

Kjersti Engan, professor, Dept. of Electrical Eng. and Computer Science, UiS

---

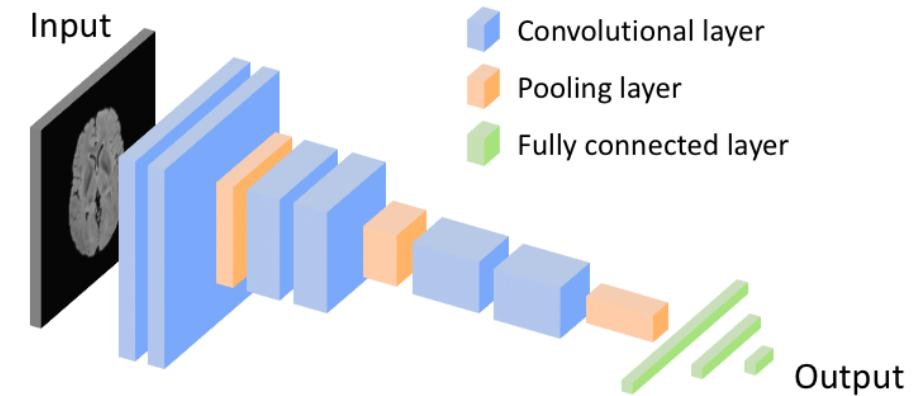
# Introduction to Deep Neural Networks for image analysis

ELE510 Image processing and computer vision 2023

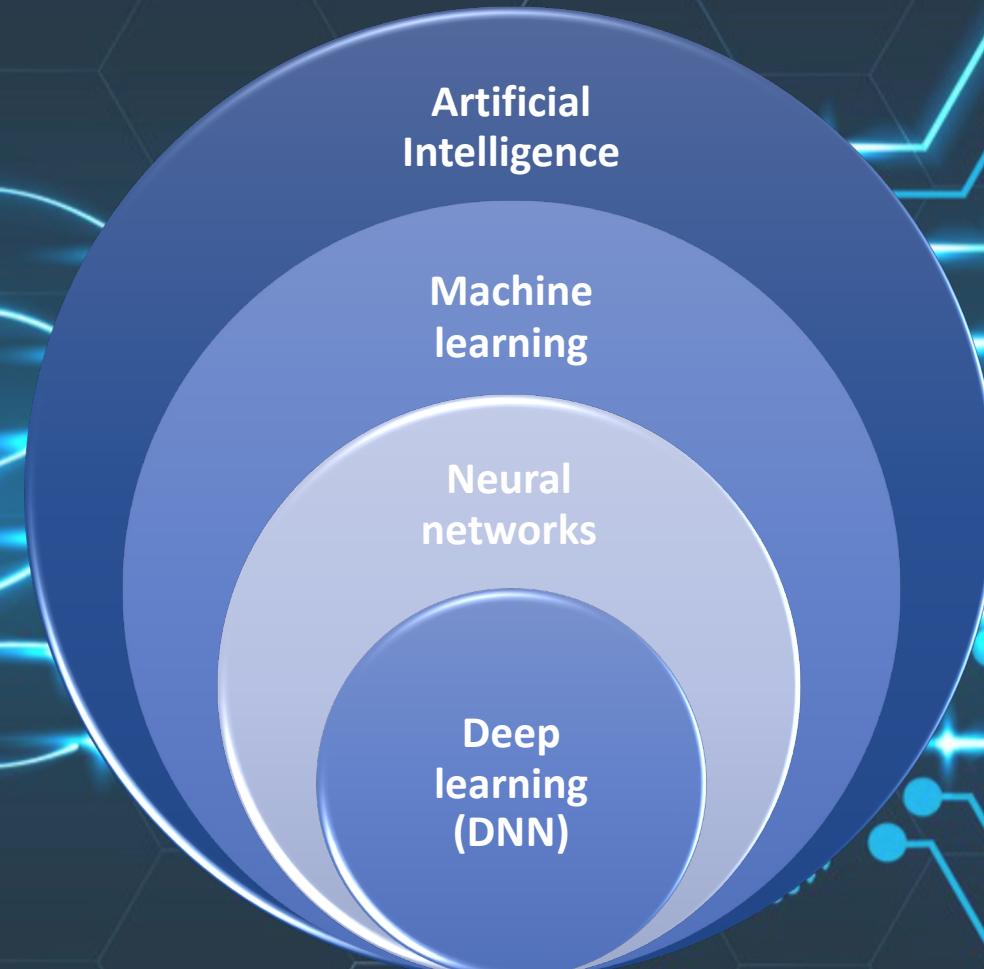
# Intro to Deep Neural Network for image analysis

Three points form the topic:

1. What is a artificial neuron?
2. How does convolutional neural network relate to spatial filtering (convolution)
3. What is supervised learning?



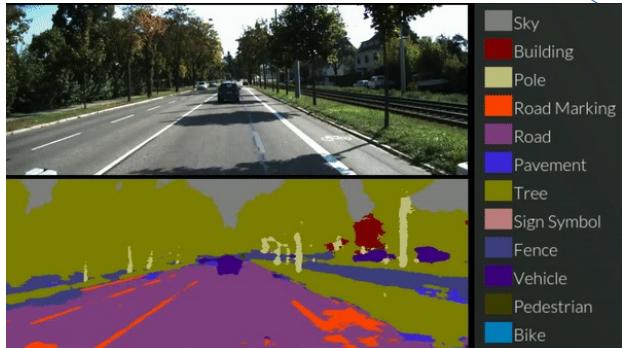




Generate an image from text - And much more..



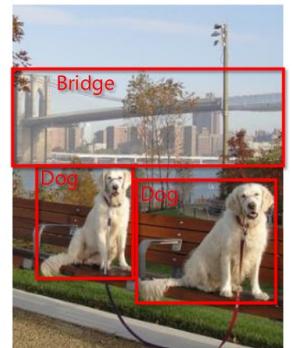
Segmentation



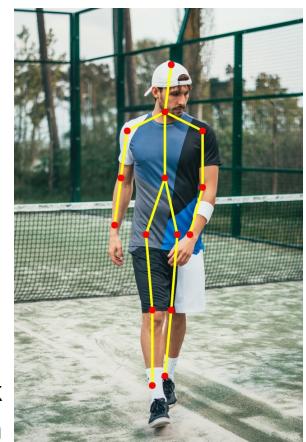
“Paint” this image with the style of another image

What is in the image?

Classification, easy these days



Where in the image is it?



## Neural Networks for images

What am I seeing?

Find a specific shape in the image

Is there a face in this image?

Landmark detection



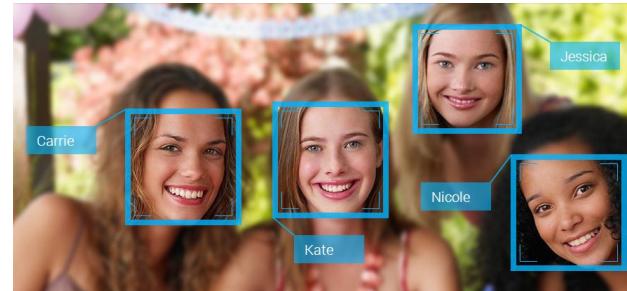
Is this person bob?

Who is in this image?

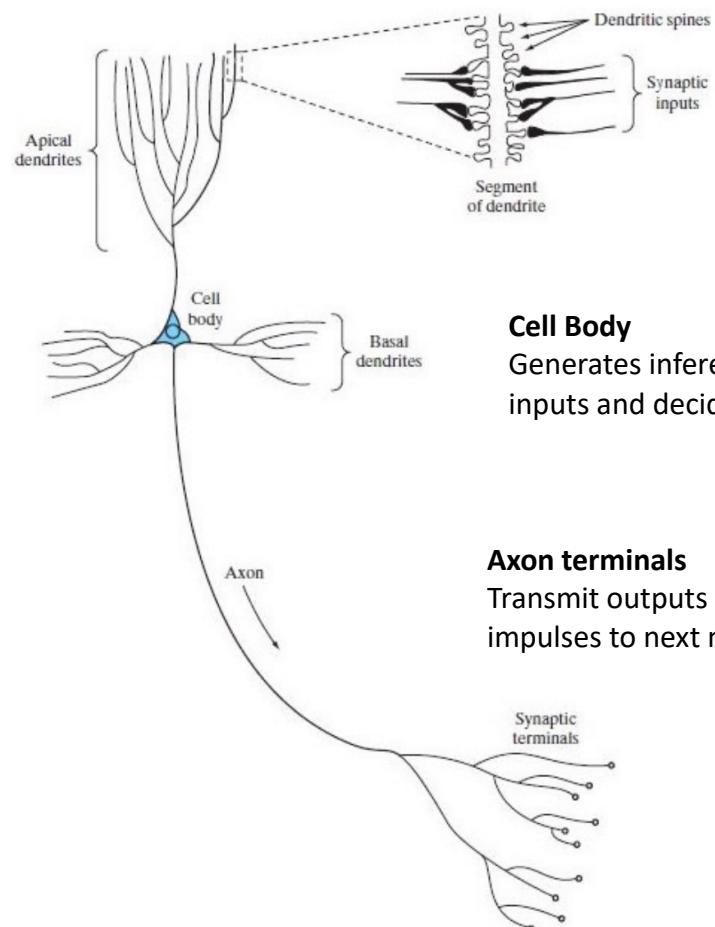
Face verification (1:1 matching)



Face recognition (1:N matching)



# Biological neurons



## Dendrites

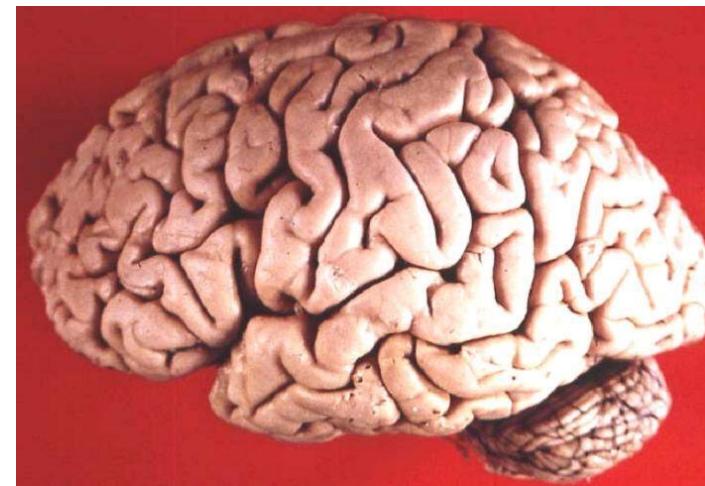
Input from other neurons in the network in form of electrical impulses

