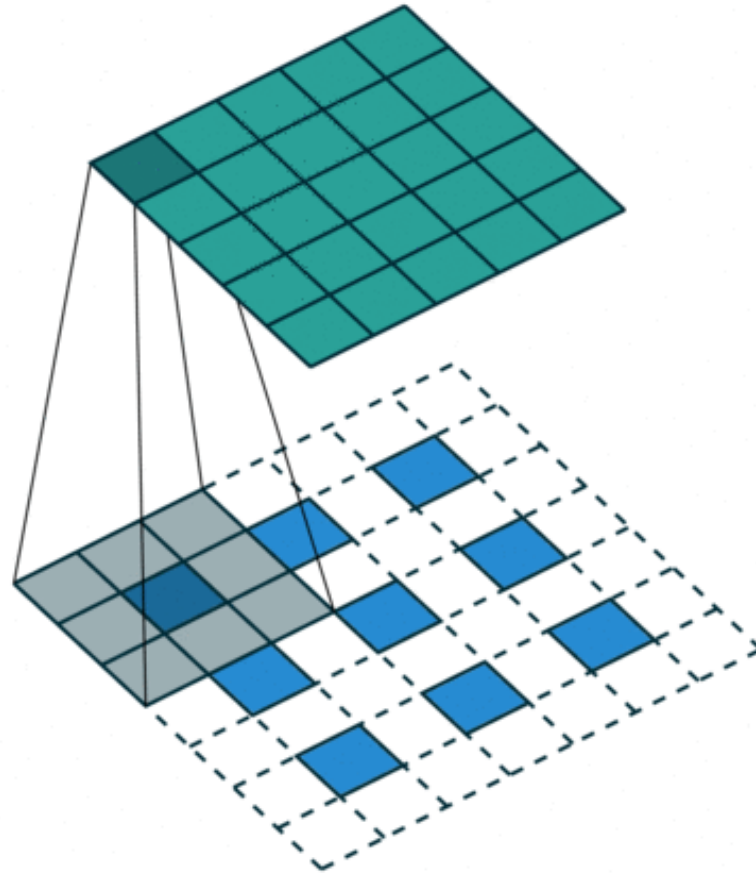
Parameters of a convolutional layer

- Kernel size



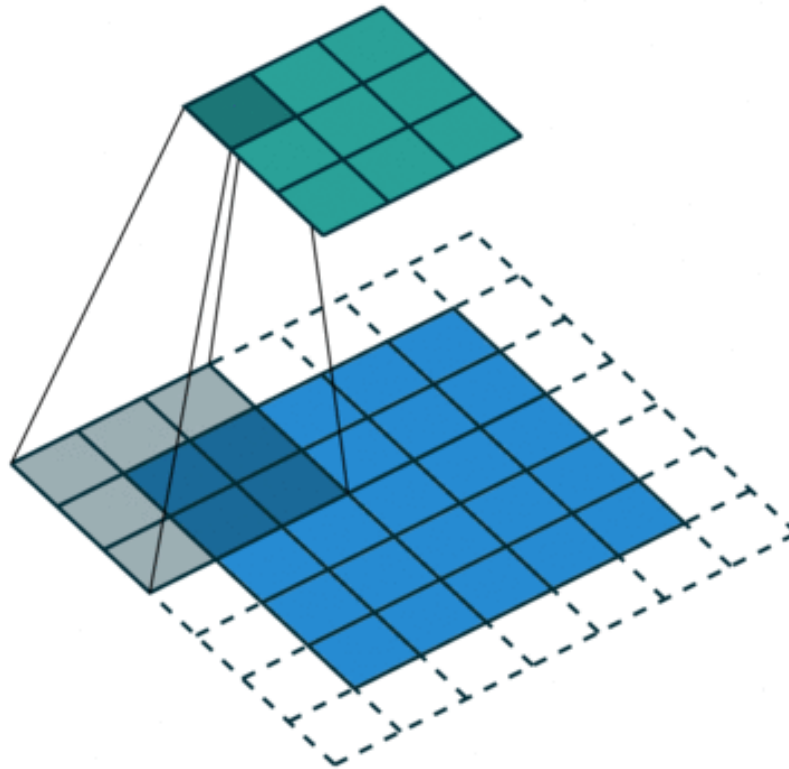
Parameters of a convolutional layer

- Stride



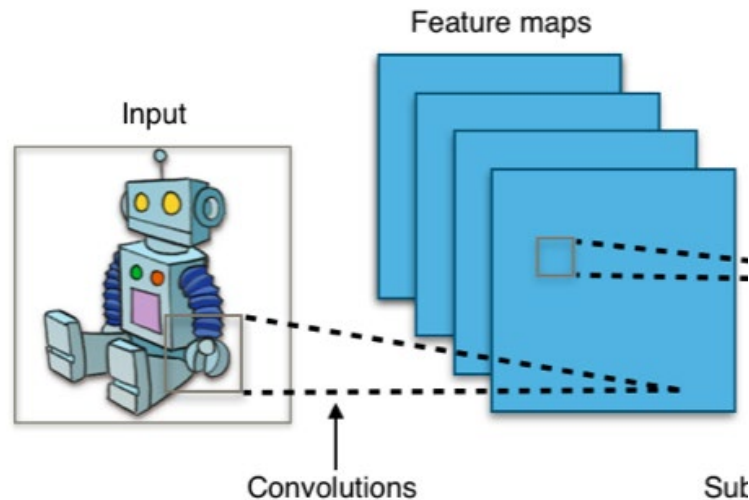
Parameters of a convolutional layer

- Padding:
 - Zero
 - Constant
 - Reflection
 - Replication

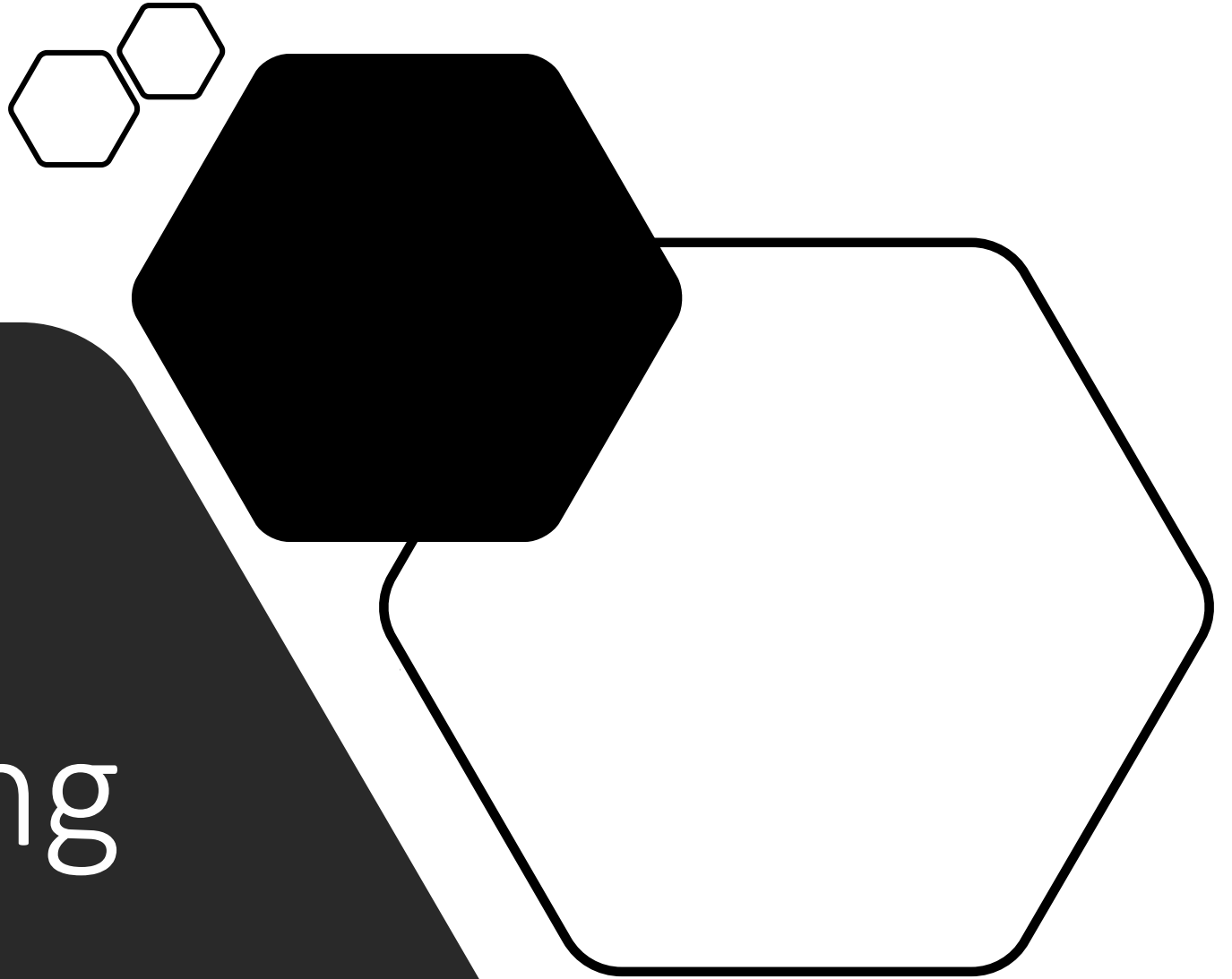


Parameters of a convolutional layer

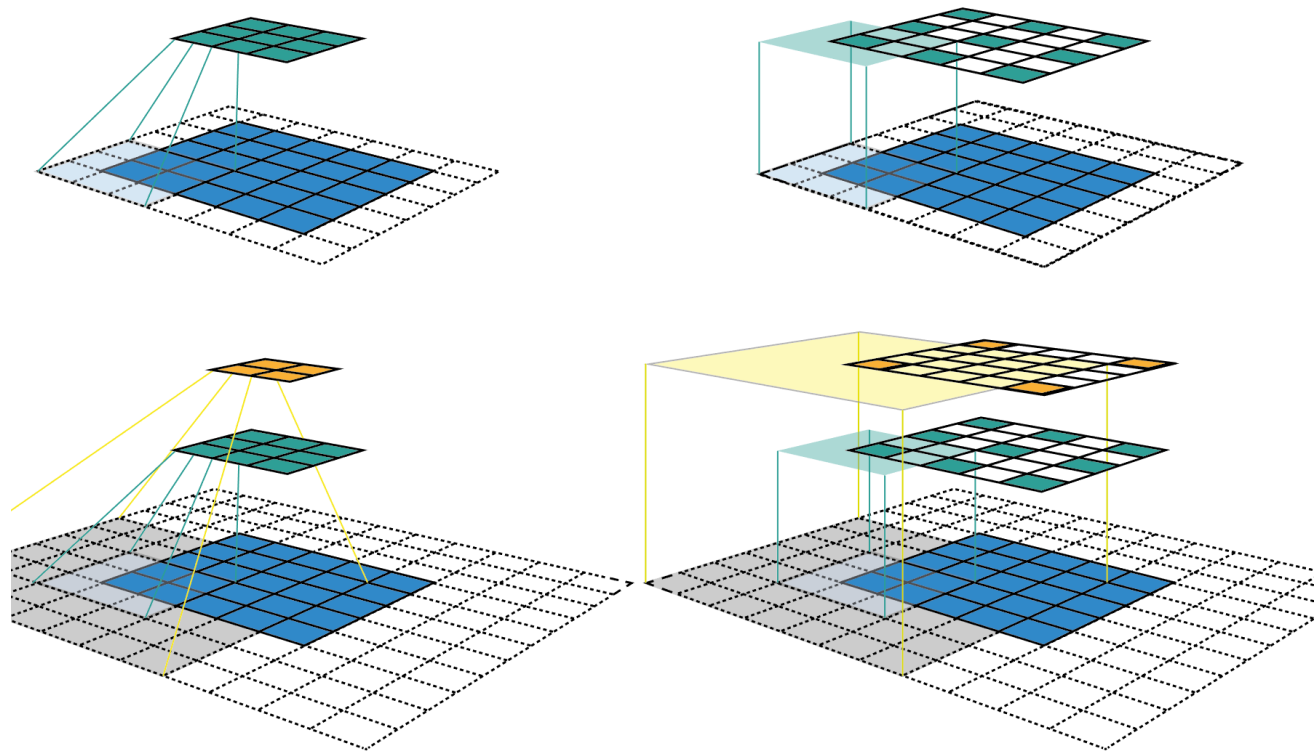
- Number of features
