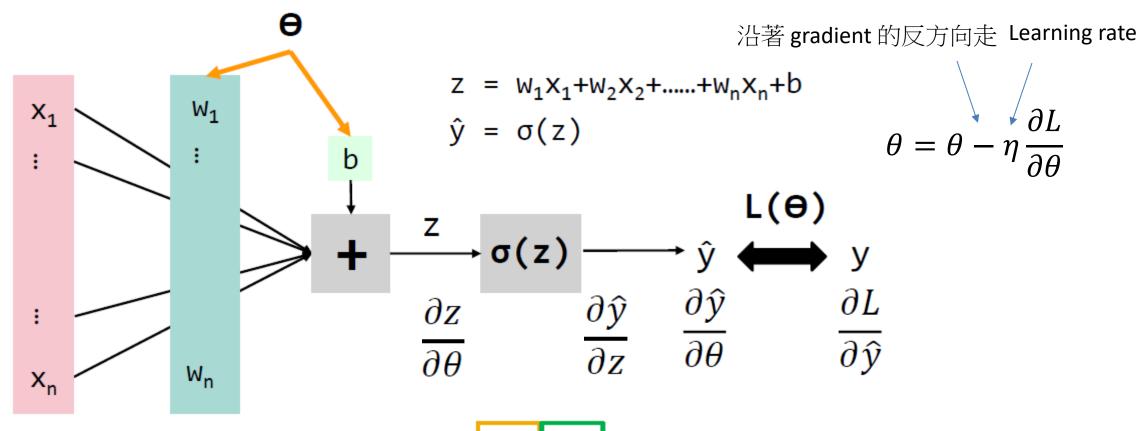
Lab1 Backpropagation and Basic Pytorch

- Backpropagation
- Dataset
- Task1
- Task2
- Regulations

- Backpropagation
- Dataset
- Task1
- Task2
- Rules

Back Propagation



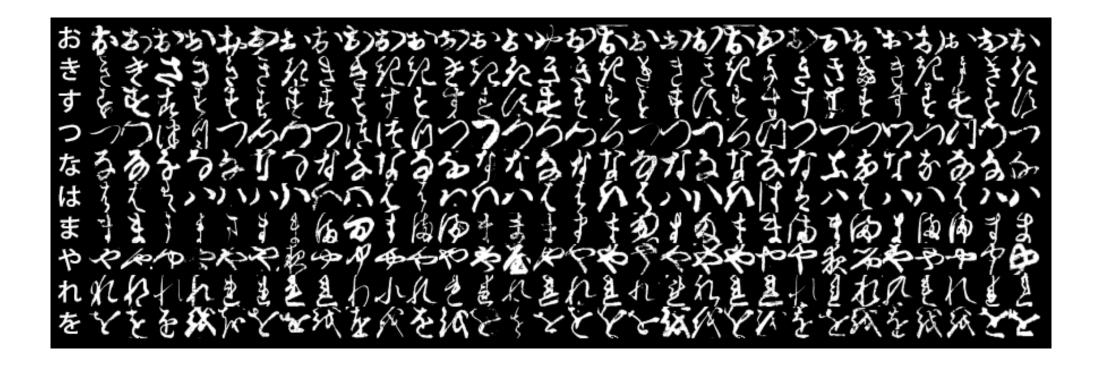
$$\frac{\partial L}{\partial \theta} = \frac{\partial L}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial \theta} = \frac{\partial L}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial z} \frac{\partial z}{\partial \theta}$$

Chain rule (又叫 Back propagation 倒傳遞)

- Backpropagation
- Dataset
- Task1
- Task2
- Rules

Dataset: Kuzushiji-MNIST

- Image Classification
- Classes: 10, image size =28*28 grayscale
- Training: 50,000 Validation: 10,000 Testing: 10,000



- Backpropagation
- Dataset
- Task1
- Task2
- Rules

Task 1 – NN from scratch

- You need to build a neural network from scratch
 - layer.py: Define the functions of each layer (include forward & backward)
 - network.py: Build a network with layers you defined in layer.py
 - Lab1_task1.ipynb : Decide the hyperparameters and run your code
 - The settings for training have been written by TA
- For task1, you are not allowed to use deep learning frameworks.

Package available in Task 1

- numpy
- pandas
- Scikit-learn
- Pytorch, tensorflow, keras

model/layer.py

```
import numpy as np
## by yourself .Finish your own NN framework
## Just an example. You can alter sample code anywhere.
class Layer(object):
    def init (self):
        pass
    def forward(self, *input):
        r"""Define the forward propagation of this layer.
        Should be overridden by all subclasses.
        raise NotImplementedError
    def backward(self, *output grad):
        r"""Define the backward propagation of this layer.
        Should be overridden by all subclasses.
        raise NotImplementedError
## by yourself .Finish your own NN framework
class FullyConnected( Layer):
   def init (self, in features, out features):
    def forward(self, input):
        return output
   def backward(self, output grad):
        return input grad
```

```
## by yourself .Finish your own NN framework
class Activation1( Layer):
    def init (self):
        pass
    def forward(self, input):
        return output
    def backward(self, output grad):
       return input grad
class SoftmaxWithloss( Layer):
    def init (self):
       pass
    def forward(self, input, target):
       return predict, your loss
    def backward(self):
        return input grad
```

