

# NYCU DL

## Lab1 - Backpropagation

TA 曾昱仁

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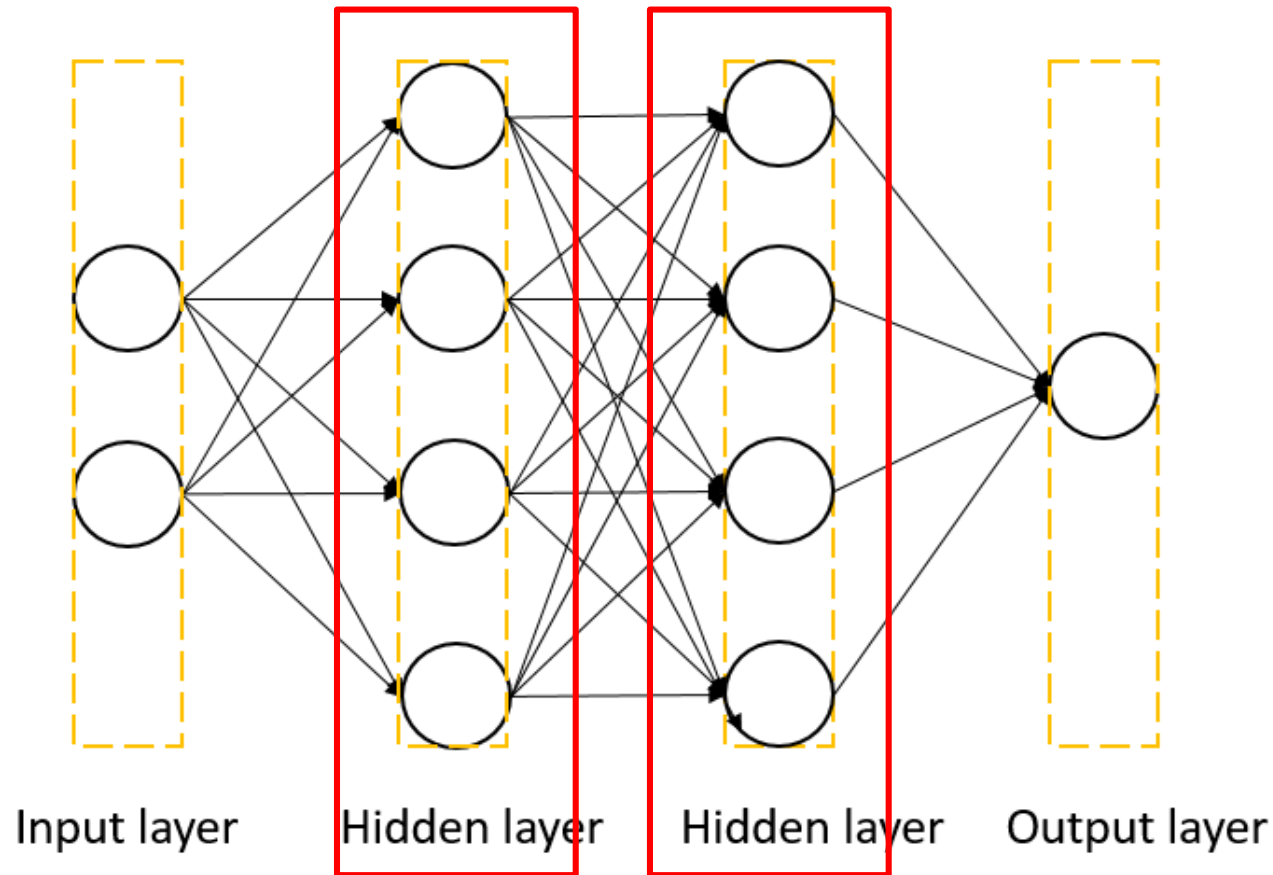
# Outline

- Lab Objective
- Lab Description
- Scoring Criteria
- Time Schedule

# Lab Objective

# Lab Objective

- In this lab, you will need to understand and implement a simple neural network with forward and backward pass using two hidden layers



