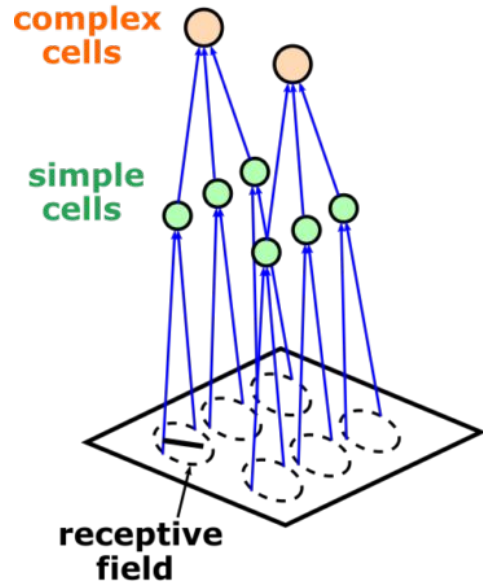
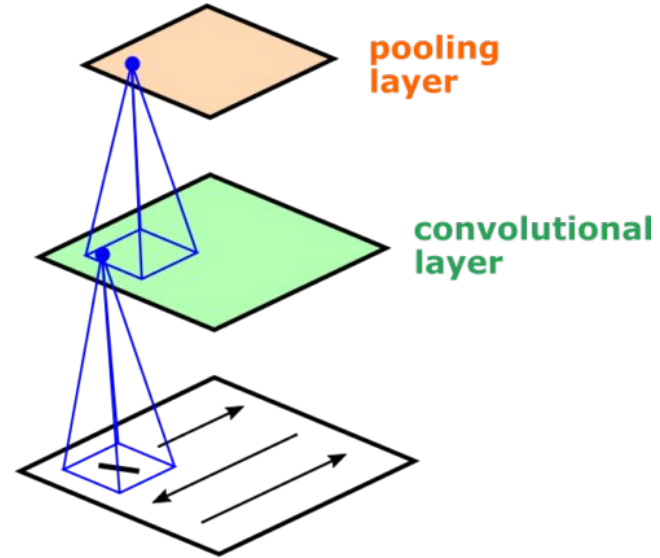


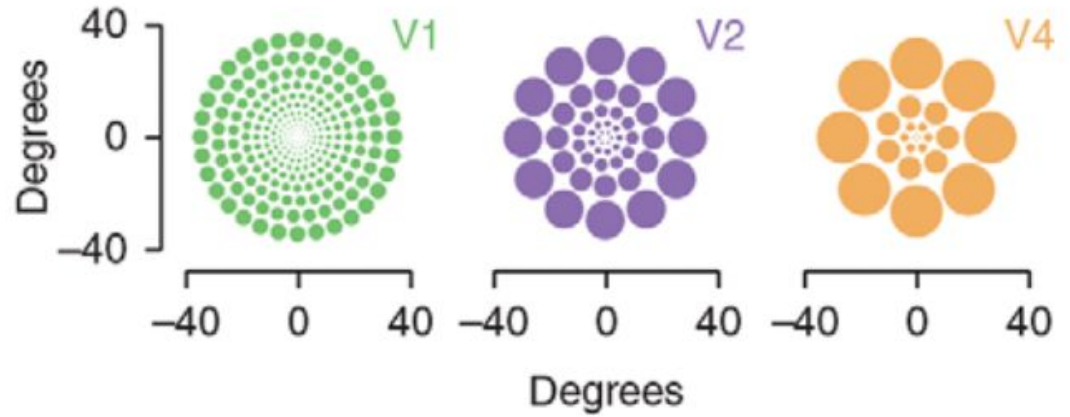
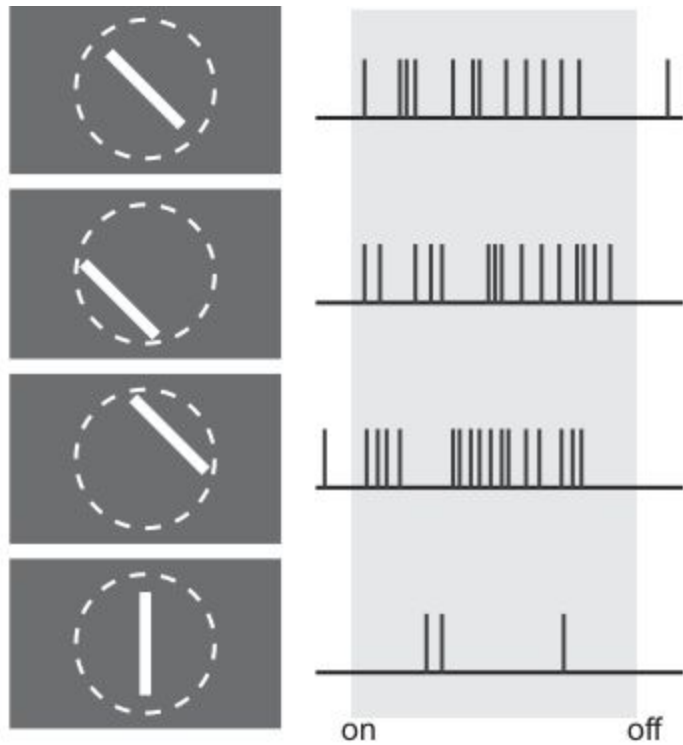
From real neurons to computational units.

