

Convolutional Neural Network

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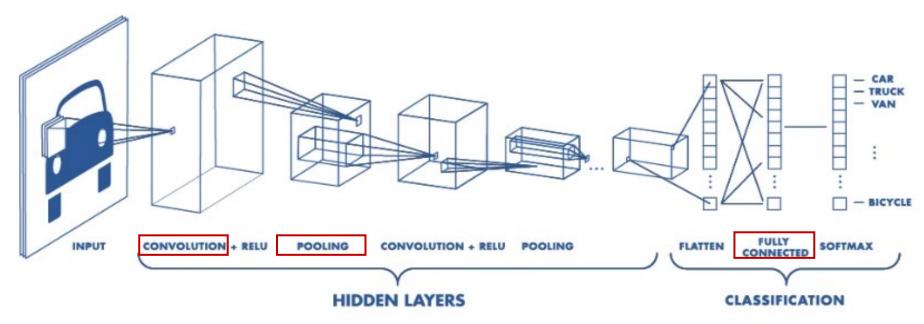
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- Convolutional Neural Network
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1. Convolutional Neural Network

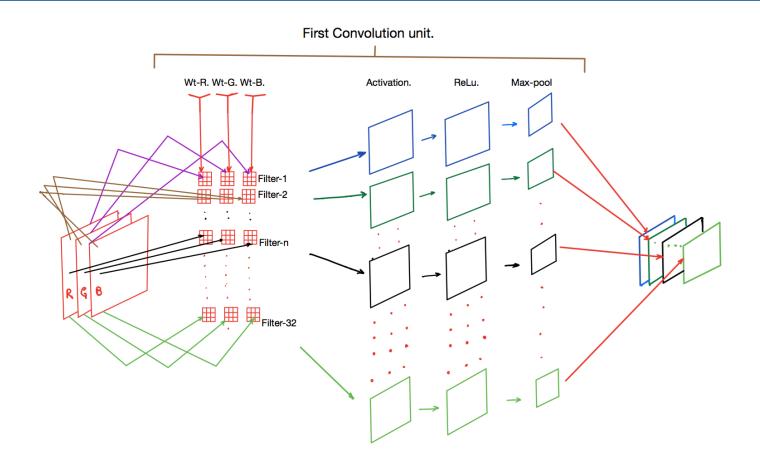




Main Components

- Convolution (Shared weights, dimension reduction)
- Nonlinearity (ReLU)
- Max pooling (translation invariance, dimension reduction for FC layer)
- Fully connected Layer (classification)





• The weights and the biases which are the components of the filters change when training the network.

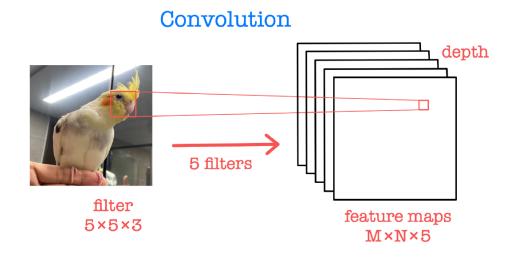
1-1. Convolution Layer

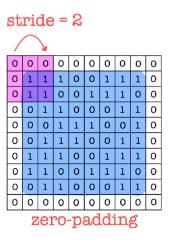


Hyperparameters on Convolution layer

- filters: number of filters (output feature maps)
- kernel size: size of kernel (convolution window: height x width)
- strides: distance between two successive kernel positions
- padding: "valid" = no padding

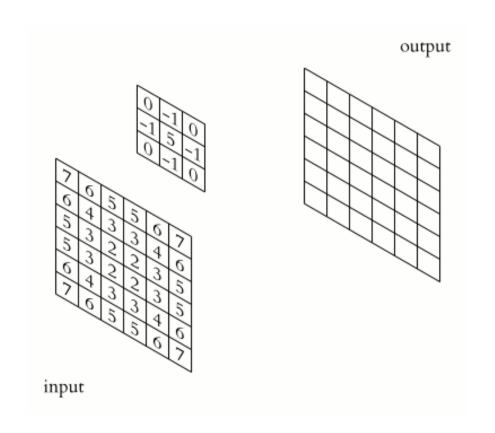
"same" = padding with zeros to make the output with the same size as the input