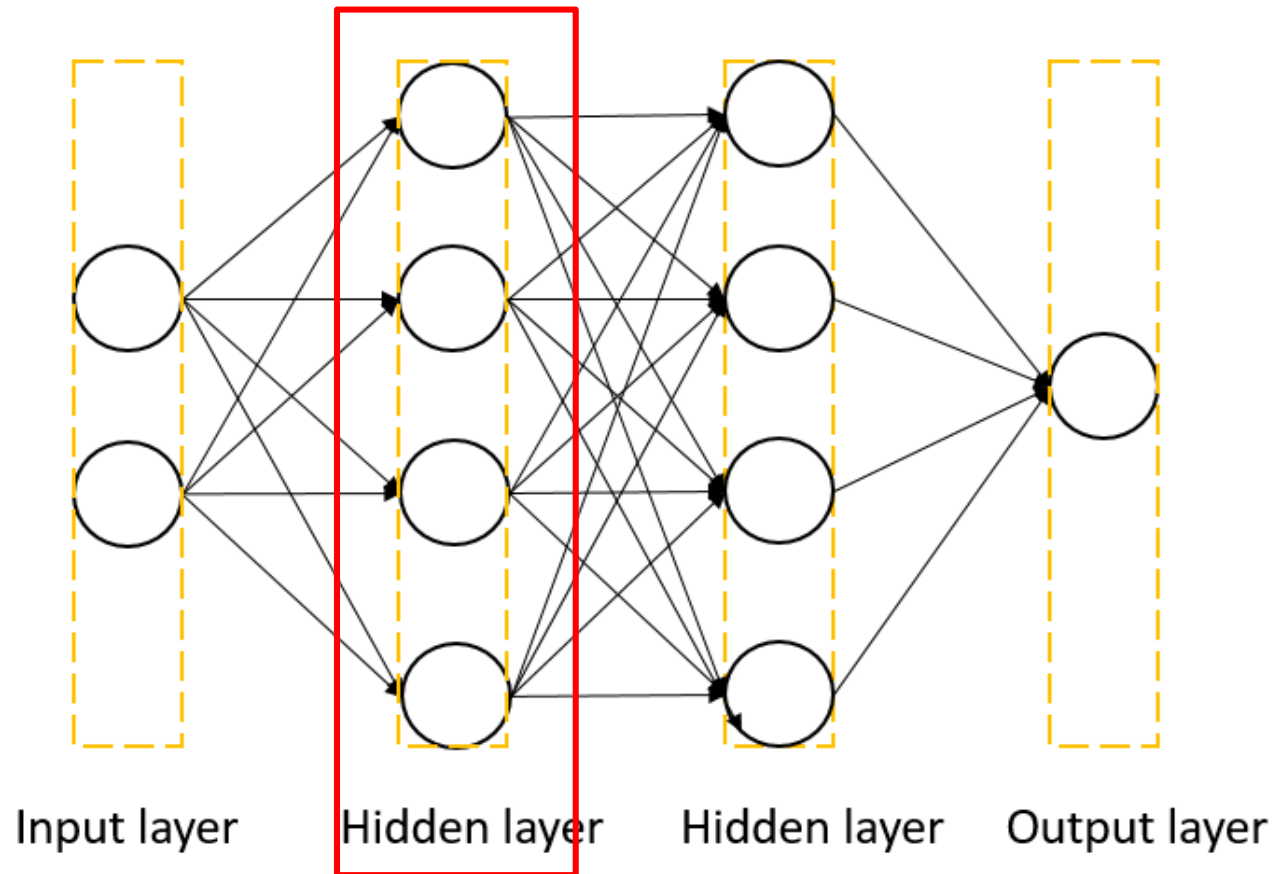
# Lab Description

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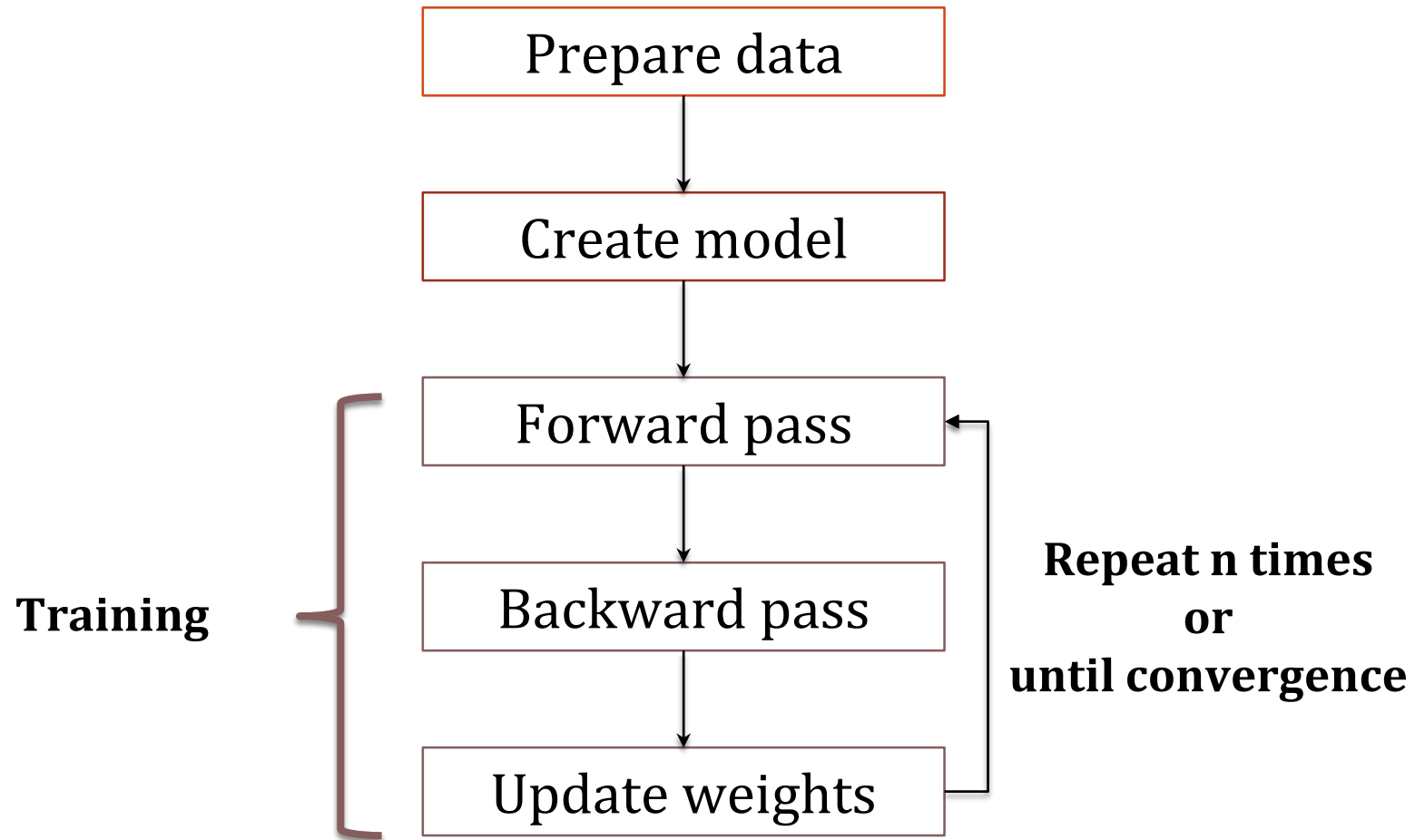
- Implement a simple neural network with two hidden layers.
- You can only use **Numpy** and other **python standard libraries**.
  - **PyTorch, TensorFlow,..... are not allowed.**
- Visualize the results in report.
  - Plot your comparison figure showing the predictions and ground truth.
  - Plot your learning curve (loss, epoch).
  - Print the accuracy of your prediction.
  - matplotlib... is allowed in this part.

# Lab Description

- Each Layer should contain at least **one transformation** (e.g. Linear, CNN,...) and **one activation function** (e.g. sigmoid, tanh....)

