



Institute of Electronics
National Yang Ming Chiao Tung University
Hsinchu, Taiwan

AI Training Course Series

Introduction to Python, Numpy and Pytorch

Lecture 0-2



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Outline

- Jupyter Notebook
- Introduction to Python
- Introduction to Numpy
- Introduction to Pytorch
- References

Jupyter Notebook on Server

Jupyter Notebook

- Jupyter Notebook:
 - 基於Web的交互式計算環境
 - 利用直譯式語言的特性，容易做到資料視覺化及逐步執行
- Jupyter Notebook on Lab server:
 - 可利用實驗室server的GPU (240, 244和245才有)
 - 可直接使用server上的dataset (ImageNet, COCO)

Step 1

- 登入server (140.113.225.245)
- 在command打：
 - jupyter notebook --generate-config

```
=====
                          Advanced Design Automation Research Laboratory
                          Advanced Computer Architecture Research Laboratory
=====
```

```
實驗室可用server IP:
```

```
140.113.225.241 / 140.113.225.242 / 140.113.225.243 / 140.113.225.244 / 140.113.225.245
```

```
實驗室專題生專用server IP:
```

```
140.113.225.245
```

```
-----
- GPU: 1 * GTX 1080Ti
=====
```

```
(base) [M108ihtseng@eng05 ~]$ jupyter notebook --generate-config
```

Step 2 (1/2)

- 進入conda環境
- 按照下圖輸入
 - ***\$ python***
 - ***>>> from notebook.auth import passwd***
 - ***>>> passwd()***

```
(torch) [TA@eng05 ~]$ python
Python 3.10.11 (main, May 16 2023, 00:28:57) [GCC 11.2.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> from notebook.auth import passwd
>>> passwd()
Enter password: ➔ 自己設定 jupyter notebook 的密碼
Verify password:
```

```
'argon2:$argon2id$v=19$m=10240,t=10,p=8$cm0GvNA6MJYRjIcqhucjw$0NZ+DIx42rGVo2Vy0mWBAnYe+FP5RidPvpcizQGBc7s'
```

```
>>> exit()
```

```
(torch) [TA@eng05 ~]$
```

每個人不一樣
根據你設定的密碼產生

要複製下來

Step 2 (2/2)

- 如果出現下圖中的Error

```
>>> from notebook.auth import passwd
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
ModuleNotFoundError: No module named 'notebook'
```

- 在conda環境下輸入以下指令，再進行上一頁的步驟
 - ***\$ conda install -c anaconda notebook***

```
(torch) [2024TA@eng05 ~]$ conda install -c anaconda notebook
```

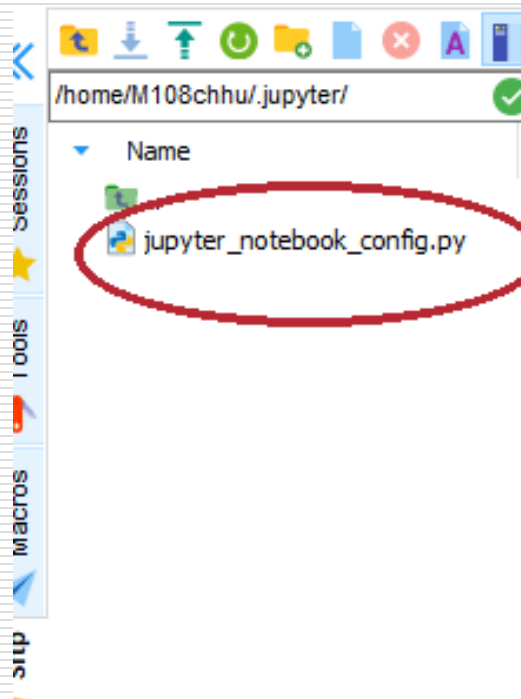
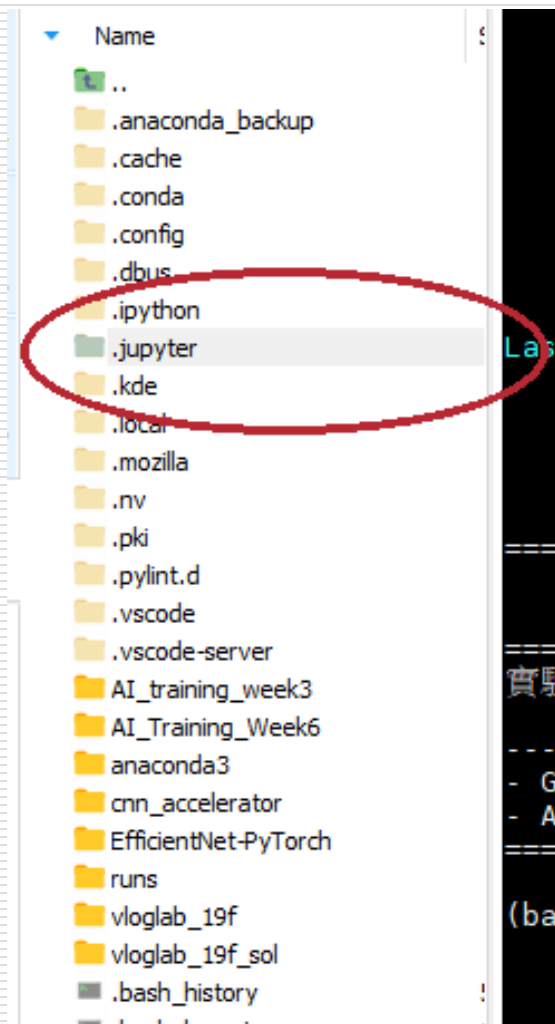
```
Proceed ([y]/n)? y

Downloading and Extracting Packages:

Preparing transaction: done
Verifying transaction: done
Executing transaction: done
(torch) [2024TA@eng05 ~]$
```

Step 3 (1/2)

- 打開jupyter_notebook_config.py



Step 3 (2/2)

- 修改以下四行並儲存 (利用 Ctrl+F 搜尋)

```
c.ServerApp.ip = '0.0.0.0'
```

※ 四行都要記得取消註解

※ 四行前面都不要留空格

```
c.ServerApp.open_browser = False
```

```
c.ServerApp.password = 'argon2:$argon2id$v=19$m=10240,t=10,p=8$FCsVGn7M2zPOubGZ1
```