- Stride / kernel size and padding consistent across feature maps of a layer.



Sub-sampling



We want to compress information



- Receptive Fields become larger and larger
- Size of the feature maps become smaller and smaller
- Features become more and more abstract (folding and non-linearities)

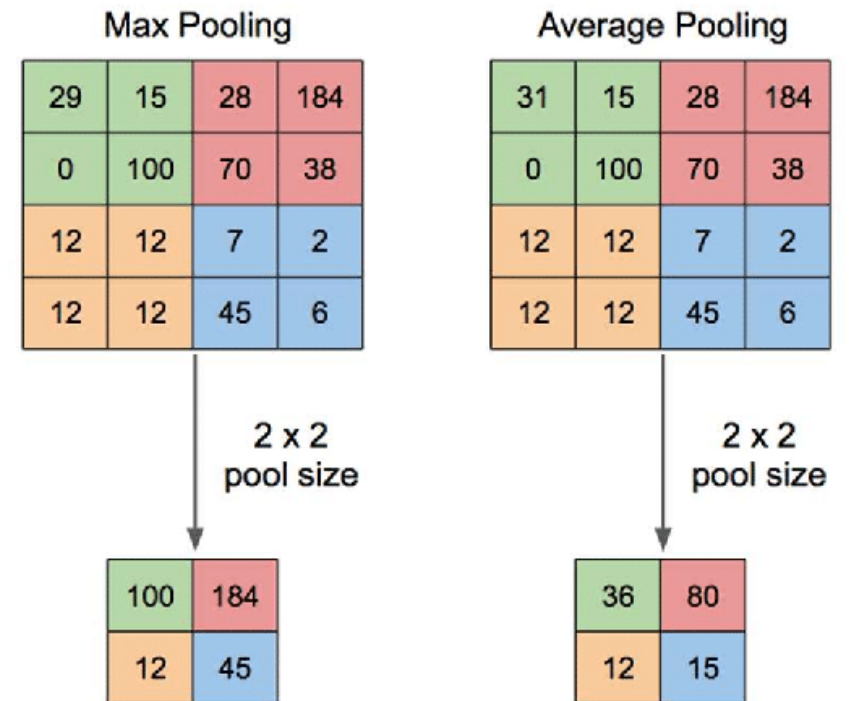
Effect of stride

- A stride greater than one allows for progressive compression of the spatial information