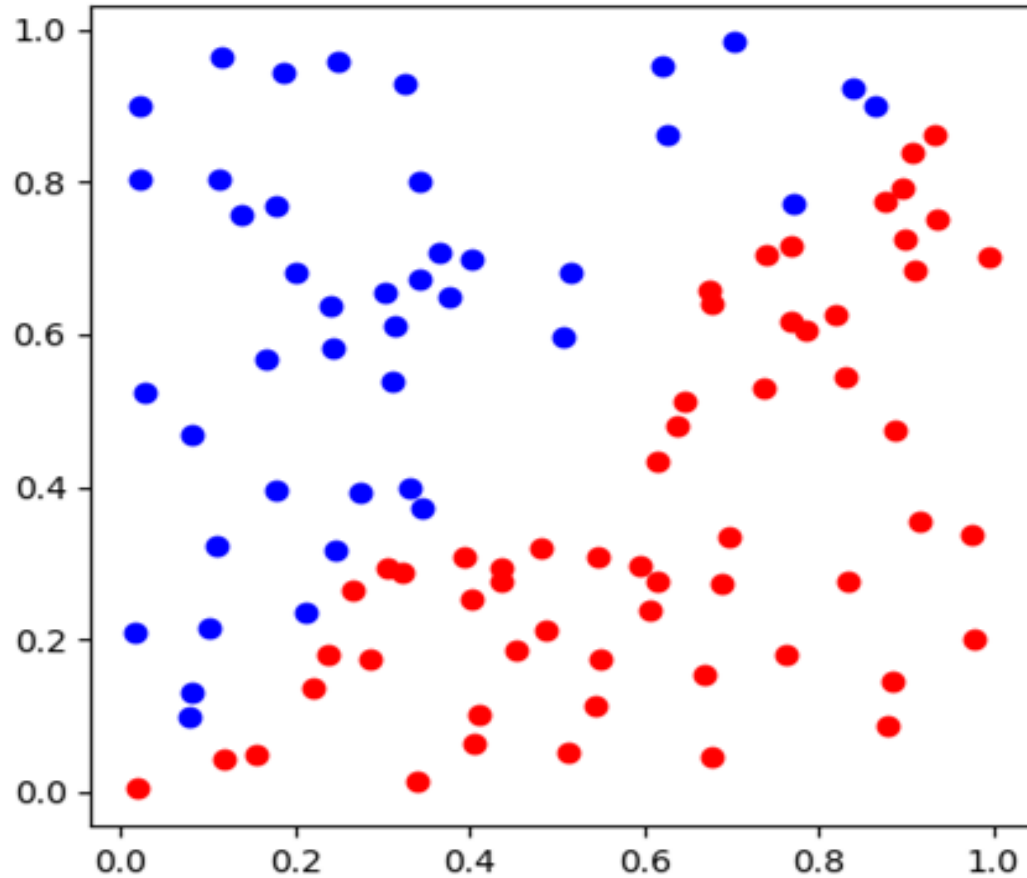


# Lab Description – Flowchart

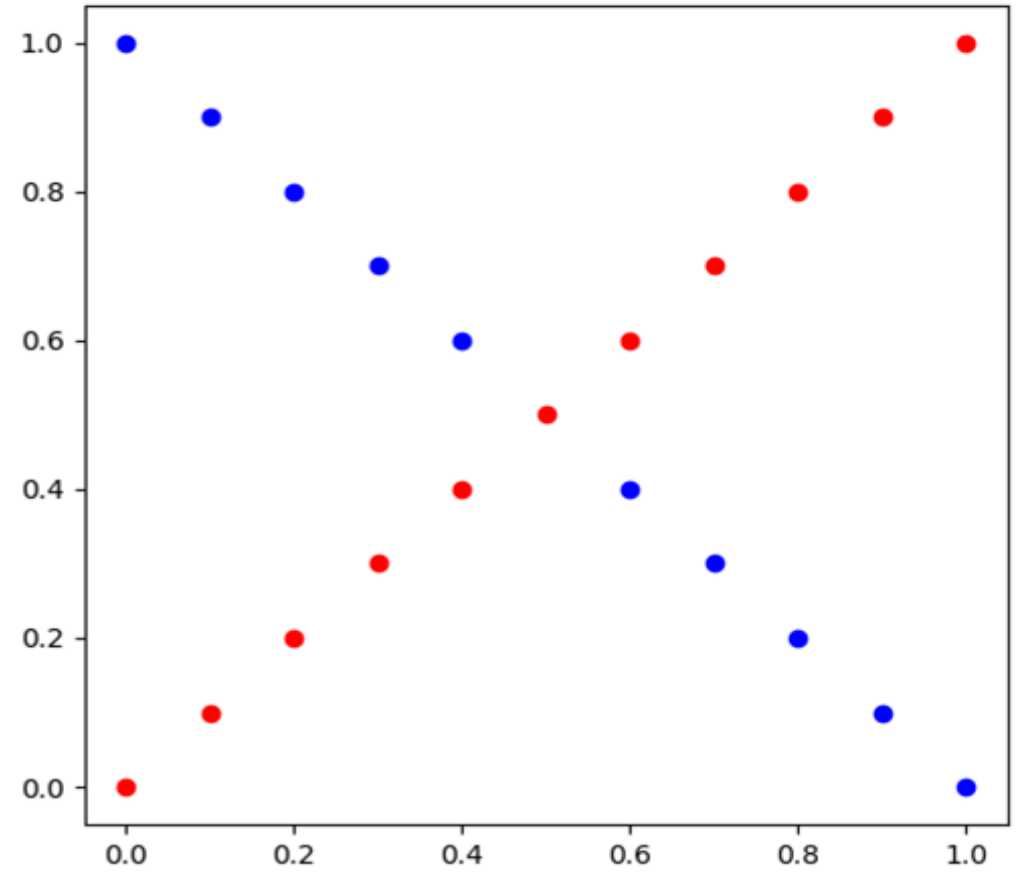




# Lab Description - Data



generate\_linear()



generate\_XOR\_easy()

# Input data generate

Training & Testing use same data

```
def generate_linear(n=100):  
    import numpy as np  
    pts = np.random.uniform(0, 1, (n, 2))  
    inputs = []  
    labels = []  
    for pt in pts:  
        inputs.append([pt[0], pt[1]])  
        distance = (pt[0]-pt[1])/1.414  
        if pt[0] > pt[1]:  
            labels.append(0)  
        else:  
            labels.append(1)  
    return np.array(inputs), np.array(labels).reshape(n, 1)
```

