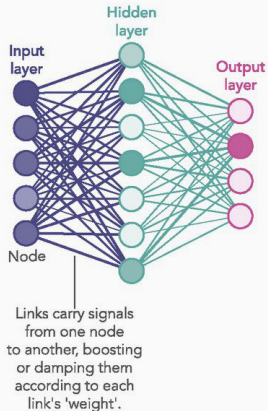


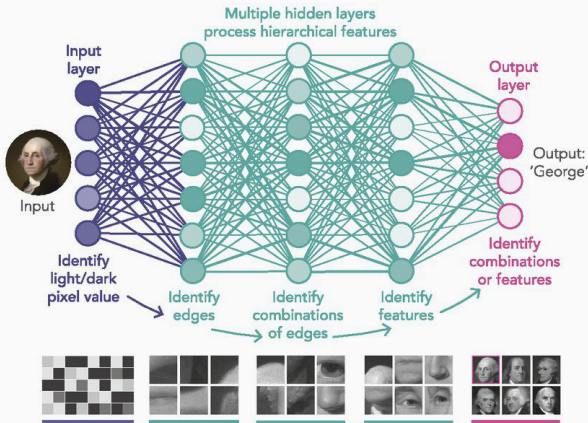
Konvolúciós neuronhálók (CNN)

Deep learning

1980S-ERA NEURAL NETWORK



DEEP LEARNING NEURAL NETWORK



<https://www.pnas.org/content/116/4/1074>

Alkalmazás: képek/szövegek osztályozása, képek szegmentálása, szöveg fordítása, hang átírása szöveggé, képek generálása, stb

Konvolúciós operátor

- ▶ diszkrét konvolúció két egydimenziós jel között:

$$(f * g)(n) = \sum_t f(t)g(n+t)$$

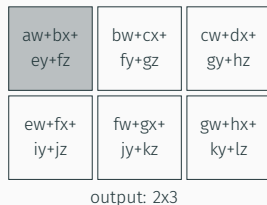
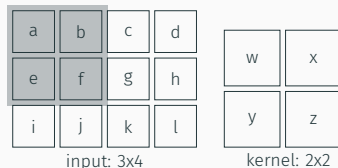
példa: $f = [a, b]$, $g = [w, x, y, z]$

$$f * g = [aw + bx, ax + by, ay + bz]$$

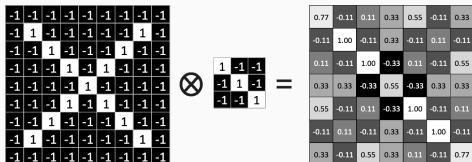
- ▶ diszkrét konvolúció két kétdimenziós jel között:

$$(K * I)(i, j) = \sum_{m, n} K(m, n)I(i + n, j + m)$$

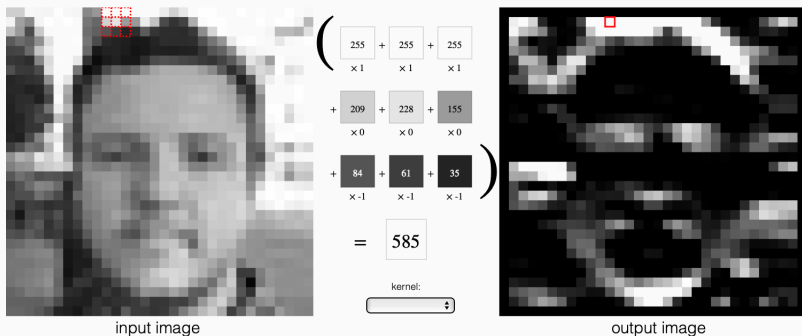
jelentés: a K konvolúciós kernel/filter/szűrő alkalmazása az I jelre/képre



Konvolúciós operátor, miéért?



https://brohrer.github.io/how_convolutional_neural_networks_work.html



JS demo: <https://setosa.io/ev/image-kernels/>

Padding, stride, channel, volume

- ▶ padding: az kép (rendszerint nullákkal való) szegélyezése
- ▶ stride: mennyivel toljuk el egy iterációban a kernelt
- ▶ channel: rendszerint R, G, B
- ▶ volume: a konvolúciós réteg (>2D) bemenete/kimenete

0	0	0	0	0	0	...
0	156	155	156	158	158	...
0	153	154	157	159	159	...
0	149	151	155	158	159	...
0	146	146	149	153	158	...
0	145	143	143	148	158	...
...