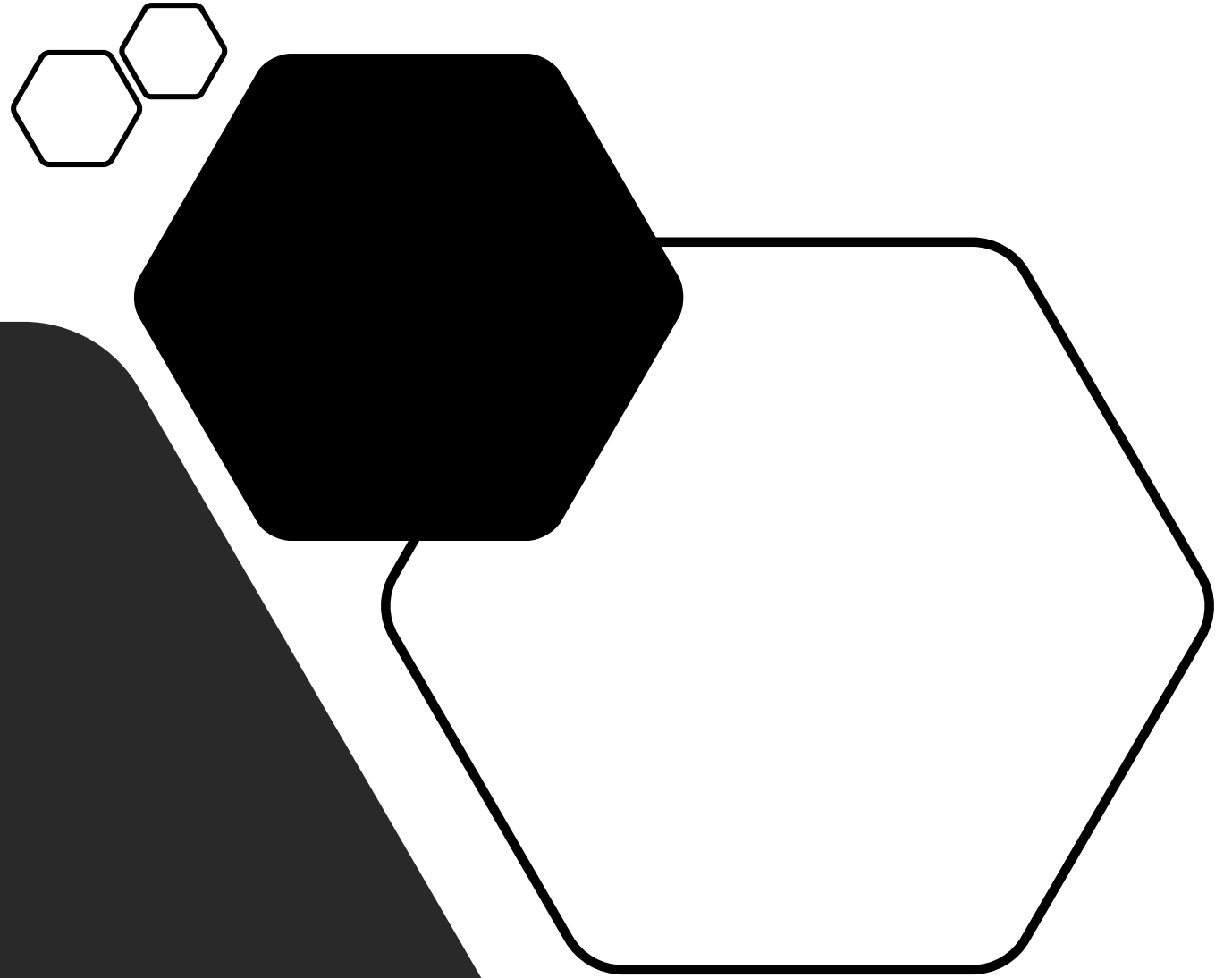


Pooling

- Mathematical Operation to group/keep only certain features:
 - Max Pooling
 - Average Pooling
- Improve translational invariance