你剛剛複製的東西

```
c.ServerApp.port = 8017
```

8000—8099挑一個
盡量不要跟別人衝到


Step 4

- 回到terminal，在base環境打 jupyter notebook
- 打開瀏覽器，網址輸入140.113.225.245:[你挑的port]
 - E.g. 140.113.225.245:8017

```
(base) [2024TA@eng05 ~]$ jupyter notebook
[2024-06-05 11:14:59.648 ServerApp] Package notebook took 0.0000s to import
[I 2024-06-05 11:14:59.657 ServerApp] Package jupyter_lsp took 0.0093s to import
[W 2024-06-05 11:14:59.657 ServerApp] A `jupyter_server_extension_points` function was not found in jupyter_lsp. Instead, a `jupyter_server_extension_paths` function was found and will be used for now. This function name will be deprecated in future releases of Jupyter Server.
[I 2024-06-05 11:14:59.662 ServerApp] Package jupyter_server_terminals took 0.0040s to import
[I 2024-06-05 11:14:59.662 ServerApp] Package jupyterlab took 0.0000s to import
[I 2024-06-05 11:14:59.697 ServerApp] Package notebook_shim took 0.0000s to import
[W 2024-06-05 11:14:59.697 ServerApp] A `jupyter_server_extension_points` function was not found in notebook_shim. Instead, a `jupyter_server_extension_paths` function was found and will be used for now. This function name will be deprecated in future releases of Jupyter Server.
[I 2024-06-05 11:15:00.492 ServerApp] Package panel.io.jupyter_server_extension took 0.7947s to import
[I 2024-06-05 11:15:00.492 ServerApp] jupyter_lsp | extension was successfully linked.
[I 2024-06-05 11:15:00.496 ServerApp] jupyter_server_terminals | extension was successfully linked.
[I 2024-06-05 11:15:00.500 ServerApp] jupyterlab | extension was successfully linked.
[W 2024-06-05 11:15:00.504 ServerApp] ServerApp.password config is deprecated in 2.0. Use PasswordIdentityProvider.hash_password.
[I 2024-06-05 11:15:00.504 ServerApp] notebook | extension was successfully linked.
[I 2024-06-05 11:15:00.722 ServerApp] notebook_shim | extension was successfully linked.
[I 2024-06-05 11:15:00.722 ServerApp] panel.io.jupyter_server_extension | extension was successfully linked.
[I 2024-06-05 11:15:00.735 ServerApp] notebook_shim | extension was successfully loaded.
[I 2024-06-05 11:15:00.737 ServerApp] jupyter_lsp | extension was successfully loaded.
[I 2024-06-05 11:15:00.738 ServerApp] jupyter_server_terminals | extension was successfully loaded.
[I 2024-06-05 11:15:00.740 LabApp] JupyterLab extension loaded from /home/2024TA/anaconda3/lib/python3.11/site-packages/jupyterlab
[I 2024-06-05 11:15:00.740 LabApp] JupyterLab application directory is /home/2024TA/anaconda3/share/jupyter/lab
[I 2024-06-05 11:15:00.740 LabApp] Extension Manager is 'pypi'.
[I 2024-06-05 11:15:00.742 ServerApp] jupyterlab | extension was successfully loaded.
[I 2024-06-05 11:15:00.745 ServerApp] notebook | extension was successfully loaded.
[I 2024-06-05 11:15:00.745 ServerApp] panel.io.jupyter_server_extension | extension was successfully loaded.
[I 2024-06-05 11:15:00.745 ServerApp] Serving notebooks from local directory: /home/2024TA
[I 2024-06-05 11:15:00.745 ServerApp] Jupyter Server 2.10.0 is running at:
[I 2024-06-05 11:15:00.746 ServerApp] http://eng05:8056/tree
[I 2024-06-05 11:15:00.746 ServerApp] http://127.0.0.1:8056/tree
[I 2024-06-05 11:15:00.746 ServerApp] Use Control-C to stop this server and shut down all kernels (twice to skip confirmation).
```

- 在用jupyter notebook的時候不要在terminal按到 ctrl+c (不要不小心關掉)
- 設定一次就可以了，下次不需要重新設定config

Jupyter Notebook (1/4)

 jupyter

Password:

輸入你設定的密碼

Jupyter Notebook (2/4)

The screenshot displays the Jupyter Notebook web interface. At the top, the 'jupyter' logo is visible, followed by a navigation bar with 'File', 'View', 'Settings', and 'Help' menus. Below this, a toolbar contains 'Files' and 'Running' tabs. The 'Files' tab is active, showing a file browser with a list of items: 'Name', 'anaconda3', and 'Anaconda3-2024.02-1-Linux-x86_64.sh'. A 'New' button in the top right corner is highlighted with a red box, and its dropdown menu is open, showing options: 'Notebook' (highlighted with a red box), 'Terminal', 'Console', 'New File', and 'New Folder'. In the center of the interface, a 'Select Kernel' dialog box is open. It contains a text input field with the value 'Python 3 (ipykernel)' and a dropdown arrow. Below the input field, there is a checkbox labeled 'Always start the preferred kernel' which is unchecked. To the right of the checkbox are two buttons: 'No Kernel' and 'Select' (highlighted with a red box).

jupyter

File View Settings Help

Files Running

Select items to perform actions on them.

/

☐ Name

☐ anaconda3

☐ Anaconda3-2024.02-1-Linux-x86_64.sh

New

Notebook

Terminal

Console

New File

New Folder

Select Kernel

Select kernel for: "Untitled.ipynb"

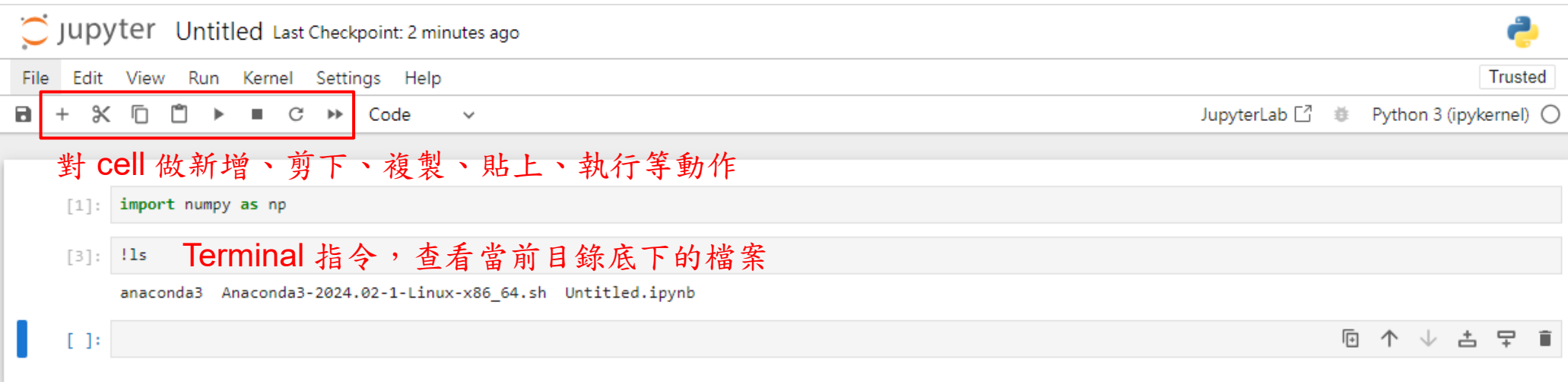
Python 3 (ipykernel)

☐ Always start the preferred kernel

No Kernel

Select

Jupyter Notebook (3/4)



The screenshot shows the JupyterLab interface. At the top, the title bar says "jupyter Untitled Last Checkpoint: 2 minutes ago". Below it is a menu bar with "File", "Edit", "View", "Run", "Kernel", "Settings", and "Help". A toolbar contains icons for creating a new file, opening a recent file, saving, running a cell, and other actions. A red box highlights the icons for creating a new cell, deleting a cell, copying, pasting, and running a cell. The main area shows a code cell with the text `[1]: import numpy as np`. Below it is a terminal window with the text `[3]: !ls` and the output `Terminal 指令，查看當前目錄底下的檔案`. The terminal also shows the shell path `anaconda3 Anaconda3-2024.02-1-Linux-x86_64.sh Untitled.ipynb`.