## Cell Body

Generates inferences from the dendrite inputs and decides what action to take

## Axon terminals

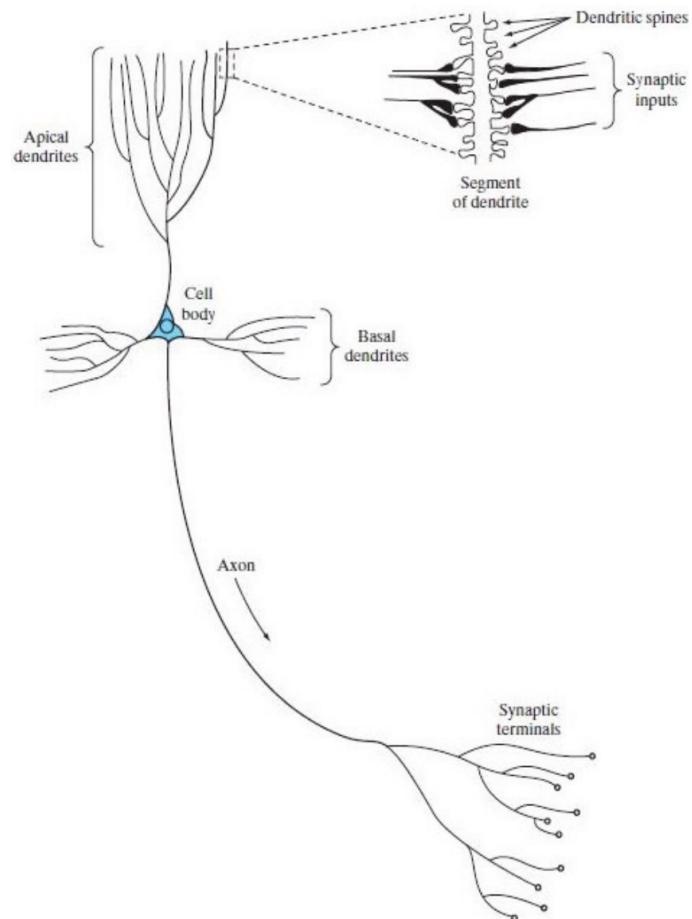
Transmit outputs in form of electrical impulses to next neuron



## Attribution:

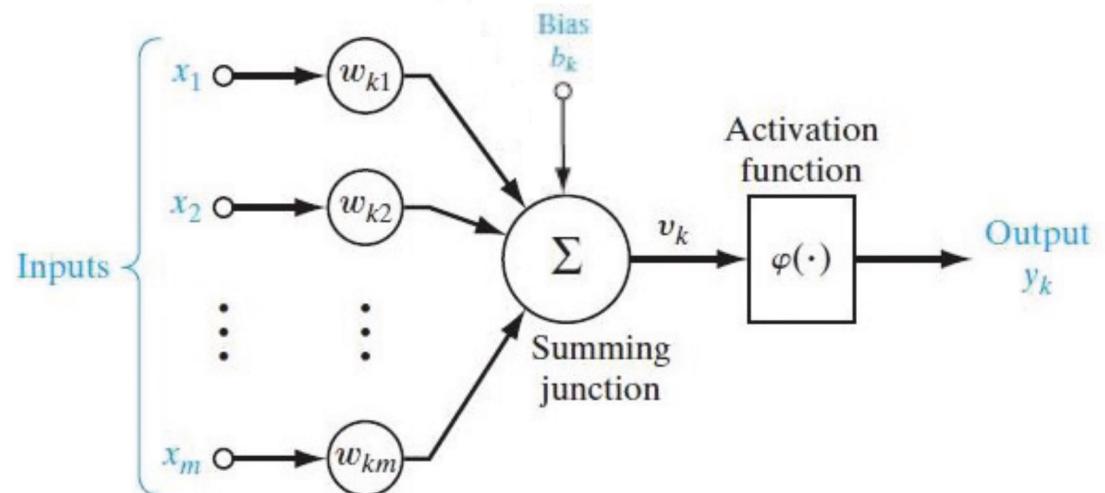
[John A Beal, PhD Dep't. of Cellular Biology & Anatomy, Louisiana State University Health Sciences Center Shreveport / CC BY](#)

## Biological neurons



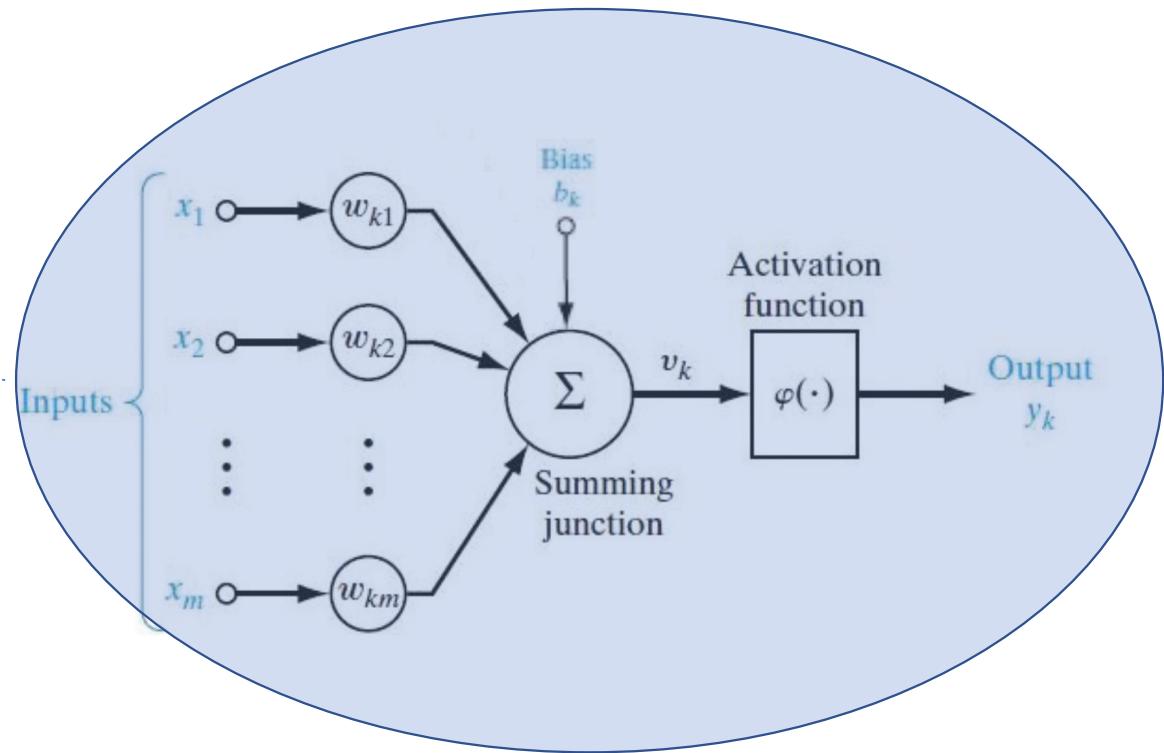
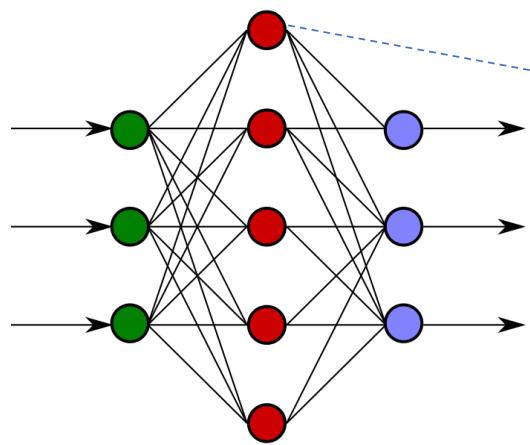
Zayegh, Amer, and Nizar Al Bassam.  
"Neural network principles and applications."  
Digital Systems. IntechOpen, 2018.

## Artificial neurons (It is only an equation)



Zayegh, Amer, and Nizar Al Bassam.  
"Neural network principles and applications."  
Digital Systems. IntechOpen, 2018.

## Neural network

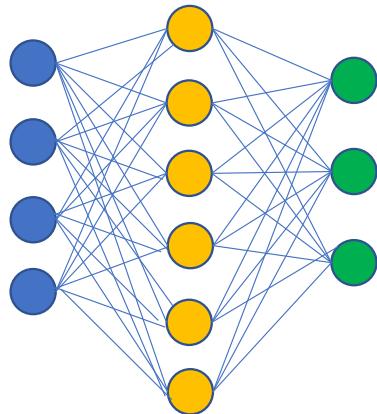


●  
Input  
layer

●  
Hidden  
layer

●  
Output  
layer

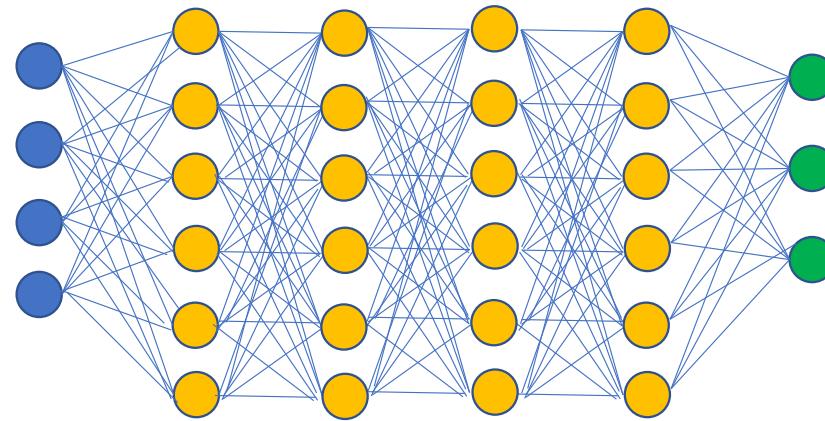
# Deep learning / deep neural networks



Simple neural network

- Input layer neuron
- Hidden layer neuron
- Output layer neuron

Neurons can have different activation functions



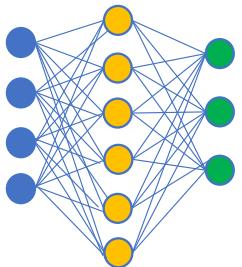
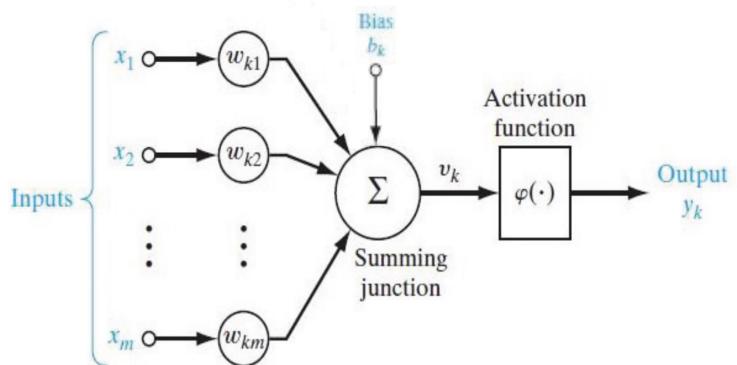
Deep neural network