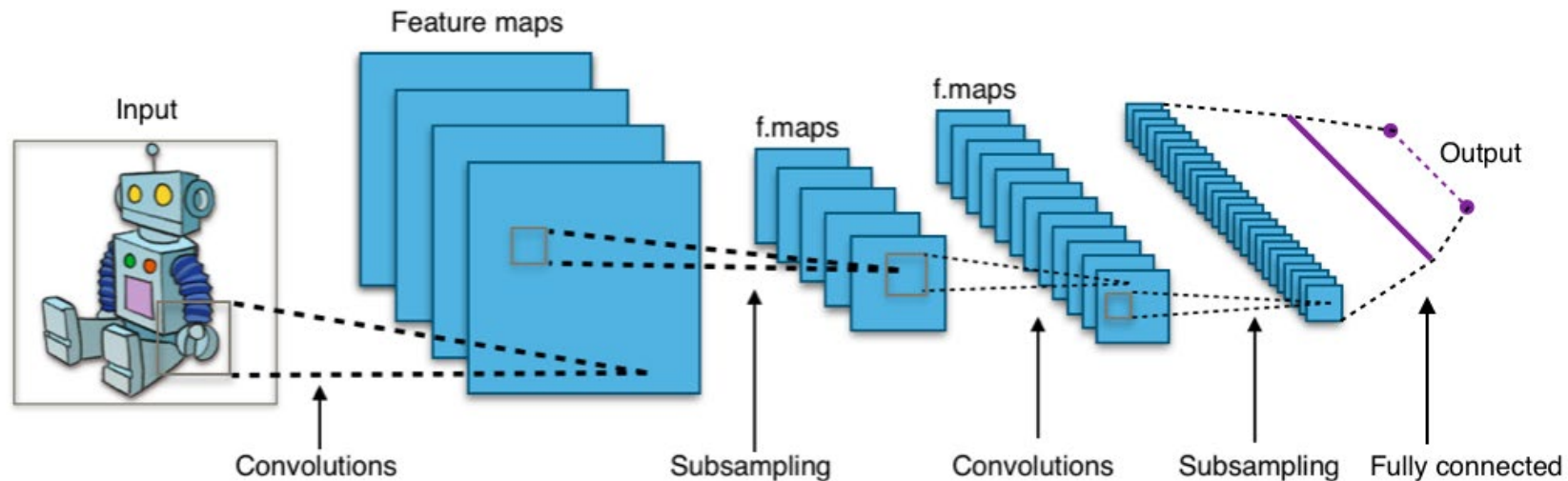


CNN overview

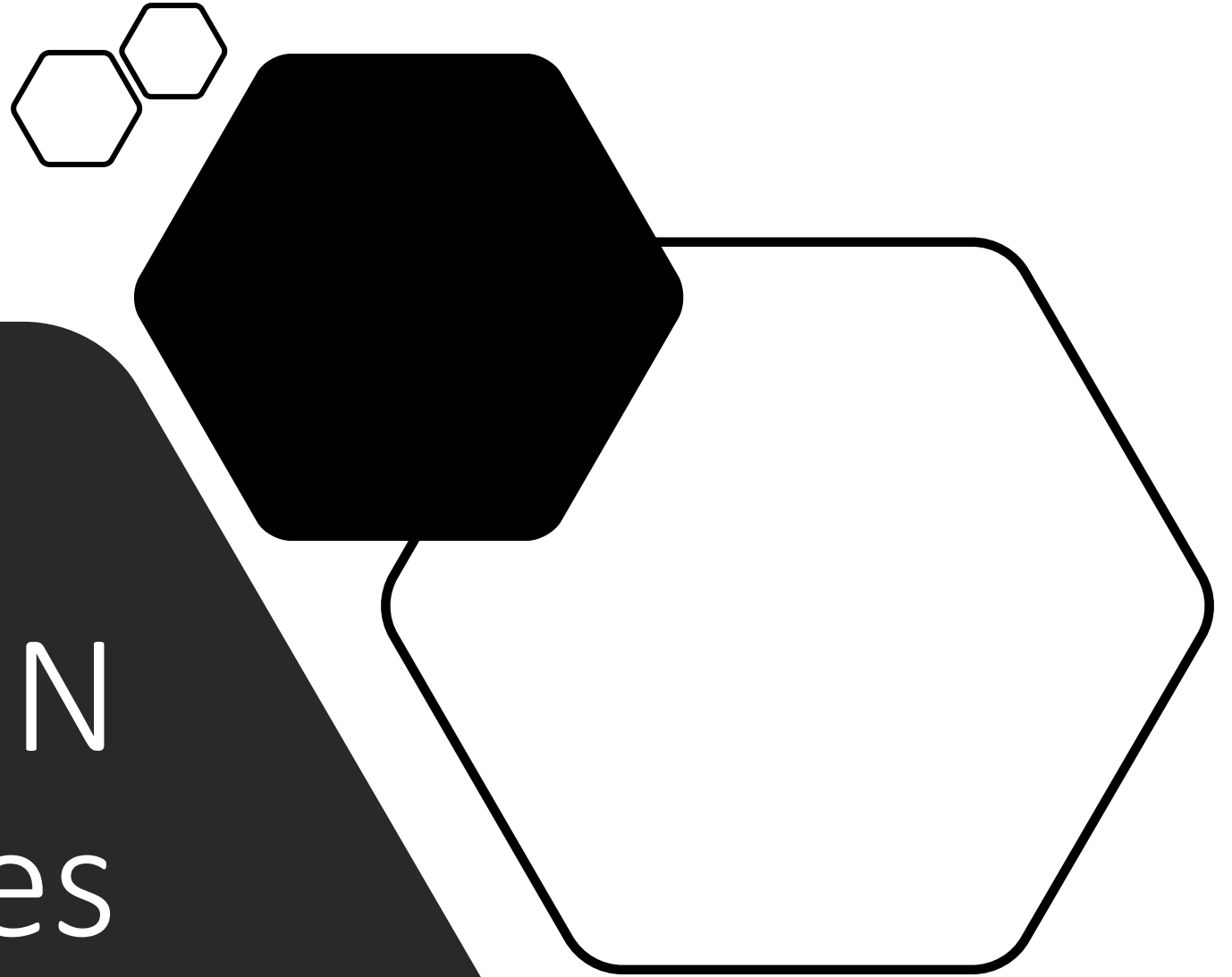


Classical CNN architecture

- For classification: last convolutional layer is concatenated and flattened
- Fully connected, then softmax

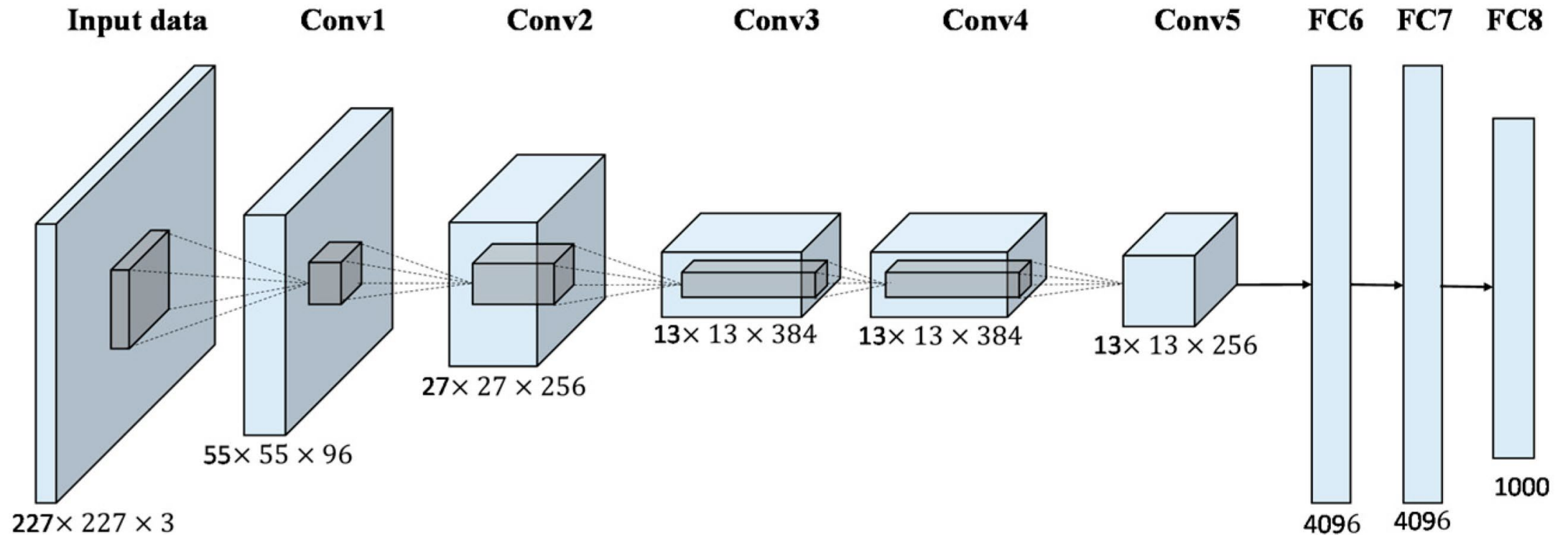


Classical CNN Architectures



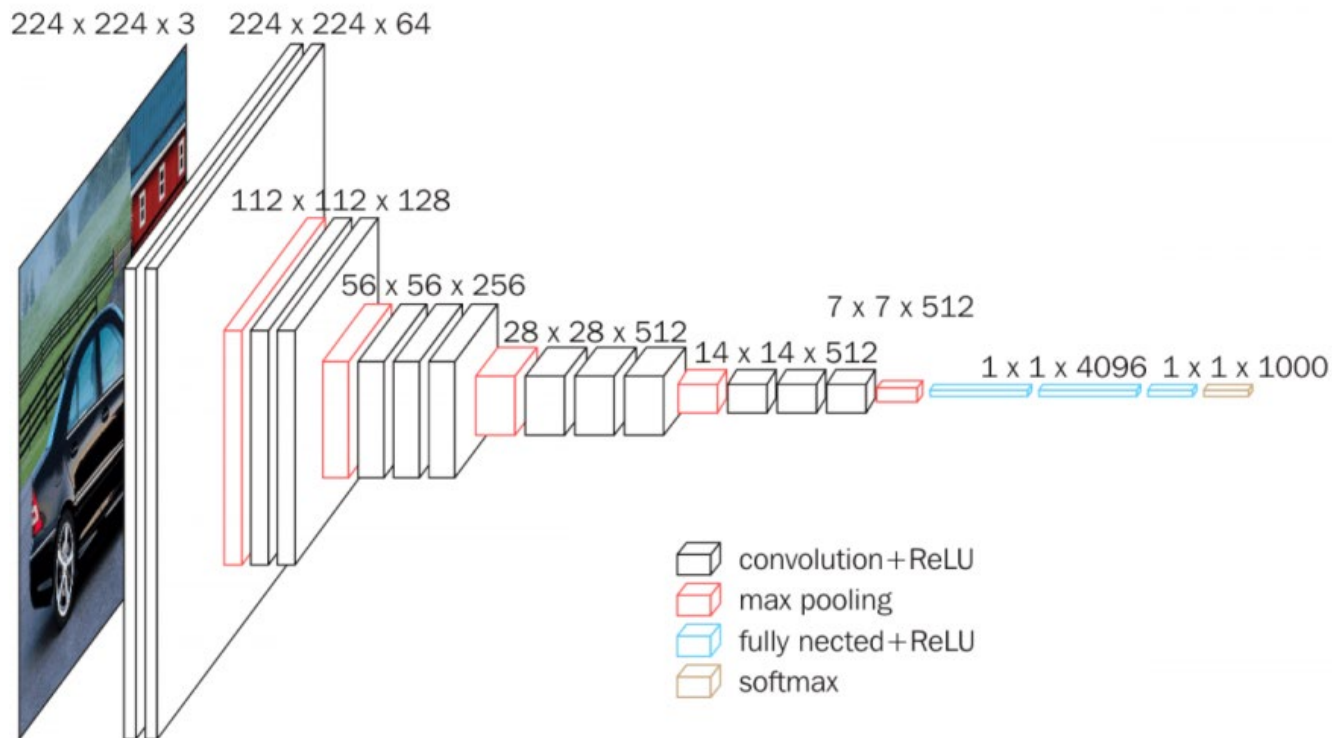
AlexNet (2012)

