



Centro Brasileiro de Pesquisas Físicas



Inteligência artificial e aplicações em física

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clearnightsrthebest.com



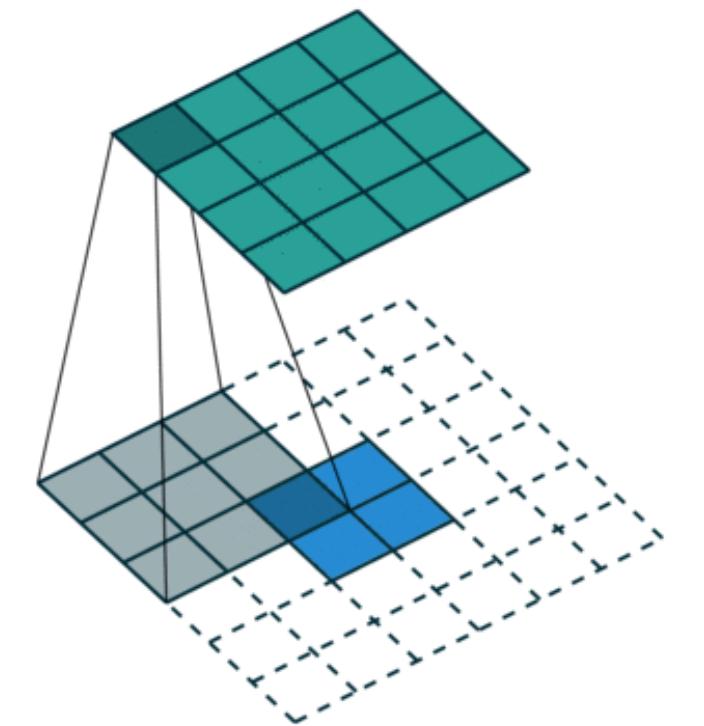
Conv Layers

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

Pixel representation of filter



Visualization of a curve detector filter



Conv Layers



Visualization of the filter on the image

0	0	0	0	0	0	0
0	40	0	0	0	0	0
40	0	40	0	0	0	0
40	20	0	0	0	0	0
0	50	0	0	0	0	0
0	0	50	0	0	0	0
25	25	0	50	0	0	0

Pixel representation of receptive field

*

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

Pixel representation of filter

Multiplication and Summation = 0



Visualization of the receptive field

0	0	0	0	0	0	30
0	0	0	0	50	50	50
0	0	0	20	50	0	0
0	0	0	50	50	0	0
0	0	0	50	50	0	0
0	0	0	50	50	0	0
0	0	0	50	50	0	0