Input Channel #1 (Red)

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

Kernel Channel #1



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3D kernel

Input Volume (+pad 1) (7x7x3)

$x[:, :, 0]$

0	0	0	0	0	0	0
0	1	2	0	0	1	0
0	2	2	0	1	0	0
0	2	2	1	2	1	0
0	0	1	1	0	2	0
0	2	1	0	2	1	0
0	0	0	0	0	0	0

$x[:, :, 1]$

0	0	0	0	0	0	0
0	0	2	0	0	1	0
0	1	1	0	2	2	0
0	0	1	1	0	2	0
0	1	2	0	2	0	0
0	0	2	0	1	0	0
0	0	0	0	0	0	0

$x[:, :, 2]$

0	0	0	0	0	0	0
0	2	2	1	0	0	0
0	2	1	0	0	1	0
0	0	2	2	2	1	0
0	1	2	1	0	2	0
0	2	1	1	1	1	0
0	0	0	0	0	0	0

Filter W0 (3x3x3)

$w0[:, :, 0]$

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

$w0[:, :, 1]$

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

$w0[:, :, 2]$

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

Bias b0 (1x1x1)

$b0[:, :, 0]$

1

Filter W1 (3x3x3)

$w1[:, :, 0]$

0	0	1
1	1	0
0	0	0

$w1[:, :, 1]$

1	0	0
1	-1	1
-1	0	0

$w1[:, :, 2]$

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

Bias b1 (1x1x1)

$b1[:, :, 0]$

0

Output Volume (3x3x2)

$o[:, :, 0]$

-6	-7	-5
-9	-6	-9
3	-5	-8

$o[:, :, 1]$

2	3	-2
7	4	1
5	5	7

$$(K * I)(x, y) = \sum_{i=1}^h \sum_{j=1}^w \sum_{k=1}^d K(i, j, k) I(x+i, y+j, k)$$

3D kernel

Input Volume (+pad 1) (7x7x3)

x[:, :, 0]						
0	0	0	0	0	0	0
0	1	2	0	0	1	0
0	2	2	0	1	0	0
0	2	2	1	2	1	0
0	0	1	1	0	2	0
0	2	1	0	2	1	0
0	0	0	0	0	0	0
x[:, :, 1]						
0	0	0	0	0	0	0
0	0	2	0	0	1	0
0	1	1	0	2	2	0
0	0	1	1	0	2	0
0	1	2	0	2	0	0
0	0	2	0	1	0	0
0	0	0	0	0	0	0
x[:, :, 2]						
0	0	0	0	0	0	0
0	2	2	1	0	0	0
0	2	1	0	0	1	0
0	0	2	2	2	1	0
0	1	2	1	0	2	0
0	2	1	1	1	1	0
0	0	0	0	0	0	0

Filter W0 (3x3x3)

w0[:, :, 0]		
1	-1	-1
-1	0	0
-1	-1	0
w0[:, :, 1]		
-1	0	1
-1	0	0
-1	-1	-1
w0[:, :, 2]		
-1	-1	1
0	0	0
1	-1	-1
Bias b0 (1x1x1)		
b0[:, :, 0]		
1		

Filter W1 (3x3x3)

w1[:, :, 0]		
0	0	1
1	1	0
0	0	0
w1[:, :, 1]		
1	0	0
1	-1	1
-1	0	0
w1[:, :, 2]		
-1	1	0
0	-1	1
1	0	-1
Bias b1 (1x1x1)		
b1[:, :, 0]		
0		

Output Volume (3x3x2)

o[:, :, 0]		
-6	-7	-5
-9	-6	-9
3	-5	-8
o[:, :, 1]		
2	3	-2
7	4	1
5	5	7

$$(K * I)(x, y) = \sum_{i=1}^h \sum_{j=1}^w \sum_{k=1}^d K(i, j, k) I(x+i, y+j, k)$$

3D kernel

Input Volume (+pad 1) (7x7x3)

$x[:, :, 0]$

0	0	0	0	0	0	0
0	1	2	0	0	1	0
0	2	2	0	1	0	0
0	2	2	1	2	1	0
0	0	1	1	0	2	0
0	2	1	0	2	1	0
0	0	0	0	0	0	0