- Trained on ImageNet (1000 classes)



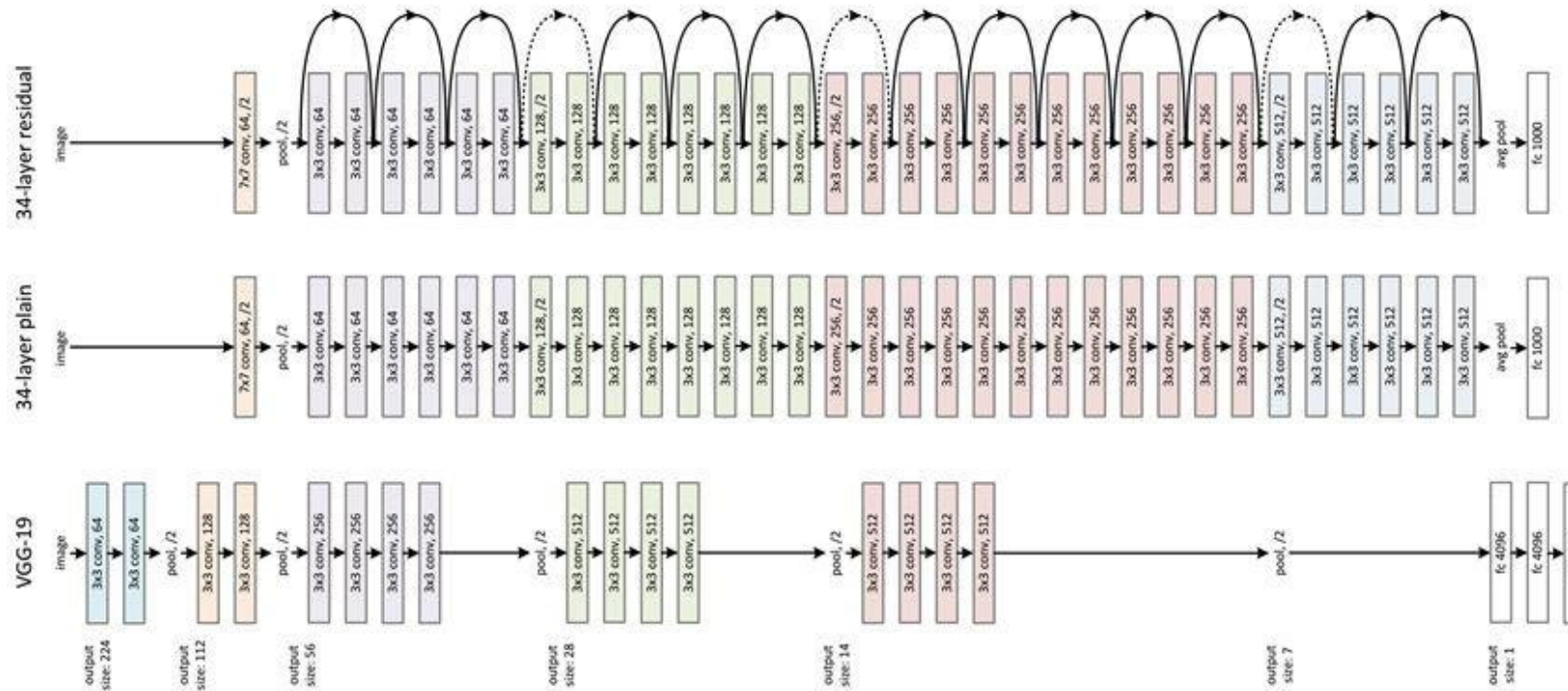
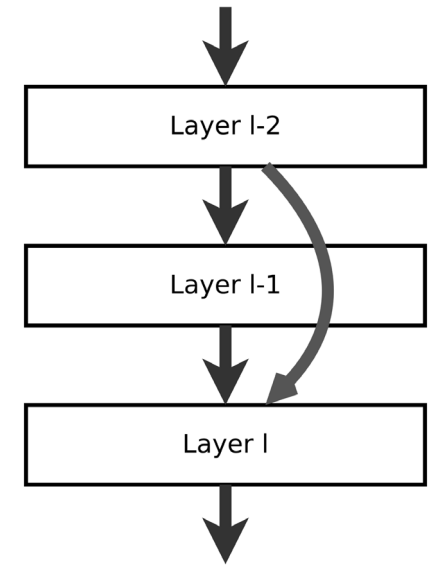
VGG (2014)

- Trained on ImageNet (1000 classes)