PRIMARY VISUAL CORTX



CONVOLUTIONAL NEURAL NETWORK





Invariant visual representation by single neurons in the human brain

R. Quian Quiroga^{1,2,†}, L. Reddy¹, G. Kreiman³, C. Koch¹ & I. Fried²

It takes a fraction of a second to recognize a person or an object even when seen under strikingly different conditions. How such a robust, high-level representation is achieved by neurons in the human brain is still unclear¹⁻⁶. In monkeys, neurons in the upper stages of the ventral visual pathway respond to complex images such as faces and objects and show some degree of invariance to metric properties such as the stimulus size, position and viewing angle^{2,4,7-12}. We have previously shown that neurons in the human medial temporal lobe (MTL) fire selectively to images of faces, animals, objects or scenes^{13,14}. Here we report on a remarkable

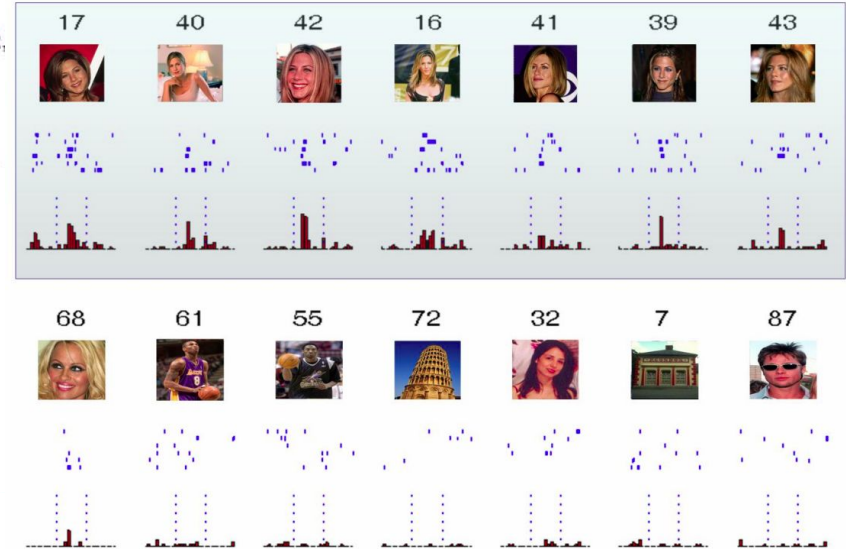
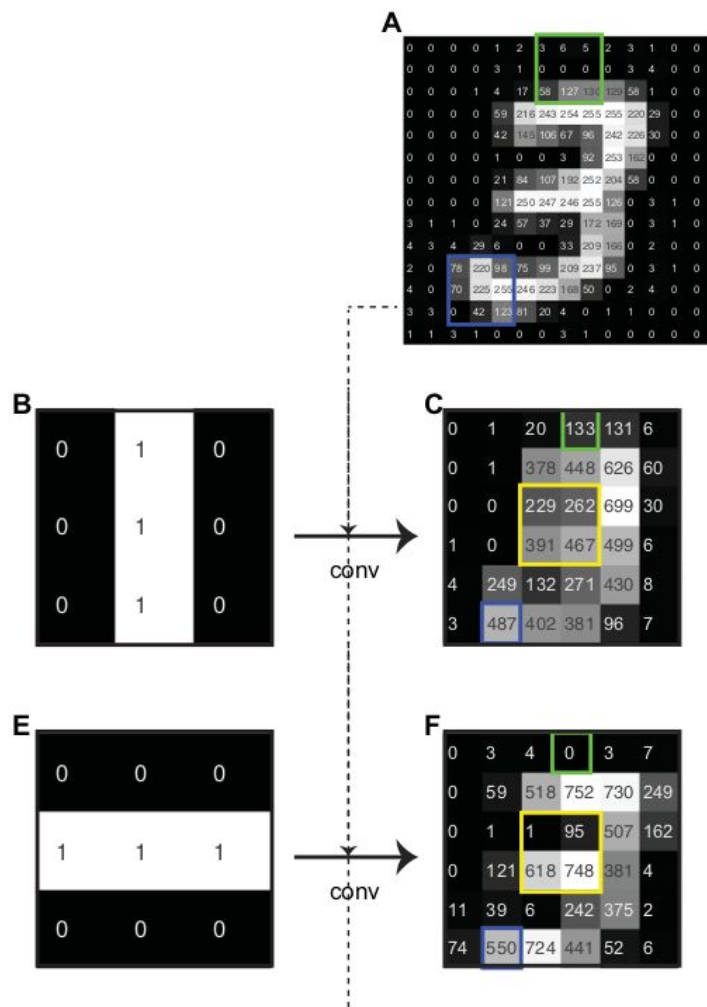
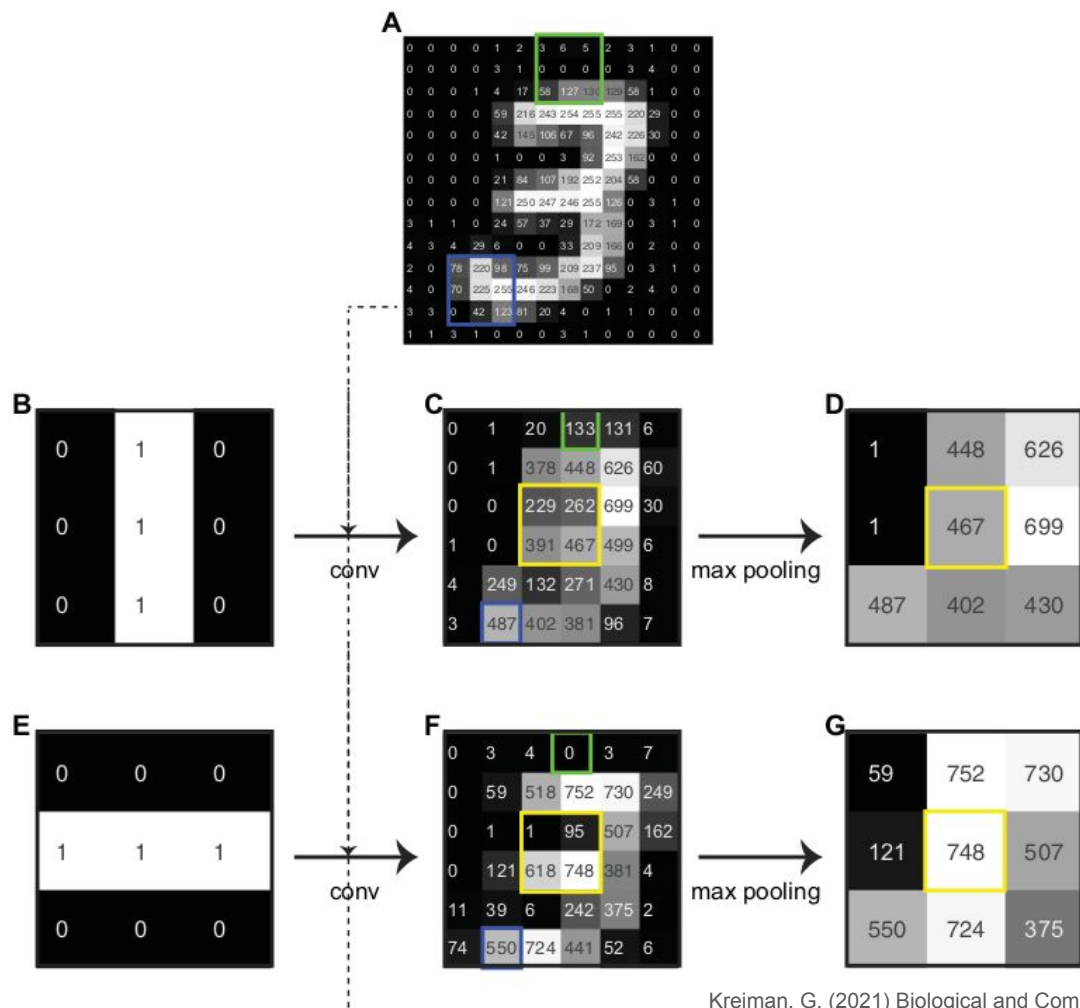


