

## Week 8

### Lecture

- CNN
  - Architecture (**INPUT – CONV – RELU – POOL - FC**)
- RNN
  - Neural work + memory
  - LSTM (Input gate, forget gate, output gate)
  - GRU (reset gate, update gate)
  - Difference between LSTM and GRU

## Tutorial

### Task 1: Using CNN for MNIST data.

Step 1: Load MNIST data and create validation set.

Step 2: Define the CNN model by using Keras.

Step 3: Optimization and evaluation.

### Task 2: Using LSTM for movie review.

Step 1: load the movie review data.

Step 2: Define LSTM model.

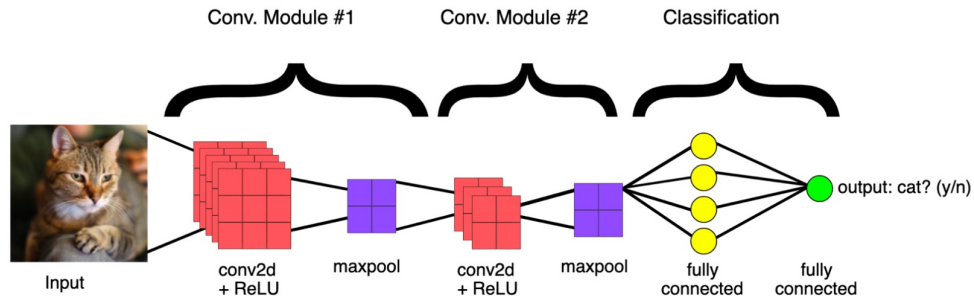
Step 3: Optimization and evaluation.

**TODO:**

**1. Create CNN for CIFAR-10**

# Deep Learning

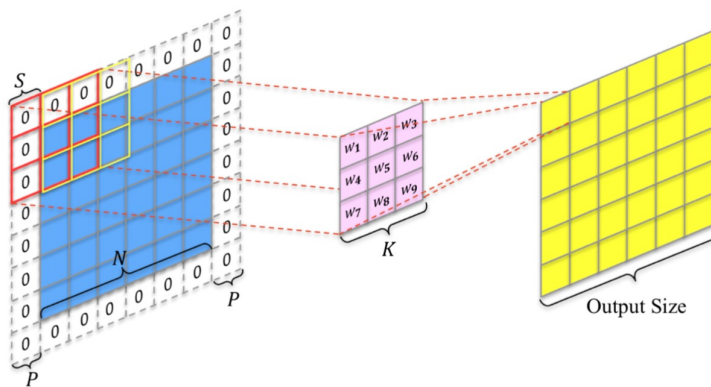
## CNN



Basic CNN structure:

- Convolutional Layer
- Pooling
- Fully-connected Layer

$$\text{Output Size} = \frac{N+2P-K}{S} + 1$$



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

1	0	-1
-1	0	0
0	0	1

-1	2		

Stride = 1

Zero padding (pad = 1)

0	0	0	0	0	0	0	0
0							0
0							0
0							0
0							0
0							0
0							0
0	0	0	0	0	0	0	0