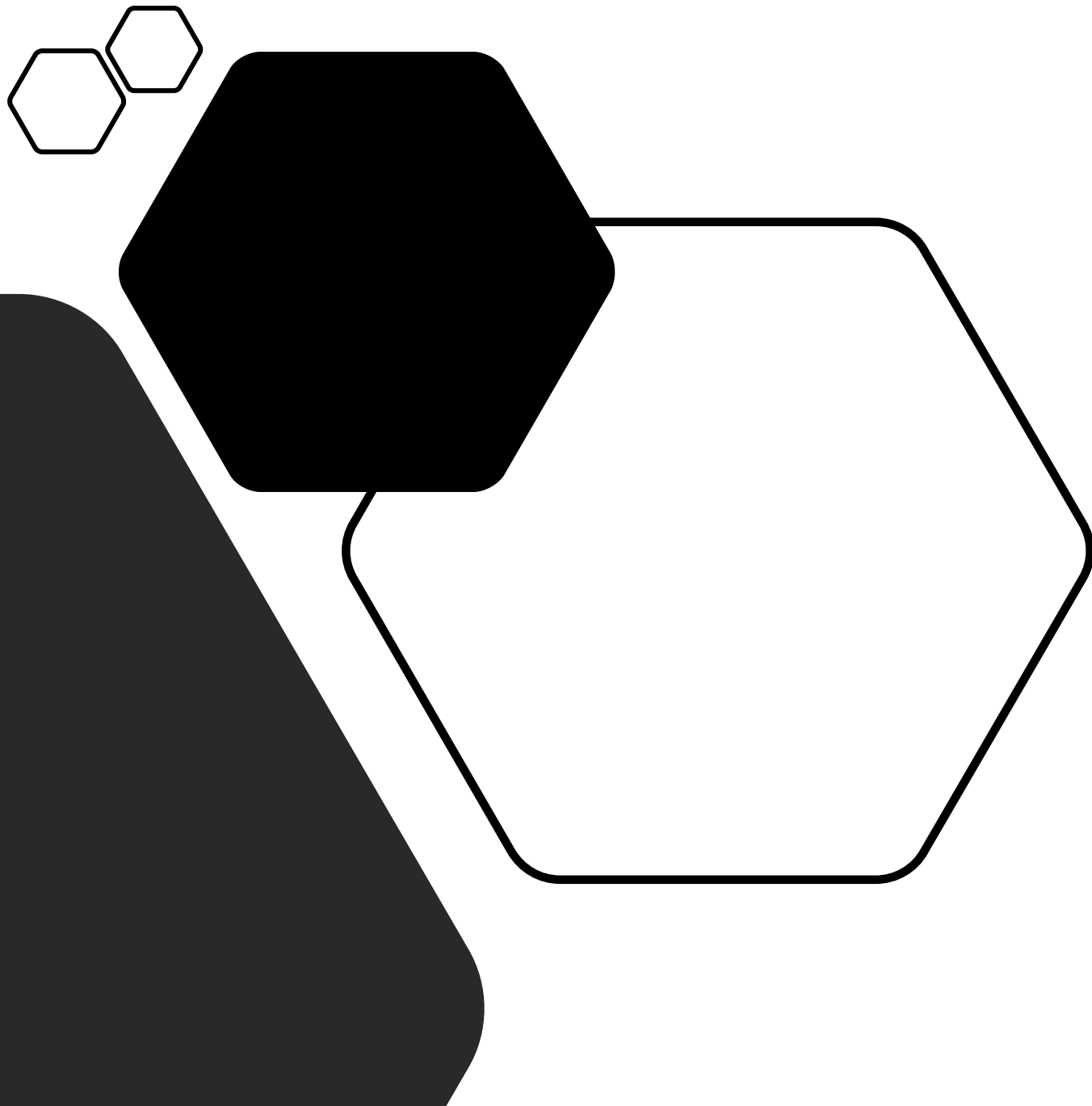


ResNet (2015)

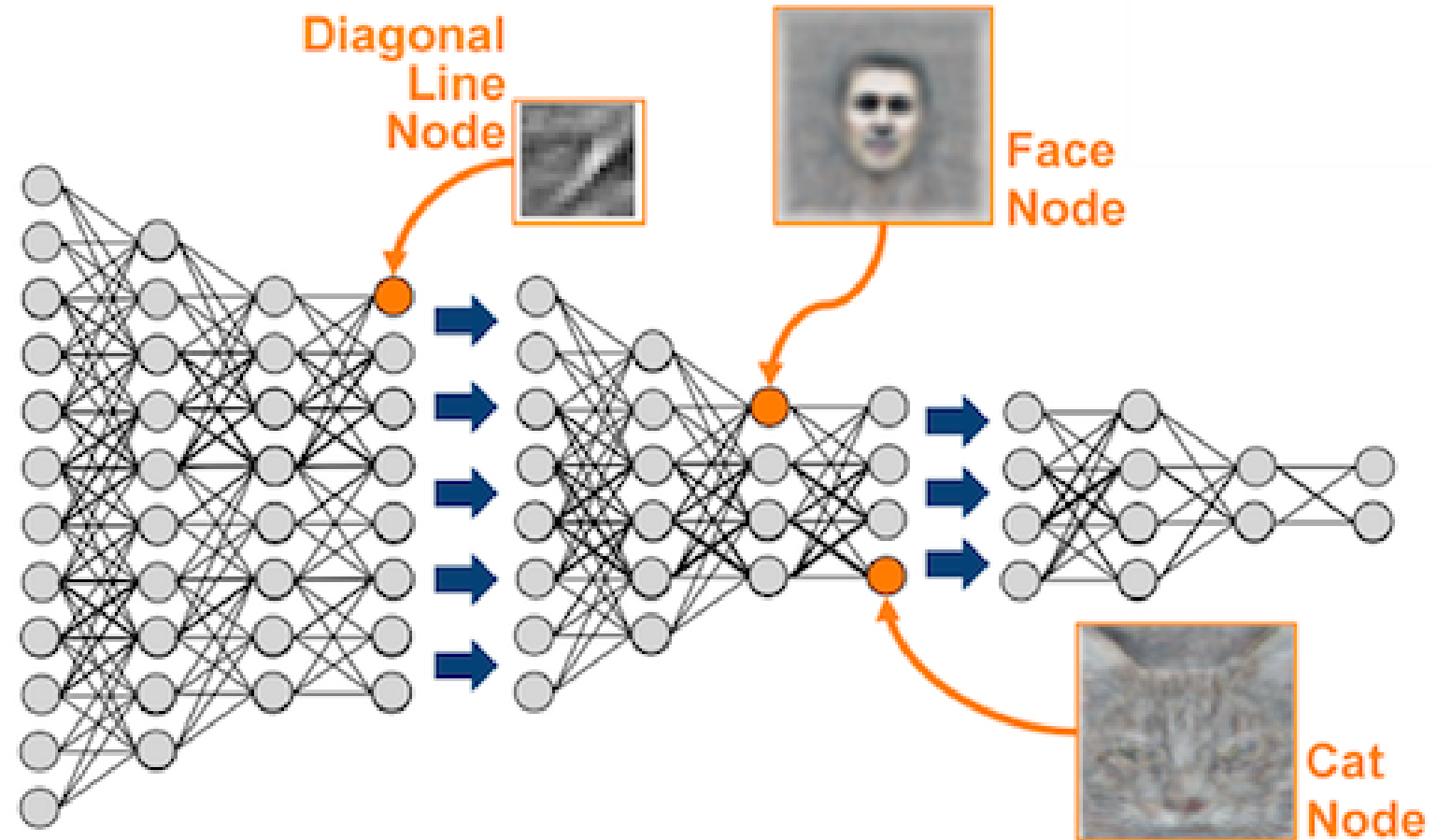
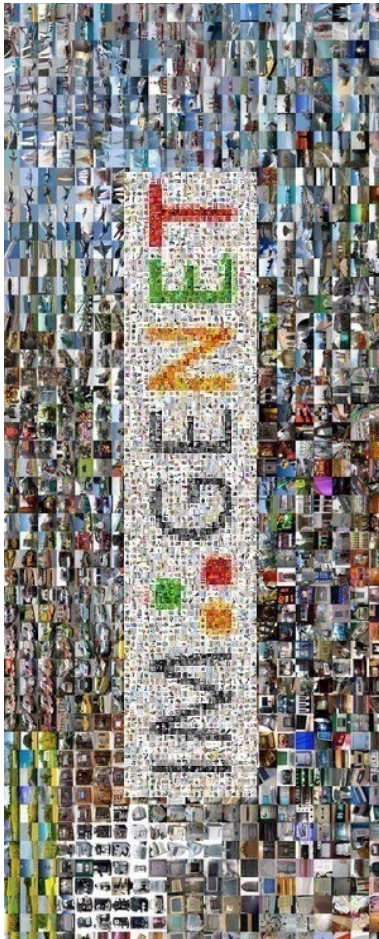
- Skip connections to improve gradient propagation
- Prevents the gradient dying out