Figure 1 | A single unit in the left posterior hippocampus activated exclusively by different views of the actress Jennifer Aniston.



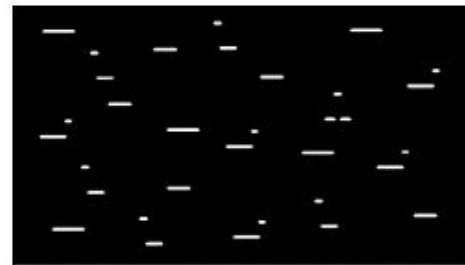
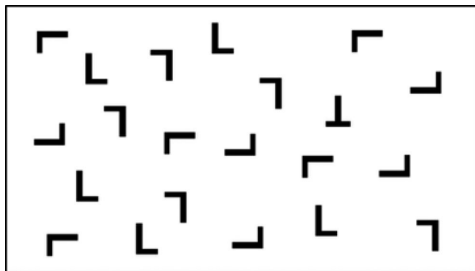


Kernel 1

$$* \begin{bmatrix} 1/4 & 0 & -1/4 \\ 1/4 & 0 & -1/4 \\ 1/4 & 0 & -1/4 \end{bmatrix}$$

Kernel 2

$$* \begin{bmatrix} -1/4 & -1/4 & -1/4 \\ 0 & 0 & 0 \\ 1/4 & 1/4 & 1/4 \end{bmatrix}$$



Imagen



Feature map 1



Feature map 2

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)

0	0	0	0	0	0	...
0	167	166	167	169	169	...
0	164	165	168	170	170	...
0	160	162	166	169	170	...
0	156	156	159	163	168	...
0	155	153	153	158	168	...
...

Input Channel #2 (Green)

0	0	0	0	0	0	...
0	163	162	163	165	165	...
0	160	161	164	166	166	...
0	156	158	162	165	166	...
0	155	155	158	162	167	...
0	154	152	152	157	167	...
...

Input Channel #3 (Blue)