- Each connection has a weight
- Each neuron has a bias

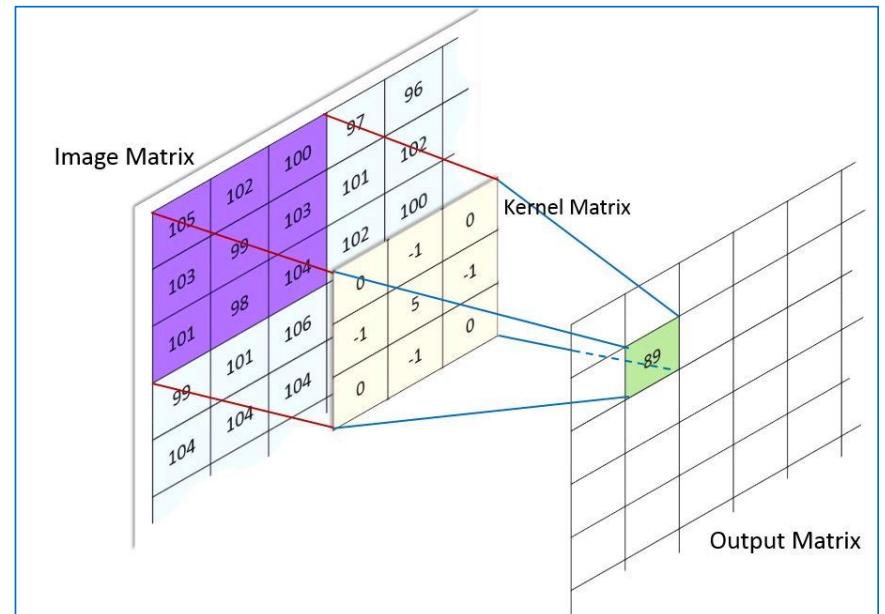
A deep neural network can have millions of parameters  
These are learned using example data

Feed-forward network  
Fully connected layers (FC)

# Convolutional layers



Fully connected layer



# Convolution

10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0

\*

1	0	-1
1	0	-1
1	0	-1

Vertical

=

0	0	30	30	0	0
0	0	30	30	0	0
0	0	30	30	0	0
0	0	30	30	0	0
0	0	30	30	0	0

1	0	-1
2	0	-2
1	0	-1

Sobel filter

3	0	-3
10	0	-10
3	0	-3

Scharr filter

$W_1$	$W_2$	$W_3$
$W_4$	$W_5$	$W_6$
$W_7$	$W_8$	$W_9$

parameterized filter

10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
10	10	10	10	0	0	0	0	0
0	0	0	0	10	10	10	10	10
0	0	0	0	10	10	10	10	10
0	0	0	0	10	10	10	10	10
0	0	0	0	10	10	10	10	10
0	0	0	0	10	10	10	10	10

\*

1	1	1
0	0	0
-1	-1	-1

Horizontal

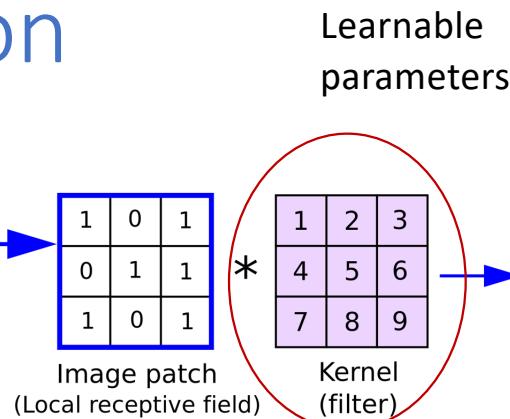
=

0	0	0	0	0	0
0	0	0	0	0	0
30	30	10	-10	-30	-30
30	30	10	-10	-30	-30
0	0	0	0	0	0

# Convolution

1	0	1	0	1	0
0	1	1	0	1	1
1	0	1	0	1	0
1	0	1	1	1	0
0	1	1	0	1	1
1	0	1	0	1	0

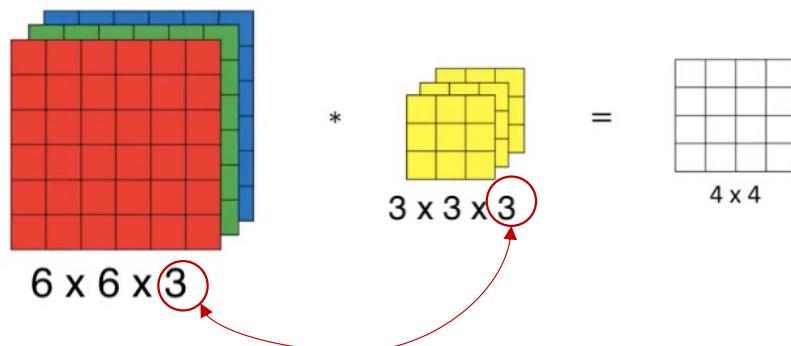
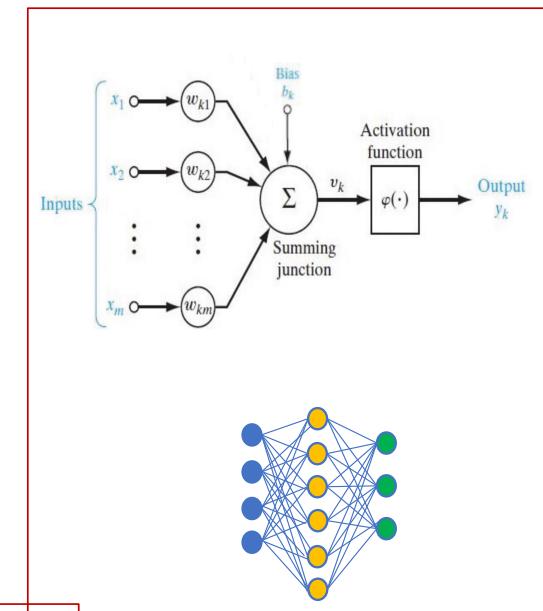
Input



31			

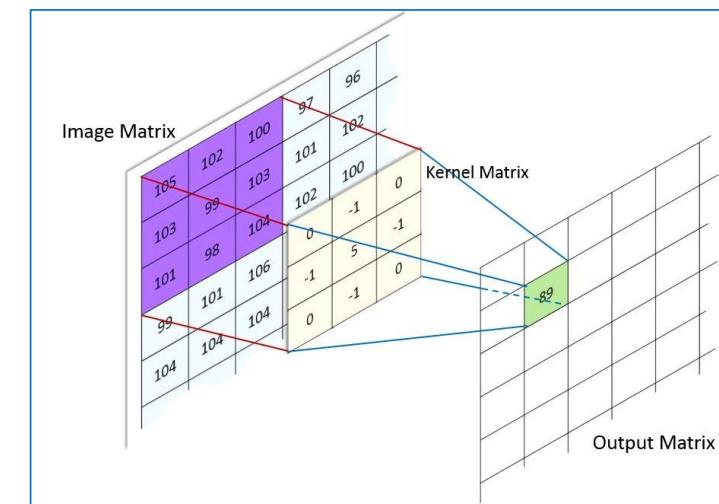
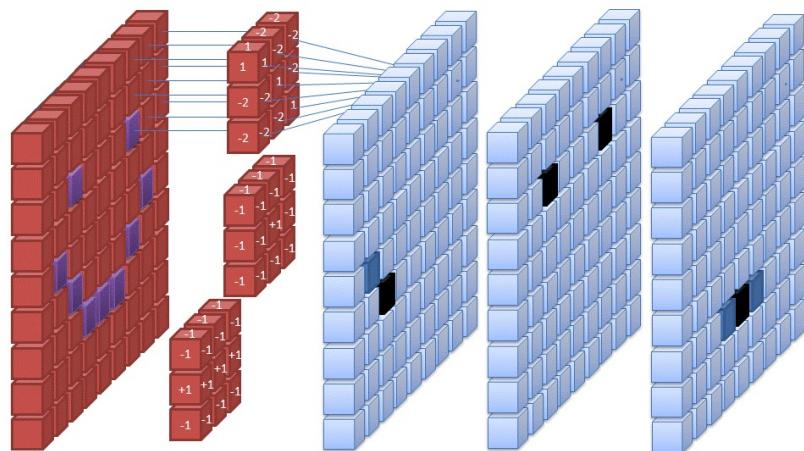
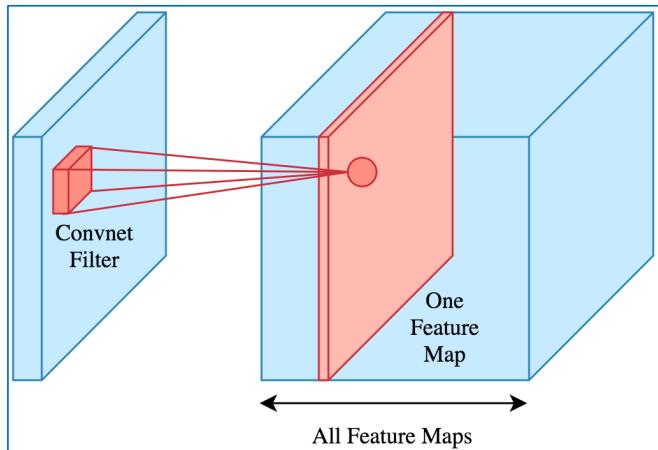
Output