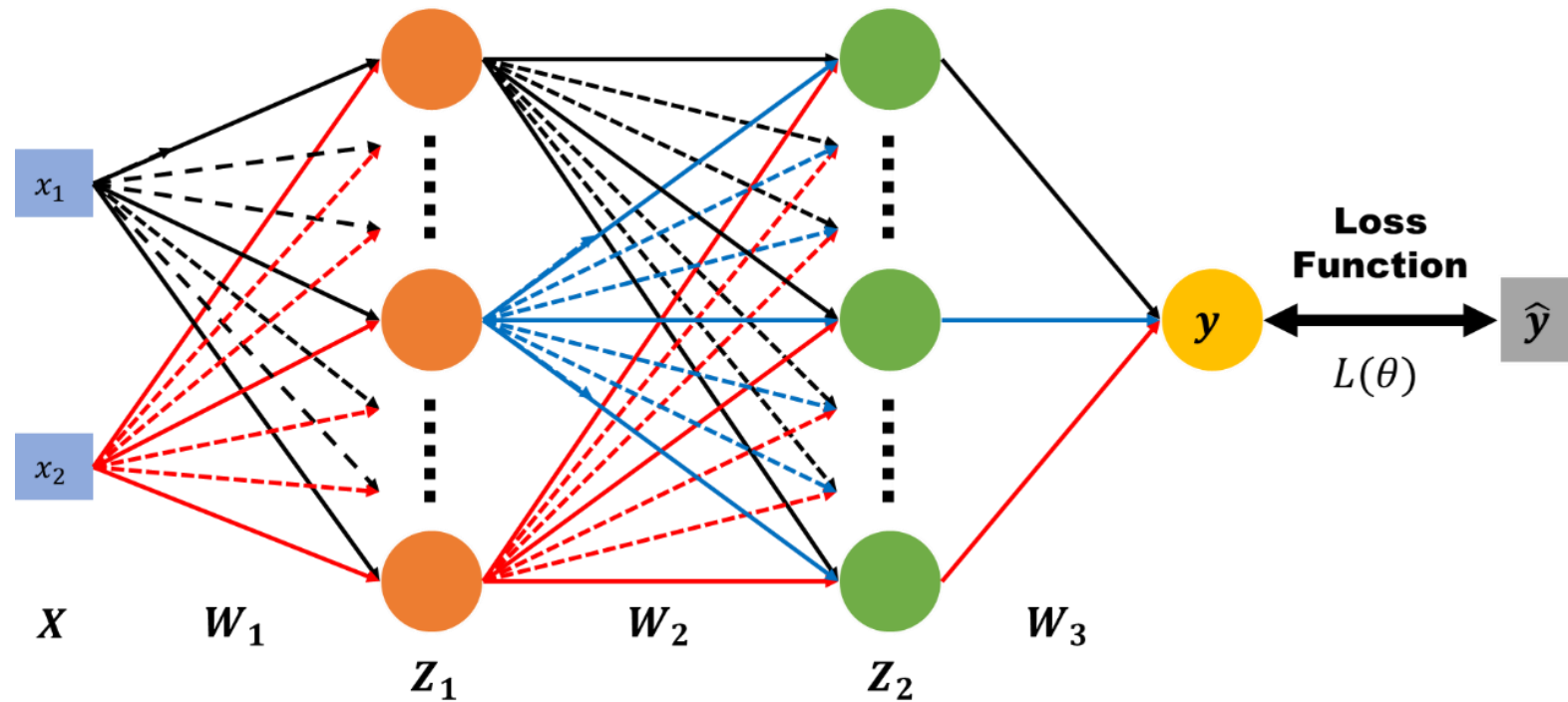
```
def generate_XOR_easy():  
    import numpy as np  
    inputs = []  
    labels = []  
  
    for i in range(11):  
        inputs.append([0.1*i, 0.1*i])  
        labels.append(0)  
  
        if 0.1*i == 0.5:  
            continue  
  
        inputs.append([0.1*i, 1-0.1*i])  
        labels.append(1)  
  
    return np.array(inputs), np.array(labels).reshape(21, 1)
```

Function usage

```
[x, y = generate_linear(n=100)]  
[x, y = generate_XOR_easy()]
```

Don't overwrite these functions!!!