對 cell 做新增、剪下、複製、貼上、執行等動作

```
[1]: import numpy as np
```

```
[3]: !ls
```

Terminal 指令，查看當前目錄底下的檔案

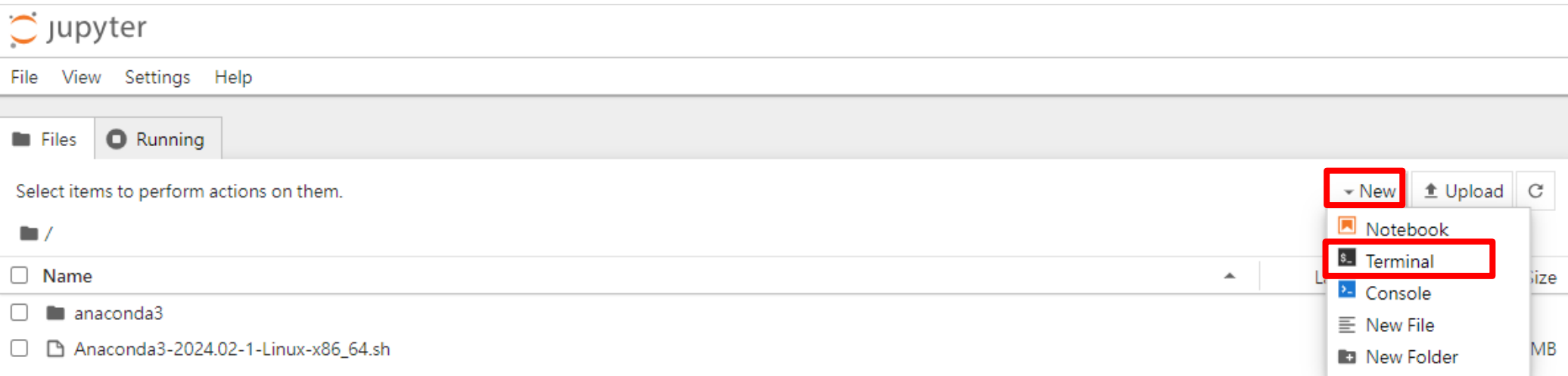
```
anaconda3 Anaconda3-2024.02-1-Linux-x86_64.sh Untitled.ipynb
```

```
[ ]:
```

- 執行cell: shift+enter
- Terminal指令：指令前面加！

Jupyter Notebook (4/4)

- Terminal on Jupyter



The screenshot shows the Jupyter Notebook web interface. At the top, there's a header with the Jupyter logo and the word 'jupyter'. Below it is a navigation bar with 'File', 'View', 'Settings', and 'Help'. The main area has two tabs: 'Files' and 'Running'. Below the tabs, there's a message 'Select items to perform actions on them.' and a file list with columns for checkboxes, names, and sizes. The file list contains three items: 'anaconda3', 'Anaconda3-2024.02-1-Linux-x86_64.sh', and 'Untitled.ipynb'. On the right side of the file list, there's a 'New' dropdown menu that is open, showing options: 'Notebook', 'Terminal', 'Console', 'New File', and 'New Folder'. The 'Terminal' option is highlighted with a red box.



File View Settings Help

```
(base) [2024TA@eng05 ~]$ ls
anaconda3  Anaconda3-2024.02-1-Linux-x86_64.sh  Untitled.ipynb
(base) [2024TA@eng05 ~]$
```

Introduction to Python

Python 特性

- 強調代碼的可讀性和簡潔的語法 (注重空格縮排)
- 直譯式語言(一行一行動態執行)
- 能夠自動管理記憶體使用
- 容易整合其他底層語言 (glue language)
- 完備的標準庫和套件
- 應用廣 (讀寫檔案、機器學習、網站開發)

語法差異

C++

需要分號結尾

用大括號決定語意

C++ for loop

```
for(int i=0; i<100; i++){  
    cout << i;  
}
```

C++ if else

```
if (x>10){  
    .....  
}else if (x>0){  
    .....  
}else{  
    .....  
}
```

Python

不須分號結尾

用縮排決定語意

Python for loop

```
for i in range(100):  
    print(i)
```

May cause indentation error

Python if else

```
if x>10:  
    .....  
elif x>0:  
    .....  
else:  
    .....
```

執行 .py File

- python example.py
- python -i example.py
 - 執行完程式後留在python控制台
 - 互動式窗口，debug用

```
(base) [M108ihtseng@eng04, ~]$ python -i example.py
Hello World!
>>> a = 1
>>> a
1
>>> exit()
(base) [M108ihtseng@eng04, ~]$
```

Variables (1/4)

- 不須事先宣告，直接用“=”賦值
- Number
 - Python3支持 int, float, complex

```
1 a, b, c = 5, 0.3, 7+6j
2 print('Type of a: ', type(a))
3 print('Type of b: ', type(b))
4 print('Type of c: ', type(c))
```

```
Type of a: <class 'int'>
Type of b: <class 'float'>
Type of c: <class 'complex'>
```

- 數值運算：+(加), -(減), *(乘), /(除,得到浮點數),
//(除,得到整數), %(取餘數), **(次方)

Variables (2/4)

- String

```
1 string_ = 'Hello World!'  
2 print(string_ + ' Ha Ha Ha')
```

```
Hello World! Ha Ha Ha
```

- Boolean
 - True, False

List (1/4)

- List
 - 由 0 或多個元素組成
- Assignment

```
list_1 = []           # 0 個元素的 list
list_2 = list(range(5)) # list_2 的元素從 0 開始到 (5-1)
list_3 = ['A', True, 123] # list 的元素可以是不同 type
print(list_1)
print(list_2)
print(list_3)
```

```
[]
[0, 1, 2, 3, 4]
['A', True, 123]
```

List (2/4)

- 串列重複

```
list_1 = [1, 2, 3, 4, 5]
print(list_1 * 2)          # 將 list_1 疊加兩層
```

[1, 2, 3, 4, 5, 1, 2, 3, 4, 5]

- 取得元素值

```
print(list_1[1])
```

2

- 切片

```
print(list_1[2:5])          # 從 list_1[2] 到 list_1[5-1]
```

[3, 4, 5]

List (3/4)

- 刪除元素

```
list_2 = [1, 2, 3, 4, 5]
del list_2[1:4]           # 刪除 list_2 的第 1 個元素到第 (4-1) 個元素
print(list_2)
```

```
[1, 5]
```

- List 資訊

```
list_3 = [5, 6, 3, 4, 1, 2, 0]
print('len:', len(list_3))
print('max:', max(list_3))
print('min:', min(list_3))
```

```
len: 7
max: 6
min: 0
```

List (4/4)

- List 函數

```
list_0 = [1, 2, 3]
list_1 = [True]
list_0.append(4)           # 在 list_0 尾端加上元素 4
print(list_0)
list_0.extend(list_1)      # 在 list_0 尾端加上 list_1
print(list_0)
list_0.insert(1, False)    # 在 list_0 的第 1 個位置加上元素
print(list_0)
a = list_0.pop()           # 取出並刪除最後一個元素
print('Take out:', a, ', Remain:', list_0)
list_0.remove(False)       # 移除 list_0 中第一個 False
print(list_0)
list_0.reverse()           # 反轉 list_0 中的元素
print(list_0)
```

```
[1, 2, 3, 4]
[1, 2, 3, 4, True]
[1, False, 2, 3, 4, True]
Take out: True , Remain: [1, False, 2, 3, 4]
[1, 2, 3, 4]
[4, 3, 2, 1]
```



Tuple

- Tuple

- 類似list，但宣告後不能修改
- 執行速度比 list 快，儲存的資料沒有被修改的風險

```
tuple_1 = (1, 2, 3)
list_1 = (4, 5, 6)
tuple_2 = tuple(list_1)           # 將 list 轉成 tuple
print(type(tuple_1), tuple_1)
print(type(tuple_2), tuple_2)
```

```
<class 'tuple'> (1, 2, 3)
<class 'tuple'> (4, 5, 6)
```

```
tuple_1.append(4)
```

```
-----
AttributeError                                Traceback (most recent call last)
<ipython-input-24-77214de2ca1a> in <module>
----> 1 tuple_1.append(4)
```

```
AttributeError: 'tuple' object has no attribute 'append'
```