model/network.py

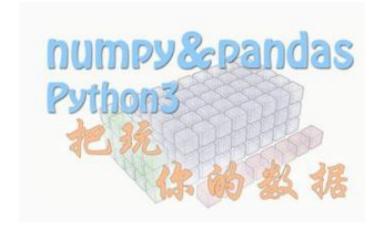
```
from .layer import *
class Network (object):
    def init (self):
        ## by yourself .Finish your own NN framework
        ## Just an example. You can alter sample code anywhere.
    def forward(self, input, target):
        ## by yourself .Finish your own NN framework
        return pred, loss
    def backward(self):
        ## by yourself .Finish your own NN framework
    def update(self, lr):
        ## by yourself .Finish your own NN framework
        ## Hint: You should update weight and bias with learning rate
```

Task1 Overview

- You need to be familiar with jupyter python and useful library
 - numpy, pandas, matplot, etc...

- If you are not familiar with python, I strongly recommend you to see 莫凡's python tutorial. (or other online materials)
 - https://morvanzhou.github.io/







Source: 2023 DL Lab1

- Backpropagation
- Dataset
- Task1
- Task2
- Rules

Task2 – Build neural network using PyTorch

- In Lab1_task2.ipynb
 - You need to rewrite the network you built in Task 1 using PyTorch
 - The settings for training have been written by TA
 - Network should be the same as Task 1
 - You may only use pytorch in Task 2

Task2 Code

Hyperparameters

```
EPOCH =
Batch_size = # 10000 should be divisible by batch_num
Learning_rate =
```

Criterion and Optimizer

```
import torch.optim as optim

criterion =
optimizer =
```

Build the network using pytorch

```
import torch.nn as nn
import torch.nn.functional as F
class Net(nn.Module):
    def init (self):
        super().__init__()
    def forward(self, input):
        return
net = Net()
```

PyTorch tutorial

- Official tutorial
 - https://pytorch.org/tutorials/
- 莫凡
 - https://mofanpy.com/tutorials/machine-learning/torch/
- AssemblyAI PyTorch Crash Course
 - https://www.youtube.com/watch?v=OlenNRt2bjg

- Backpropagation
- Dataset
- Task1
- Task2
- Rules

Assignment Regulation

- Please use Google Colab to finish this lab
- Please refer to "Colab_Tutorial_2024.pptx" for further details
- You don't need to use GPU in this lab

Join Kaggle competition

- Competition URL:
 - https://www.kaggle.com/t/9a54f52550064f998d346bdff61dd47c

2024 DL Lab1 Backpropagation and basic pytorch



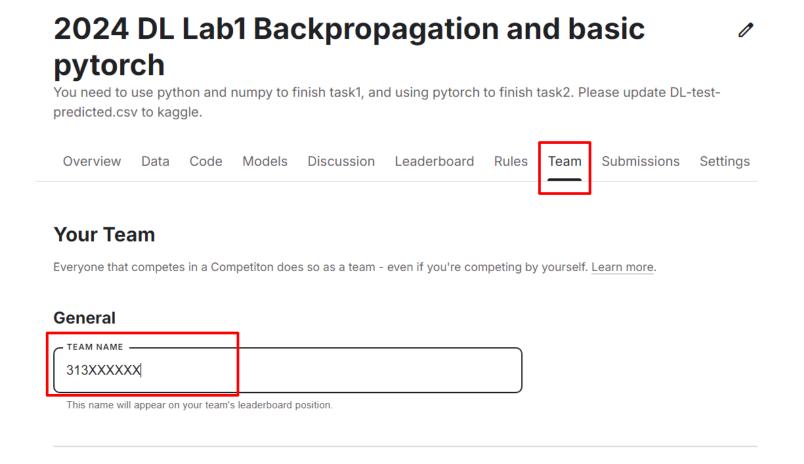
You need to use python and numpy to finish task1, and using pytorch to finish task2. Please update DL-test-predicted.csv to kaggle.



- Maximum daily submission limit: 20
- Change your team name to your student ID

Change your team name to your student ID

If TA can not find your ID, you will loss 30% of score for this lab directly



Reminder

- Submit Deadline: 2 weeks (2024/9/23 11:59 PM)
- You need to submit your code and result to New E3

- Hand in your code and in the following format (5 files)
 - Lab1_task1_studentid.ipynb
 - network.py
 - layer.py
 - Lab1_task2_studentid.ipynb
 - Lab1_report_studentid.pdf

Grading policy

- Lab1
 - Submit your homework to E3 (60%)
 - Task1 (40%) (validation accuracy should >= 85%, put the screenshot in your report)
 - Task2 (20%) (validation accuracy should >= 85%, put the screenshot in your report)
 - Performance (30%) (submit to Kaggle, test accuracy should >= 82%)
 - The results you upload to Kaggle should be generated only by task1
 - Report (10%)
- Please do not plagiarize (0 points will be calculated if caught)

Report

- How to improve the accuracy (list your method)
 - Your network?
 - Loss function?
 - Activation function?
 - Hyperparameters?
 - Etc...
- What differences do you find between the results of Task1 and Task2?