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

Kernel Channel #1



308

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

Kernel Channel #2



-498

0	1	1
0	1	0
1	-1	1

Kernel Channel #3



164

+

+

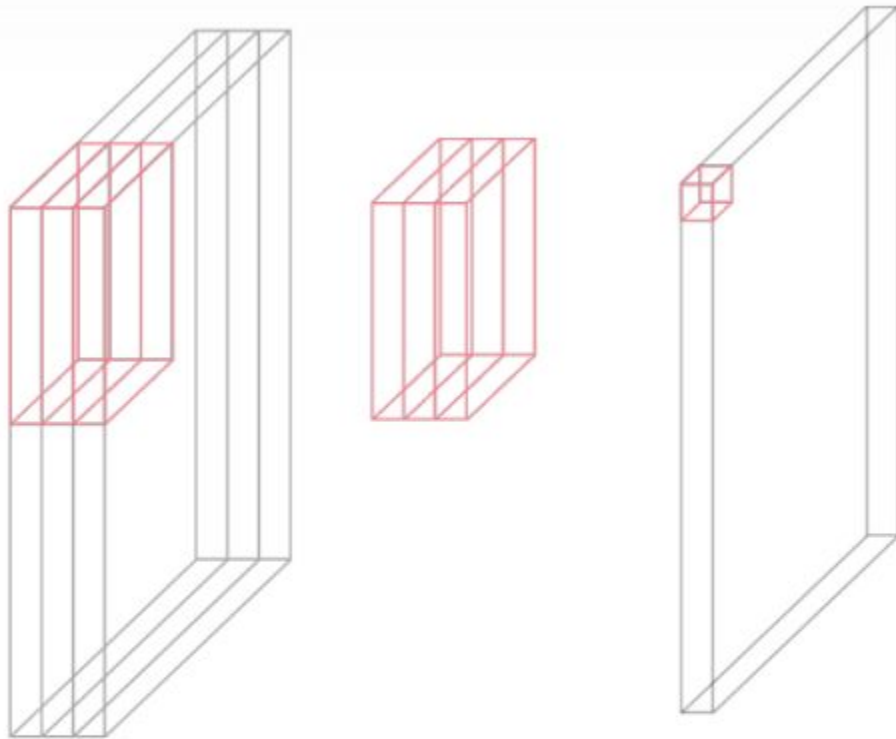


Bias = 1

+ 1 = -25

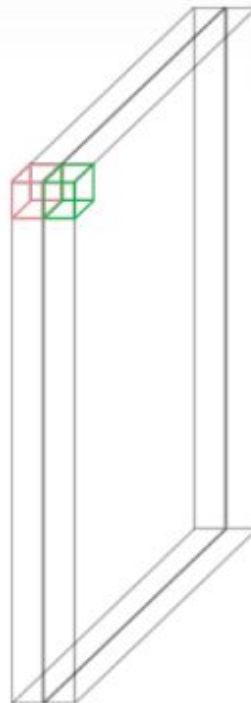
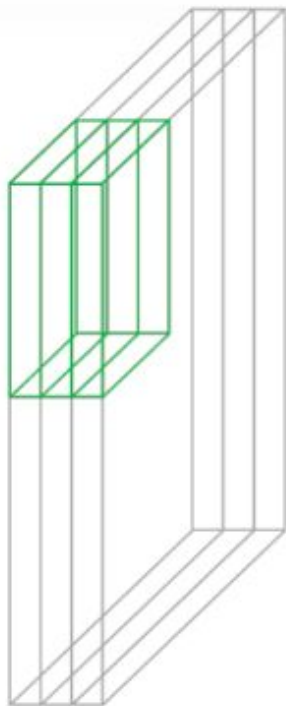
Output

-25				...
				...
				...
				...
...



