# MACHINE LEARNING IN PHYSICS TUTORIAL 4 / CNN

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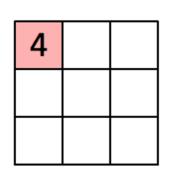
A standard CNN comprises three types of processing layers:

1. convolution, 2. pooling, and 3. classification.

#### 1. Convolution

<b>1</b> <sub>×1</sub>	1,0	1,	0	0
0,0	1,	1,0	1	0
<b>0</b> <sub>×1</sub>	0,×0	1,	1	1
0	0	1	1	0
0	1	1	0	0

**Image** 



Convolved Feature

1	<b>1</b> <sub>×1</sub>	<b>1</b> <sub>×0</sub>	<b>0</b> <sub>×1</sub>	0
0	1,0	1,	1,0	0
0	<b>0</b> <sub>×1</sub>	1,0	1,	1
0	0	1	1	0
0	1	1	0	0

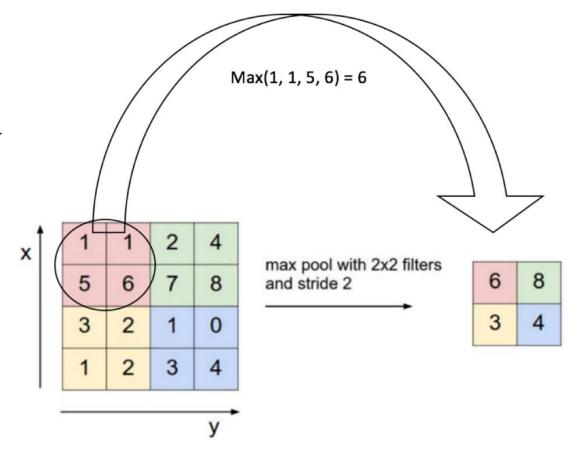
**Image** 

4	3	

Convolved Feature

### 2. Pooling

MaxPool2d AvgPool2d



Rectified Feature Map

#### 3. Classification

$$p(k|x) = \frac{p(x|k) \pi(k)}{\sum_{j=0}^{K-1} p(x|j) \pi(j)}$$

In our galaxy classification example,

- ightharpoonup K = 7,
- $\triangleright \pi(k) = 1/K$

Since this is a multi-class problem, we'll train a model,  $f_k(x)$ , with K outputs that satisfy

$$\sum_{k=0}^{K-1} f_k(x) = 1$$

by minimizing the empirical risk

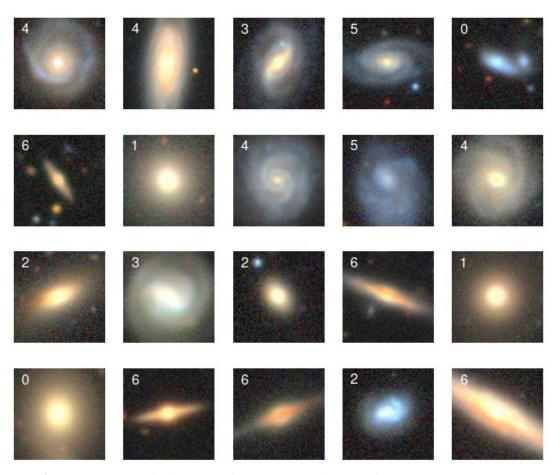
$$R(\omega) = -\frac{1}{N} \sum_{i=1}^{N} \log f_{y_i}(x_i)$$

where  $y_i$  is the class label associated with image  $x_i$ .

# **TUTORIAL 4**

## **Tutorial 4**

Goal: classify galaxies into 7 morphology classes using a CNN.



https://astronn.readthedocs.io/en/stable/galaxy10.html

### **CNN Model**

Our model comprises 4 layers, each consisting of 3 operations, followed by a linear function and a softmax. ReLU() MaxPool2d(2,2)Conv2d(4,6,3,1,1)

ReLU()

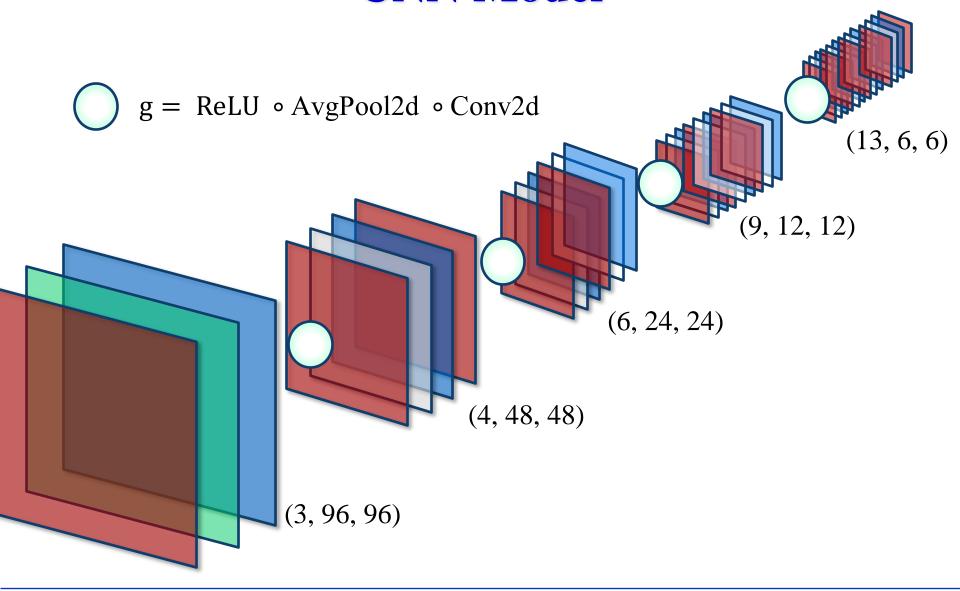
MaxPool2d(2,2)

Conv2d(3,4,3,1,1)

 $\mathbf{x}$ 

 $f(x,\omega)$ Softmax Linear(468,7) ReLU() MaxPool2d(2,2)Conv2d(9,13,3,1,1) ReLU() MaxPool2d(2,2)Conv2d(6,9,3,1,1)

## **CNN Model**



# **CNN Model: Training**

### Samples

1. Training sample size: 10,000

2. Validation sample size: 1,600

3. Testing sample size: 1,000

### Training hyperparameters

1. Number of iterations: 10,000 (200 epochs)

2. Batch size: 200

3. Learning rate: 10<sup>-3</sup>

### **CNN Model: Results**

Accuracy: 71.2%