Pixel representation of the receptive field

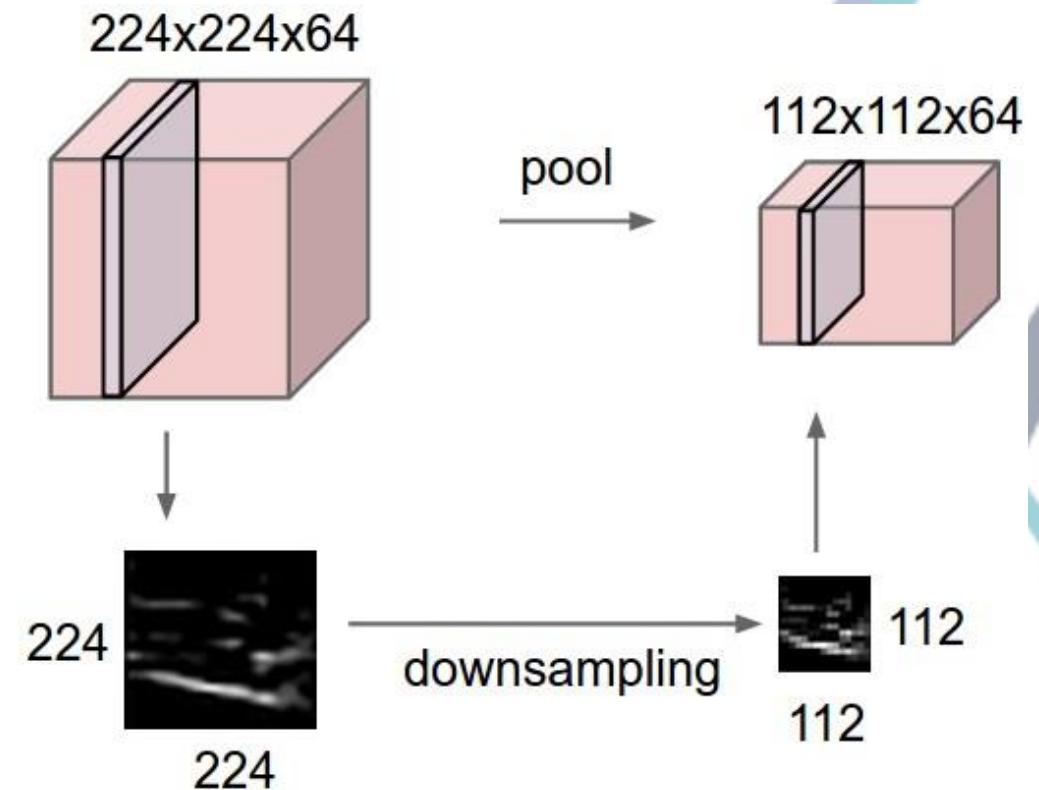
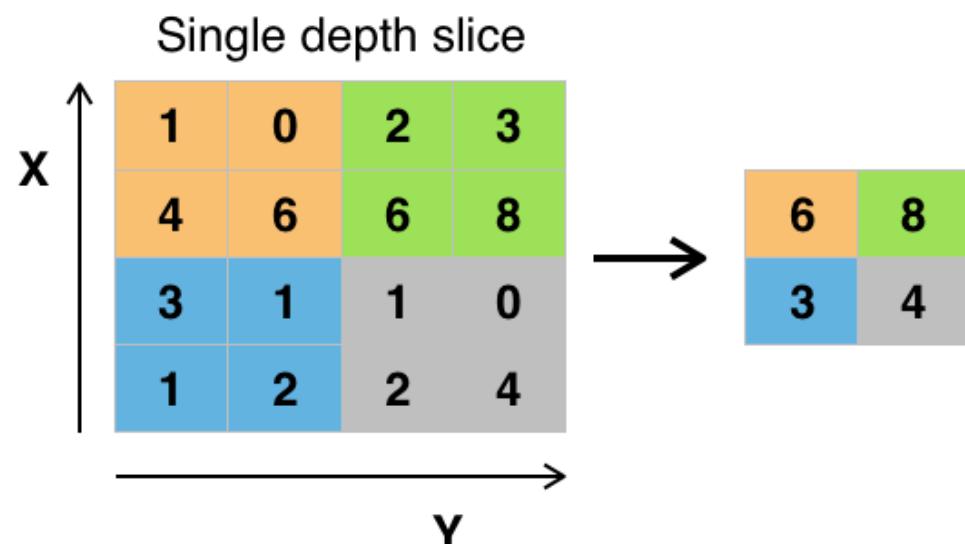
*

