

MACHINE LEARNING DAY 2

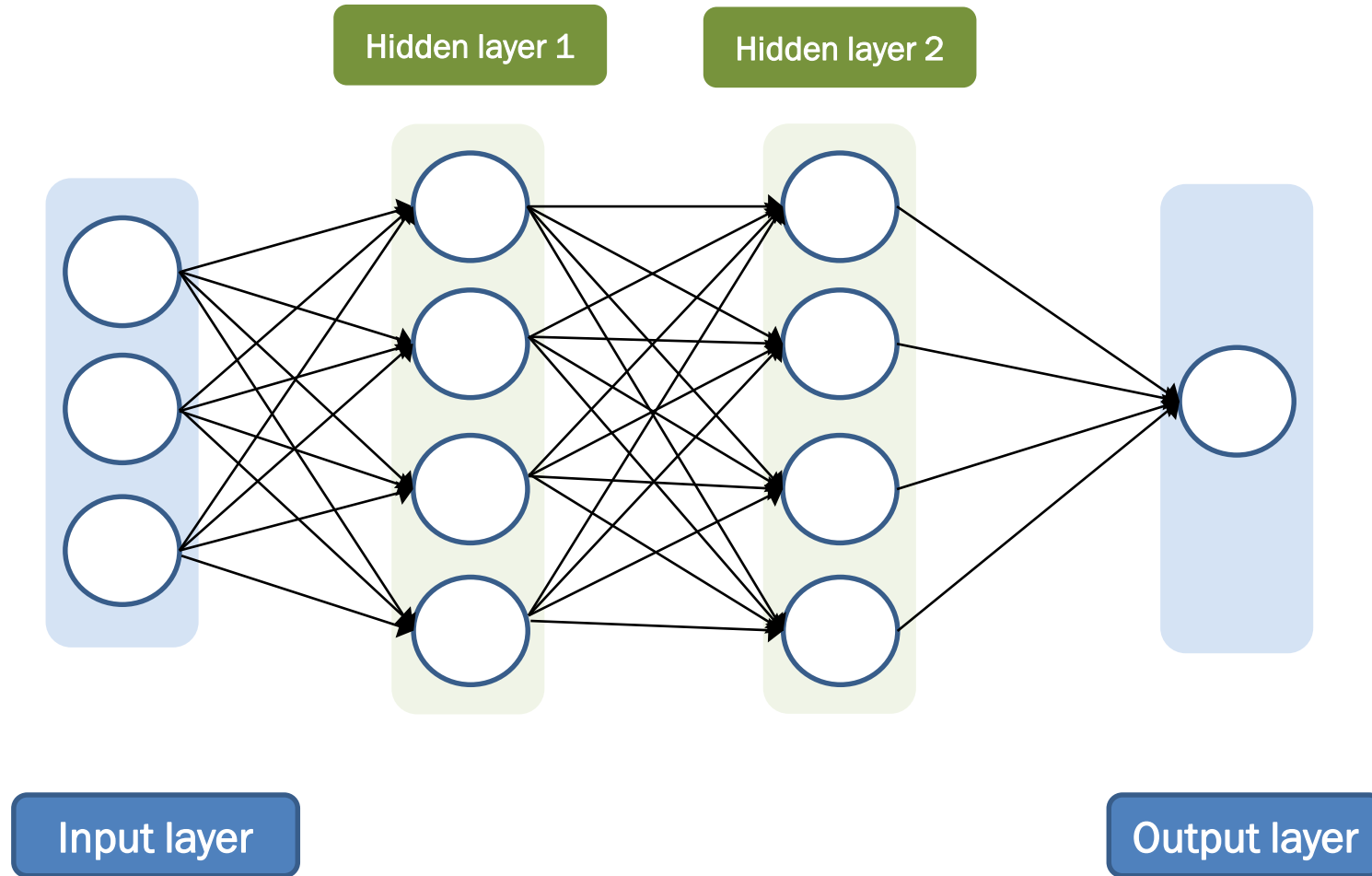
DEEP LEARNING

Session IV: Convolutional Neural Network

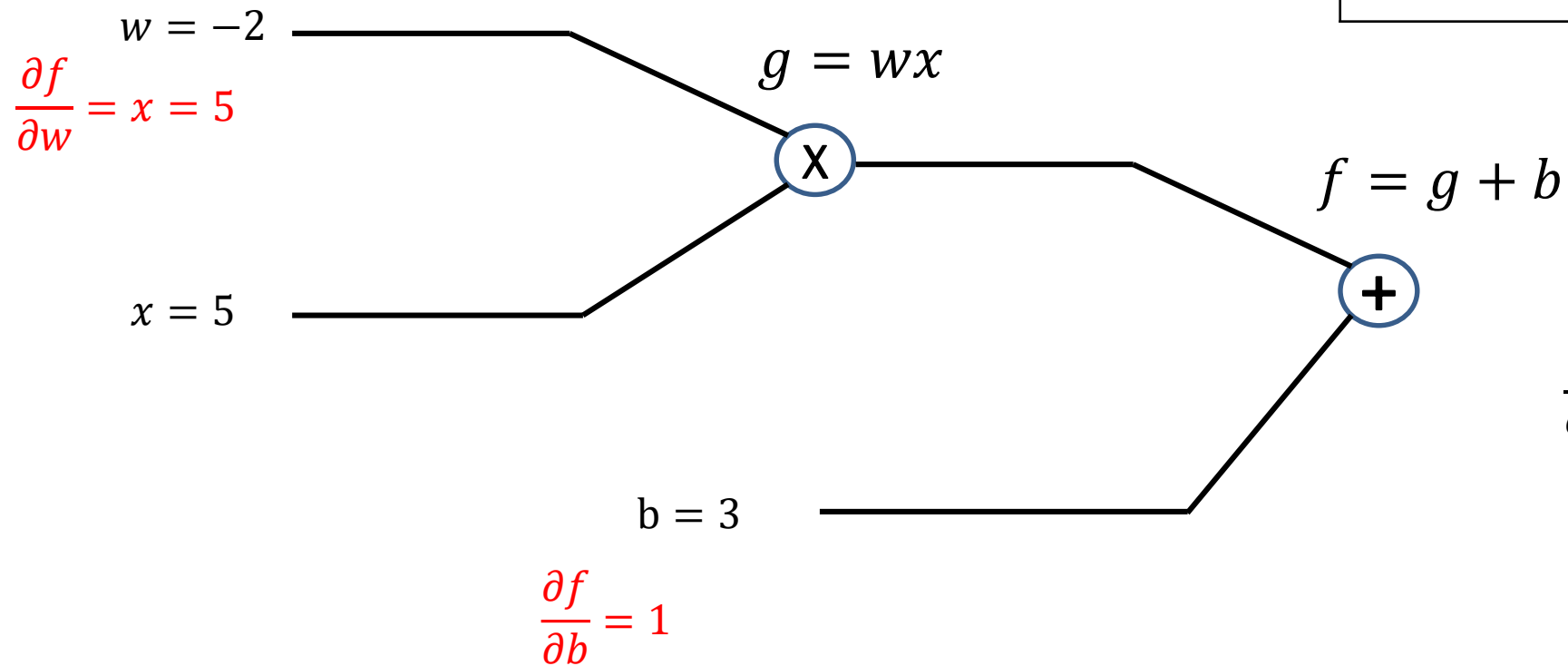
Session IV

- Backward propagation
- Model capacity / overfitting
- MNIST classification
- Vanishing gradient problem
- *Lab 4A: Multi-Layer Perceptron (MNIST)*
- Issue with MLP
- Convolutional Neural Network
- Techniques
- *Lab 4B: Convolutional Neural Network (CIFAR10)*

Multi-Layer Perceptron



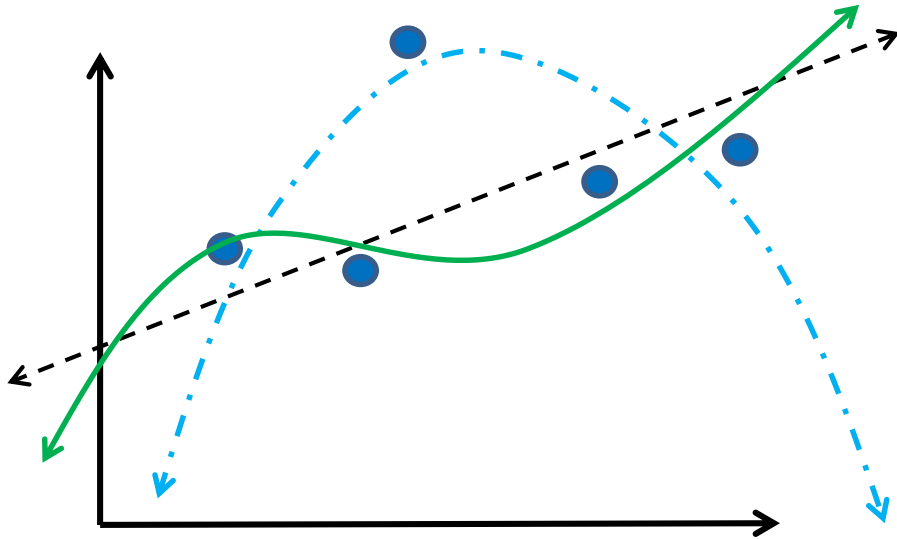
Backward propagation



$g = wx$	$f = g + b$
$\frac{\partial g}{\partial w} = x, \frac{\partial g}{\partial x} = w$	$\frac{\partial f}{\partial g} = 1, \frac{\partial f}{\partial b} = 1$

$$\frac{\partial f}{\partial w} = \frac{\partial f}{\partial g} \frac{\partial g}{\partial w} = 1x$$