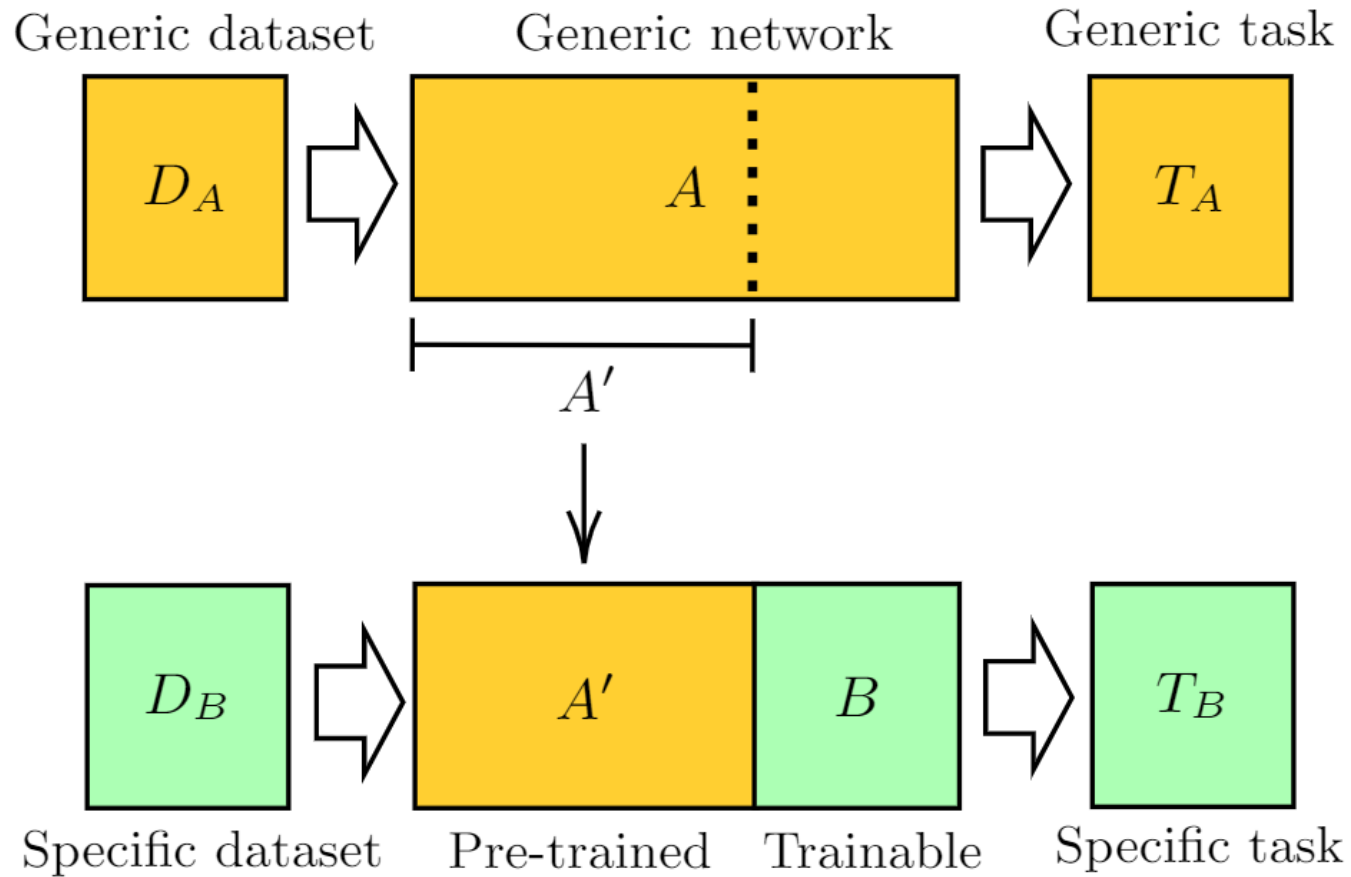
Transfer Learning



Learning generic features

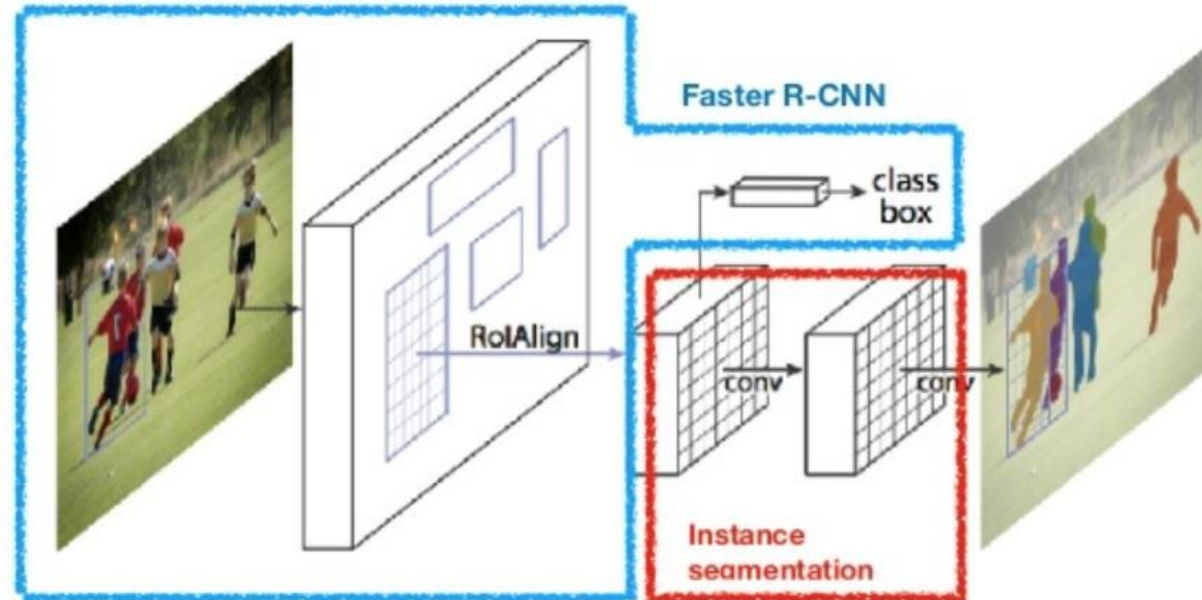


Re-using pre-trained networks

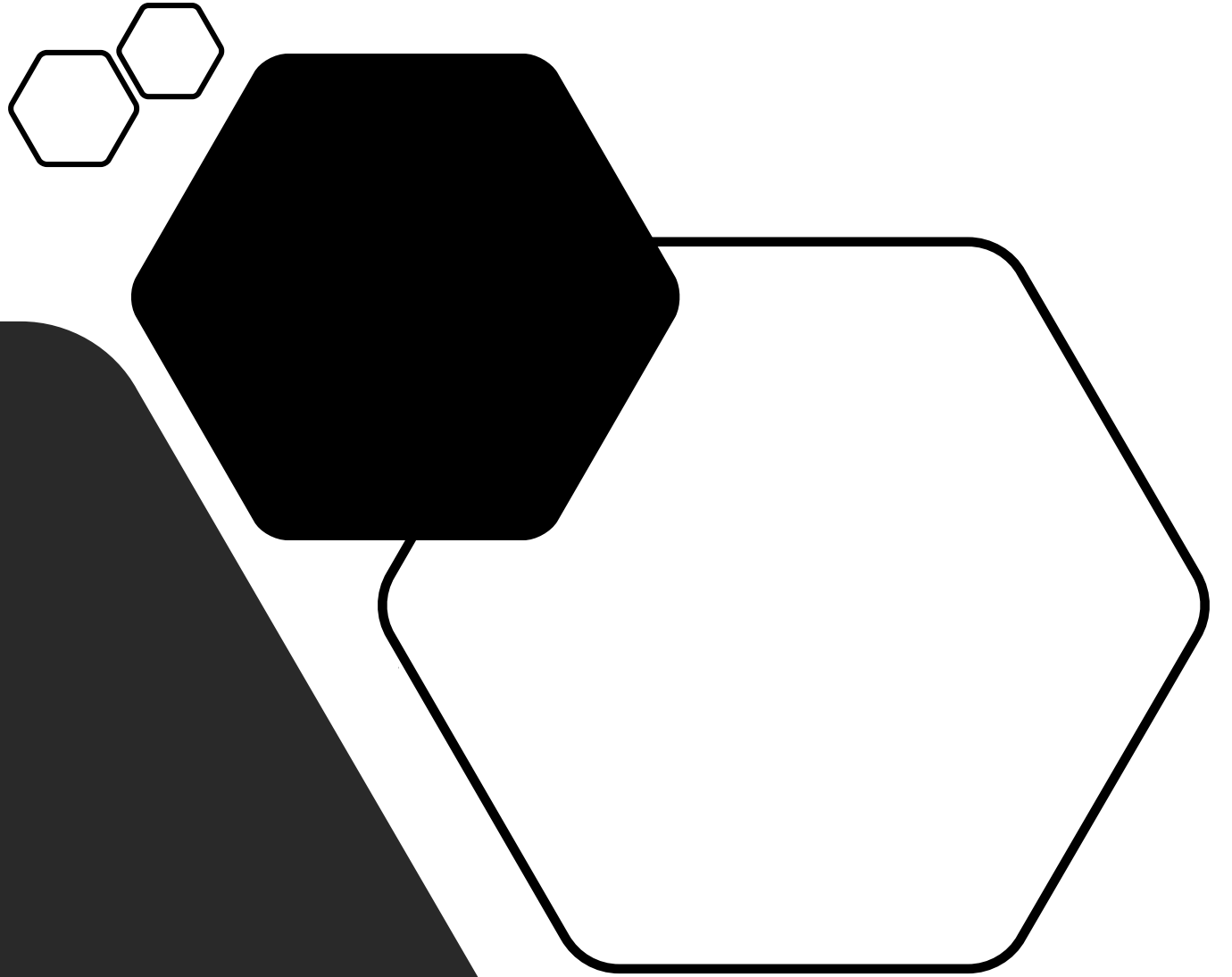


Going forward: Mask-RCNN

- Every patch of image can correspond to an object

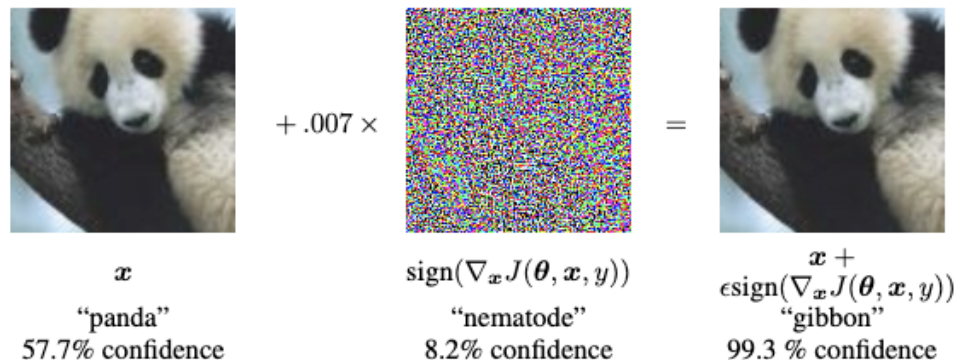


Limitations of CNNs



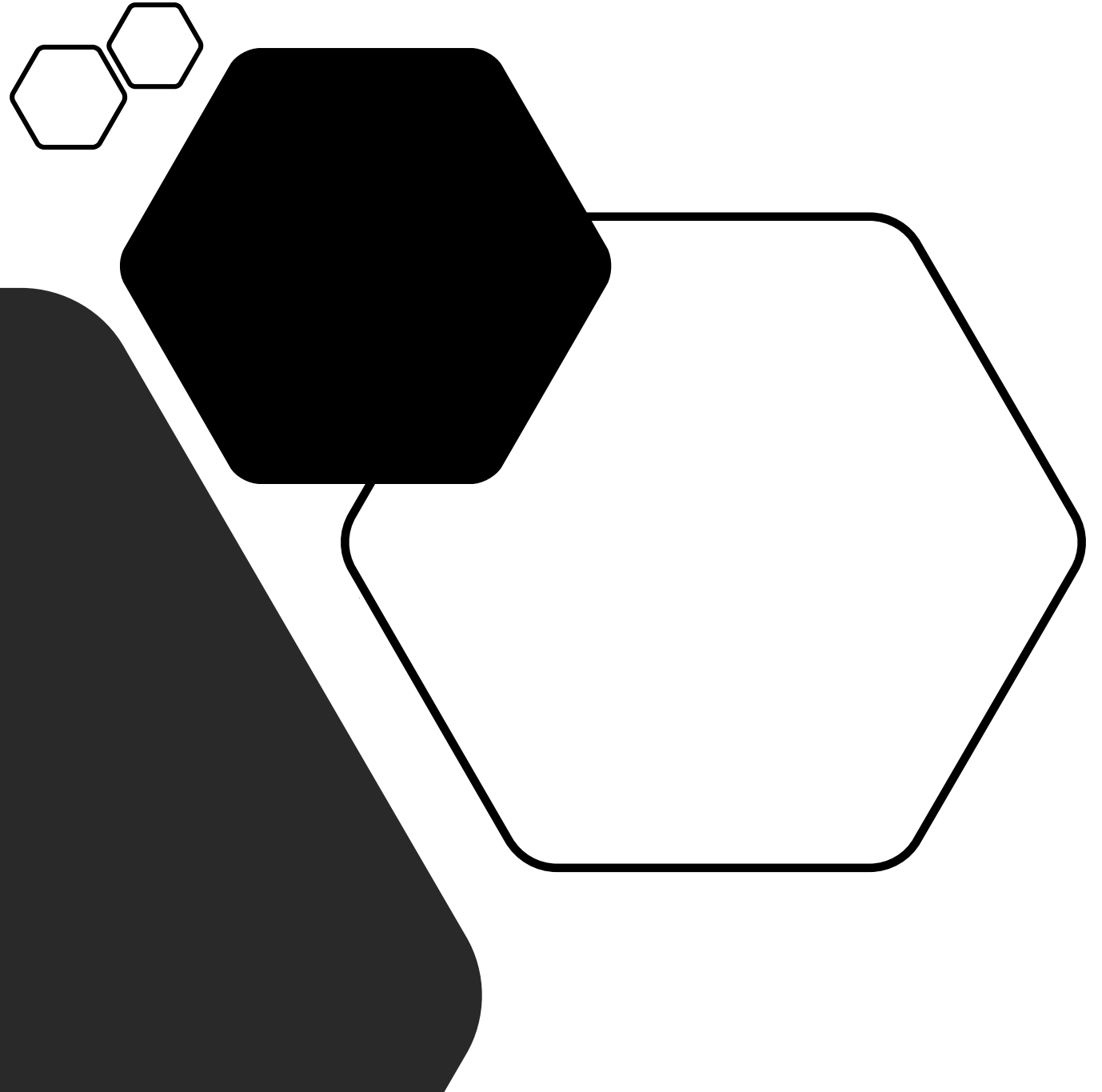
Dataset vs natural images

- A dataset gives you examples of images + classes
- Not all images and classes are accounted for (data is limited)
- How well does it work in the real world?



"a young boy is holding a baseball bat."

Lab 10 – 1



Lab 10 – part 1

- CIFAR 10:
 - 60000 32x32 images
 - 10 classes
- Use CNNs to solve this
- Use colab if necessary

airplane



automobile



bird



cat



deer



dog



frog



horse



ship



truck