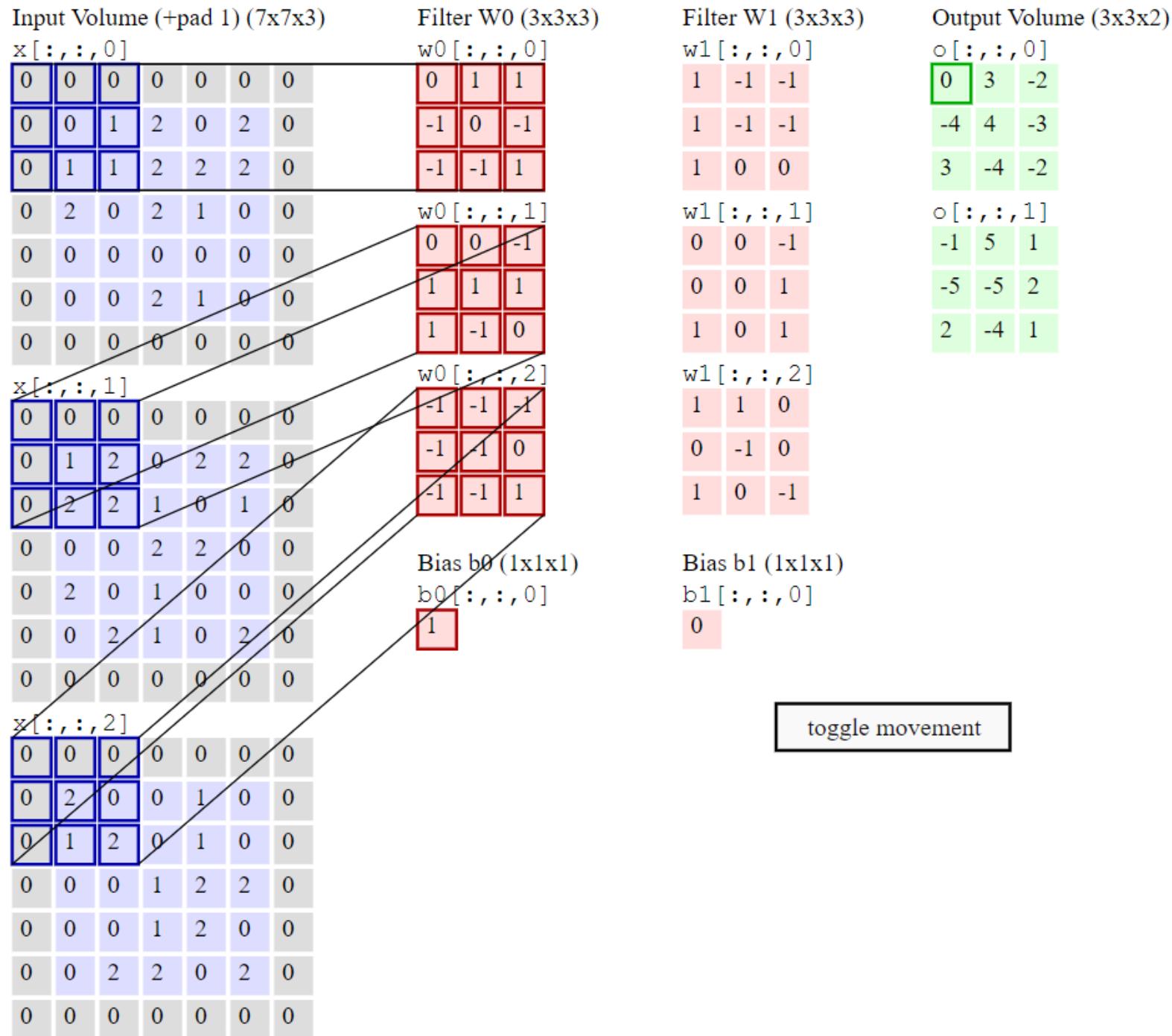
0	0	0	0	0	0	30	0
0	0	0	0	0	30	0	0
0	0	0	30	0	0	0	0
0	0	0	30	0	0	0	0
0	0	0	30	0	0	0	0
0	0	0	30	0	0	0	0
0	0	0	0	0	0	0	0

Pixel representation of filter

Multiplication and Summation = $(50*30)+(50*30)+(50*30)+(20*30)+(50*30) = 6600$ (A large number!)

Pooling





Input Volume (+pad 1) (7x7x3)

 $x[:, :, 0]$

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

Filter W0 (3x3x3)

 $w0[:, :, 0]$

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

 $w0[:, :, 1]$

0	0	-1
1	1	1
1	-1	0

 $w0[:, :, 2]$

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

Bias b0 (1x1x1)
 $b0[:, :, 0]$

1

 $x[:, :, 1]$

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

 $x[:, :, 2]$

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

Filter W1 (3x3x3)

 $w1[:, :, 0]$

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

 $w1[:, :, 1]$

0	0	-1
0	0	1
1	0	1

 $w1[:, :, 2]$

1	1	0
0	-1	0
1	0	-1

Bias b1 (1x1x1)