improves performance by keeping information at the borders

### Max pooling

- Filter size: (2,2)
- Stride: (2,2)
- Pooling ops:  $\max(\cdot)$

-1	2	0	0
0	1	3	-2
0	0	-1	4
3	-1	-2	-2

2	3
3	4

Subsample map

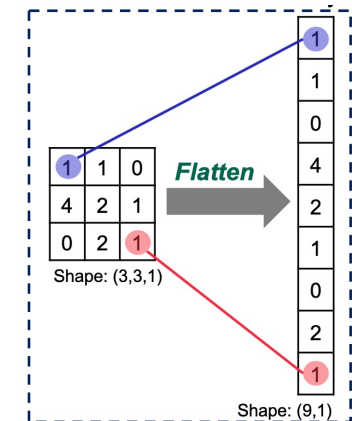
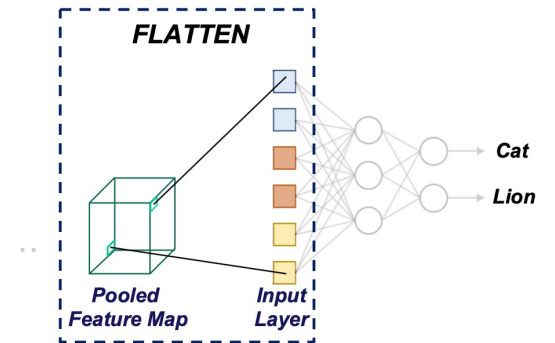
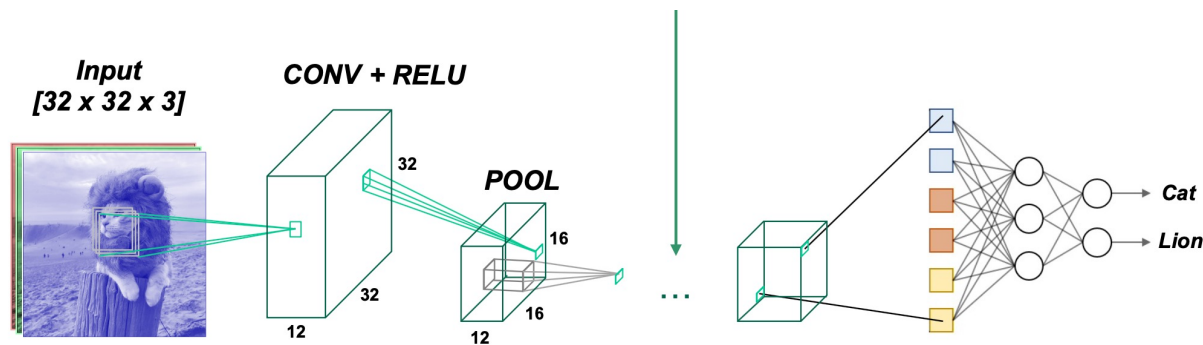
Feature map

# Deep Learning

## CNN

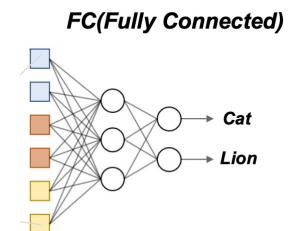
### **FLATTEN (Flattening) Layer**

In between the convolutional layer and the fully connected layer, there is a '**Flatten**' layer. Flattening transforms a multi-dimensional matrix of features into a vector that can be fed into a fully connected neural network classifier.



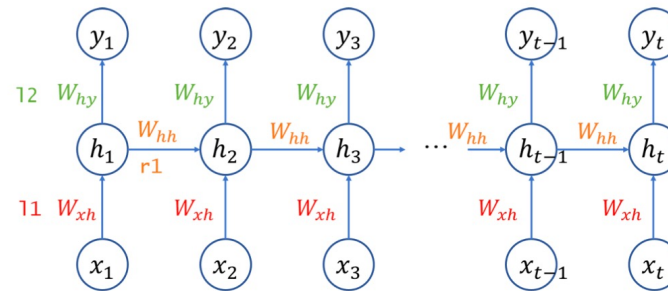
### **FC (Full Connected) layer**

Compute the class scores, each of the 2 categories correspond to a class score, such as among the 2 categories of our dataset. As with ordinary Neural Networks and as the name implies, each neuron in this layer will be connected to all the numbers in the previous volume