Model capacity



$$y = w_1x + b$$

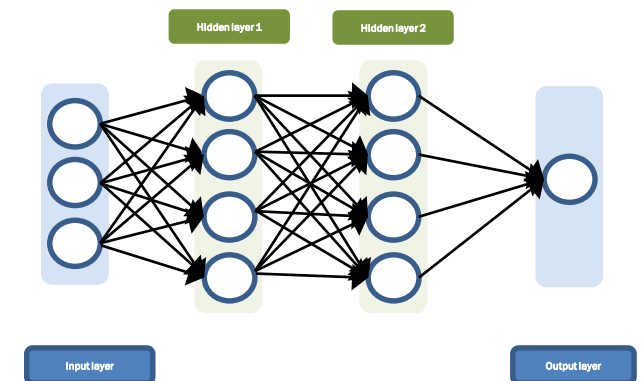
$$y = w_2x^2 + w_1x + b$$

$$y = w_3x^3 + w_2x^2 + w_1x + b$$

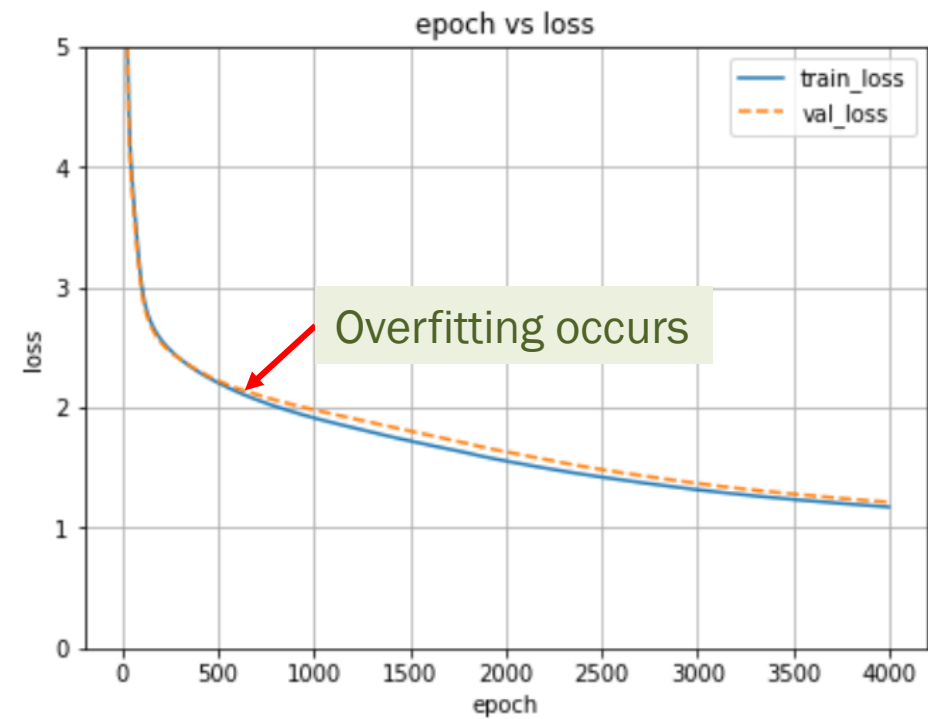
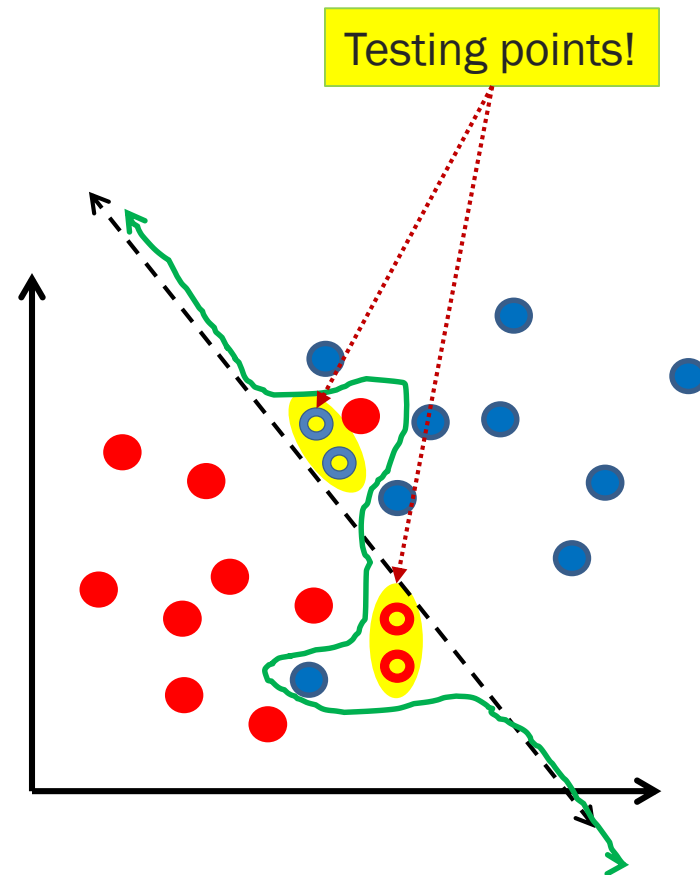
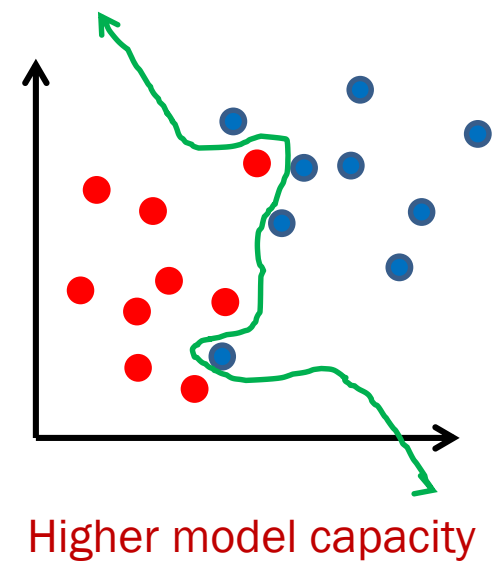
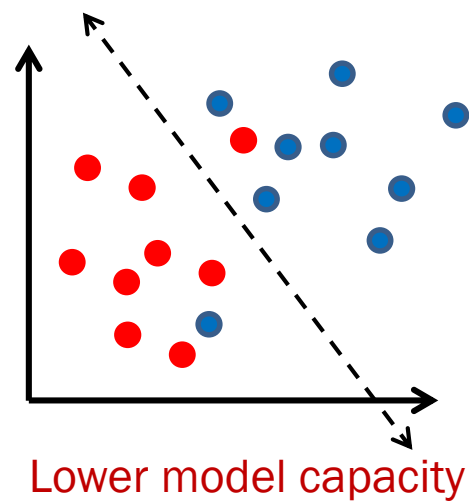
Higher model capacity

The more hidden layers and units,
the higher model capacity it will have.

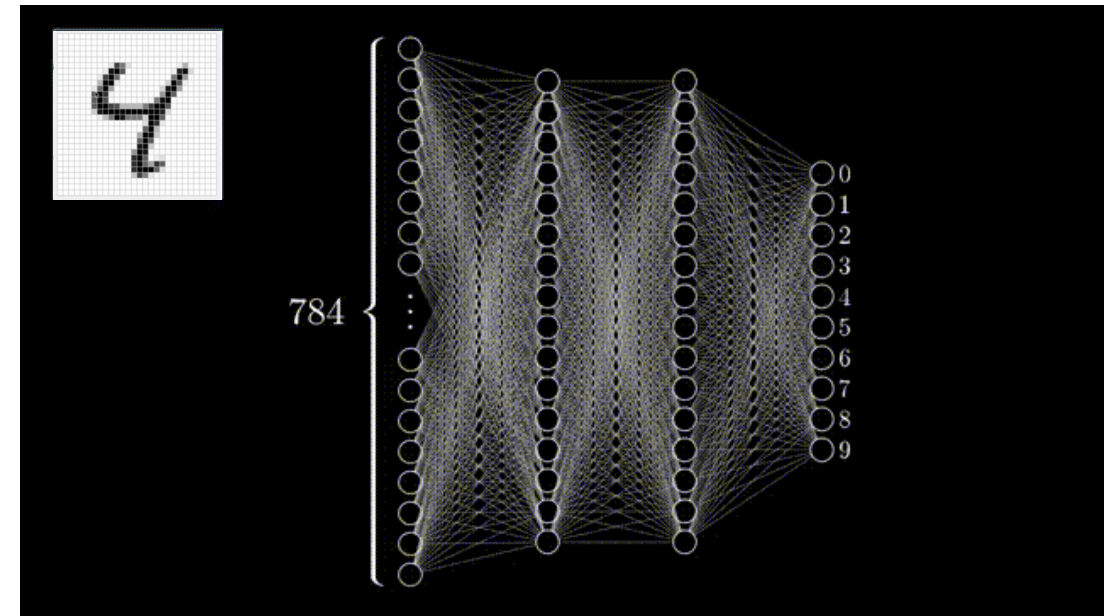
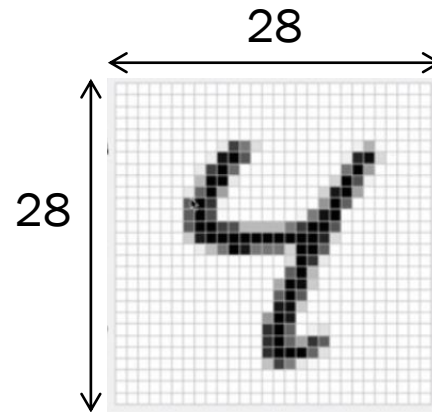
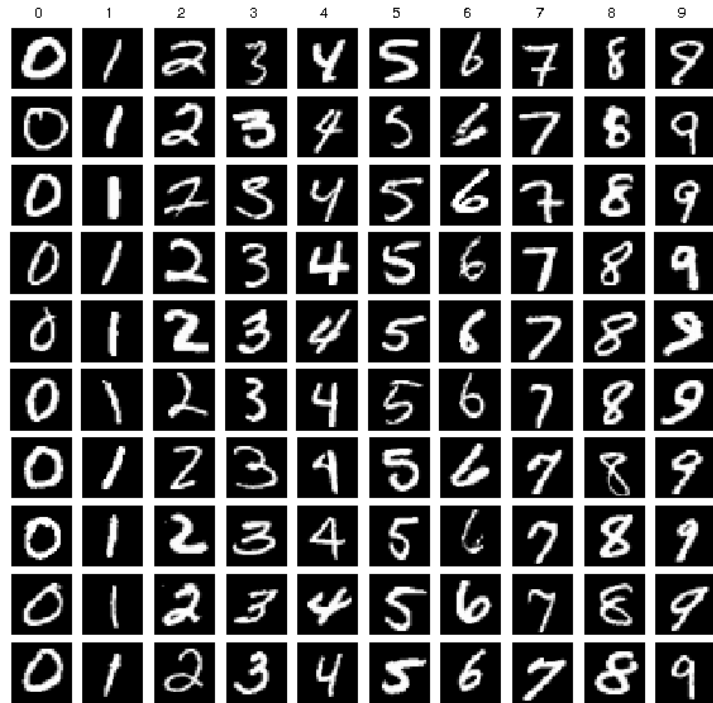
Does it guarantee better accuracy?