 $b1[:, :, 0]$

0

Output Volume (3x3x2)

 $o[:, :, 0]$

0	3	-2
-4	4	-3
3	-4	-2

 $o[:, :, 1]$

-1	5	1
-5	-5	2
2	-4	1

toggle movement

Input Volume (+pad 1) (7x7x3)

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

Filter W0 (3x3x3)

w0 [:, :, 0]
0 1 1
-1 0 -1
-1 -1 1
0 0 -1
1 1 1
1 -1 0
-1 -1 -1
-1 -1 0
-1 -1 1

Filter W1 (3x3x3)

w1 [:, :, 0]
1 -1 -1
1 -1 -1
1 0 0
0 0 -1
0 0 1
0 0 1
1 1 0
1 0 1
1 0 -1

Output Volume (3x3x2)

o [:, :, 0]
0 3 -2
-4 4 -3
3 -4 -2
o [:, :, 1]
-1 5 1
-5 -5 2
2 -4 1

x [:, :, 1]

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

w0 [:, :, 1]

w0 [:, :, 1]
-1 -1 -1
-1 -1 0
-1 -1 1

w1 [:, :, 1]

w1 [:, :, 1]
1 1 0
0 -1 0
1 0 -1

x [:, :, 2]

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

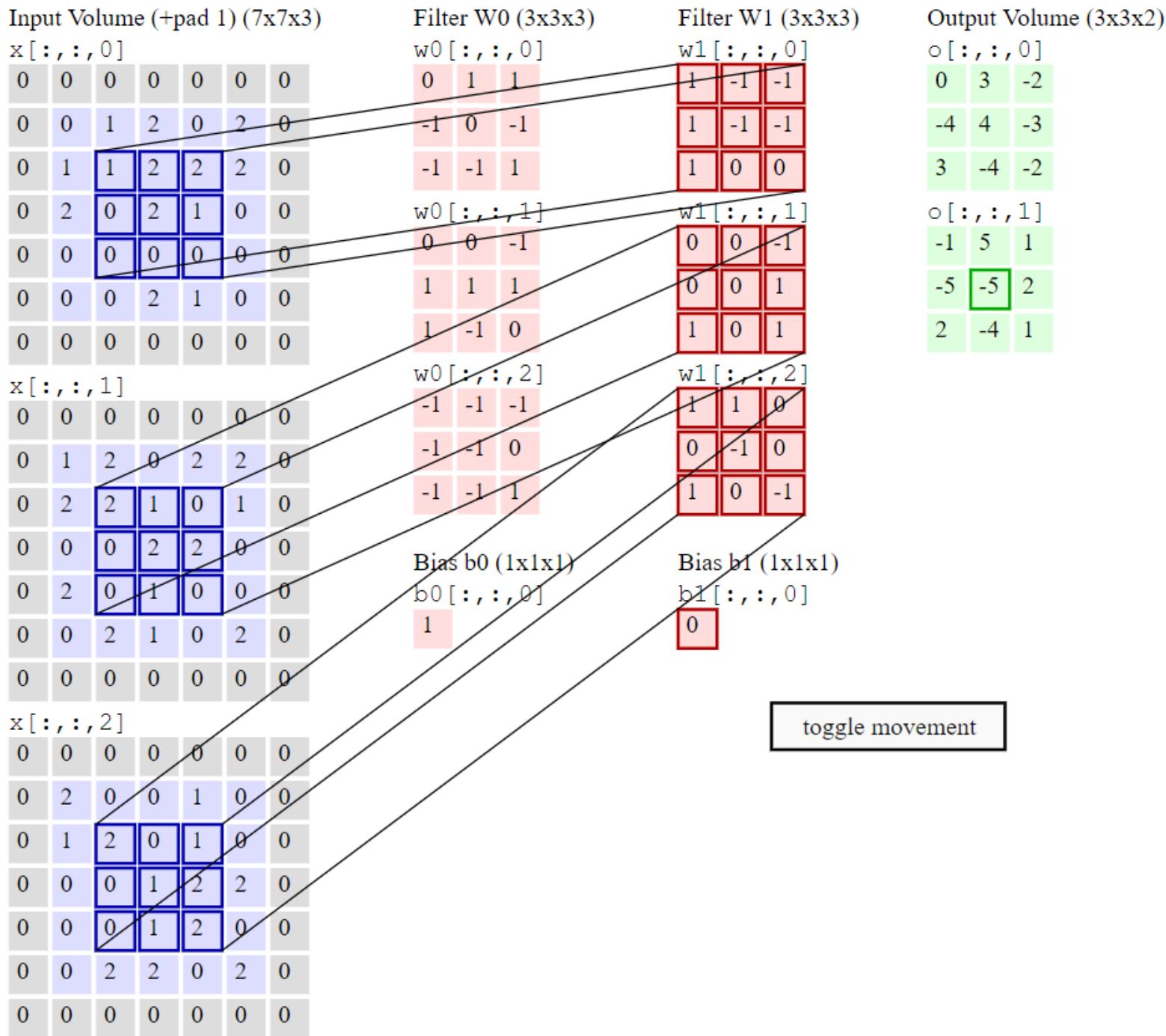
Bias b0 (1x1x1)