# Deep Learning

## RNN vs. LSTM



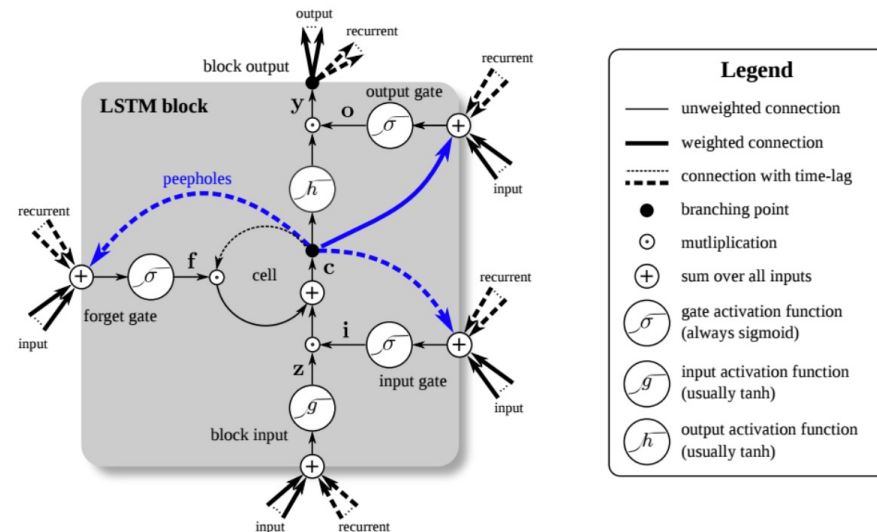
$$h_t = \tanh(l1(x_t) + r1(h_{t-1}))$$

$$y_t = l2(h_t)$$

The output of hidden layer are stored in memory. Memory can be considered as another input.

So information cycles through a loop.

1. current input
2. what it has learned from the inputs it received previously.



LSTM is an extension for RNN.

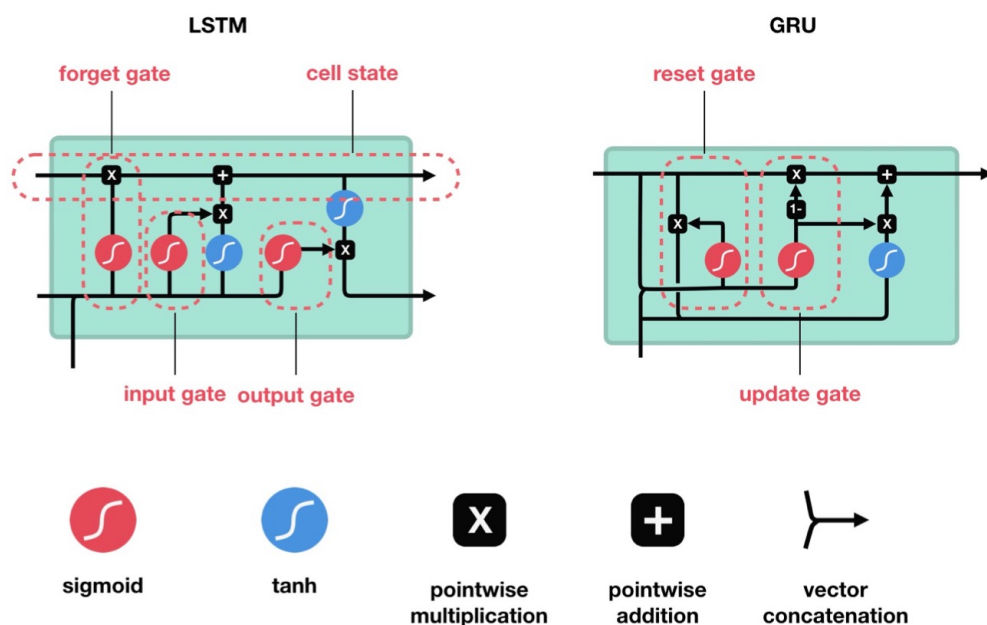
The memory in LSTM can be seen as a gated cell (store or delete based on the importance it assigns to the information).

Three gates:

1. input gate: whether or not let new input in
2. forget gate: delete information
3. output gate: output at the current time step

# Deep Learning

## RNN vs. LSTM



GRU first computes an **update gate** based on **current input word vector** and **hidden state**

Compute reset gate similarly but with different weights

- If reset gate unit is  $\sim 0$ , then this ignores previous memory and only stores the new word information

Final memory at time step combines current and previous time steps