# Lab Description – Architecture



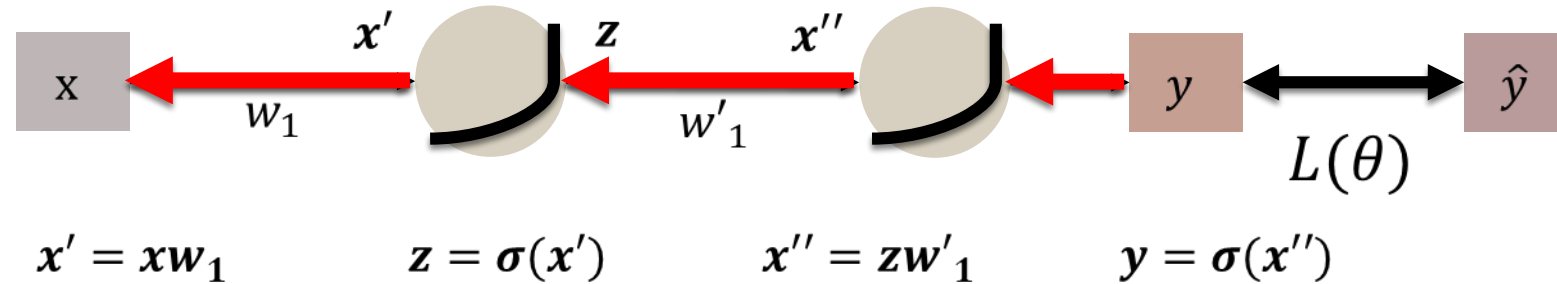
$X : [x_1, x_2]$

$y : \text{outputs}$

$\hat{y} : \text{ground truth}$

$W_1, W_2, W_3 : \text{weight matrix of network layers}$

# Lab Description – Backward



## Chain rule

$$y = g(x) \quad z = h(y)$$

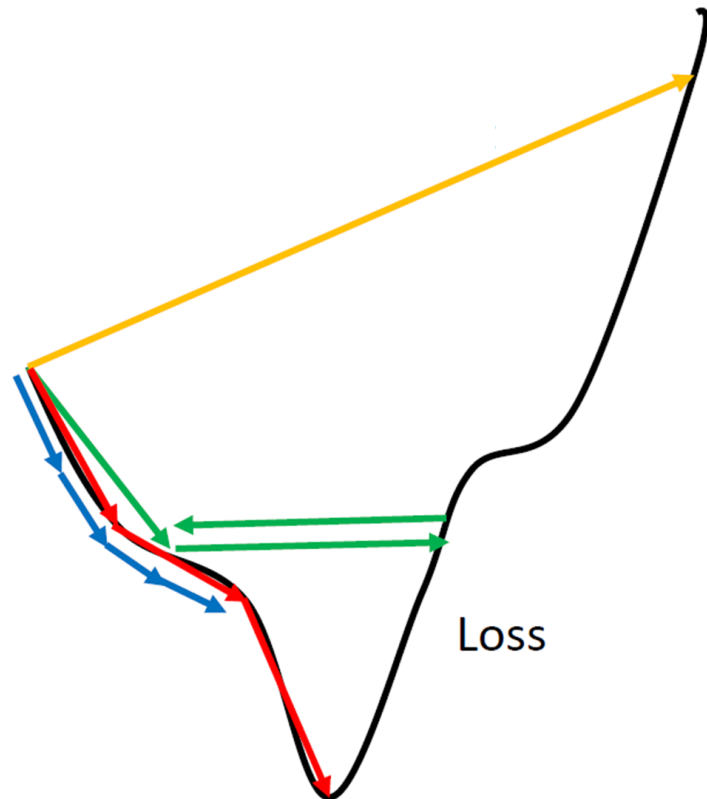
$$\mathbf{x} \xrightarrow{g()} \mathbf{y} \xrightarrow{h()} \mathbf{z}$$

$$\frac{dz}{dx} = \frac{dz}{dy} \frac{dy}{dx}$$

$$\begin{aligned} \frac{\partial L(\theta)}{\partial w_1} &= \frac{\partial y}{\partial w_1} \frac{\partial L(\theta)}{\partial y} \\ &= \frac{\partial y}{\partial x''} \frac{\partial L(\theta)}{\partial y} \frac{\partial x''}{\partial w_1} \\ &= \frac{\partial y}{\partial x''} \frac{\partial L(\theta)}{\partial y} \frac{\partial x''}{\partial z} \frac{\partial z}{\partial w_1} \\ &= \frac{\partial y}{\partial x''} \frac{\partial L(\theta)}{\partial y} \frac{\partial x''}{\partial z} \frac{\partial z}{\partial x'} \frac{\partial x'}{\partial w_1} \end{aligned}$$