b0 [:, :, 0]
1

Bias b1 (1x1x1)

b1 [:, :, 0]
0

toggle movement



Input Volume (+pad 1) (7x7x3)

 $x[:, :, 0]$

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

Filter W0 (3x3x3)

 $w0[:, :, 0]$

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

Filter W1 (3x3x3)

 $w1[:, :, 0]$

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

Output Volume (3x3x2)

 $o[:, :, 0]$

0	3	-2
-4	4	-3
3	-4	-2

 $o[:, :, 1]$

-1	5	1
-5	-5	2

2	-4	1
---	----	---

 $x[:, :, 1]$

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

 $w0[:, :, 2]$

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

 $w1[:, :, 2]$

1	1	0
0	-1	0
1	0	-1

Bias b0 (1x1x1)

 $b0[:, :, 0]$

1

Bias b1 (1x1x1)

 $b1[:, :, 0]$

0

 $x[:, :, 2]$

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

toggle movement

The simplest example I know

```
from tensorflow.keras import models
from tensorflow.keras.models import Sequential
from tensorflow.keras import layers
from tensorflow.keras.layers import Convolution2D
from tensorflow.keras.layers import MaxPooling2D
from tensorflow.keras.layers import Dense, Dropout, Flatten, Activation

model = Sequential()
model.add(Conv2D(32, kernel_size=(5, 5), strides=(1, 1),
                activation='relu',
                input_shape=input_shape))
model.add(MaxPooling2D(pool_size=(2, 2), strides=(2, 2)))
model.add(Conv2D(64, (5, 5), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dense(1000, activation='relu'))
model.add(Dense(num_classes, activation='softmax'))
```