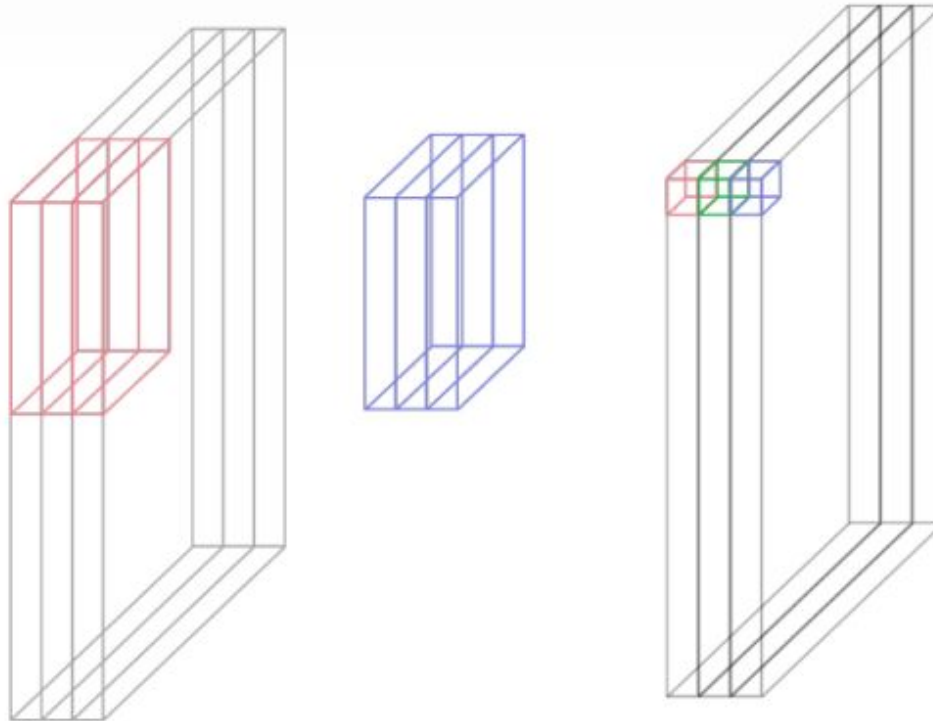
Parámetros a aprender en la capa





Parámetros a aprender en la capa



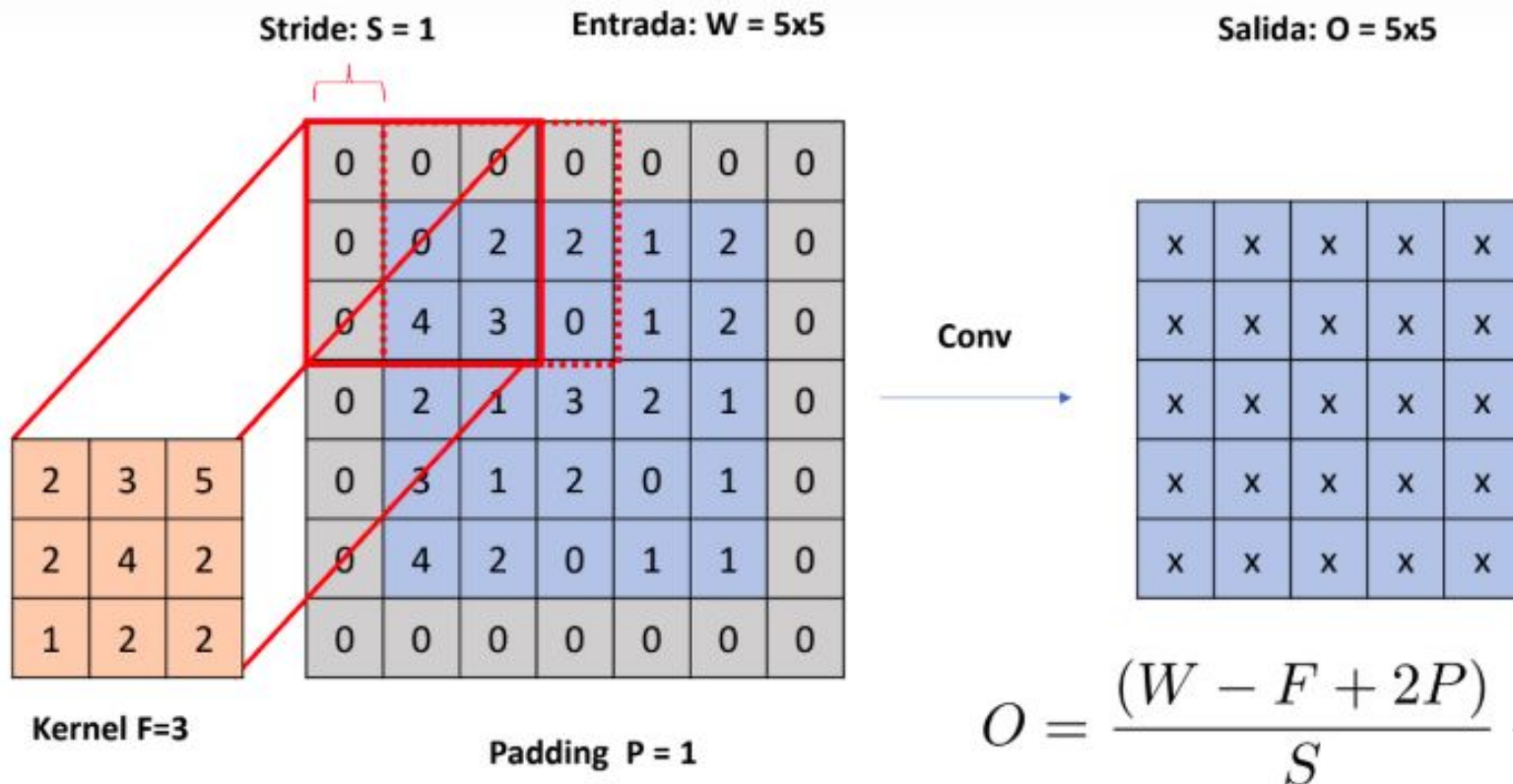


Parámetros a aprender en la capa



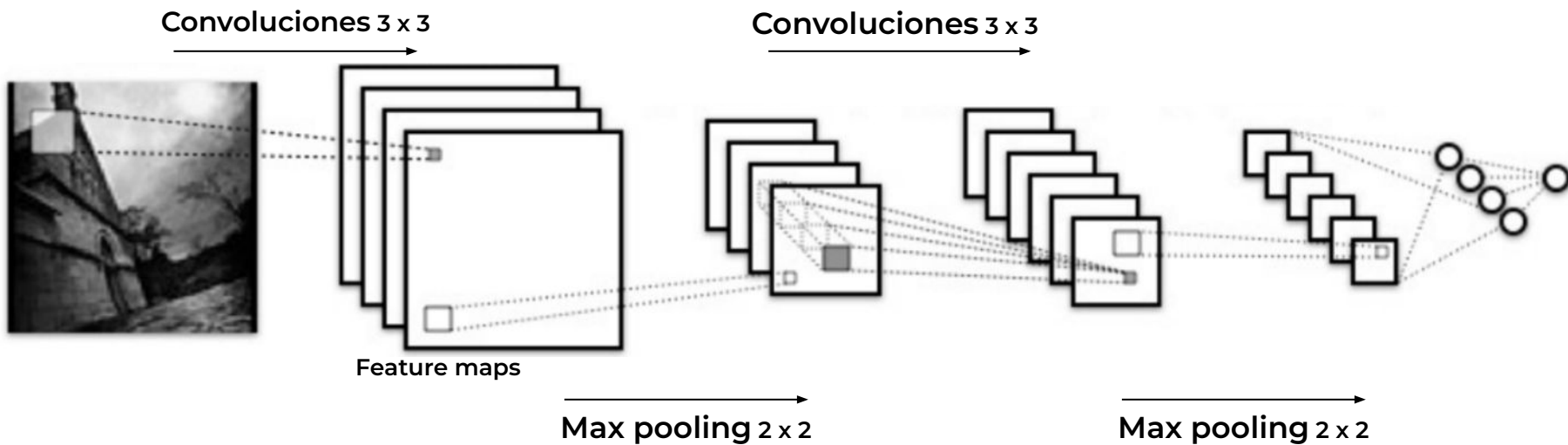
Hiperparámetros de una capa convolucional