Dictionary

- Dictionary
 - list, tuple 都是用數字取得元素值 (ex. list_1[0]) , dictionary 可以用 key 取值
 - 語法 : {key1: value1, key2: value2, ...}

```
1 stu_1 = {'name': 'Jack', 'edge': 13}
2 print(stu_1['name'])
```

Jack

Range

- 用來創建整數列表
- range(結束值)
 - `range(5)` => `[0,1,2,3,4]`
 - `range(0)` => `[]`
- range(起始值, 結束值)
 - `range(1, 5)` => `[1,2,3,4]`
- range(起始值, 結束值, 間隔)
 - `range(0,10,2)` => `[0, 2, 4, 6, 8]`
 - `range(10,1,-2)` => `[10, 8, 6, 4, 2]`

Enumerate

- Enumerate 用於將數據列表組合為一個索引序列，可以同時列出數據和數據索引，通常用於 for loop

```
seasons = ['spring', 'summer', 'fall', 'winter']  
for idx, season in enumerate(seasons):  
    print(idx, season)
```

```
0 spring  
1 summer  
2 fall  
3 winter
```

```
for idx, season in enumerate(seasons, start=1):  
    print(idx, season)
```

```
1 spring  
2 summer  
3 fall  
4 winter
```

Control Flow

- if 條件1 :

...

elif 條件2 :

...

else:

...

- for x in range(5):

...

- while:

...

- break

- continue

For Loop (1/2)

```
1 animal_list = ['cat', 'dog', 'fish']  
2 for i in animal_list:  
3     print(i)
```

```
cat  
dog  
fish
```

```
1 for i in reversed(animal_list):  
2     print(i)    # 列印出順序顛倒的 list
```

```
fish  
dog  
cat
```

For Loop (2/2)

```
1 for i ,animal in enumerate(animals_list):  
2     print(i,':', animal)
```

```
0 : cat  
1 : dog  
2 : fish
```

```
1 color_list = ['yellow', 'white', 'red']  
2 for animal, color in zip(animals_list, color_list):  
3     print(animal, 'is', color)
```

```
cat is yellow  
dog is white  
fish is red
```

Advanced Usage of List

- 搭配 for loop

```
1 Even_numbers = [x for x in range(5) if x%2 ==0]
2 Even_squares = [x*x for x in Even_numbers]
```

```
[0, 2, 4]
[0, 4, 16]
```

- Indexing

```
List[-1] # 擷取 List 最後一個元素
List[2:] # 擷取 List 第三個到最後一個元素
```

- 創建多維 list

```
1 L = [[0 for _ in range(2)] for _ in range(3)]
2 # L為一個 3*2 的 List
3 print(L)
```

```
[[0, 0], [0, 0], [0, 0]]
```



Import Module

- import `module1`
 - `module1.func()`
- from `module1` import `func`
 - `func()`
- import `module1` as `m1`
 - `m1.func()`

```
import torch
import torch.nn as nn
from PIL import Image
```

Def (1/2)

- def func_name (parameters, ...):

```
def printinfo( name, age ):  
  
    print ( "名字: ", name)  
    print ( "年龄: ", age)  
    return
```

```
printinfo( "runoob", 50 )
```

```
名字:  runoob  
年龄:  50
```

```
def printinfo( arg1, *vartuple ):  
  
    print ( "输出: ")  
    print ( "arg1= ", arg1)  
    for var in vartuple:  
        print (var)  
    return
```

```
printinfo( 10 )  
printinfo( 70, 60, 50 )
```

```
输出:  
arg1=  10  
输出:  
arg1=  70  
60  
50
```



Def (2/2)

- 參數傳遞
 - 不可變(類似pass by value): number 、 string 、 tuple
 - 可變(類似pass by reference): list 、 dictionary

```
def ChangeInt( a ):  
    a = 10  
  
b = 5  
ChangeInt(b)  
print(b)
```

5

```
def ChangeList( mylist ):  
    mylist.append([1,2,3,4])  
mylist = [10,20,30]  
ChangeList( mylist )  
print (mylist)
```

[10, 20, 30, [1, 2, 3, 4]]

Global/Local Variables (1/2)

- 在函數內建立的變數為local variable, 在函式外建立的變數為global variable
- 若只是使用 global variable，則直接用就好
- Local variable的遮蔽效果:

```
def test():  
    a = 2    #local variable  
    print(a)
```

```
a = 1 #global variable  
test()  
print(a)
```

```
2  
1
```

```
def test():  
    b = 2    #local variable  
    print(a) # print global a
```

```
a = 1 #global variable  
test()  
print(a)
```

```
1  
1
```

Global/Local Variables (2/2)

- 如果要在函式裡修改global variable的值，要用關鍵字 `global`

```
def test():  
    global a  
    a = 2      # modify global variable a  
    print(a)  # print global variable a
```

```
a = 1 #global variable  
test()  
print(a)
```

2

2

Class

- Class class_name:
 - def __init__(self,parameter,...): 建構子 constructor
 - › 用來宣告和初始化 class 中的變數
 - › self.parameter = ... 宣告這個變數的屬性

```
class Shape:
```

```
    def __init__(self, width=0, height=0):  
        self.width = width  
        self.height = height
```

```
    def area(self):  
        return self.width*self.height
```

→ member function

```
my_obj = Shape()  
print(my_obj.width)  
print(my_obj.height)  
print("Area = ",my_obj.area())  
print("=====")  
my_obj2 = Shape(10)  
print(my_obj2.width)  
print(my_obj2.height)  
print("Area = ",my_obj2.area())
```

```
0  
0  
Area = 0  
=====  
10  
0  
Area = 0
```

Class Inheritance (1/3)

- Class class_name(父類別):
 - def __init__(self,parameter,...):
 - › super().__init__(父類別的parameter)

```
class Shape:

    def __init__(self, width=0, height=0):
        self.width = width
        self.height = height

    def area(self):
        return self.width*self.height

class Prism(Shape): Prim 繼承 Shape

    def __init__(self, width, height,length = 0):
        super().__init__(width, height)
        self.length = length
    def volumn(self):
        return self.width*self.height*self.length
```

Class Inheritance (2/3)

```
my_obj3 = Shape(10, 20)
print(my_obj3.width)
print(my_obj3.height)
print("Area = ",my_obj3.area())
my_obj4 = Prism(10,20,30)
print(my_obj4.width)
print(my_obj4.height)
print(my_obj4.length)
print("Area = ",my_obj4.volumn())
```

10	
20	
Area =	200
10	
20	
30	
Area =	6000

Class Inheritance (3/3)

```
import torch.nn as nn
import torch.nn.function as F
           submodule

class Model(nn.Module): # 繼承 nn.module
    def __init__(self): # Define what modules you need
        super(Model, self).__init__()
        self.fc1 = nn.Linear(2,3)

    def forward(self, x): # Define how to pass data
        out = self.fc1(x)
        out = F.relu(out)
        return out
```

Time

- `import time`
- `ticks = time.time()` → 從1970年1月1日午夜到現在經過幾秒

- Ex :

```
ticks1 = time.time()
```

```
.....
```

一些計算過程

```
.....
```

```
ticks2 = time.time()
```

`second = ticks2 - ticks1` → 計算中間的指令需要幾秒

File I/O

- input()

```
1 str = input('請輸入你的年齡: ')\n2 print('你的年齡是: ', str)
```

```
請輸入你的年齡: 20\n你的年齡是:  20
```