The simplest example I know

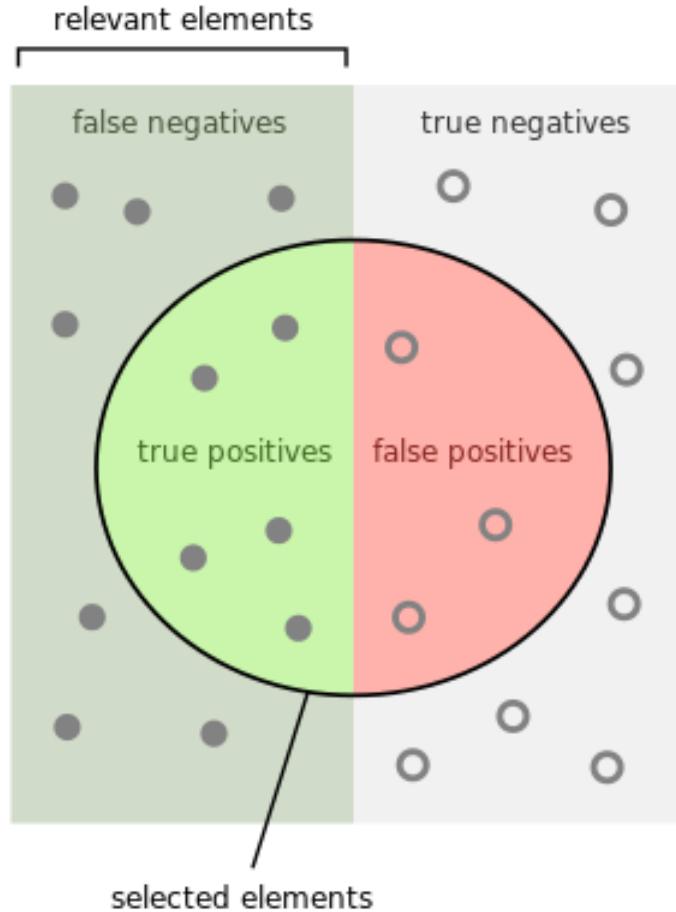
```
batch_size = 128
num_classes = 10
epochs = 10

# input image dimensions
img_x, img_y = 28, 28

# load the MNIST data set, which already splits into train and test sets
# for us
(x_train, y_train), (x_test, y_test) = mnist.load_data()

model.fit(x_train, y_train,
          batch_size=batch_size,
          epochs=epochs,
          verbose=1,
          validation_data=(x_test, y_test),
          callbacks=[history])
score = model.evaluate(x_test, y_test, verbose=0)
```

Results Metrics



How many relevant items are selected?

$$\text{Recall} = \frac{\text{true positives}}{\text{relevant elements}}$$

Sensitivity, TPR, Recall, Completeness

$$\text{TPR} = \frac{\text{TP}}{\text{P}} = \frac{\text{TP}}{\text{TP} + \text{FN}} = 1 - \text{FNR}$$

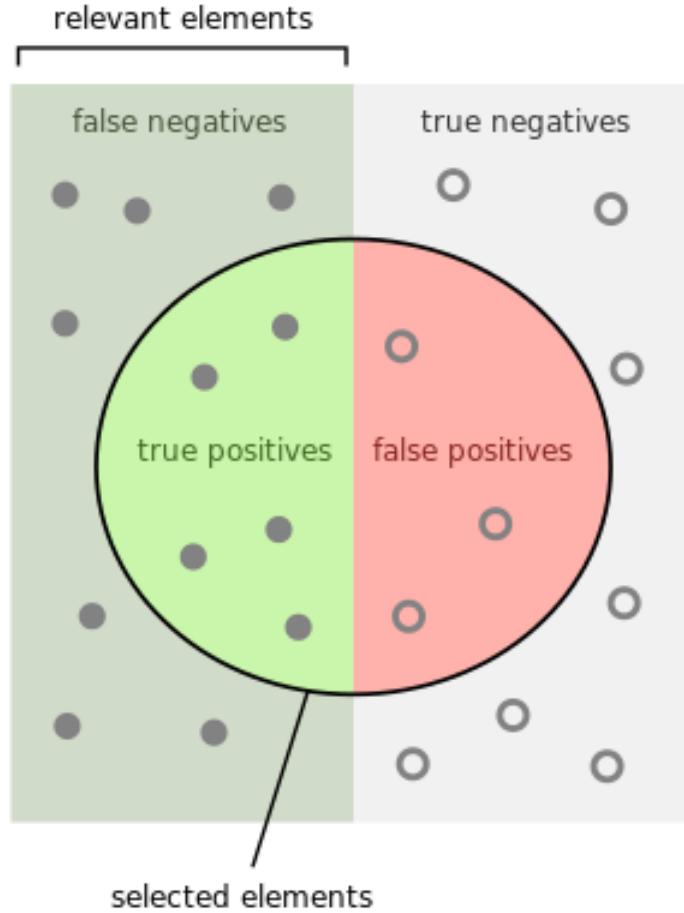


False Alarm rate=

FPR, False Alarm Rate

$$\text{FPR} = \frac{\text{FP}}{\text{N}} = \frac{\text{FP}}{\text{FP} + \text{TN}} = 1 - \text{TNR}$$

Results Metrics



Precision, Purity

How many selected items are relevant?