Overfitting



Classification problem: MNIST

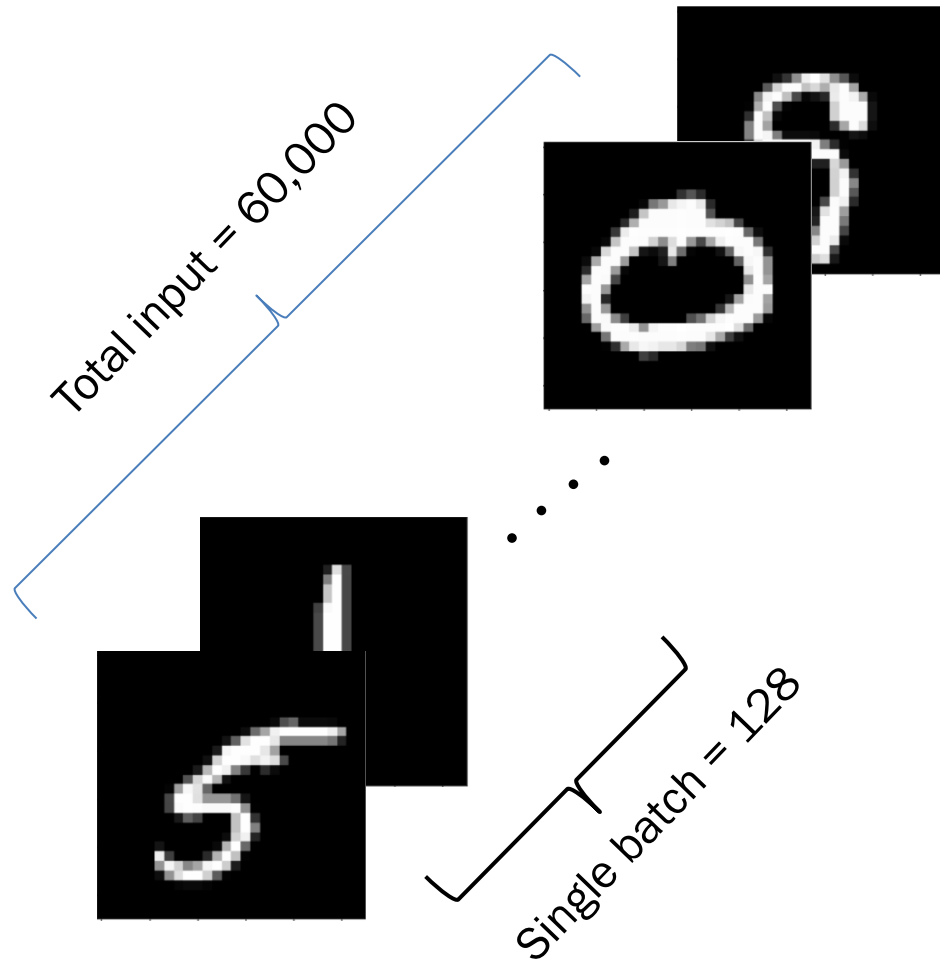


Handwritten data

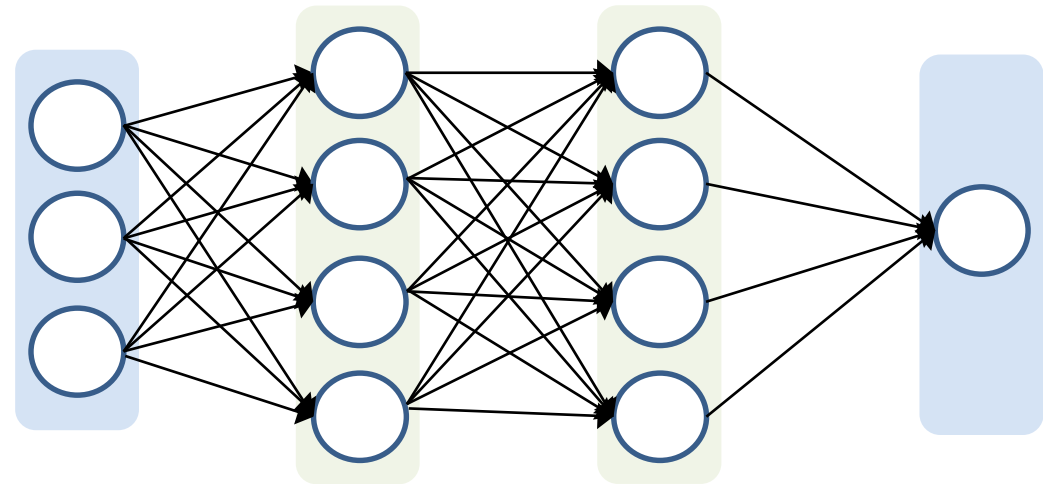
60K train set and 10K test set

Each image has a size of 28x28 (=784)

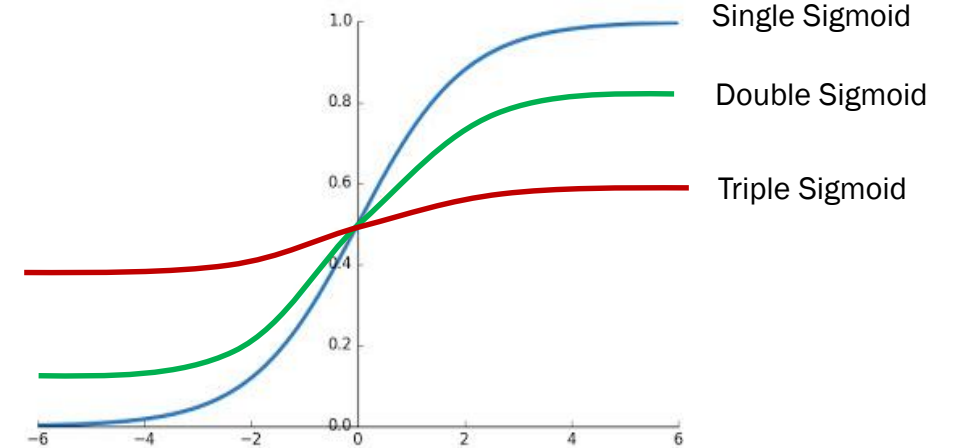
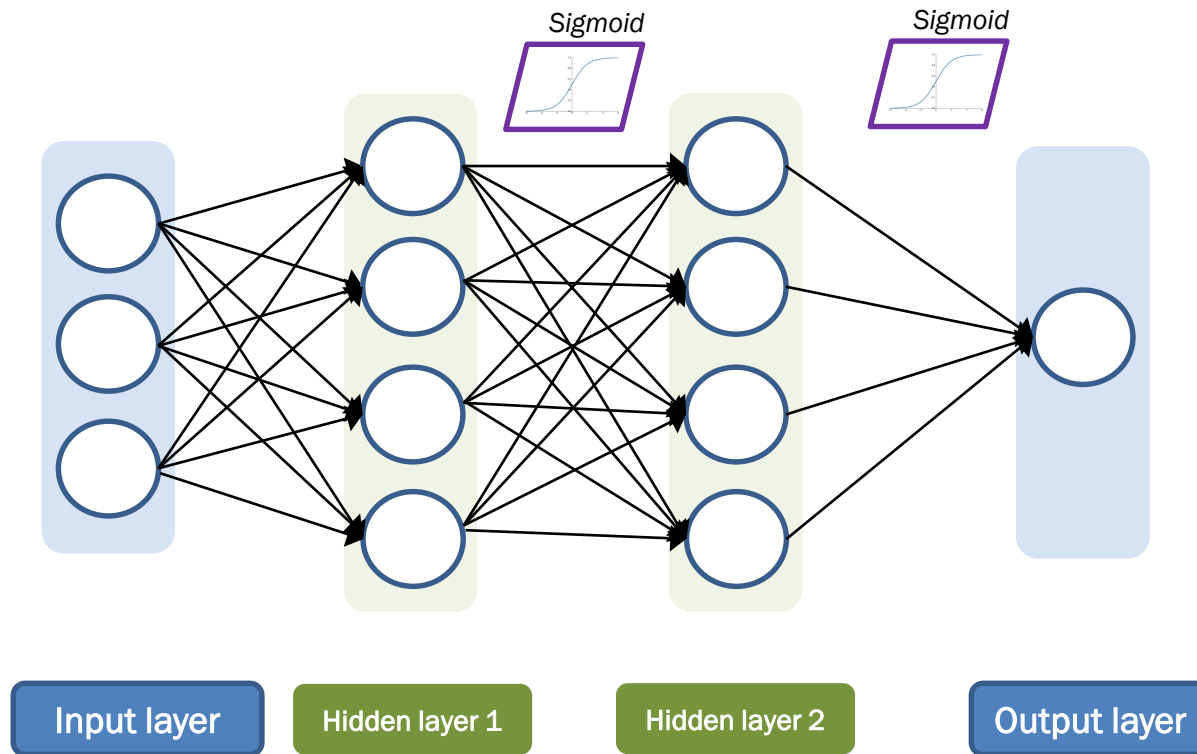
Batch, Iteration, Epoch