Activation function comes here (for each “pixel”)



Stride  $s = 1$ ,  
padding  $p=0$ , ( why is output smaller?)  
1 filter

# Convolutional layers

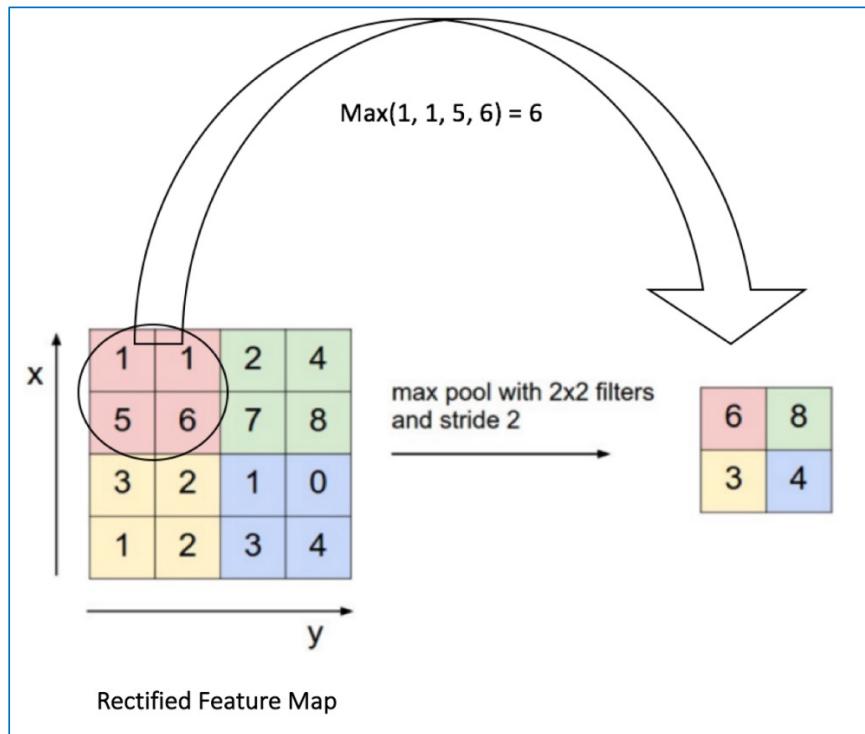


Many convolutional filters are used, give a range of feature maps. The filter coefficients are LEARNED (machine learning).

[https://commons.wikimedia.org/wiki/File:3\\_filters\\_in\\_a\\_Convolutional\\_Neural\\_Network.gif](https://commons.wikimedia.org/wiki/File:3_filters_in_a_Convolutional_Neural_Network.gif)

<https://brilliant.org/wiki/convolutional-neural-network/>

## Pooling layer



max pooling is done, reducing the size, but keeping the high outputs

<https://brilliant.org/wiki/convolutional-neural-network/>



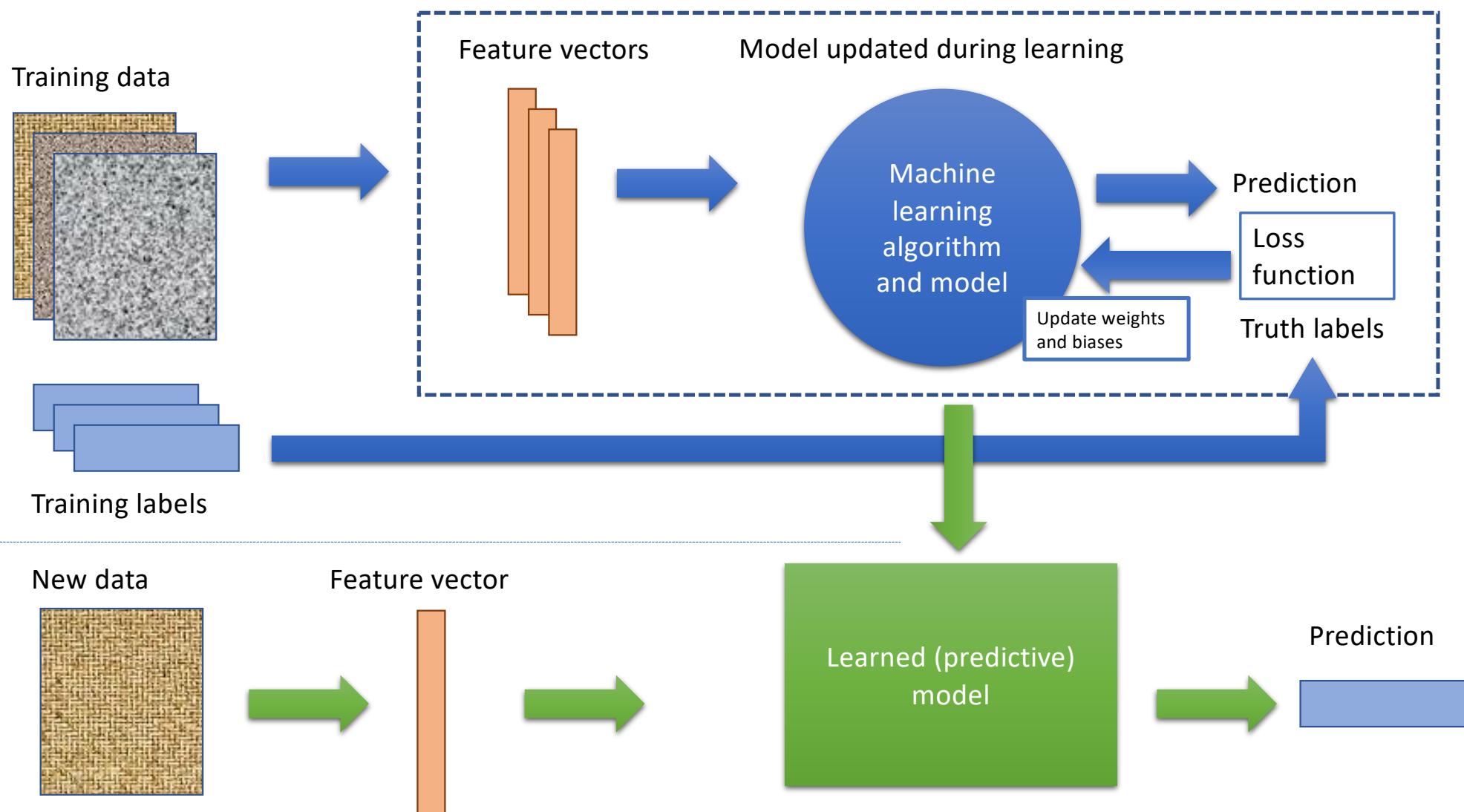
How do we learn the kernels/filters?