- file = open(filename, mode)

模式	描述
r	只讀,不創建,預設模式
r+	讀寫,不創建
w	只寫,如果該文件不存在,則創建。
w+	可讀可寫,如果該文件不存在,則創建。
a	只寫,附加方式打開,不會覆蓋,如果該文件不存在,則創建。
a+	可讀可寫,附加方式打開,不會覆蓋,如果該文件不存在,則創建。



Print

- % 開頭
- .format()

語法	說明
<code>%s</code>	以 <code>str()</code> 函數輸出文字。
<code>%f</code>	以浮點數方式輸出數值。
<code>%d</code>	以十進位整數方式輸出數值。
<code>%e</code> 、 <code>%E</code>	以科學記號輸出數值。
<code>%o</code>	以八進位整數方式輸出數值。
<code>%x</code> 、 <code>%X</code>	以十六進位整數方式輸出數值。
<code>%c</code>	以字元方式輸出。
<code>%r</code>	以 <code>repr()</code> 函數輸出文字。
<code>%%</code>	輸出 <code>%</code> 百分比符號。

% vs. .format()

1.

```
str = '%s am %d years old ! ' % ('I',20)
print(str)
str = '{} am {} years old ! ' .format('I',20)
print(str)
```

```
I am 20 years old !
I am 20 years old !
```

2.

```
str = '(%10s)' % 'HiHi'
print(str)
str = '({:>10})' .format('HiHi')
print(str)
```

```
(      HiHi)
(      HiHi)
```

3.

```
str = '(%-10s)' % 'HiHi'
print(str)
str = '({:10})' .format('HiHi')
print(str)
```

```
(HiHi      )
(HiHi      )
```

4.

```
str = '%.5s ! ' % 'Hello , World'
print(str)
str = '{:.5} ! ' .format('Hello , World')
print(str)
```

```
Hello !
Hello !
```

5.

```
str = 'PI is (%07.3f) ' % 3.14159
print(str)
str = 'PI is ({:07.3f}) ' .format(3.14159)
print(str)
```

```
PI is (003.142)
PI is (003.142)
```

Parser (1/5)

- `import argparse`

```
ArgumentParser(prog=None, usage=None, description=None, epilog=None)
```

- `prog`:

- › program 的名字，default 是檔名。

- `usage`:

- › 字串，主要是告知使用者說應該怎麼使用你寫的program

- › 保持None的話就會自動根據你設的參數產生相對應的說明字串。

- `description`:

- › 字串，簡短說明程式資訊，會在所有參數說明的前面。

- `epilog`:

- › 字串，補充程式資訊，會在所有參數說明的後面。

Parser (2/5)

- Example

```
from argparse import ArgumentParser
parser=ArgumentParser(prog="prog",usage="Tutorial",description="input var1",epilog="type: int")
parser.print_help()
```

usage: Tutorial → *usage*

input var1 → *description*

optional arguments:

-h, --help show this help message and exit → 參數說明，自動產生

type: int → *epilog*

Parser (3/5)

- Parser.add_argument() 增加參數

- Name or flags

- › 參數的名稱，可以用縮寫，但要有全名。例如:--target, -t

- dest

- › 當parse_args()解析完後的參數名稱。

- default

- › 預設的參數值

- type

- › 參數值的型態

- help

- › 參數的說明

- required

- › 參數值是否必須

```
parser.add_argument("--var1",  
                    "-v1",  
                    dest = "var1",  
                    help = "first variable",  
                    default = 0,  
                    type = int)
```


Parser (4/5)

- Positional argument:

- 依照輸入順序放進你宣告的引數變數中
- 沒有前綴“-”
- `parser.add_argument(“pos”, help = “pos_arg”)`

```
$ python parser1.py 2
```

- Optional argument:

- 有前綴“-”
- `parser.add_argument(“-o”, “--opt”, help = “opt_arg”)`

```
$ python parser1.py --opt 2
```

- `args = parser.parse_args()` ➡ 解析添加的參數
- `a = args.opt` ➡ 將參數提出來使用

Parser (5/5)

```
parser.add_argument("--var1",    → 增加參數
                    "-v1",
                    dest = "var1",
                    help = "first variable",
                    default = 0,
                    type = int)

parser.print_help()
args = parser.parse_args() → 解析參數
print("var1 =", args.var1) → print 參數 var1
```

```
(base) [U110tyhsiao@eng04, ~/AI_training_2022]$ python parser_test.py -v1 100
usage: Tutorial

input var1

optional arguments:
  -h, --help            show this help message and exit
  --var1 VAR1, -v1 VAR1
                        first variable

type: int
var1 = 100
```

Introduction to Numpy

Numpy Array vs. List

- Numpy Array:
 - homogeneous
 - 高效能的多維陣列(multi-dimensional array)數學函式庫
 - › 需要import
 - 平行處理、科學運算較快
- List:
 - homogeneous or heterogeneous
 - python內建資料型別

Ndarray Attribute

- Import numpy as np
- A.ndim ➡ array A 的維度
- A.shape ➡ 每個維度的大小
- A.size ➡ array A 的總元素量
- A.dtype ➡ 元素的型態
- A.itemsize ➡ 每個元素的大小
是多少個byte

```
import numpy as np
A = np.array([[1,2,3],[4,5,6]])
print("ndim:", A.ndim)
print("shape:", A.shape)
print("size:", A.size)
print("dtype:", A.dtype)
print("itemsize", A.itemsize)
```

```
ndim: 2
shape: (2, 3)
size: 6
dtype: int64
itemsize 8
```

Generate an Nddarray (1/2)

- np.array
- np.arange(起始值,結束值,間隔)

```
a = np.array([[1,2,3],[4,5,6]])  
b = np.arange(10)  
c = np.arange(0,10,1.5, dtype=np.float64)  
print("a:", a)  
print("b:", b)  
print("c:", c)
```

```
a: [[1 2 3]  
     [4 5 6]]  
b: [0 1 2 3 4 5 6 7 8 9]  
c: [0.  1.5 3.  4.5 6.  7.5 9. ]
```

Generate an Nddarray (2/2)

- `np.empty((a,b))` ➡ 沒有初始值的 $a*b$ array
- `np.zeros((a,b))` ➡ 所有元素都為 0 的 $a*b$ array
- `np.ones((a,b))` ➡ 所有元素都為 1 的 $a*b$ array
- `np.linspace(起始值,結束值,中間要產生多少個元素)`
- `np.random.randn(2,3)` ➡ 常態分佈的 $2*3$ array

```
empty = np.empty((2,3))
print(empty)
zeros = np.zeros((2,3))
print(zeros)
ones = np.ones((2,3))
print(ones)
lin = np.linspace(3,5,9)
print(lin)
random = np.random.randn(2,3)
print(random)
```

```
[[1.12433272  0.79871433  0.27558333]
 [0.02951496  1.83307077  0.44965672]] } empty

[[0.  0.  0.]
 [0.  0.  0.]] } zeros

[[1.  1.  1.]
 [1.  1.  1.]] } ones

[3.    3.25  3.5   3.75  4.    4.25  4.5   4.75  5.   ] } lin

[[ 1.03707036 -0.36387992 -0.4434428 ]
 [ 1.08279876  1.79948769  0.66261707]] } random
```