$x[:, :, 1]$

0	0	0	0	0	0	0
0	0	2	0	0	1	0
0	1	1	0	2	2	0
0	0	1	1	0	2	0
0	1	2	0	2	0	0
0	0	2	0	1	0	0
0	0	0	0	0	0	0

$x[:, :, 2]$

0	0	0	0	0	0	0
0	2	2	1	0	0	0
0	2	1	0	0	1	0
0	0	2	2	2	1	0
0	1	2	1	0	2	0
0	2	1	1	1	1	0
0	0	0	0	0	0	0

Filter W0 (3x3x3)

$w0[:, :, 0]$

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

$w0[:, :, 1]$

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

$w0[:, :, 2]$

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

Bias b0 (1x1x1)

$b0[:, :, 0]$

1

Filter W1 (3x3x3)

$w1[:, :, 0]$

0	0	1
1	1	0
0	0	0

$w1[:, :, 1]$

1	0	0
1	-1	1
-1	0	0

$w1[:, :, 2]$

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

Bias b1 (1x1x1)

$b1[:, :, 0]$

0

Output Volume (3x3x2)

$o[:, :, 0]$

-6	-7	-5
-9	-6	-9
3	-5	-8

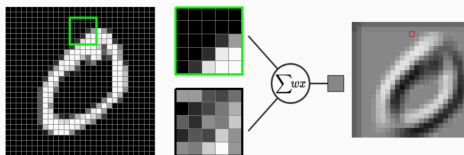
$o[:, :, 1]$

2	3	-2
7	4	1
5	5	7

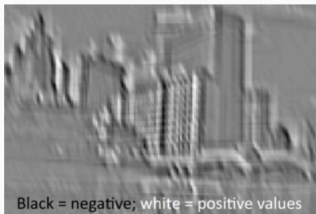
$$(K * I)(x, y) = \sum_{i=1}^h \sum_{j=1}^w \sum_{k=1}^d K(i, j, k) I(x+i, y+j, k)$$

Konvolúciós réteg

- ▶ cél: térbeli jellemzők kinyerése
- ▶ bemenet: $W_1 \times H_1 \times D_1$
- ▶ kernel: $F \times F \times K$
 - ▶ hiperparaméter: stride (S)
 - ▶ hiperparaméter: padding (P)
- ▶ kimenet: $W_2 \times H_2 \times D_2$
 - ▶ $W_2 = (W_1 - F + 2P) / S + 1$
 - ▶ $H_2 = (H_1 - F + 2P) / S + 1$
 - ▶ $D_2 = K$
- ▶ a kernelek a megtanulandó paraméterek közé tartoznak
- ▶ alkalmazás: kép, hang, idősor, stb feldolgozása

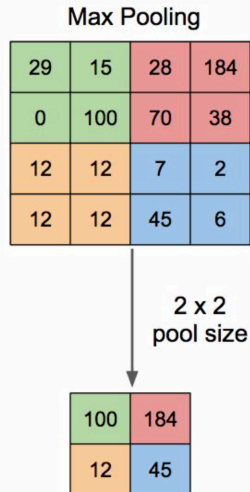


- ▶ cél: nonlinearitás
- ▶ $ReLU(x) = \max(0, x)$
- ▶ alternatívák: LeakyReLU, Param ReLU, Rand ReLU, exponential linear unit, stb.
- ▶ a nonlináris réteg kimenete: activation map

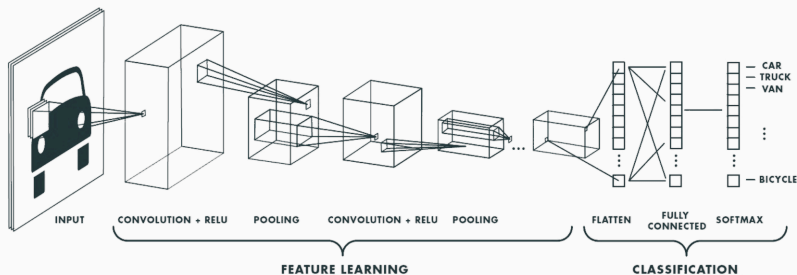


Mintavételező réteg

- ▶ cél: a bemenet méretének csökkentése
- ▶ bemenet: $W_1 \times H_1 \times D_1$
- ▶ subsampling/pooling:
 - ▶ hiperparaméter: méret ($F \times F$)
 - ▶ hiperparaméter: stride (S)
 - ▶ pooling függvény: max, vagy avg
- ▶ kimenet: $W_2 \times H_2 \times D_2$
 - ▶ $W_2 = (W_1 - F) / S + 1$
 - ▶ $H_2 = (H_1 - F) / S + 1$
 - ▶ $D_2 = D_1$