Machine learning



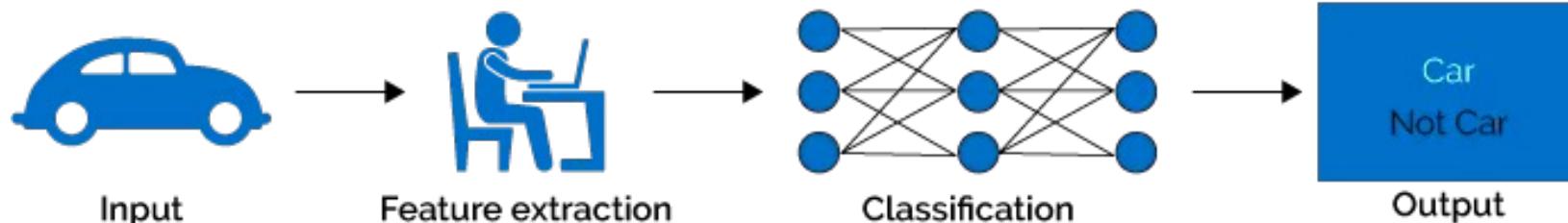
just a little teaser

## Supervised learning

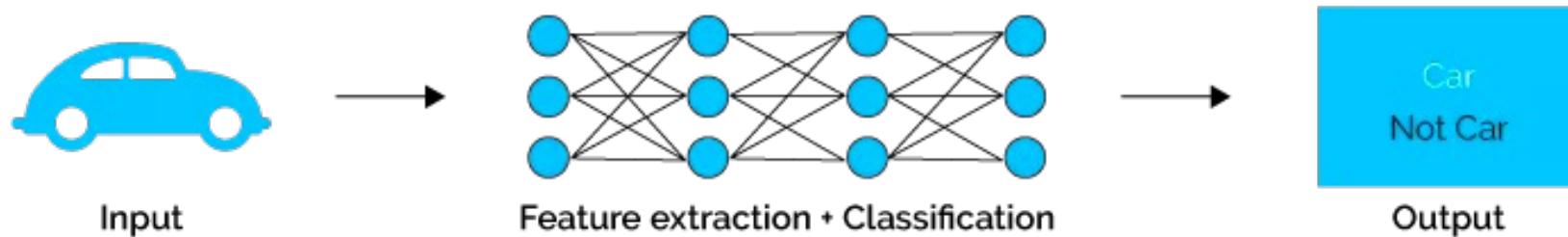


## Machine Learning

Need lots of data



## Deep Learning



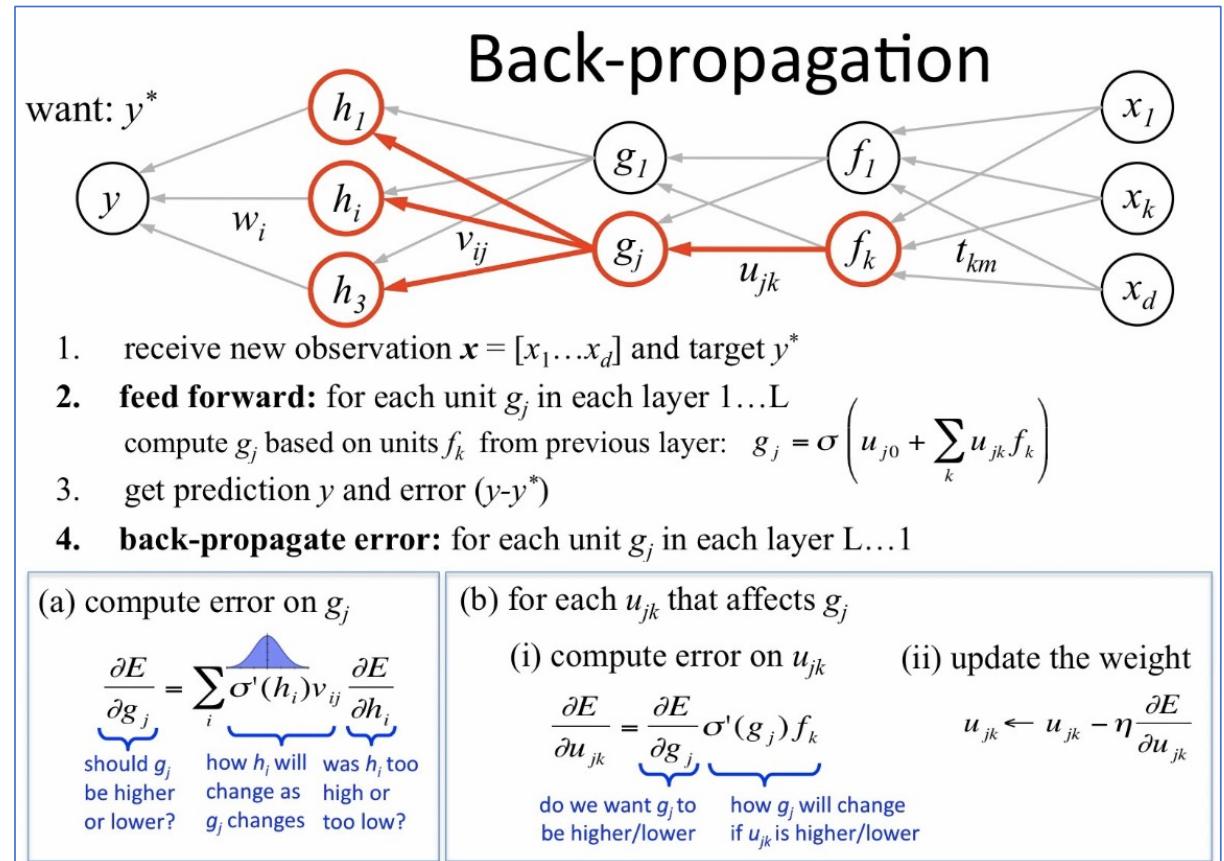
Need even more data

<https://thedataScientist.com/what-deep-learning-is-and-isnt/>

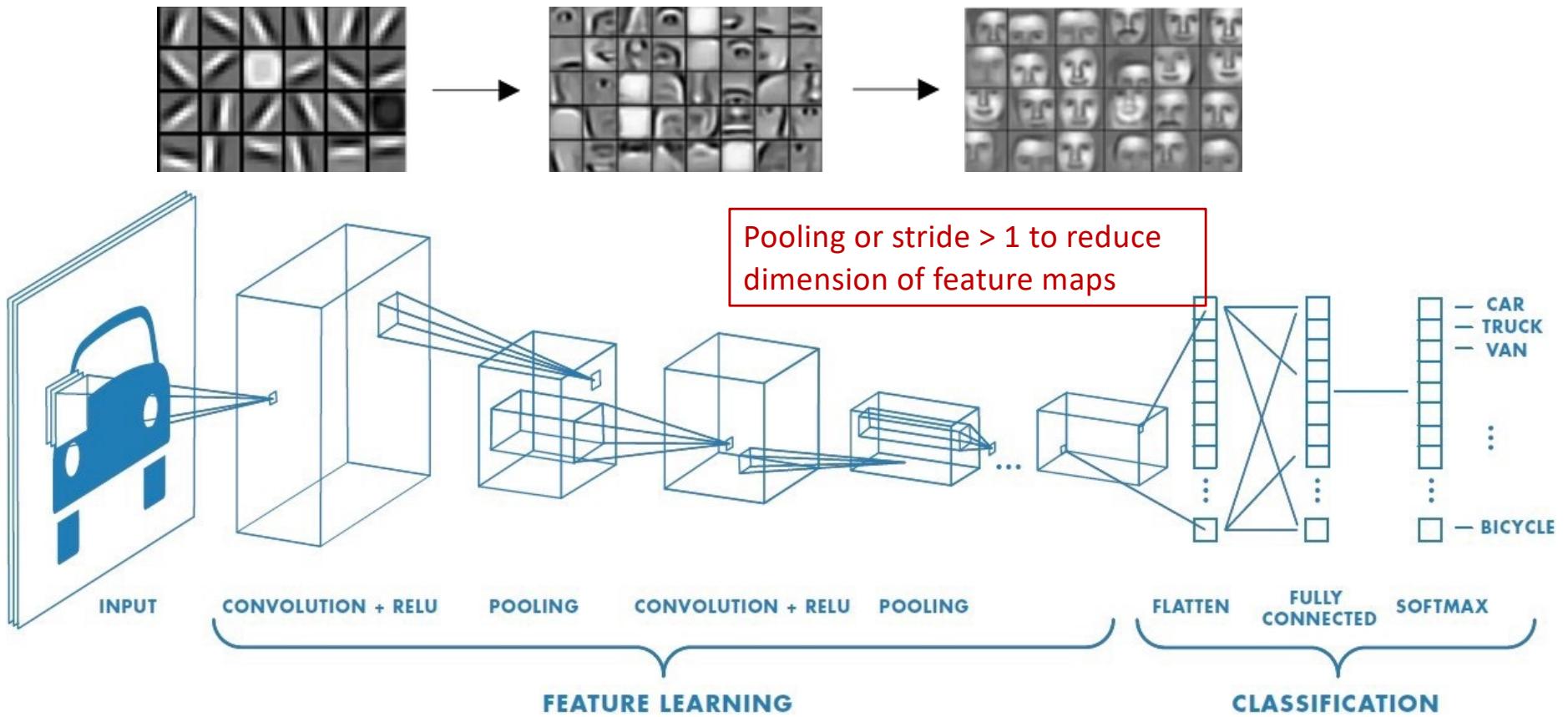
# Supervised learning – how?

- Loss function
  - Feed forward to compute loss
  - Backpropagation of the error
  - Update weights to reduce error
  - Repeated many epochs

# The magic force lies in a relevant loss function

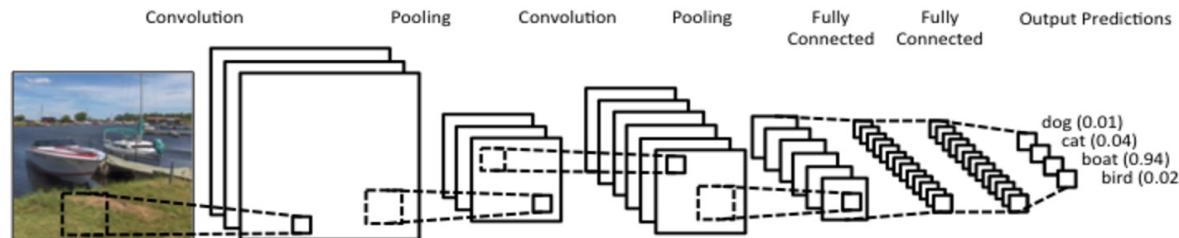


attribution: Victor Lavrenko  
<https://www.youtube.com/watch?v=An5z8IR8asY> 6 min. video



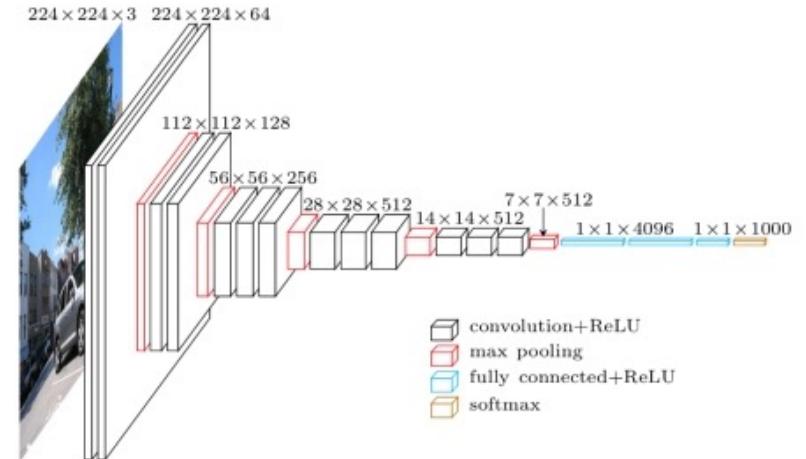
<https://towardsdatascience.com/a-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way-3bd2b1164a53>  
<https://datahacker.rs/edge-detection/>

# Deep Neural Networks – works “magically” good for many tasks



And really; it is all math...