Relación entrada/salida



$$O = \frac{(W - F + 2P)}{S} + 1$$

Estructura de una CNN

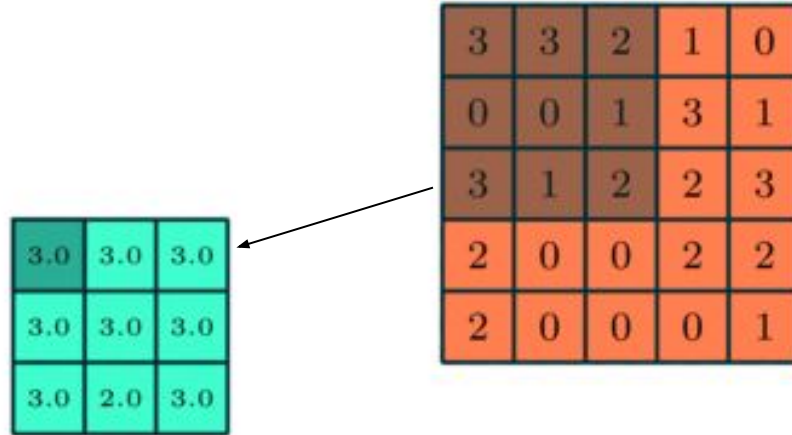


12	20	30	0
8	12	2	0
34	70	37	4
112	100	25	12

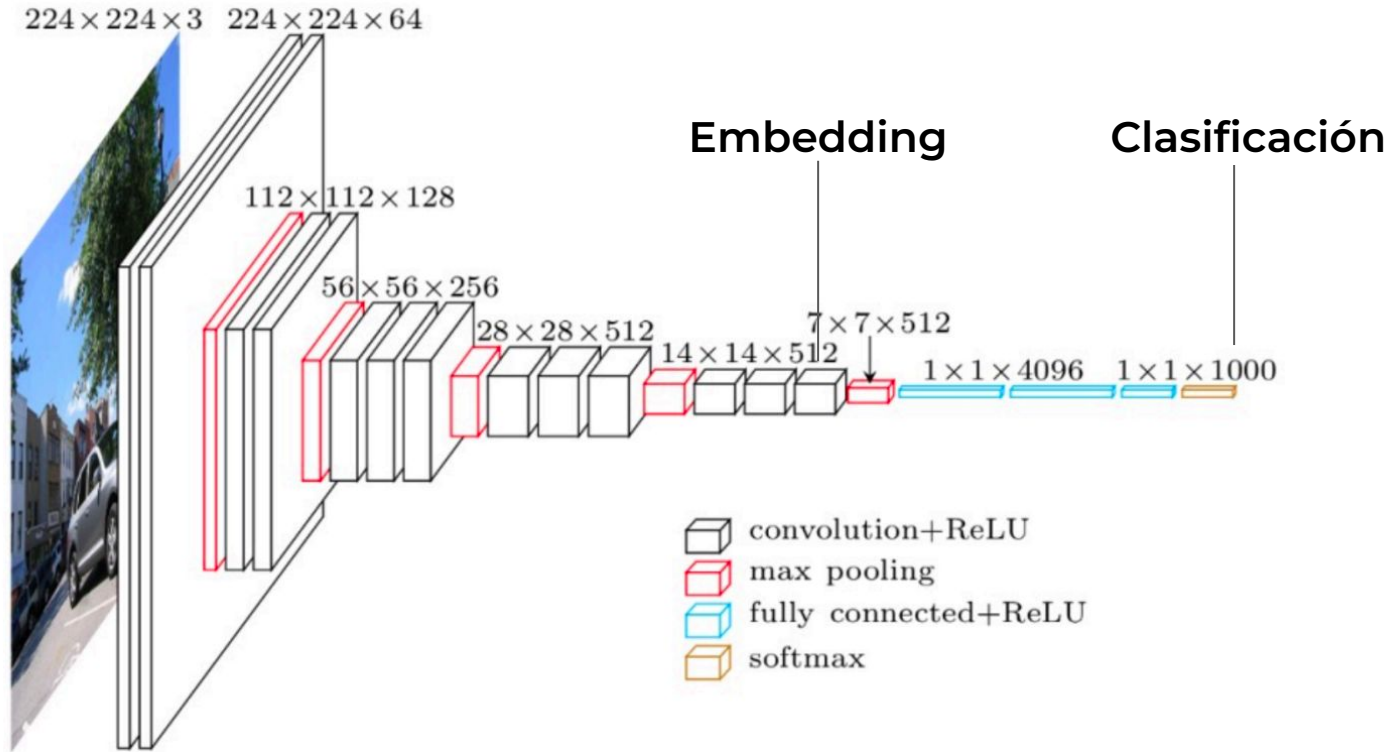
Max pooling 2 x 2

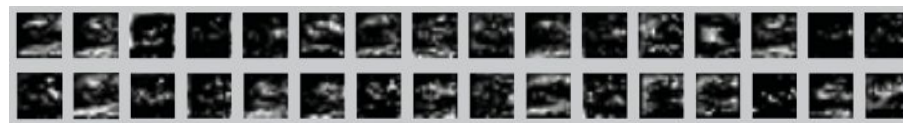
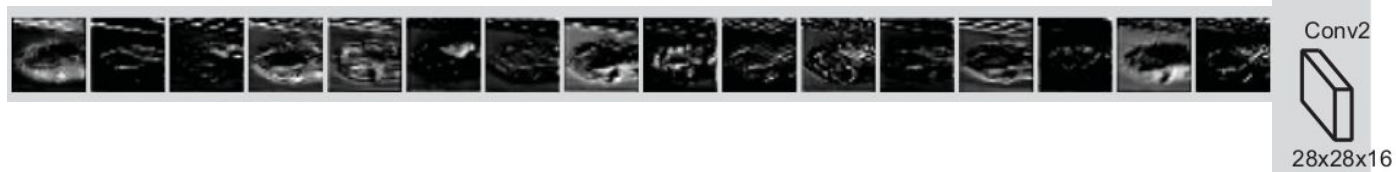
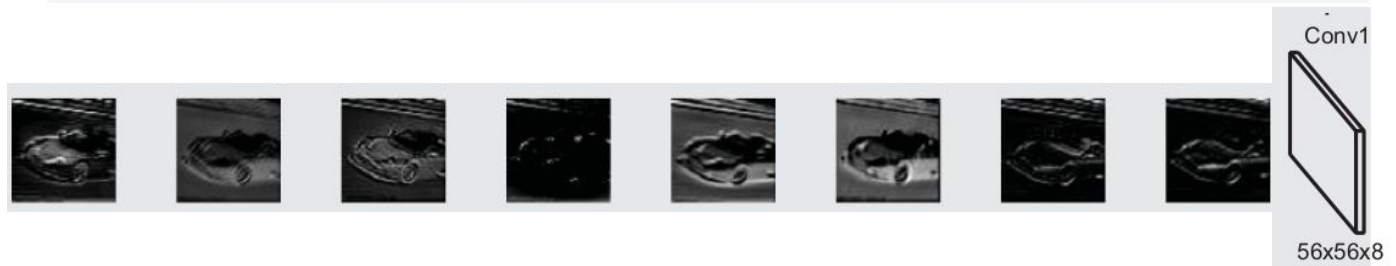
20	30
112	37