Common dtype

bool_	布尔型数据类型 (True 或者 False)
int_	默认的整数类型 (类似于 C 语言中的 long, int32 或 int64)
intc	与 C 的 int 类型一样, 一般是 int32 或 int 64
intp	用于索引的整数类型 (类似于 C 的 ssize_t, 一般情况下仍然是 int32 或 int64)
int8	字节 (-128 to 127)
int16	整数 (-32768 to 32767)
int32	整数 (-2147483648 to 2147483647)
int64	整数 (-9223372036854775808 to 9223372036854775807)
float_	float64 类型的简写
float16	半精度浮点数, 包括: 1 个符号位, 5 个指数位, 10 个尾数位
float32	单精度浮点数, 包括: 1 个符号位, 8 个指数位, 23 个尾数位
float64	双精度浮点数, 包括: 1 个符号位, 11 个指数位, 52 个尾数位



Array vs. List

- `np.array(list_name)` ➡ list 轉 numpy array
- `list(array_name)` vs. `array_name.tolist()`
- `reshape` ➡ 改變 array 的形狀

```
a1 = np.arange(12)
print(a1)
print(a1.shape)
a2 = a1.reshape(3,4)
print(a2)
print(a2.shape)
```

```
[ 0  1  2  3  4  5  6  7  8  9 10 11]
(12,)
[[ 0  1  2  3]
 [ 4  5  6  7]
 [ 8  9 10 11]]
(3, 4)
```

- `dtype`
 - `A = np.array([1,2,3],dtype=np.float64)`
- `astype` ➡ 轉換資料型別
 - `B=A.astype(np.float64)`

Basic Operations

- $+$, $-$, $*$, $/$, $**$

- 一定要同維度的兩個array ➡ 同位置的元素做運算
- array跟特定純量做運算 ➡ 純量跟每個元素做運算

```
a = np.array([4,5,6])
b = np.arange(1,4)
c = a + b
d = a - b
e = a * b
f = a / b
g = a ** b
print(a) [4 5 6]
print(b) [1 2 3]
print(c) [5 7 9]
print(d) [3 3 3]
print(e) [ 4 10 18]
print(f) [4.  2.5 2. ]
print(g) [ 4 25 216]
```

```
a = np.array([4,5,6])
b = a + 2
c = a - 2
d = a * 2
e = a / 2
f = a ** 2
print(a) [4 5 6]
print(b) [6 7 8]
print(c) [2 3 4]
print(d) [ 8 10 12]
print(e) [2.  2.5 3. ]
print(f) [16 25 36]
```

- 矩陣相乘：A.dot(B)

Common Usage

- $A[A \leq 0] = 0$
- `np.sum(A)`
 - `A.sum(axis=0)` ↓
 - `A.sum(axis=1)` →
- `np.cumsum()`
 - 前面元素的累加
- `np.max()`
- `np.min()`
- `np.mean()`
- `B=np.exp(A)` ➡ e的指數次方
- `B=np.sqrt(A)`

```
A = np.array([[1,2,3],[-1,-2,-3]])
A[A<0] = 0
print(A)
print(A.sum(axis=0))
print(A.sum(axis=1))
print(np.cumsum(A))
```

```
[[1 2 3]
 [0 0 0]]
[1 2 3]
[6 0]
[1 3 6 6 6 6]
```

Shape of Array (1/2)

- `np.reshape(a,b,-1)` ➡ -1 表示自動計算維度
- `np.ravel()` ➡ flattened
- `A.T` ➡ A的轉置矩陣
- `np.vstack((A,B))` ➡ 縱向疊加
- `np.hstack((A,B))` ➡ 橫向疊加
- `np.vsplit(A,num)` ➡ 縱向拆分
- `np.hsplit(A,num)` ➡ 橫向拆分
- `np.concatenate((a,b), axis=dim)` ➡ 沿dim維度疊加

Shape of Array (2/2)

```
a = np.arange(8)
print(a)
a = a.reshape(2,2,-1)
print(a)
print(a.shape)
a = np.ravel(a)
print(a)
```

```
[0 1 2 3 4 5 6 7]
```

```
[[[0 1]
   [2 3]]
```

```
[[4 5]
 [6 7]]]
```

```
(2, 2, 2) → a.shape
```

```
[0 1 2 3 4 5 6 7]
```

```
b = np.arange(3)
print("vstack:")
print(np.vstack((b,b)))
print("hstack:")
print(np.hstack((b,b)))
```

```
vstack:
```

```
[[0 1 2]
 [0 1 2]]
```

```
hstack:
```

```
[0 1 2 0 1 2]
```

Indexing & Slicing (1/2)

- A [第0個維度的index/slicing, 第1個維度...]
 - $A[a:b]$
 - › a : 起始索引值, b : 結束索引值+1
 - › a 的位置不填: default 0, b 的位置不填: default size+1
 - $A[a:b:c]$
 - › a : 起始索引
 - › b : 結束索引+1
 - › c : 選取資料間隔, 以索引值可以被此值整除的元素, 預設為1
 - $A[:]$
 - › 選取所有此維度的元素
 - $A[::-1]$
 - › 此維度的元素順序顛倒
- Ex. $A[::-1, 1]$ ➡ A 的第1個column且順序顛倒

Example

```
A = np.array([[1, 2, 3], [4, 5, 6]])  
print(A[::-1, 1])
```

```
[5 2]
```

Introduction to Pytorch

Pytorch & Tensor

- Pytorch 是 Facebook 於 2017 年所開源的深度學習框架，因其語法簡潔、直觀的特性深受歡迎，已成為目前深度學習熱門框架之一。
- Tensor
 - 多維度的矩陣
 - Pytorch 的基本元素
 - 用法與 numpy 類似
 - Pytorch 可以在 GPU 上執行，numpy 只能在 CPU 上執行

Generate a Tensor (1/2)

- `import torch`
- `A = torch.empty(a,b)`

```
A=torch.empty(5,4)
print(A)
```

```
tensor([[1.9631e-36, 0.0000e+00, 6.8664e-44, 7.9874e-44],
        [6.3058e-44, 6.7262e-44, 7.7071e-44, 6.3058e-44],
        [6.8664e-44, 7.0065e-44, 1.1771e-43, 6.8664e-44],
        [7.5670e-44, 8.1275e-44, 6.7262e-44, 7.1466e-44],
        [8.1275e-44, 7.2868e-44, 7.4269e-44, 6.4460e-44]])
```

- `A = torch.rand(a,b)`
 - 均匀分布
- `A = torch.randn(a,b)`
 - 常態分布

```
A=torch.rand(5,4)
print(A)
```

```
tensor([[0.2574, 0.7742, 0.8416, 0.8914],
        [0.8516, 0.2400, 0.4267, 0.6256],
        [0.3021, 0.5501, 0.3705, 0.9150],
        [0.3561, 0.2540, 0.7119, 0.8962],
        [0.8289, 0.1838, 0.4516, 0.7829]])
```

Generate a Tensor (2/2)

- `A = torch.zeros(a,b, dtype=torch.long)`
- `A = torch.ones(a,b)`
- `A = torch.tensor([2.2,3.3])`
- `A.type()`

```
import torch

a = torch.zeros(2,3, dtype = torch.long)
print(a.type())
b = torch.randn(2,3)
print(b.type())
```

```
torch.LongTensor
torch.FloatTensor
```



Torch.dtype

Data type	dtype	Tensor types
32-bit floating point	<code>torch.float32</code> or <code>torch.float</code>	<code>torch.*.FloatTensor</code>
64-bit floating point	<code>torch.float64</code> or <code>torch.double</code>	<code>torch.*.DoubleTensor</code>
16-bit floating point	<code>torch.float16</code> or <code>torch.half</code>	<code>torch.*.HalfTensor</code>
8-bit integer (unsigned)	<code>torch.uint8</code>	<code>torch.*.ByteTensor</code>
8-bit integer (signed)	<code>torch.int8</code>	<code>torch.*.CharTensor</code>
16-bit integer (signed)	<code>torch.int16</code> or <code>torch.short</code>	<code>torch.*.ShortTensor</code>
32-bit integer (signed)	<code>torch.int32</code> or <code>torch.int</code>	<code>torch.*.IntTensor</code>
64-bit integer (signed)	<code>torch.int64</code> or <code>torch.long</code>	<code>torch.*.LongTensor</code>