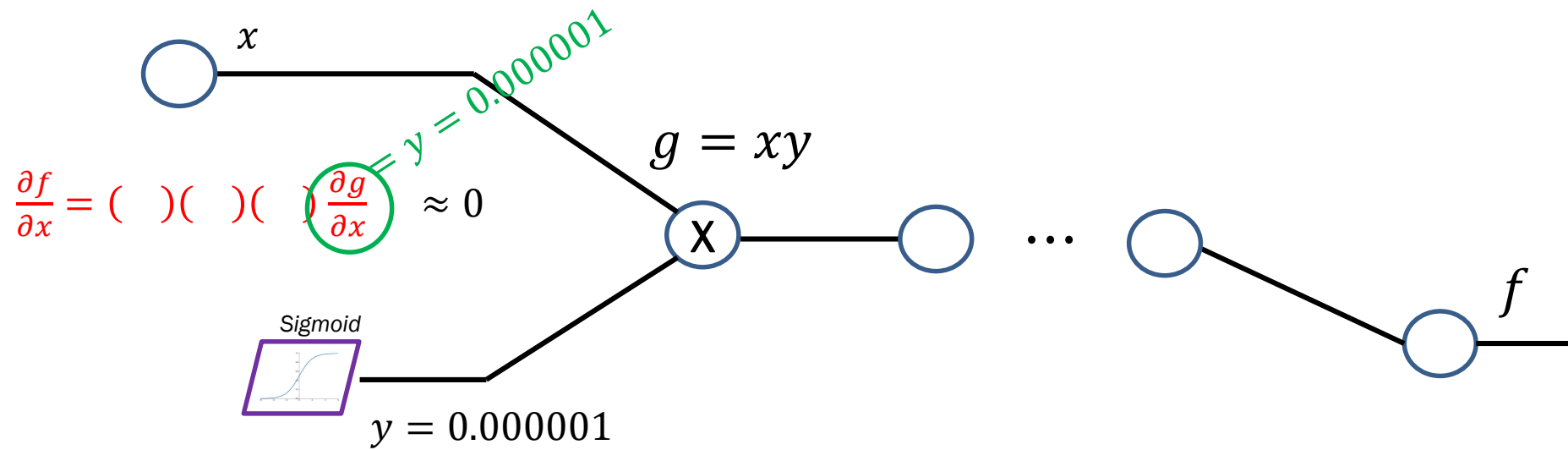
Epoch: one complete run of total input
Batch size: the amount of input for each iteration
 $\# \text{ of iteration} = \# \text{ total input} / \text{batch size}$



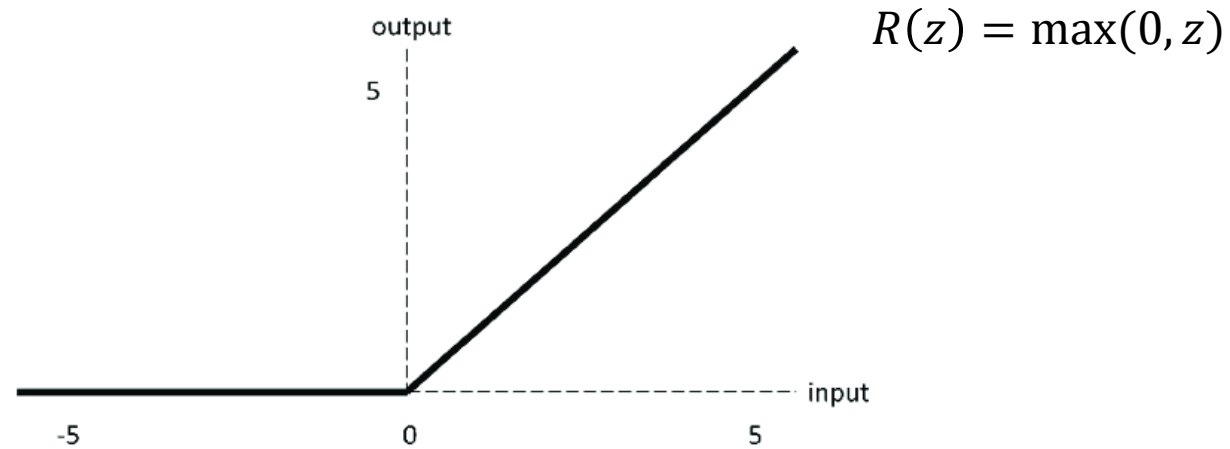
Activation function: Sigmoid problem



Vanishing gradient problem



Activation functions: Rectified Linear Unit (ReLU)



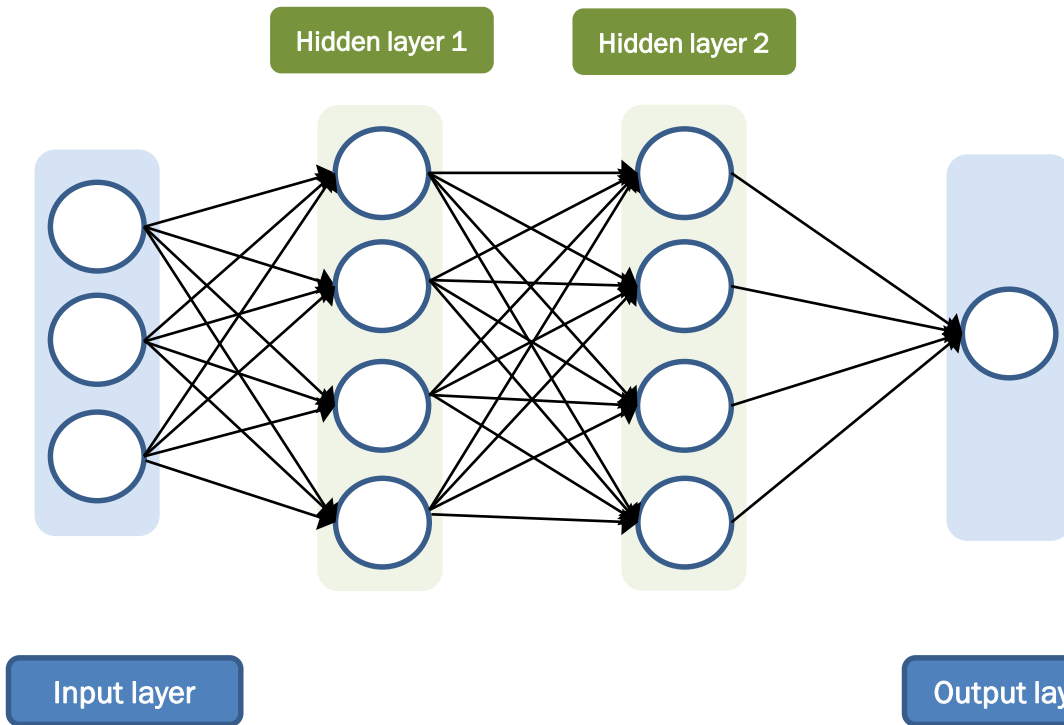
Lab 4A: Linear regression – MLP:MNIST

1. Check the model define (Linear, MLP)
2. Check the result by varying learning rate
3. Check the result with different number of Epoch
4. Check the result with different number of layers
5. Check the result with different activation functions



**Break
room**

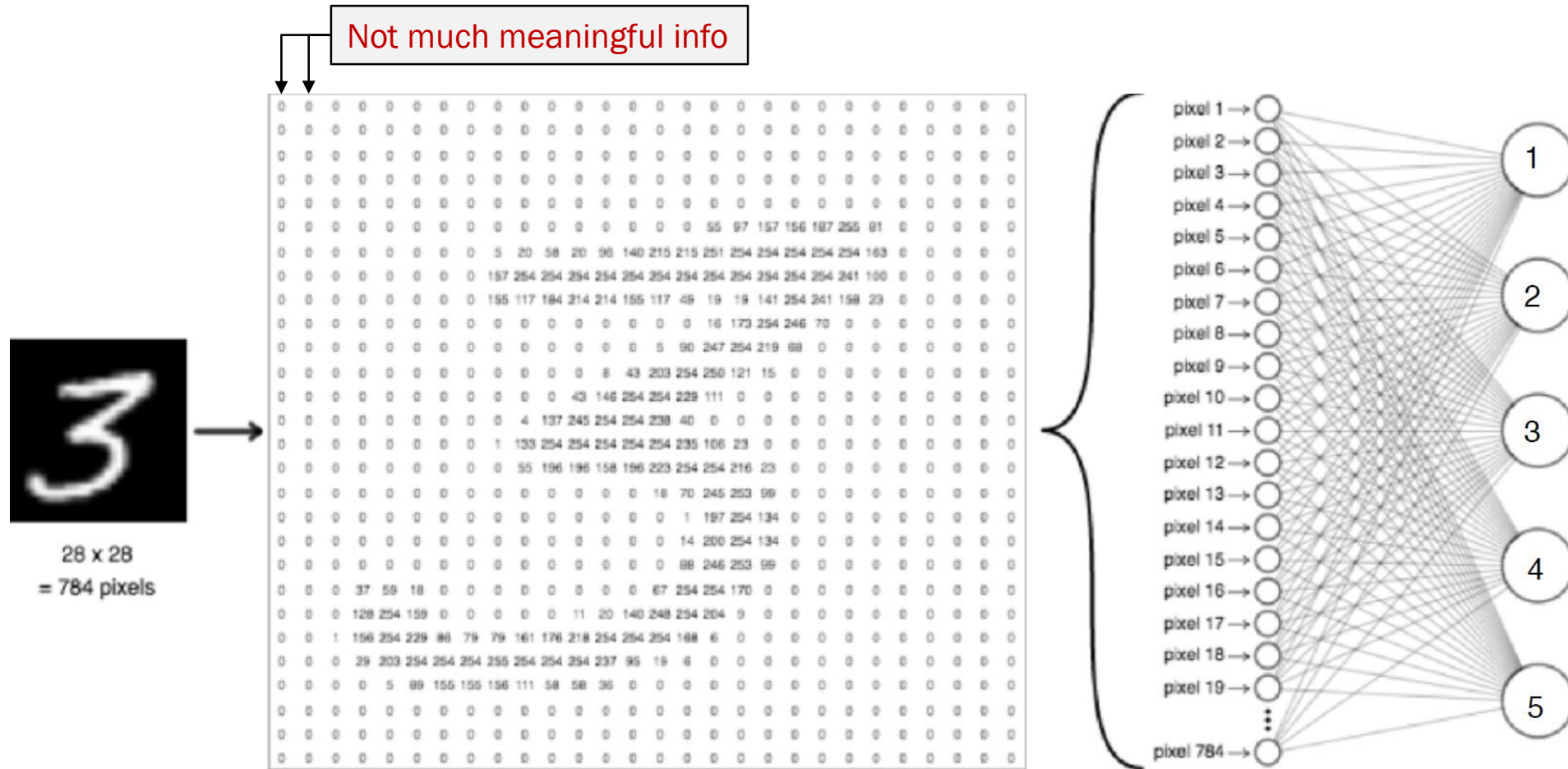
Issue with MLP



A neuron is connected with every neuron in next layer (fully connected)