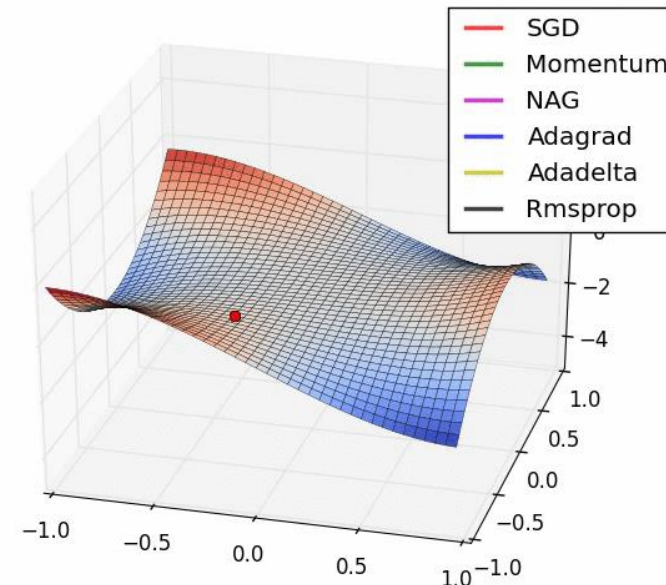
# Lab Description – Gradient descent

**Network Parameters**  $\theta = \{w_1, w_2, w_3, w_4, \dots\}$



$$\begin{aligned}\theta^1 &= \theta^0 - \rho \nabla L(\theta^0) \\ \theta^2 &= \theta^1 - \rho \nabla L(\theta^1) \\ \theta^3 &= \theta^2 - \rho \nabla L(\theta^2)\end{aligned}$$

$\rho$  : Learning rate



# Lab Description - Prediction

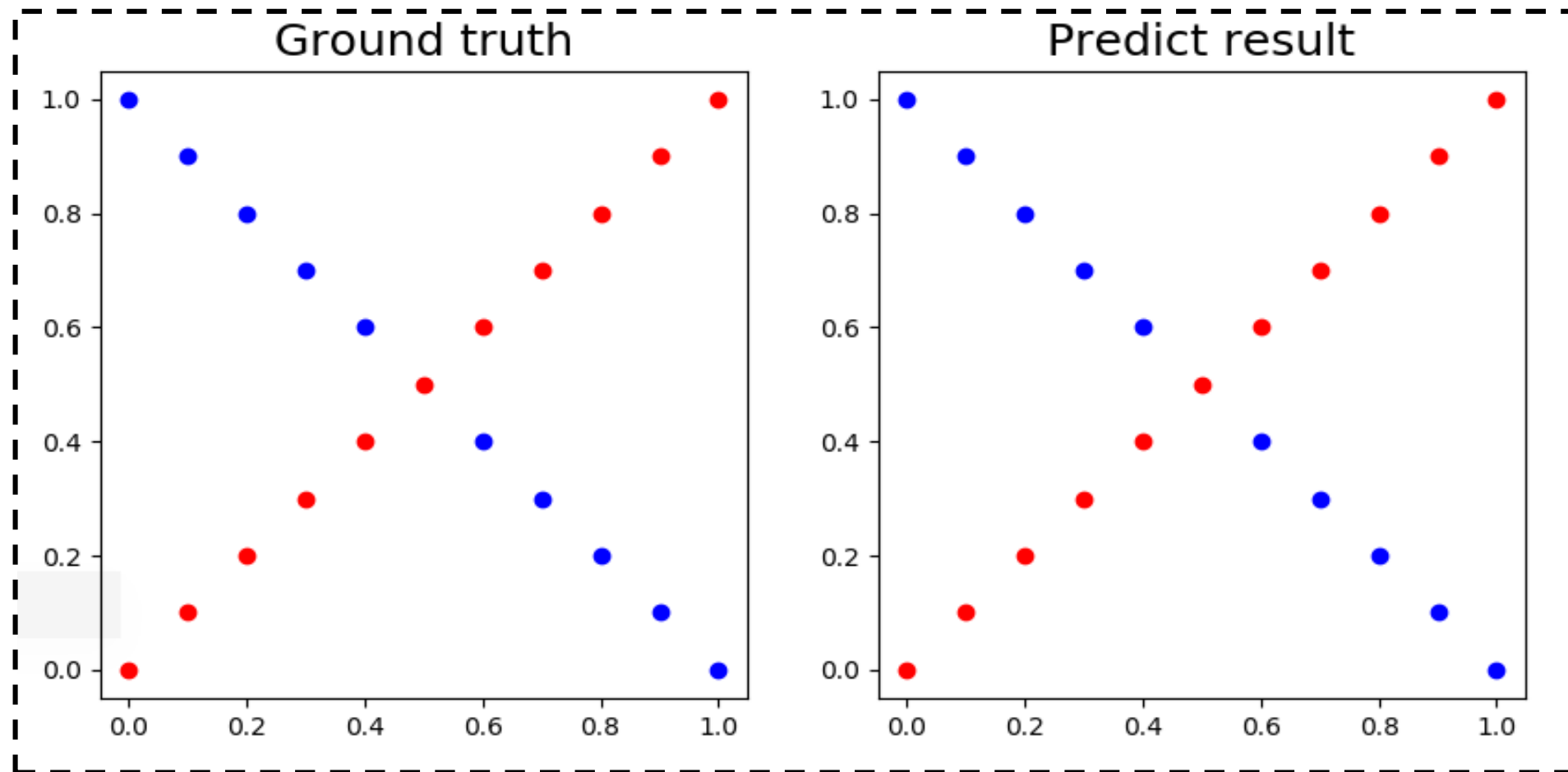
- In the training, you need to print loss
- In the testing, you need to show your predictions, also the accuracy

```
epoch 10000 loss : 0.16234523253277644
epoch 15000 loss : 0.2524336634177614
epoch 20000 loss : 0.1590783047540092
epoch 25000 loss : 0.22099447030234853
epoch 30000 loss : 0.3292173477217561
epoch 35000 loss : 0.40406233282426085
epoch 40000 loss : 0.43052897480298924
epoch 45000 loss : 0.4207525735586605
epoch 50000 loss : 0.3934759509342479
epoch 55000 loss : 0.3615008372106921
epoch 60000 loss : 0.33077879872648525
epoch 65000 loss : 0.30333537090819584
epoch 70000 loss : 0.2794858089741792
epoch 75000 loss : 0.25892812312991587
epoch 80000 loss : 0.24119780823897027
epoch 85000 loss : 0.22583656353511342
epoch 90000 loss : 0.21244497028971704
epoch 95000 loss : 0.2006912468389013
```

```
[[0.01025062]
 [0.99730607]
 [0.02141321]
 [0.99722154]
 [0.03578171]
 [0.99701922]
 [0.04397049]
 [0.99574117]
 [0.04162245]
 [0.92902792]
 [0.03348791]
 [0.02511045]
 [0.94093942]
 [0.01870069]
 [0.99622948]
 [0.01431959]
 [0.99434455]
 [0.01143039]
 [0.98992477]
 [0.00952752]
 [0.98385905]]
```