1-2. Convolution Operation

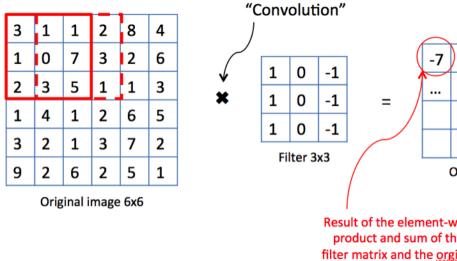


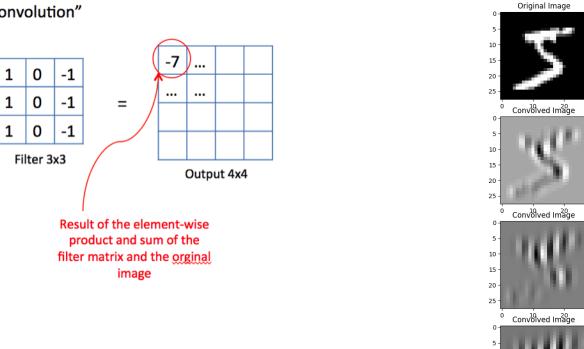


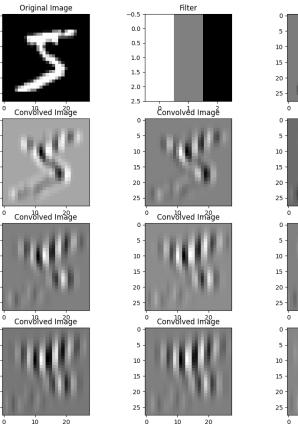
•
$$g(x,y) = w * f(x,y) = \sum_{i=-a}^{a} \sum_{j=-b}^{b} w(i,j) f(x-i,y-j)$$

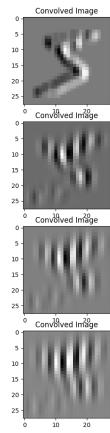
1-2. Convolution Operation - Kernel







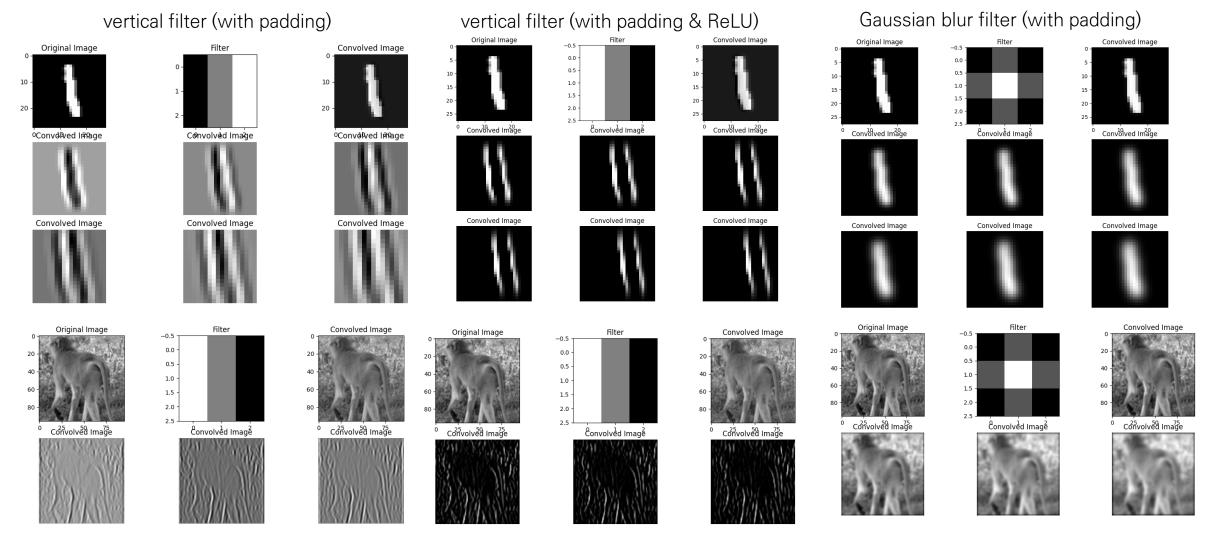




- Note that the size of the output may be different from the size of the input.
- The output of the filter will be large if the input has the feature that the filter can detect.