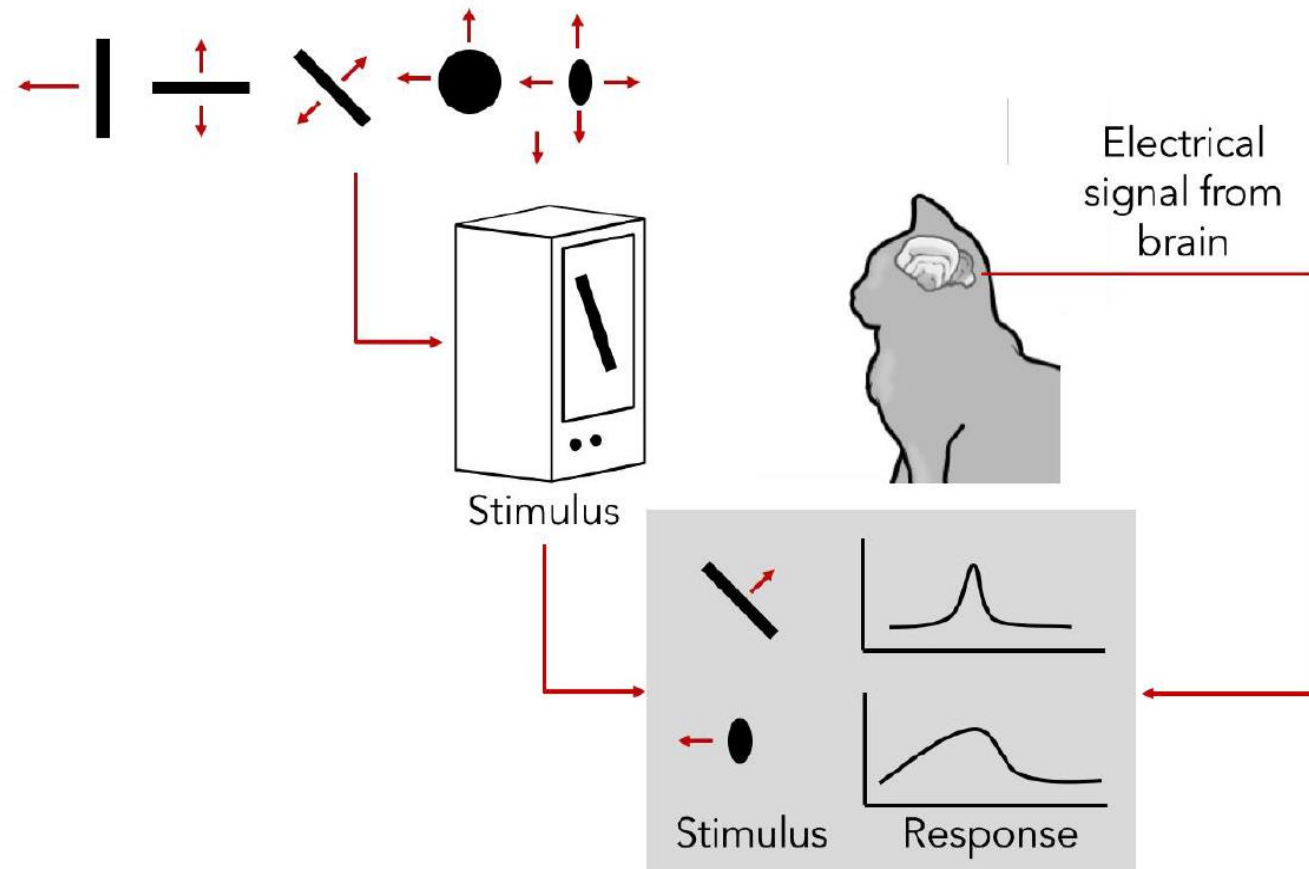
of parameters increases explosively

Issue with MLP



Some of parameters are meaningless!

How to recognize an image?



Hierarchical structure

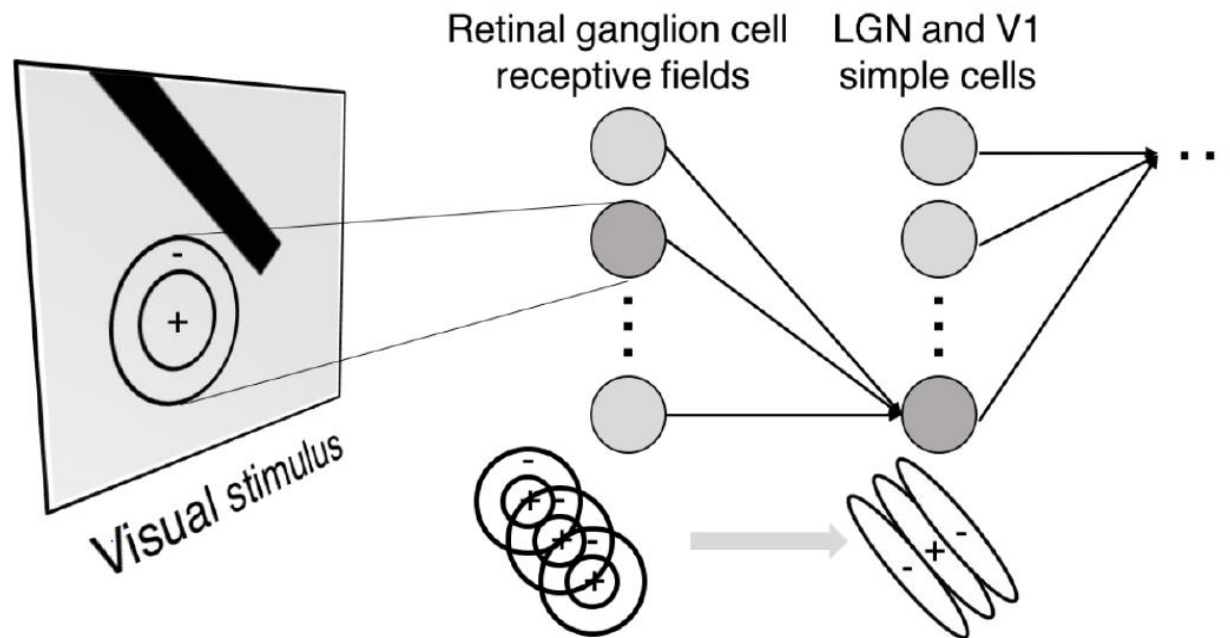


Illustration of hierarchical organization in early visual pathways by Lane McIntosh, copyright CS231n 2017