CUDA

- **CUDA** (**C**ompute **U**nified **D**evice **A**rchitecture) 是由 NVIDIA 所推出的一種整合技術，是該公司對於 GPGPU 的正式名稱。
- torch.cuda
 - 實現了與CPU張量相同的功能，但使用GPU進行計算

Torch.cuda

- torch.cuda.is_available
- torch.cuda.device_count()
- torch.cuda.get_device_name(0)
- torch.cuda.set_device(device)
- torch.cuda.current_device()

```
import torch
print(torch.cuda.is_available())
print(torch.cuda.device_count())
print(torch.cuda.get_device_name(1))
print(torch.cuda.current_device())
```

True

2

GeForce RTX 2080 Ti

0

Torch.device

- torch.device('cpu')
- torch.device('cuda:0')
- 將 tensor 搬到 CUDA 上:
 - tensor.to(device)
 - tensor.cuda()

```
device = torch.device("cuda:1" if torch.cuda.is_available() else "cpu")  
a = torch.ones(5, device = device)  
print(a)
```

```
tensor([1., 1., 1., 1., 1.], device='cuda:1')
```

```
b = torch.zeros(5)  
cuda0 = torch.device("cuda:0")  
b = b.to(cuda0)  
print(b)
```

```
tensor([0., 0., 0., 0., 0.], device='cuda:0')
```

Tensor to Numpy & Numpy to Tensor

- Tensor to Numpy:
 - `A = B.numpy()`
- Numpy to Tensor:
 - `B = torch.tensor(A)`
 - `B = torch.from_numpy(A)`
- Device 轉換
 - `cpu()`, `cuda()`
- 類型轉換
 - `double()`, `float()`

Shape Manipulation

- `X.squeeze(d)` ➡ 壓縮第d維
- `X.unsqueeze(d)` ➡ 將第d維的維度設成1(多出一個維度)
- `X.transpose(d1,d2)` ➡ 兩個維度交換
- `X.permute(d1,d2,d3,...)` ➡ 多個維度交換

```
a = torch.Tensor(1, 2, 3)
print(a.shape)
print(a.squeeze(0).shape)
print(a.unsqueeze(0).shape)
print(a.transpose(1, 0).shape)
print(a.permute(2, 0, 1).shape)
```

```
torch.Size([1, 2, 3])
torch.Size([2, 3])
torch.Size([1, 1, 2, 3])
torch.Size([2, 1, 3])
torch.Size([3, 1, 2])
```


Common Usage

- `A.size()`
- `torch.add(a, b)`, `torch.mul(a,b)`
- `torch.mm(a, b)`
- `torch.abs(a)`
- `A.view (-1,a)`
- `torch.cat((A,B), dim)` ➡ 相當於`np.concatenate`
- `A.sum()`
- `A.data` (回傳相同的tensor，但`requires_grad = False`)

Linear Regression (1/7)

擷取自台大李宏毅老師投影片

Step 1: Model

$$y = b + w \cdot x_{cp}$$

A set of
function

Model

$f_1, f_2 \dots$

w and b are parameters
(can be any value)

$$f_1: y = 10.0 + 9.0 \cdot x_{cp}$$

$$f_2: y = 9.8 + 9.2 \cdot x_{cp}$$

$$f_3: y = -0.8 - 1.2 \cdot x_{cp}$$

..... infinite

$f($



$) =$

CP after
evolution

y

Linear model:

$$y = b + \sum w_i x_i$$

$x_i: x_{cp}, x_{hp}, x_w, x_h \dots$

feature

w_i : weight, b : bias

Linear Regression (2/7)

Step 2: Goodness of Function

Training Data:
10 pokemons

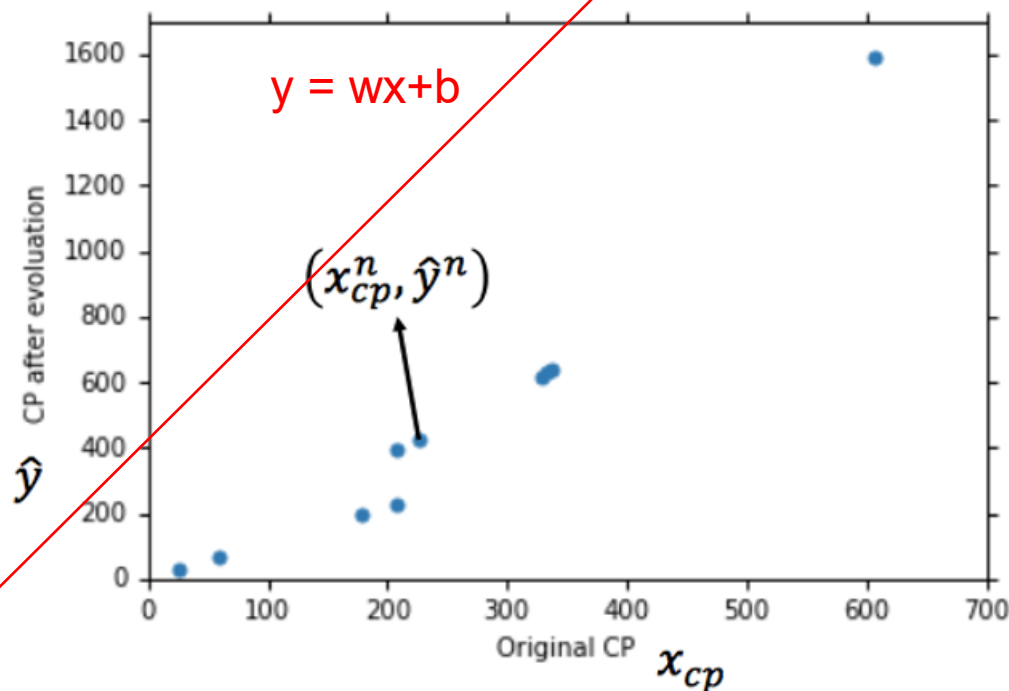
$$(x^1, \hat{y}^1)$$

$$(x^2, \hat{y}^2)$$

⋮

$$(x^{10}, \hat{y}^{10})$$

This is real data.