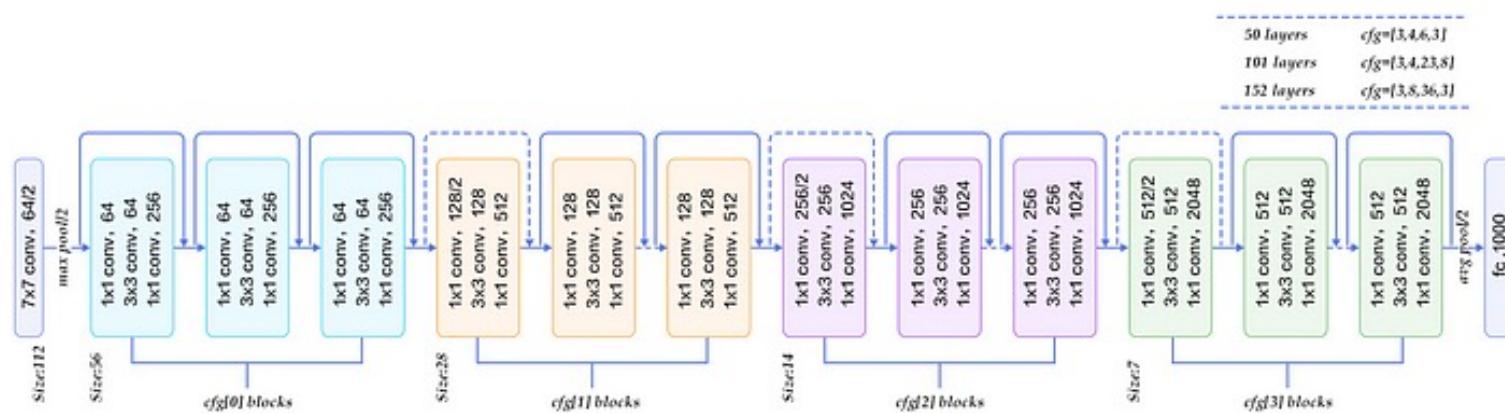
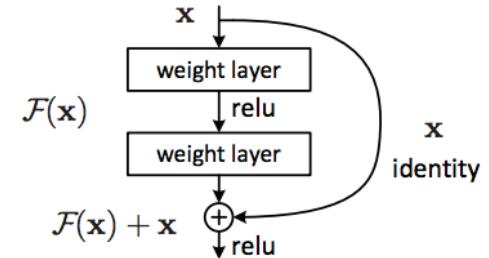
Backpropagation of the error, finding the **derivatives** for each layer, **gradient decent** for updating parameter (weights) of the network ...



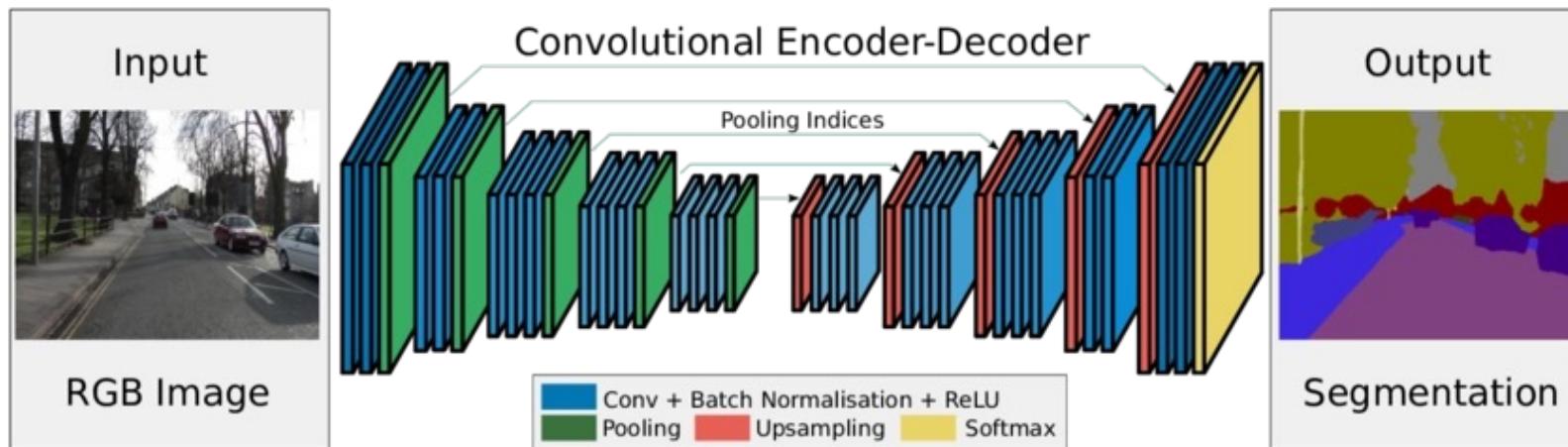
VGG16 network (16 layers  
+ 5 pooling layers)

## ResNet50

ResNet, short for Residual Networks is a classic neural network used as a backbone for many computer vision tasks. Introduced the skip-connections

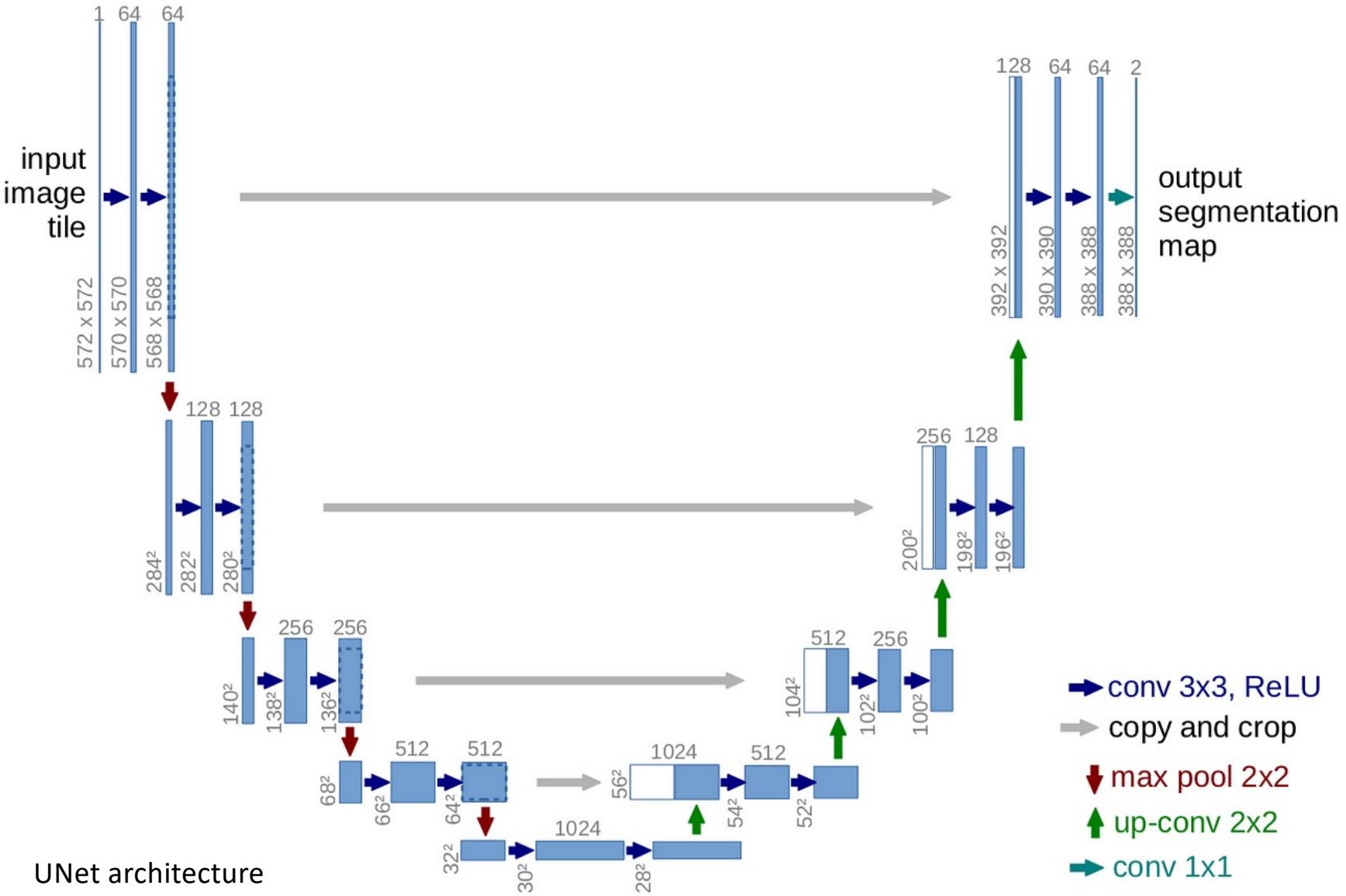


**SegNet:** SegNet was introduced as an architecture that can perform real-time semantic segmentation on high-resolution images.