Simple cells: Response to light orientation

Complex cells: Response to light orientation and movement

Hypercomplex cells: Response to movement with an end point

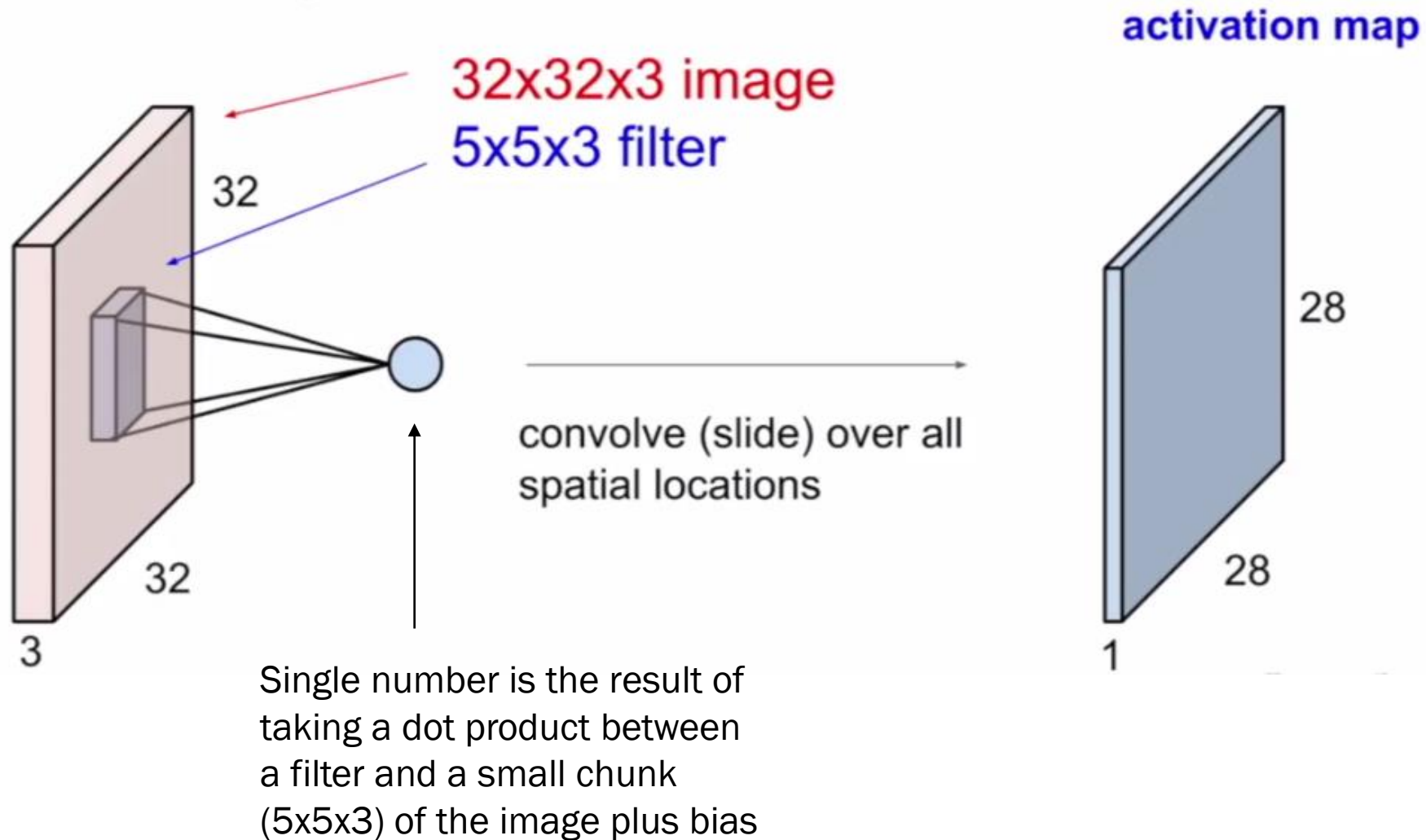


No response

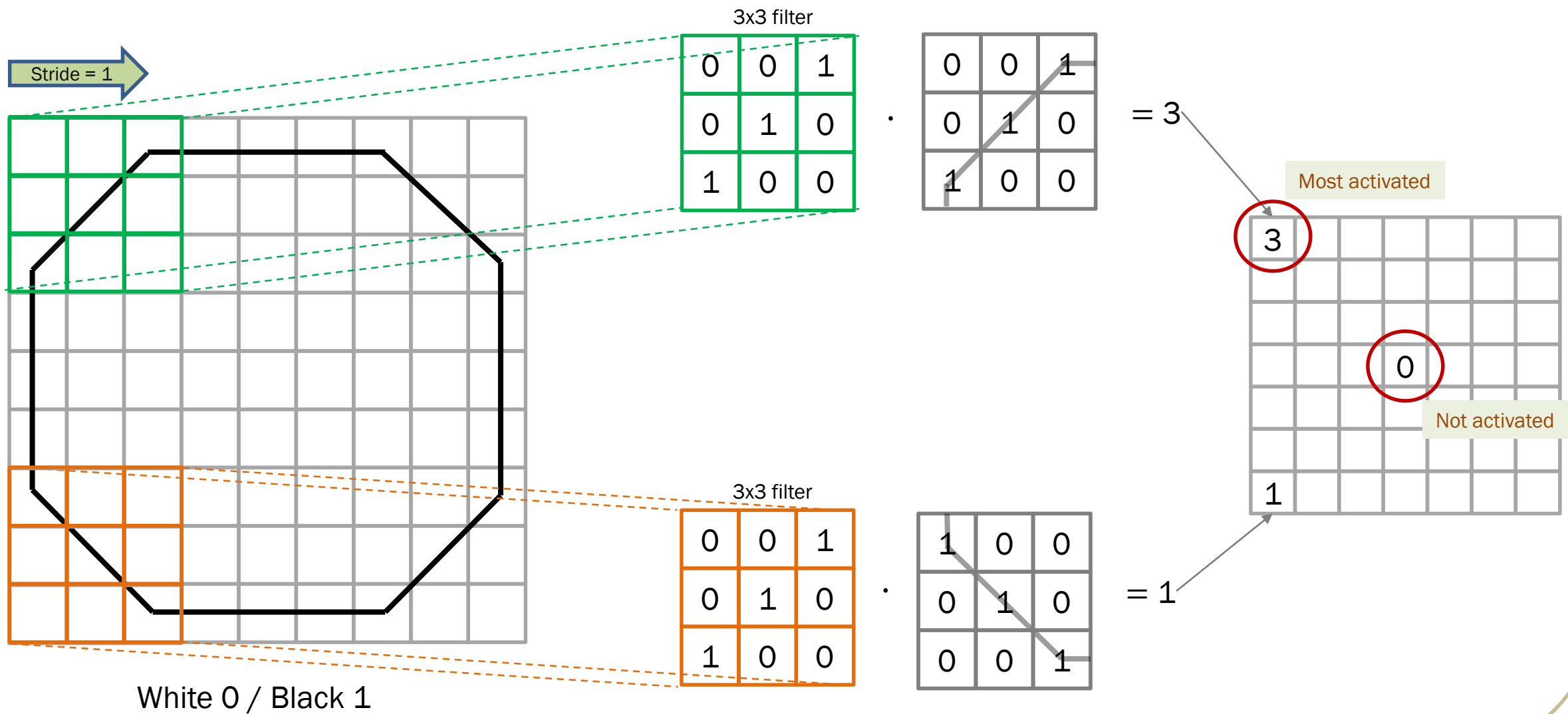


Response
(end point)