# Lab Description - Prediction

- Visualize the predictions and ground truth at the end of the training process



# Scoring Criteria



# Scoring Criteria

- Report (40%)
- Demo(60%)
  - Experimental results (40%)
    - You have to achieve **at least 90% of accuracy to get the demo score.**
  - Questions (20%)
- Extra (10%)
  - Implement different optimizers. (2%)
  - Implement different activation functions. (3%)
  - Implement convolutional layers. (5%)

# Report format

1. Introduction (20%)
2. Experiment setups (30%):
  - A. Sigmoid functions
  - B. Neural network
  - C. Backpropagation
3. Results of your testing (20%)
  - A. Screenshot and comparison figure
  - B. Show the accuracy of your prediction
  - C. Learning curve (loss, epoch curve)
  - D. Anything you want to present
4. Discussion (30%)
  - A. Try different learning rates
  - B. Try different numbers of hidden units
  - C. Try without activation functions
  - D. Anything you want to share
5. Extra (10%)
  - A. Implement different optimizers. (2%)
  - B. Implement different activation functions. (3%)
  - C. Implement convolutional layers. (5%)

# Time Schedule

# Important Date

- Assignment Deadline: 3/26 (Tue.) 11:59 a.m.
- Demo date: 3/26 (Tue.)
- Zip all files in one file
  - Report (.pdf)
  - Source code
- name it like 「DL\_LAB1\_yourstudentID\_name.zip」
  - ex: 「DL\_LAB1\_312554018\_曾显仁.zip」
  - **If the zip file name or the report spec have format error, you will get a penalty of 5 points.**

# Time schedule

	LAB1 Back-Propagation	LAB2 CNN	LAB3 CNN	LAB4 RNN+VAE	LAB5 MaskGIT	LAB6 Generative Models
Announce	3/12 (Tabc)	3/26 (Tabc)	4/2 (Tabc)	4/11 (Rn56)	5/7 (Tabc)	5/21 (Tabc)
DEMO	3/26 (Tabc)	4/11 (Rn56)	4/11 (Rn56)	5/7 (Tabc)	TBD	No demo

# Demo schedule

- Demo date: 3/26 (Tue.)
  - The Lab 2 announcement is scheduled for 3/26 initially, and the demo starting at around 7 pm.
  - Each person will be allocated approximately 5 minutes.
  - If the scheduled time is inconvenient, please contact the respective TA to arrange an alternative demo session.

link

	Google meet link		
	TA	曾昱仁	OK/Please come in
14:00	student 1		
14:05	student 2		
14:10	student 3		
14:15			
14:20			
14:25			
14:30			

# Reference

1. <http://www.denizyuret.com/2015/03/alec-radfords-animations-for.html>
2. [http://speech.ee.ntu.edu.tw/~tlkagk/courses\\_ML17\\_2.html](http://speech.ee.ntu.edu.tw/~tlkagk/courses_ML17_2.html)