Generikus CNN



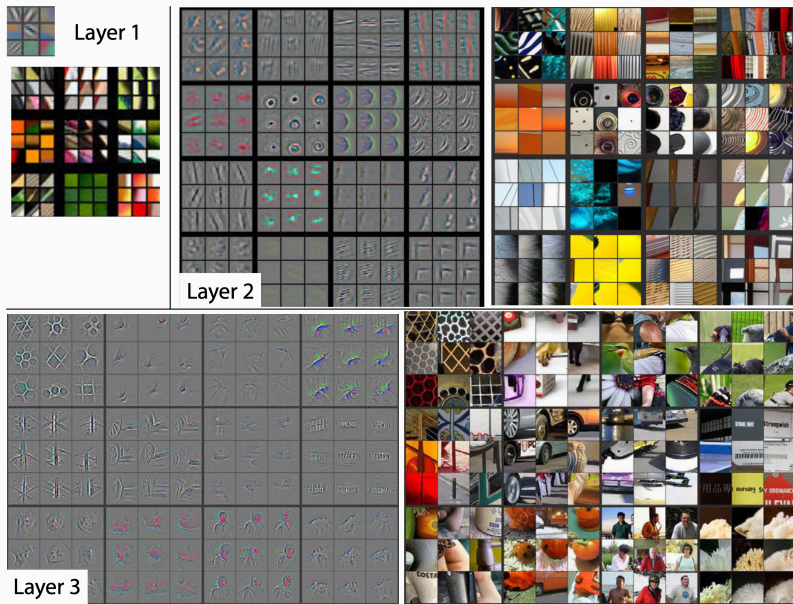
Demo:

<https://cs.stanford.edu/~karpthy/convnetjs/demo/mnist.html>

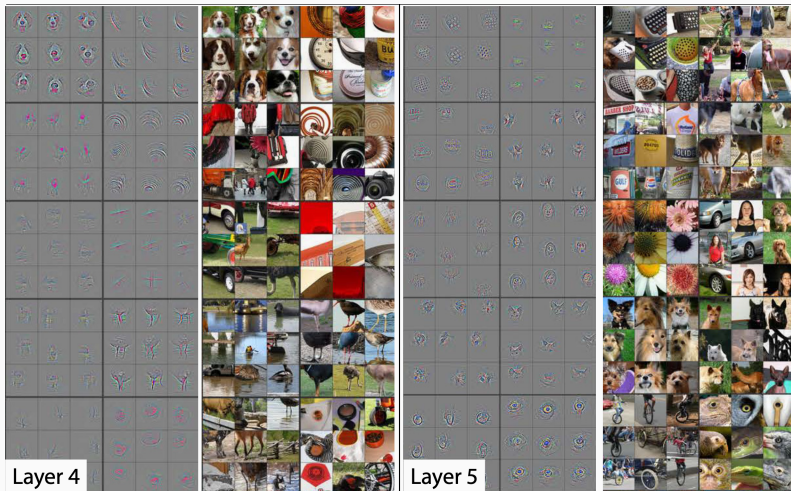
<https://www.cs.ryerson.ca/~aharley/vis/conv/flat.html>

<https://www.mathworks.com/help/deeplearning/ug/create-simple-deep-learning-network-for-classification.html>