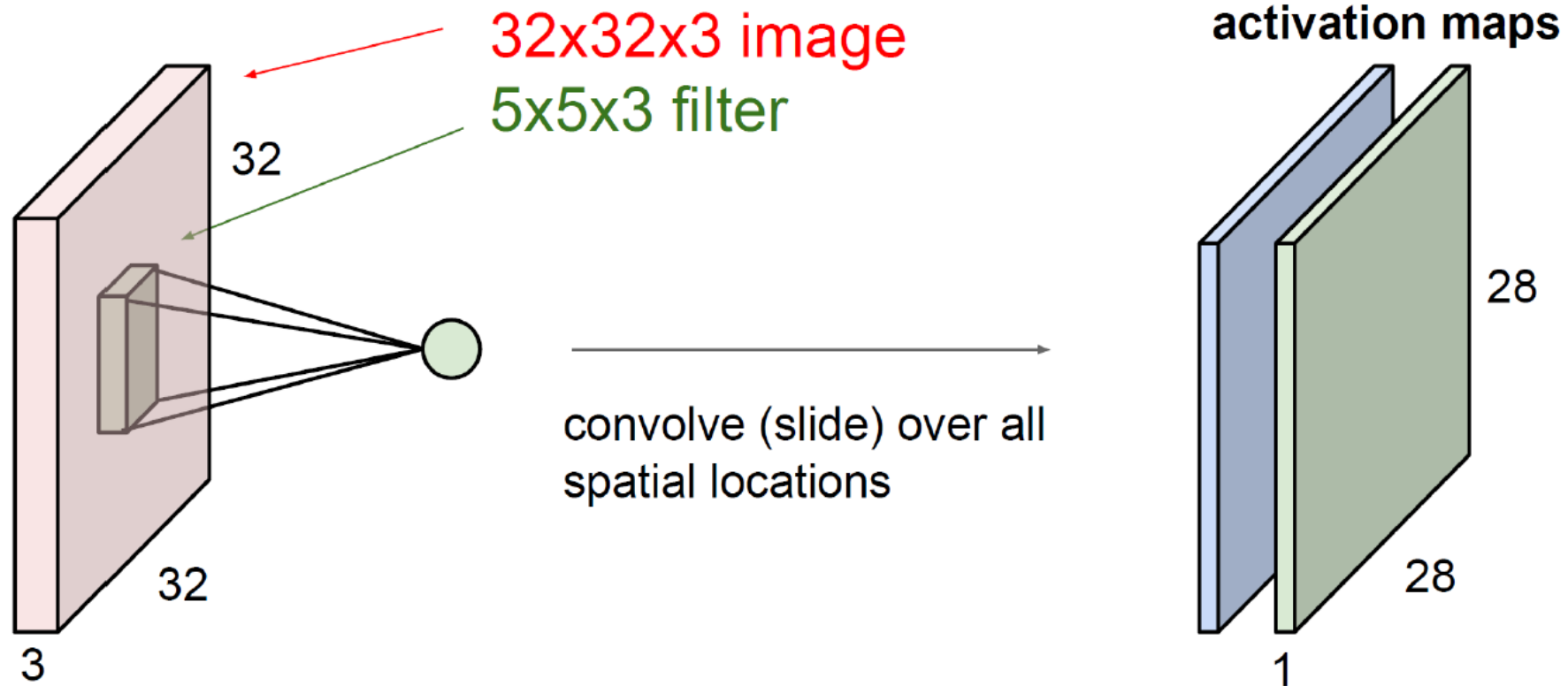
Convolutional layer



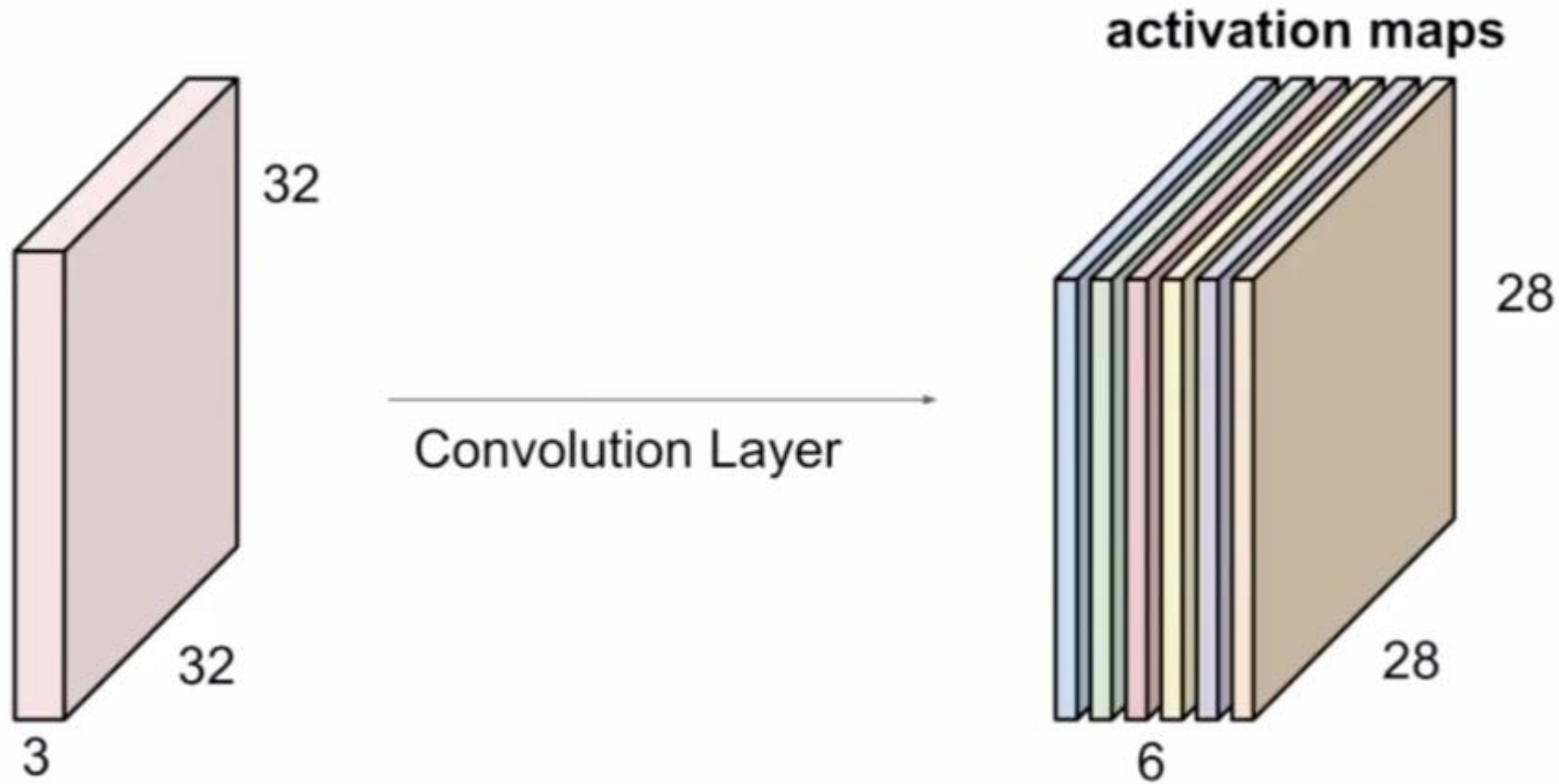
Convolutional operation



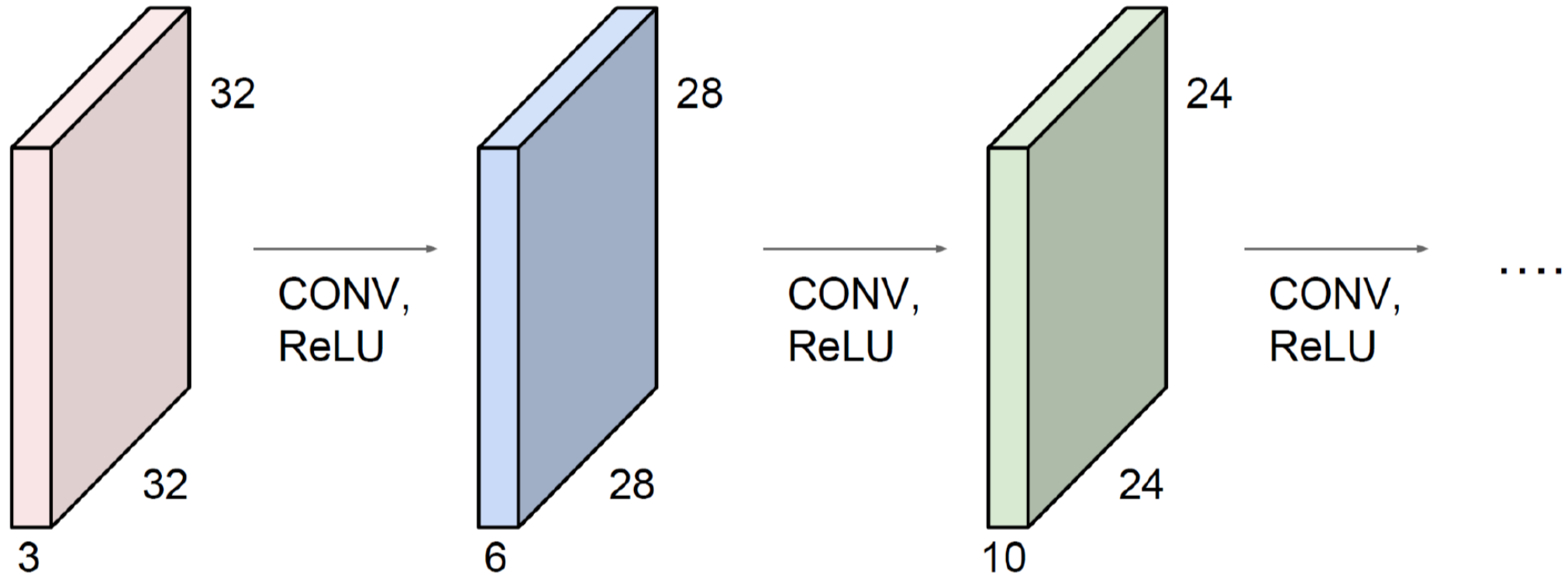
Convolutional layer



Convolutional layer



Convolutional Neural Network



Pooling

1	1	2	4
5	6	3	8
3	2	0	1
1	2	7	5

4x4

Max pooling

filter: 2x2

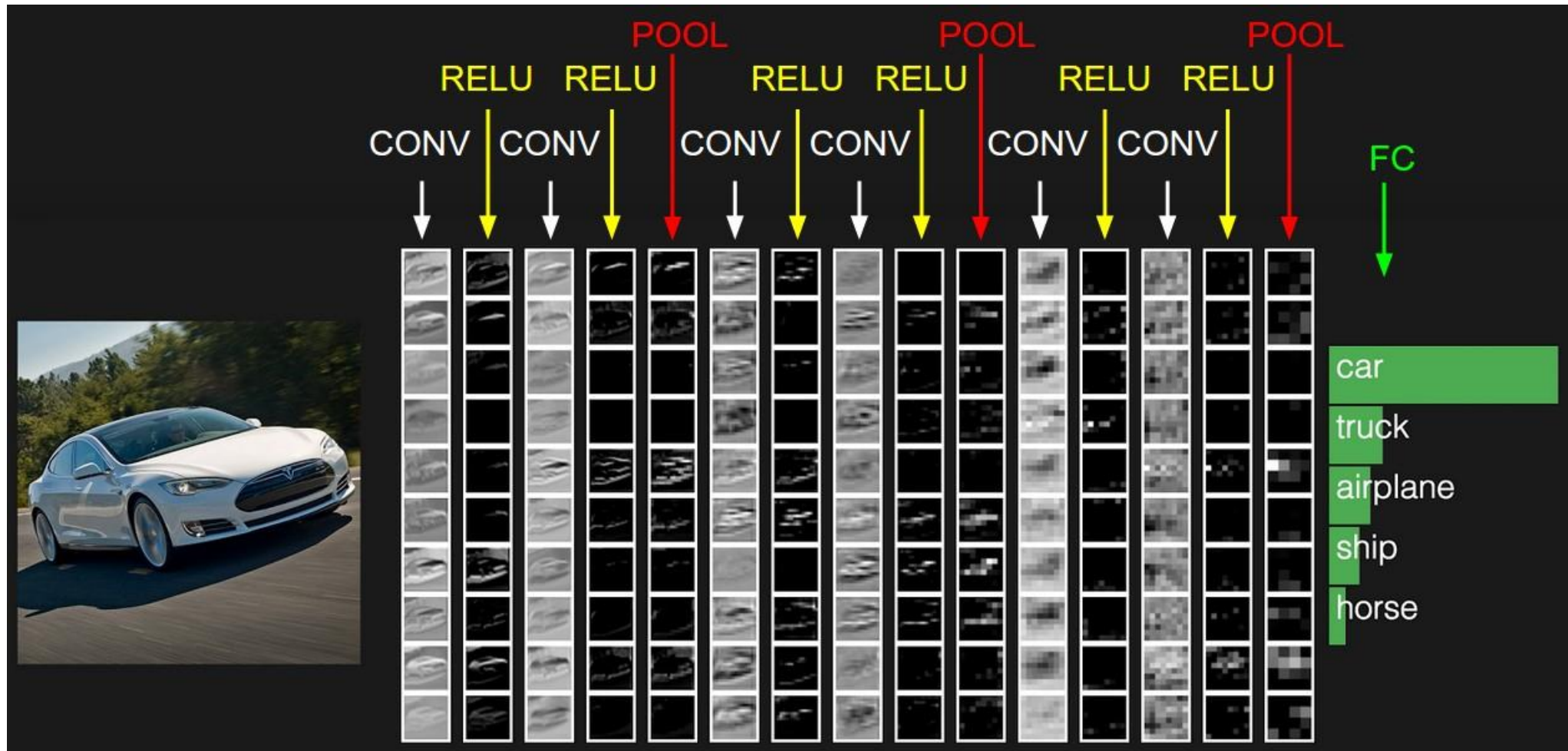
Stride: 2



6	8
3	7

2x2

ConvNet Architecture

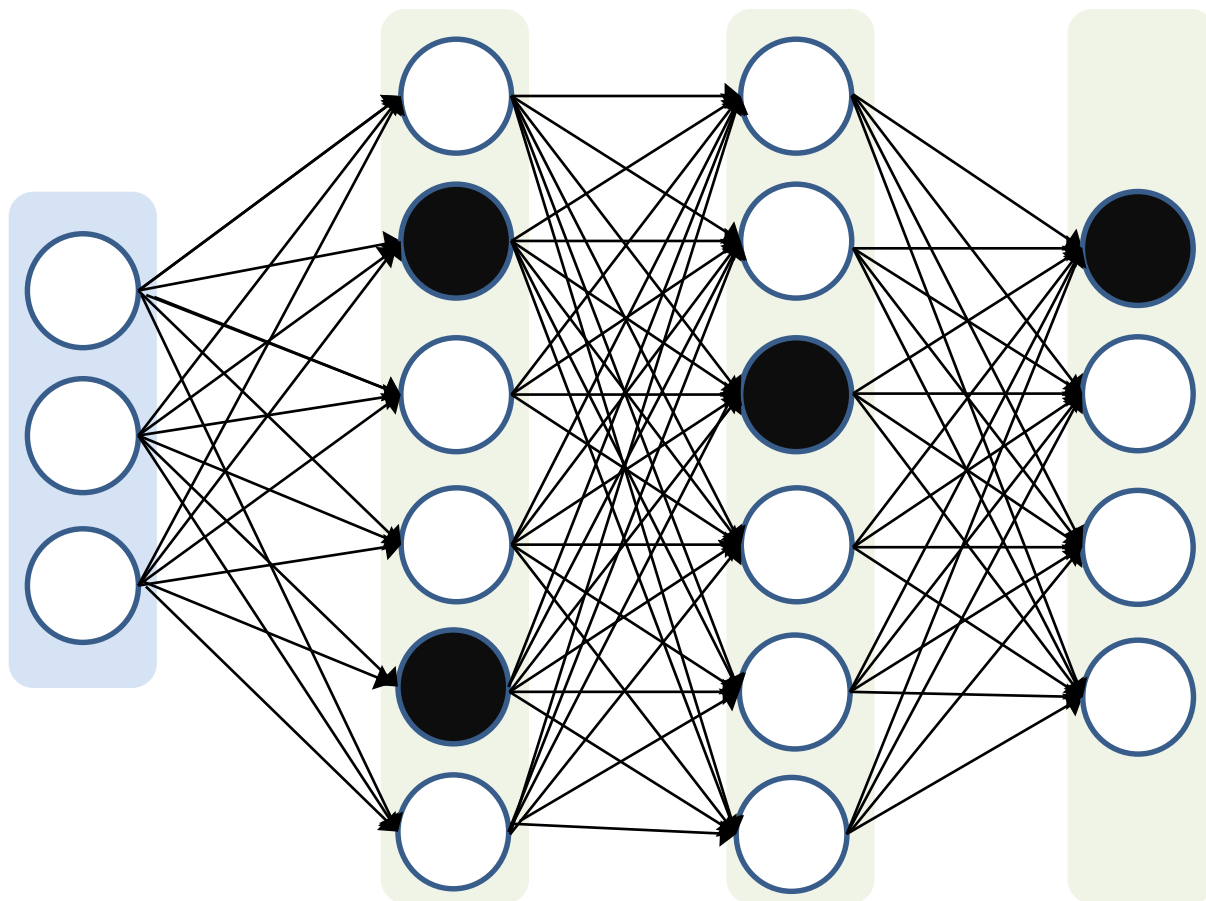


Hyperparameter?