1-2. Convolution Operation - Some Examples of Kernel





^{*} The code is uploaded on GitHub. (convolution_operation_visualization.py)

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2. Pooling Layer



Max Pooling **Average Pooling** Take the highest value from Calculate the average the area covered by the value from the area covered by the kernel Example: Kernel of size 2 x 2; stride=(2,2) Convolved Convolved 0 0 2 0, 2 0, Feature Feature (4×4) (4×4) 0 0 2 `.3 Ú, 2 3 2`. œ. 0 3 0 3 Output Output 3 Max Average values values

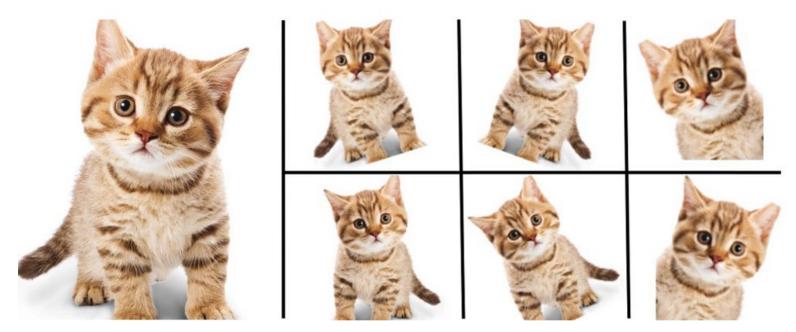
- Pooling layers reduce the dimensions of data by combining the outputs of neuron clusters at one layer into a single neuron in the next layer.
- Max pooling: Take the highest value from the area covered by the kernel
- Average pooling layer: Calculate the average value from the area covered by the kernel

Data Augmentation



1. Data Augmentation



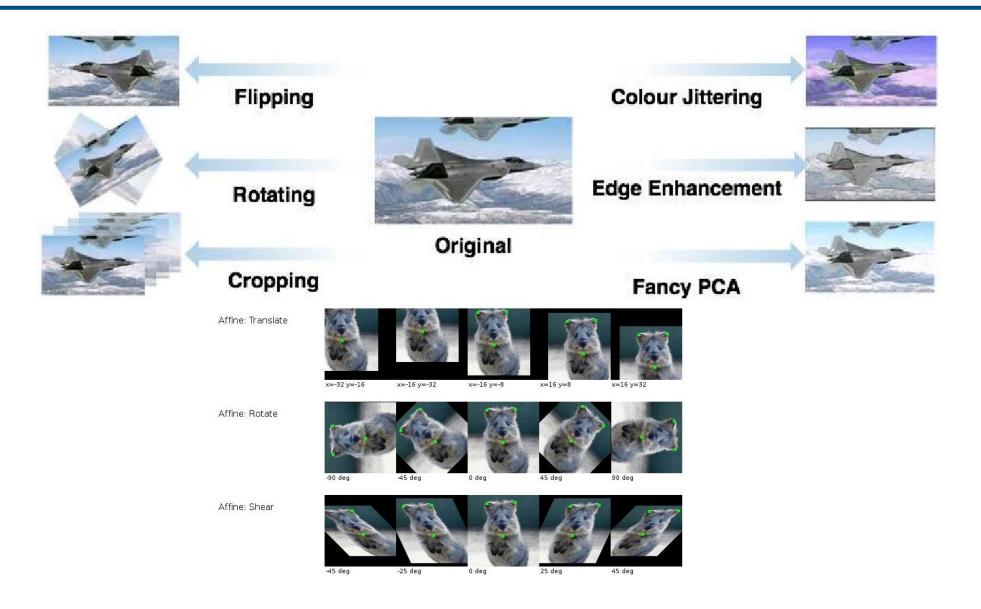


Enlarge your Dataset

One way of reducing overfitting.