Max pooling



Arquitectura de VGG-16





980: volcano



344: hippopotamus



538: dome



380: African elephant



837: sunglasses



340: zebra

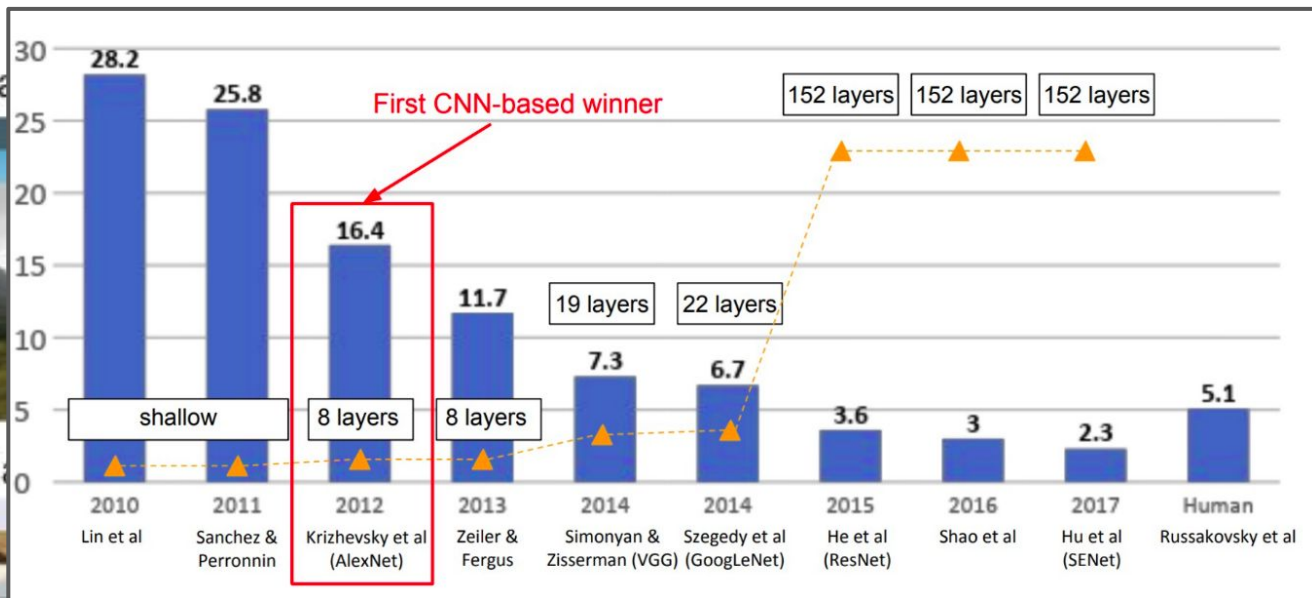


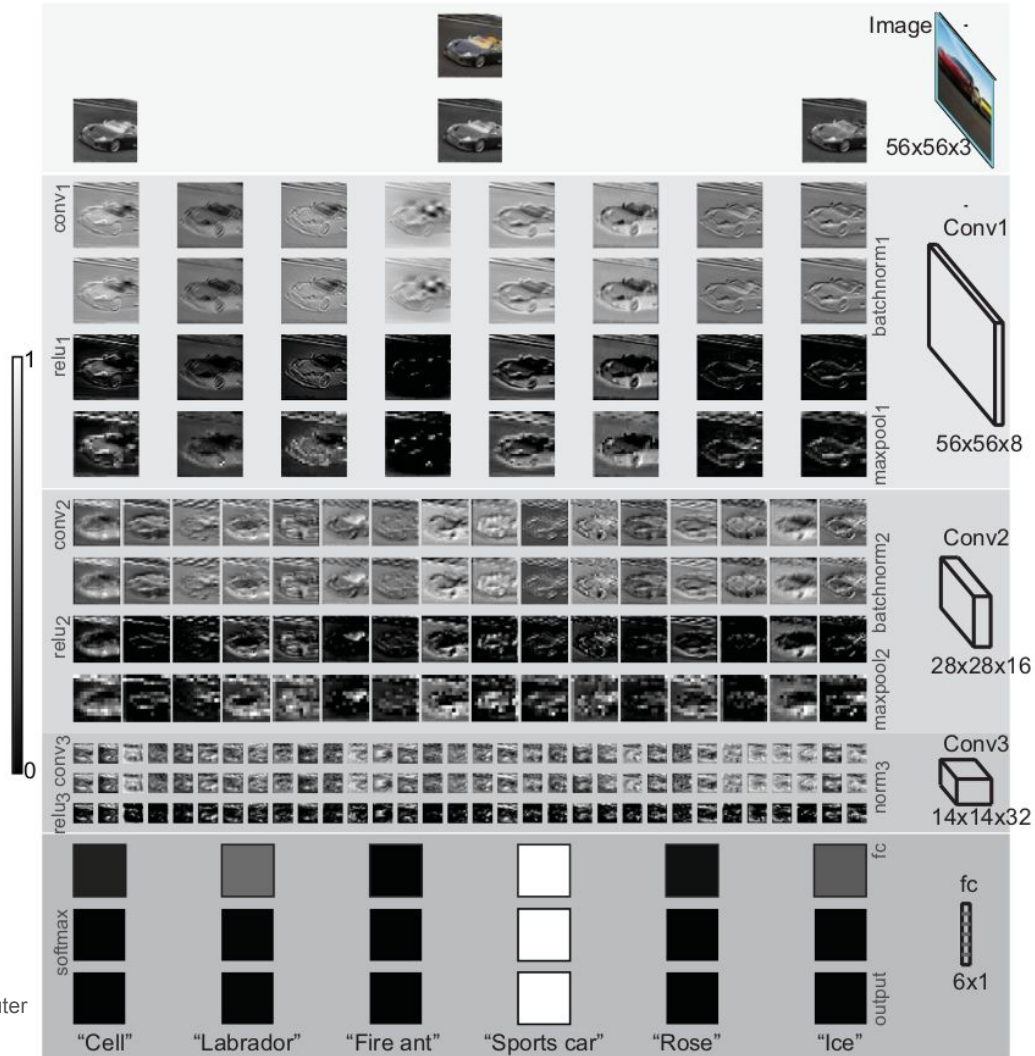
372: baboon



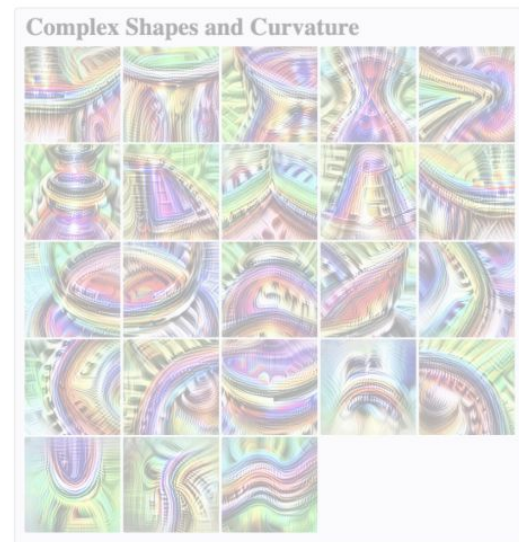
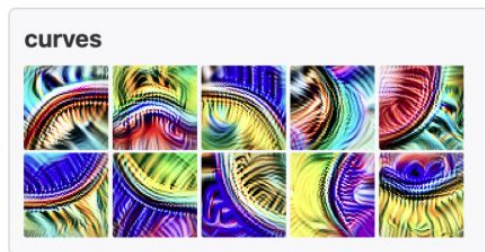
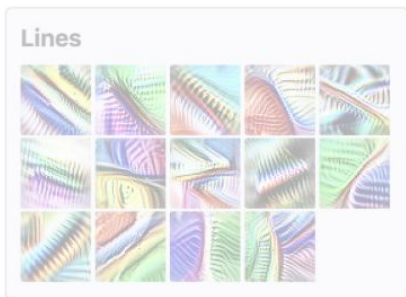
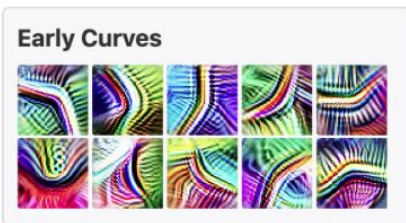
483: castle







Kreiman, G. (2021) Biological and Computer Vision. Cambridge University Press.



Windows (4b:237)
excite the car detector
at the top and inhibit
at the bottom.



Car Body (4b:491)
excites the car
detector, especially at
the bottom.



Wheels (4b:373) excite
the car detector at the
bottom and inhibit at
the top.



● positive (excitation)
● negative (inhibition)



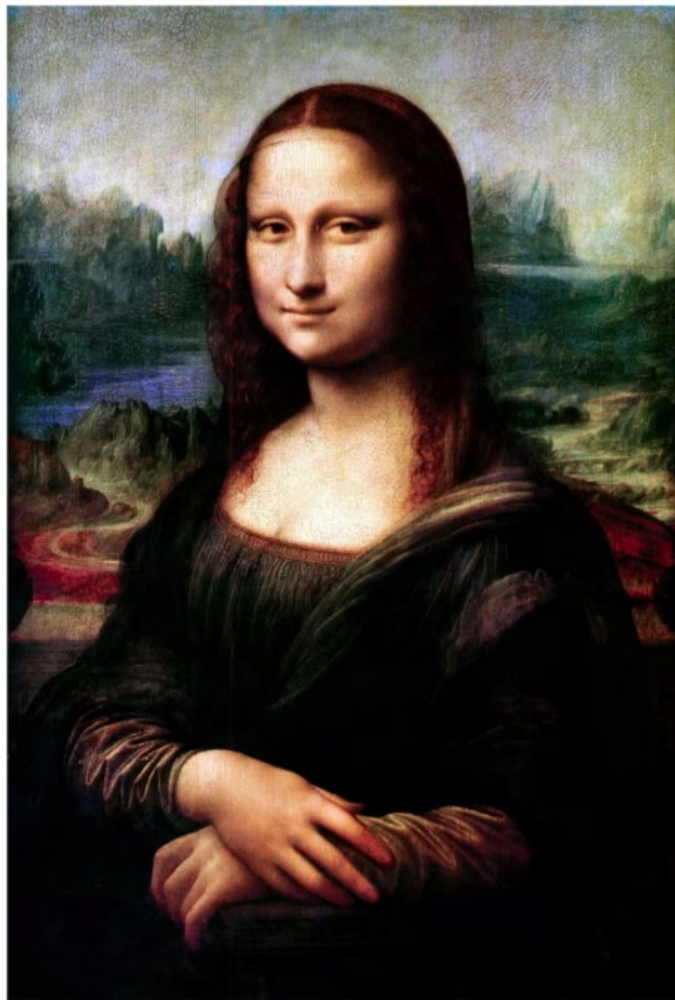
A car detector (4c:447)
is assembled from
earlier units.

```
1 import torch
2 import torch.nn as nn
3 import torch.nn.functional as F
4 from torchvision import datasets, transforms
5 from torch.utils.data import DataLoader
6
7 class CNN(nn.Module):
8     def __init__(self):
9         super().__init__()
10        self.conv1 = nn.Conv2d(3, 32, 3, padding=1)
11        self.conv2 = nn.Conv2d(32, 64, 3, padding=1)
12        self.pool = nn.MaxPool2d(2)
13        self.fc1 = nn.Linear(64 * 8 * 8, 128)
14        self.fc2 = nn.Linear(128, 10)
15    def forward(self, x):
16        x = self.pool(F.relu(self.conv1(x)))
17        x = self.pool(F.relu(self.conv2(x)))
18        x = x.view(x.size(0), -1)
19        x = F.relu(self.fc1(x))
20        return self.fc2(x)
21
```

NORMAL test.py [+]

utf-8 | unix | python 11%

6:1



Menor nivel de detalle



Alto nivel de detalle