- Non-learnable parameters

Category	Parameters
Model capacity	# of hidden layers # of hidden units Activation function
Regularization	Dropout rate Batch normalization L2 regularization Xavier initialization
Optimizing	Optimizer Learning rate # of Epoch Batch_size
Device	CPU/GPU
Post processing	Saving/filename

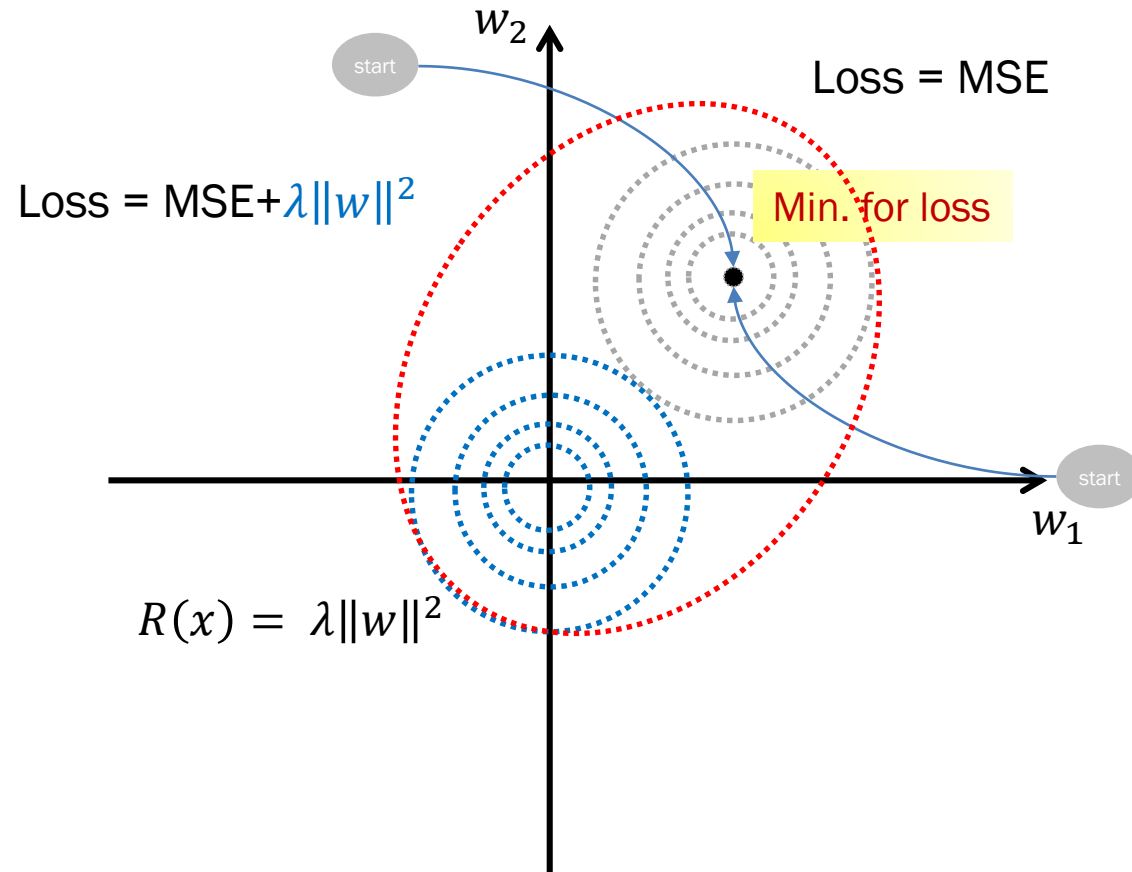
Dropout



...

Intentionally turn off
nodes with probability
when training

L2 Regularization

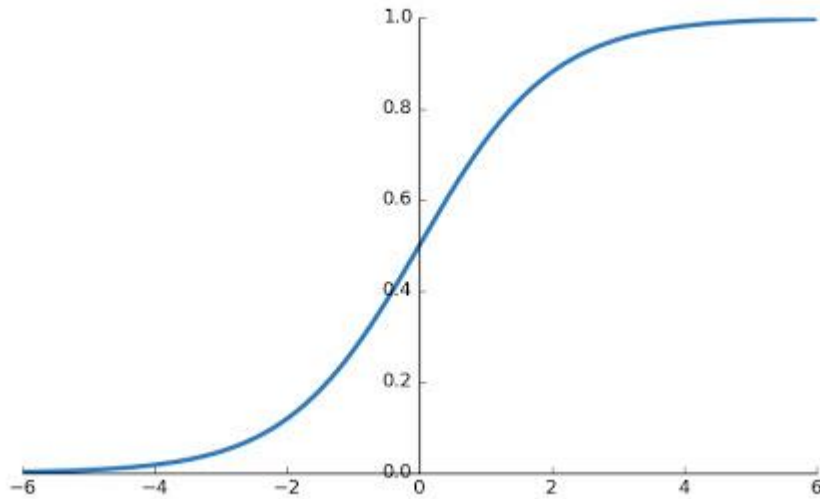


Xavier initialization

Models are sensitive to weight initialization

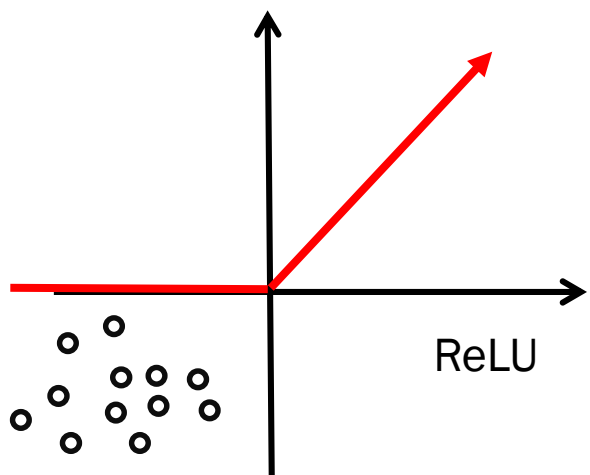
Keeping the shape of initialization valid to initiate parameters with better values randomizing the initial weights, so that the inputs of each activation function fall within the sweet range of the activation function.

Ideally, none of the neurons should start with a trapped situation.

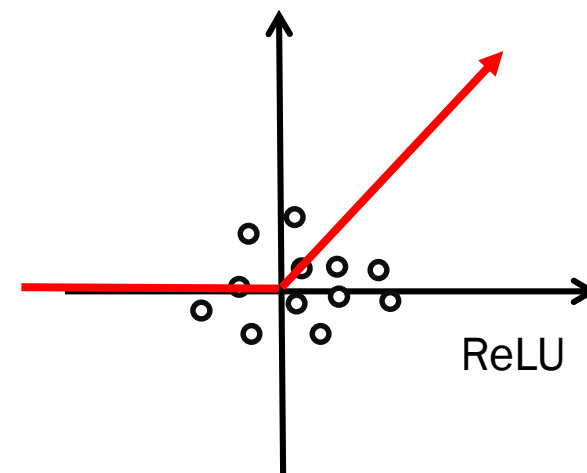
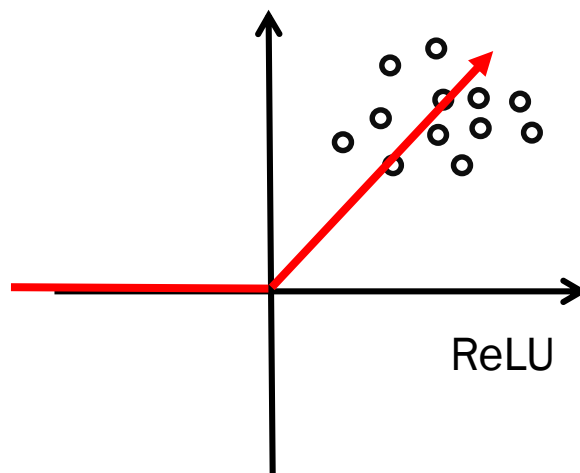


Batch normalization

Normalize distribution of each input feature in each layer across each minibatch to Normal distribution $N \sim (0,1)$



Nothing happens!



Better(greater) learning rate
Faster convergence

CIFAR10

airplane



automobile



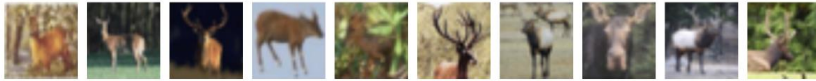
bird



cat



deer



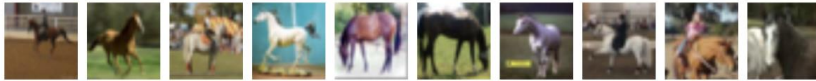
dog



frog



horse



ship



truck



- 60K 32x32 colour images
- 10 classes (6 K images per class)
- 50K training images
- 10K test images

<https://www.cs.toronto.edu/~kriz/cifar.html>

Lab 4B: Convolutional Neural Network (CIFAR10)

1. Check the model define (Conv2D)
2. Check the result by varying learning rate
3. Check the result with different number of Epoch
4. Check the result with more networks
5. Check the result with different activation functions



**Break
room**

Session ends:

Thank you!