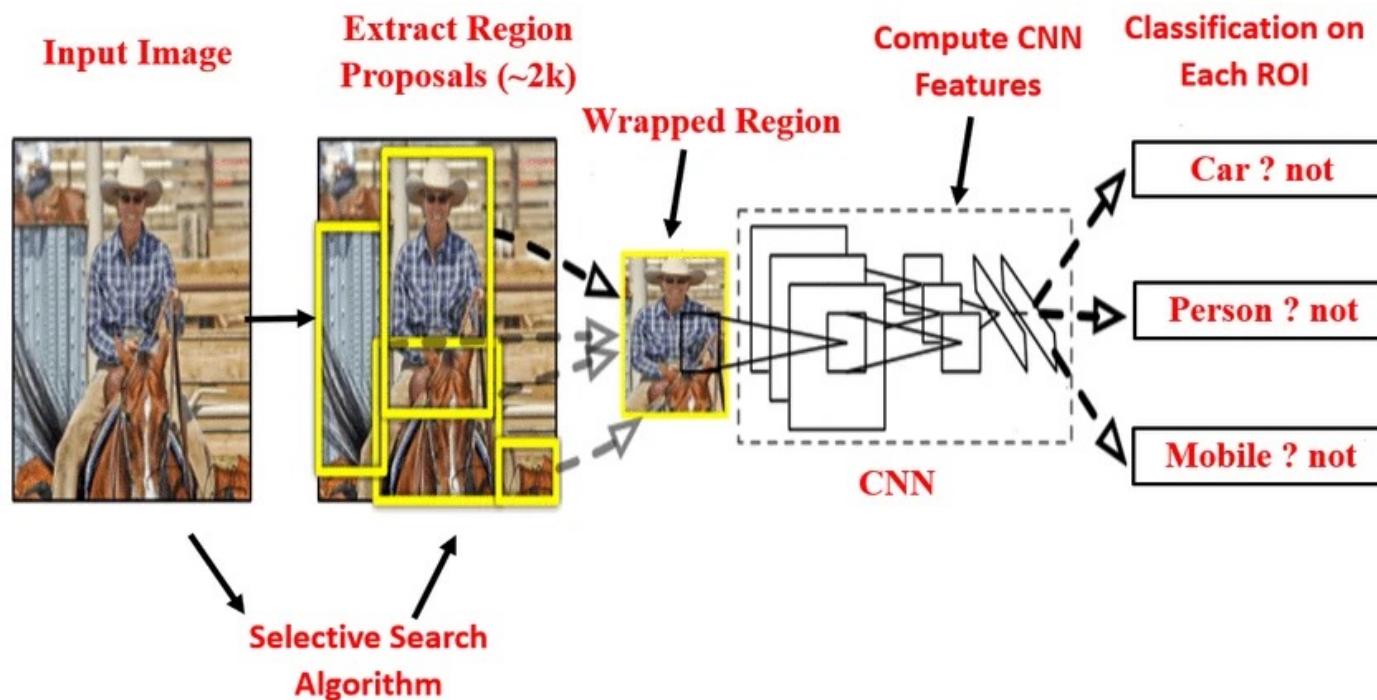


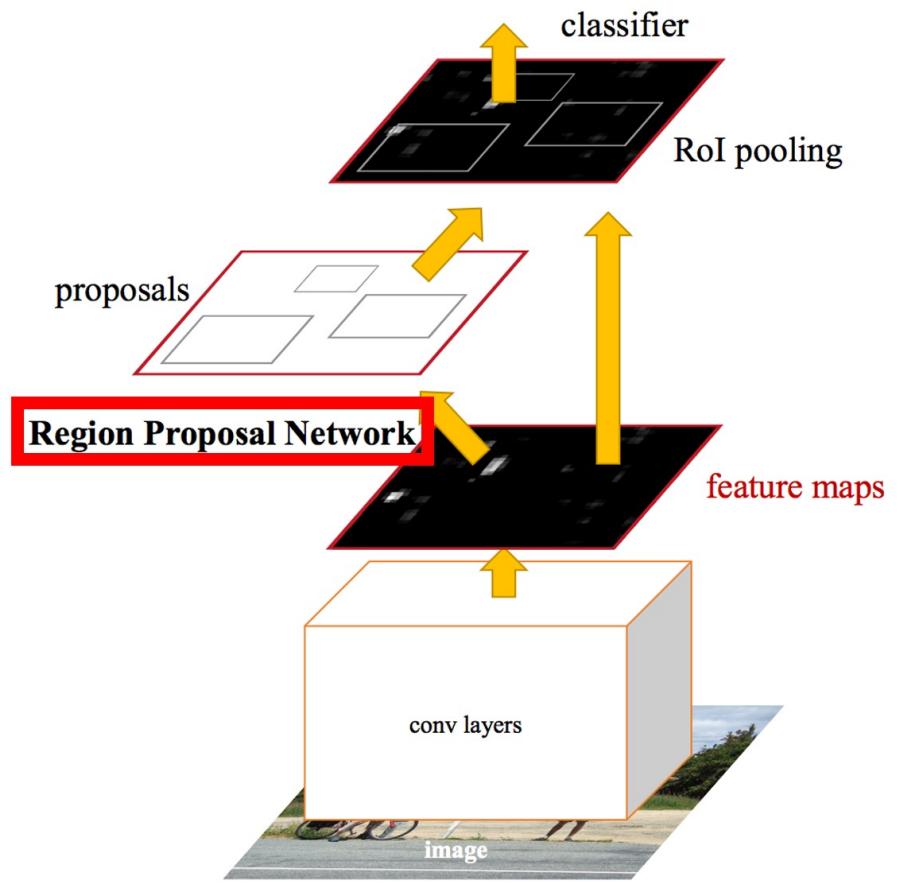
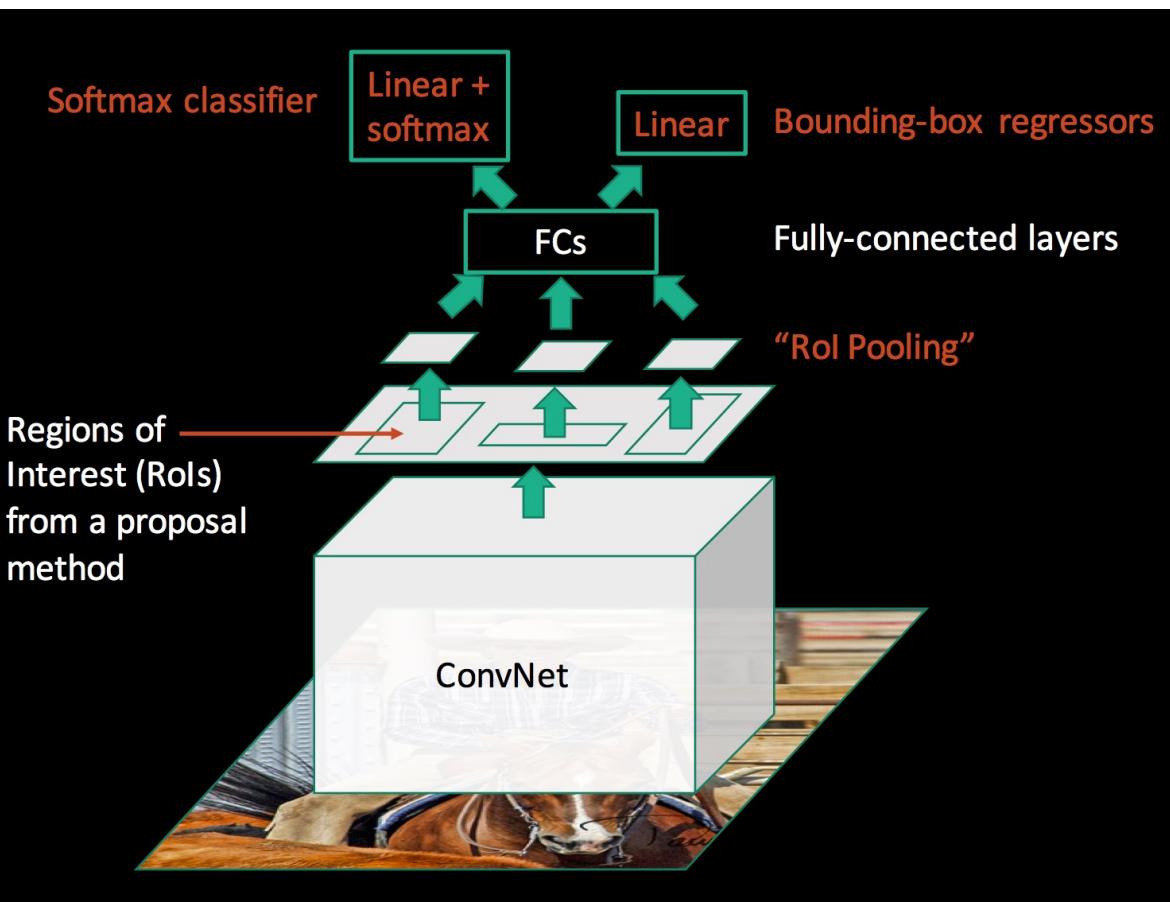
<https://arxiv.org/pdf/1511.00561.pdf>

<https://www.superannotate.com/blog/guide-to-convolutional-neural-networks>



**RCNN:** The [Region-based Convolutional Neural Network \(RCNN\)](#) was proposed by Ross Girshick et al. in 2014 and was one of the first successful deep learning approaches to object detection.