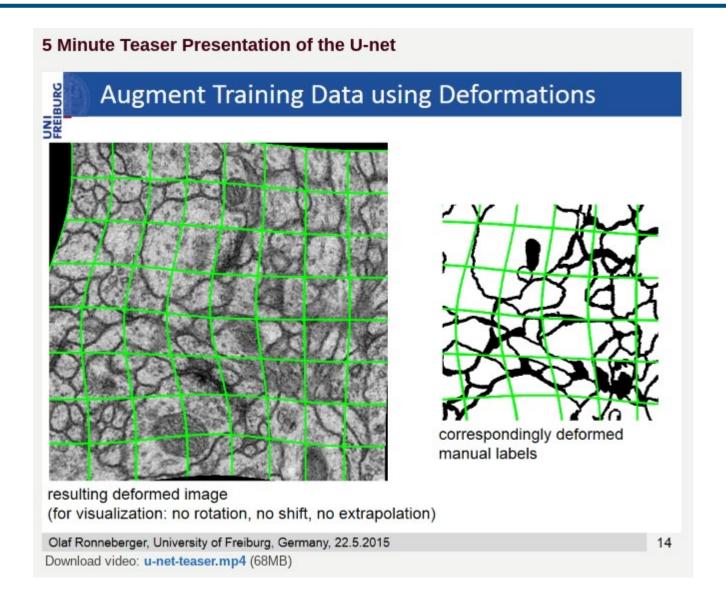
1. Data Augmentation

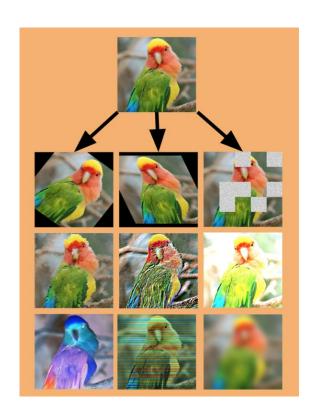




1. Data Augmentation









1. ImageNet



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14,197,122 images, 21841 synsets indexed

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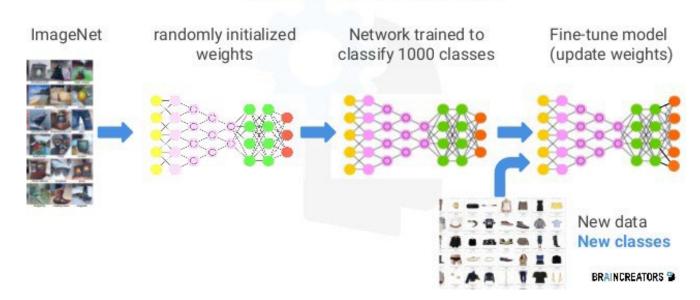
ImageNet is an image database organized according to the WordNet hierarchy (currently only the nouns), in which each node of the hierarchy is depicted by hundreds and thousands of images. Currently we have an average of over five hundred images per node. We hope ImageNet will become a useful resource for researchers, educators, students and all of you who share our passion for pictures.

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2. Transfer Learning



Transfer Learning



• The core idea of transfer learning is to get better initial weights to extract good visual features





```
import torchvision.models as models
model = models.resnet50(weights=models.ResNet50_Weights.IMAGENET1K_V1).to(device)

for name, param in model.named_parameters():
    if param.requires_grad:
        print(f"Trainable parameter: {name}")
    else:
        print(f"Non-trainable parameter: {name}")
```

- import torchvision
 for name in dir(torchvision.models)
 print(name)
 - Check pretrained models in torchvision without resnet50

- Load pretrained model at torchvision.models
- Check model's layer name & trainable / non-trainable parameters

Question



• GitHub에 공유된 코드는 직접 제작한 모델입니다. 바로 위 페이지의 코드를 참고하여 transfer learning을 직접 구현해보고, 결과를 notion에 공유해주세요.