A konvolúciós rétegek vizualizációja

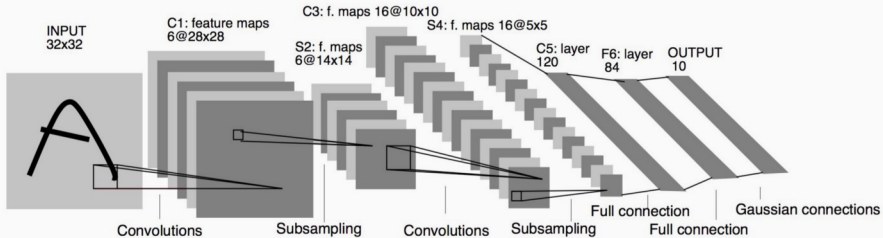


A konvolúciós rétegek vizualizációja



<https://cs.nyu.edu/~fergus/papers/zeilerECCV2014.pdf>

Architektúrák: LeNet

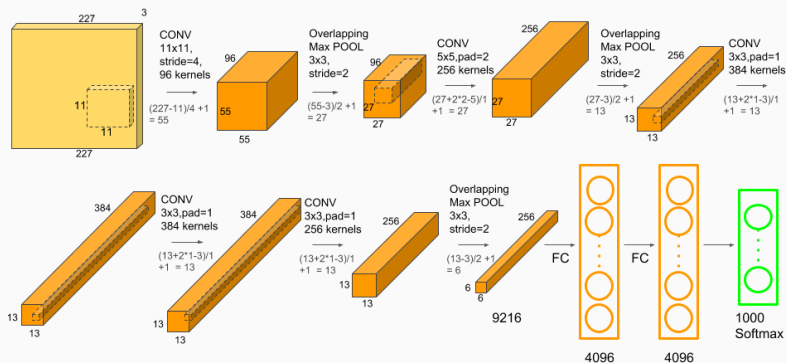


<http://yann.lecun.com/exdb/publis/pdf/lecun-01a.pdf>

- ▶ Yan LeCun, 1989: konvolúciós rétegek + backpropagation
- ▶ irányítósámok, banki csekkek feldolgoása

Architektúrák: AlexNet

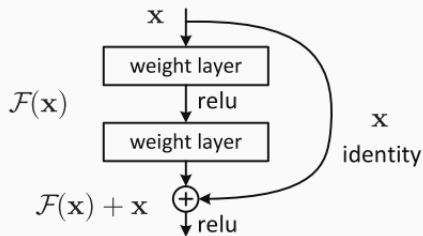
- ▶ IMAGENET Large Scale Visual Recognition Challenge, <http://image-net.org/challenges/LSVRC/2012/>
- ▶ GPU + sok adat + backprop
- ▶ Alex Krizhevsky, 2012, AlexNet, $\approx 60\text{M}$ paraméter, 6 nap tanítás 2 drb GTX 580 3GB GPU



<https://www.learnopencv.com/understanding-alexnet/>

demo: objclass / matlab

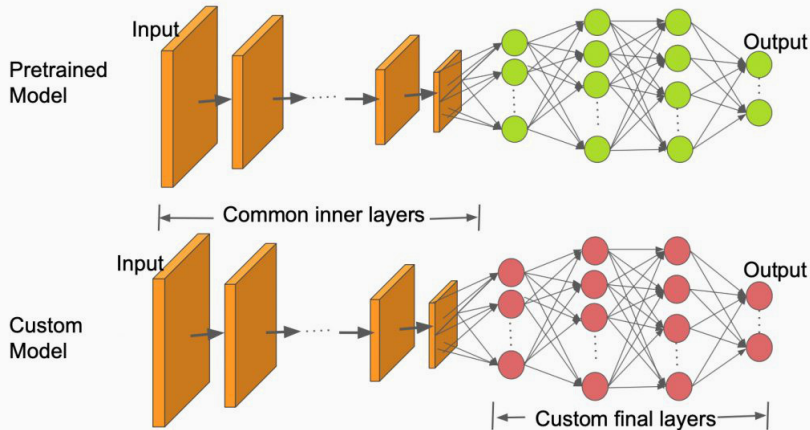
- ▶ a rétegek egyszerű felfűzése nem vezet feltétlenül jobb eredményekhez
- ▶ residual block:



<https://towardsdatascience.com/an-overview-of-resnet-and-its-variants-5281e2f56035>

- ▶ "konvexebbé" teszi a hibafüggvényt <https://arxiv.org/pdf/1712.09913.pdf>

Transfer learning



<https://www.learnopencv.com/image-classification-using-transfer-learning-in-pytorch/>

Python demo:

<https://github.com/fastai/course-v3/blob/master/nbs/dl1/lesson1-pets.ipynb>

- ▶ Python: Pytorch (Facebook)
- ▶ Python: Tensorflow (Google)
- ▶ Matlab: Deep Learning Toolbox (Mathworks)
- ▶ https://en.wikipedia.org/wiki/Comparison_of_deep-learning_software
- ▶ átjárás: Open Neural Network Exchange (ONNX) formátum