RetinaNet is a one-stage object detection model

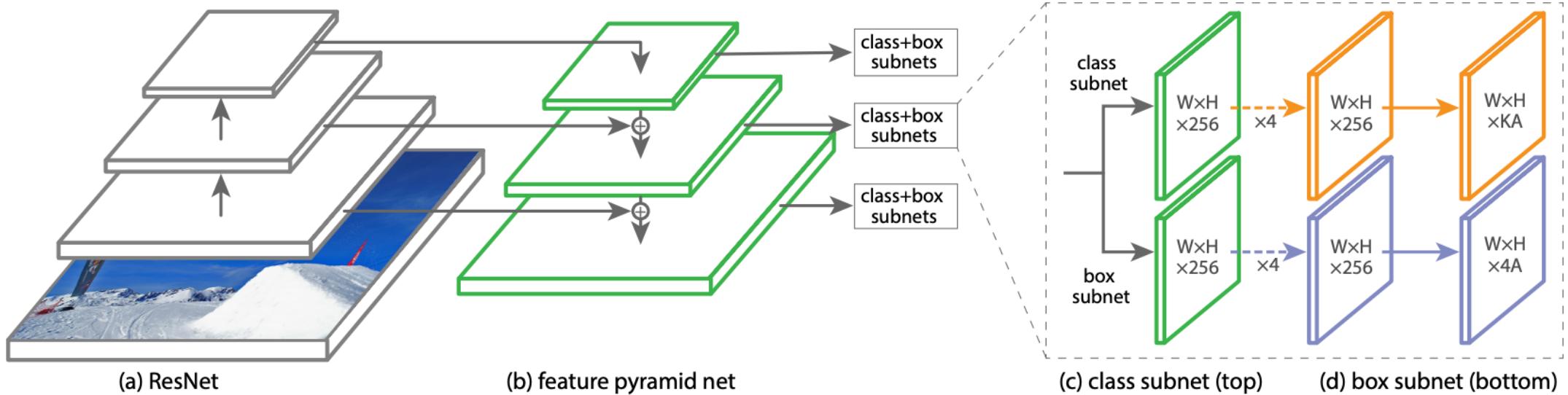


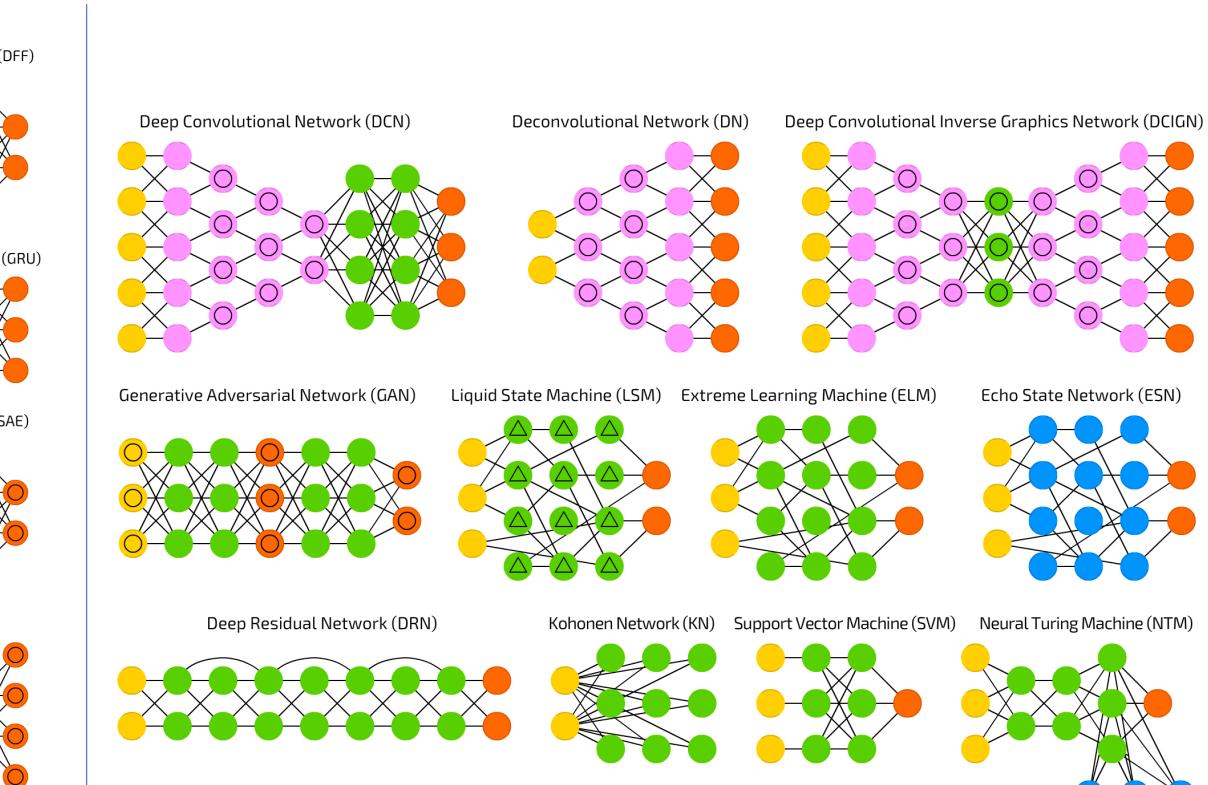
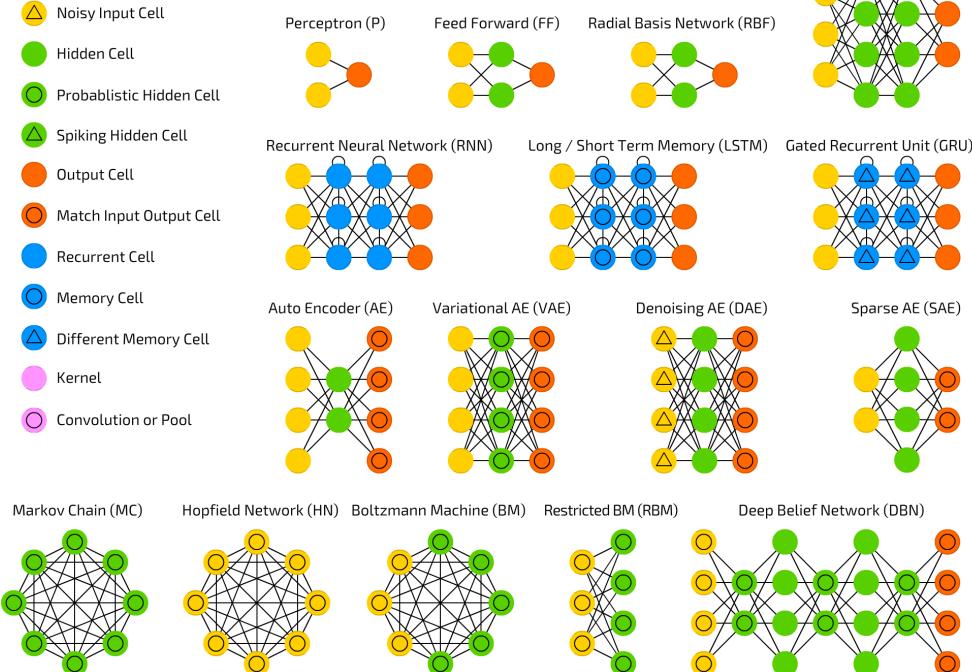
Figure 3. The one-stage **RetinaNet** network architecture uses a Feature Pyramid Network (FPN) [20] backbone on top of a feedforward ResNet architecture [16] (a) to generate a rich, multi-scale convolutional feature pyramid (b). To this backbone RetinaNet attaches two subnetworks, one for classifying anchor boxes (c) and one for regressing from anchor boxes to ground-truth object boxes (d). The network design is intentionally simple, which enables this work to focus on a novel focal loss function that eliminates the accuracy gap between our one-stage detector and state-of-the-art two-stage detectors like Faster R-CNN with FPN [20] while running at faster speeds.

*A mostly complete chart of*

# Neural Networks

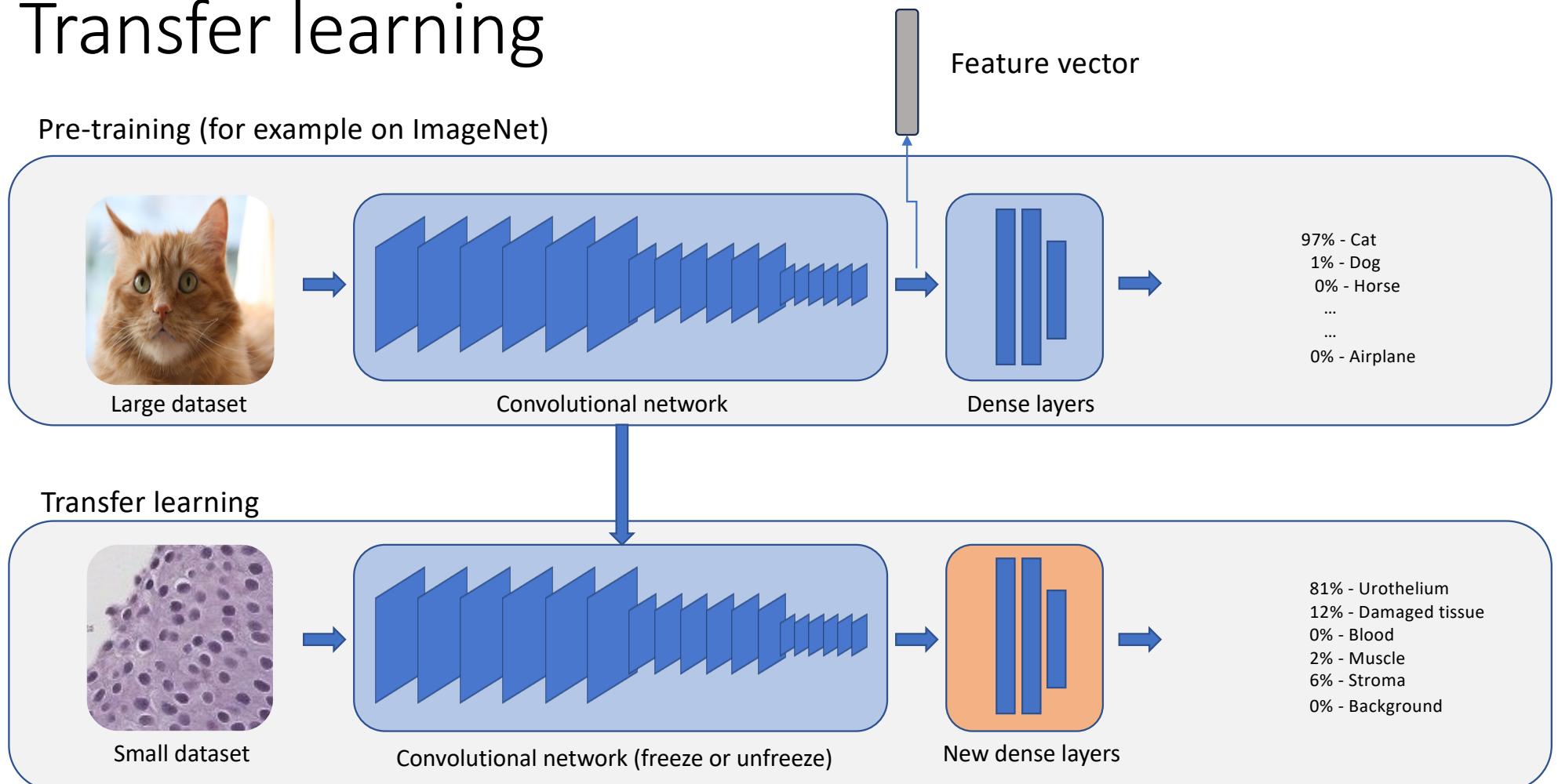
©2016 Fjodor van Veen - asimovinstitute.org

- Backfed Input Cell
- Input Cell
- △ Noisy Input Cell
- Hidden Cell
- Probabilistic Hidden Cell
- △ Spiking Hidden Cell
- Output Cell
- Match Input Output Cell
- Recurrent Cell
- Memory Cell
- △ Different Memory Cell
- Kernel
- Convolution or Pool



<https://towardsdatascience.com/the-mostly-complete-chart-of-neural-networks-explained-3fb6f2367464>

# Transfer learning



# Intro to Deep Neural Network for image analysis

Three points form the topic:

## 1. What is a artificial neuron?

- ✓ An equation, linear combination of many inputs with learnable weights, followed by nonlinearity (activation function)

## 2. How does convolutional neural network relate to spatial filtering (convolution)

- ✓ Closely related. The weights are in forms of filter kernels that are滑过 over the image as in spatial filter ( followed by non-linearity) . Filters are learned, and there are many.

## 3. What is supervised learning?

- ✓ Have example data with known “labels”. Send through network, compare output to known label, send “loss” back into netwrok and change all weights tiny bit to improve. Repeat many times.