Precision = 

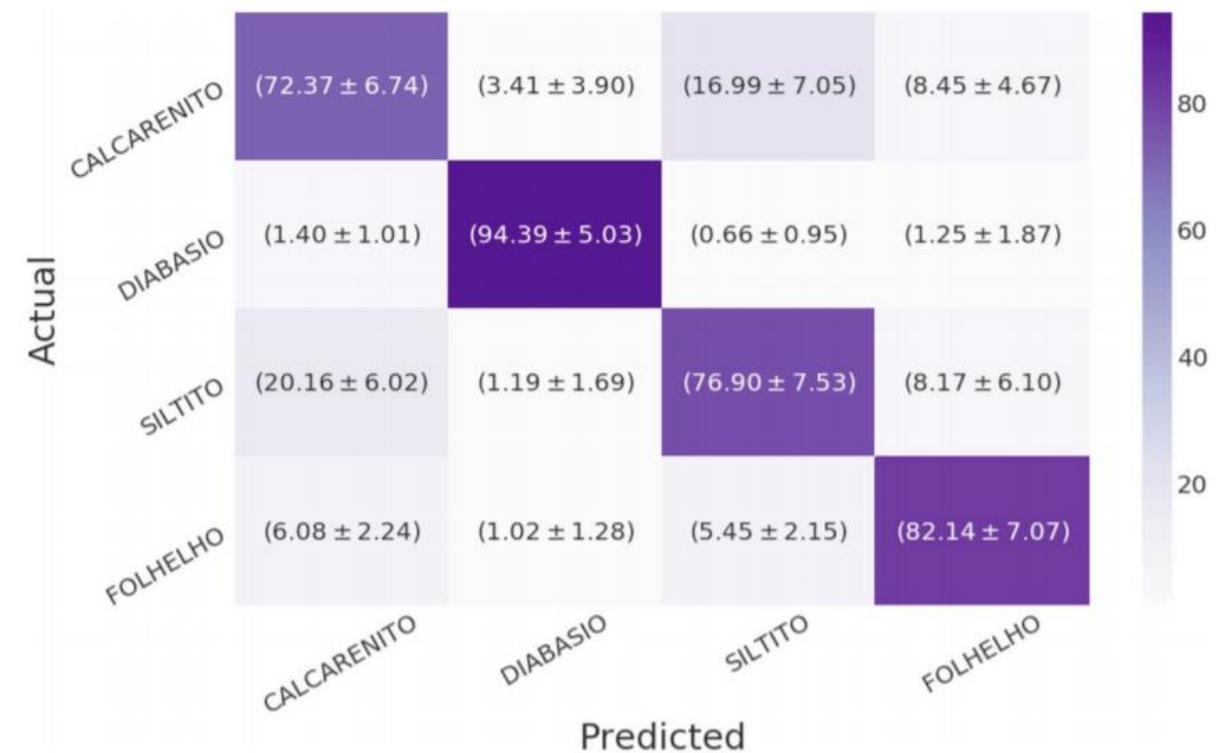
Sensitivity, TPR, Recall, Completeness

How many relevant items are selected?

Recall = 

Confusion Matrix

		Actual class		
		Cat	Dog	Rabbit
Predicted class	Cat	5	2	0
	Dog	3	3	2
	Rabbit	0	1	11



Evaluating your model

True Positive (TP):

- Reality: A wolf threatened.
- Shepherd said: "Wolf."
- Outcome: Shepherd is a hero.

False Positive (FP):

- Reality: No wolf threatened.
- Shepherd said: "Wolf."
- Outcome: Villagers are angry at shepherd for waking them up.

False Negative (FN):

- Reality: A wolf threatened.
- Shepherd said: "No wolf."
- Outcome: The wolf ate all the sheep.

True Negative (TN):

- Reality: No wolf threatened.
- Shepherd said: "No wolf."
- Outcome: Everyone is fine.



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