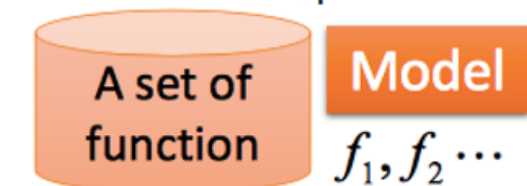


Source: <https://www.openintro.org/stat/data/?data=pokemon>

Linear Regression (3/7)

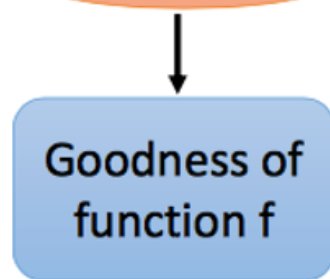
Step 2: Goodness of Function

$$y = b + w \cdot x_{cp}$$



Loss function L :

Input: a function, output:
how bad it is



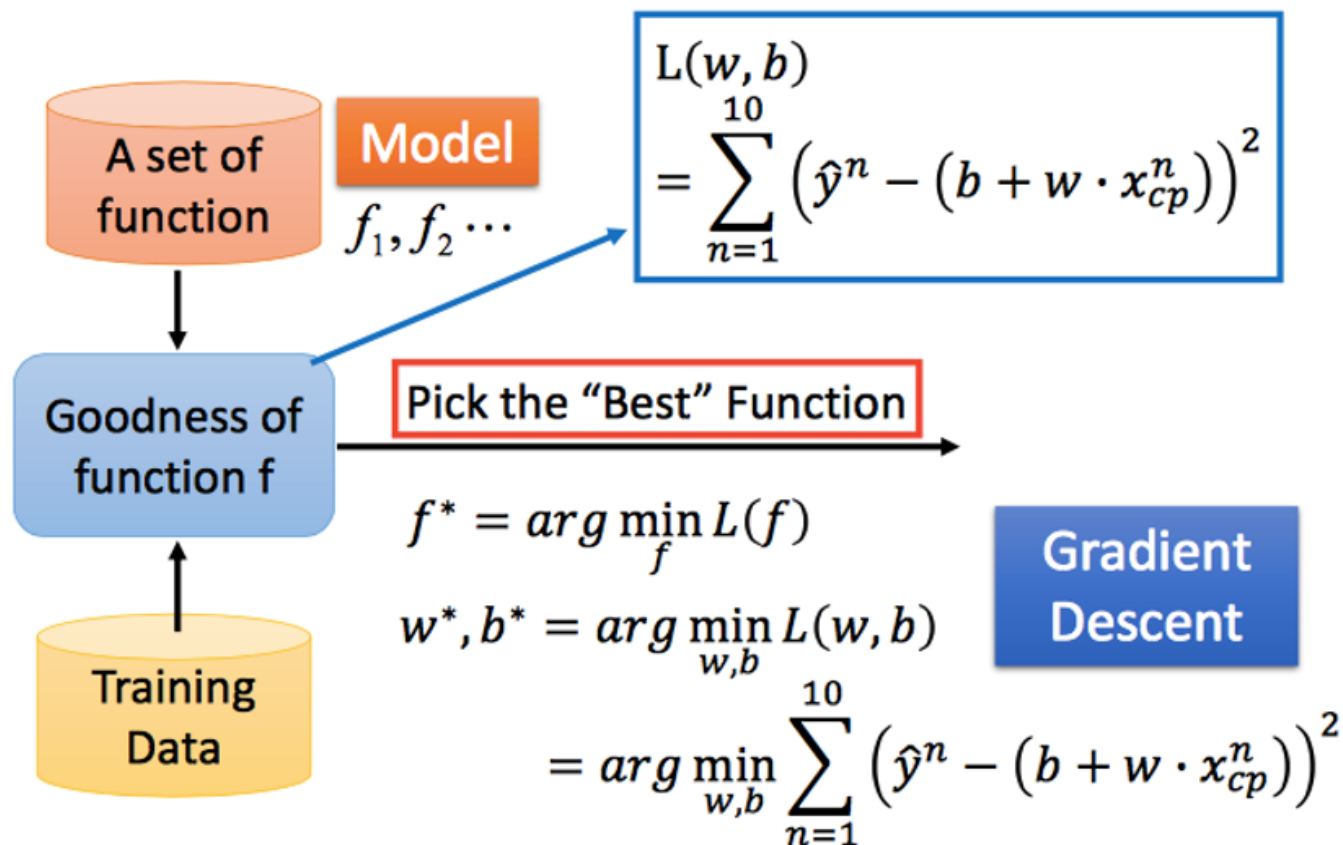
$$L(f) = \sum_{n=1}^{10} \left(\hat{y}^n - \underbrace{f(x_{cp}^n)}_{\text{Estimated } y \text{ based on input function}} \right)^2$$

Sum over examples

$$L(w, b) = \sum_{n=1}^{10} \left(\hat{y}^n - (b + w \cdot x_{cp}^n) \right)^2$$

Linear Regression (4/7)

Step 3: Best Function



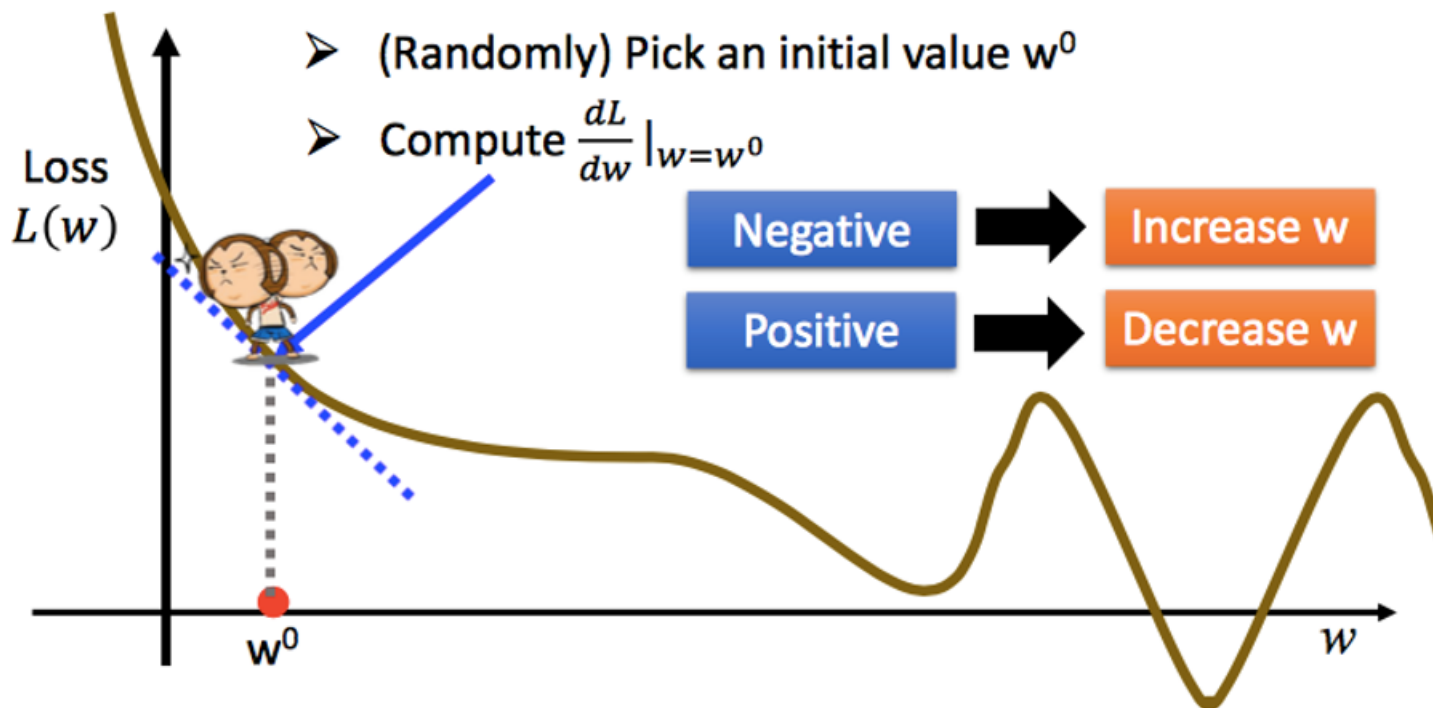
Linear Regression (5/7)

<http://chico386.pixnet.net/album/photo/171572850>

Step 3: Gradient Descent

$$w^* = \arg \min_w L(w)$$

- Consider loss function $L(w)$ with one parameter w :



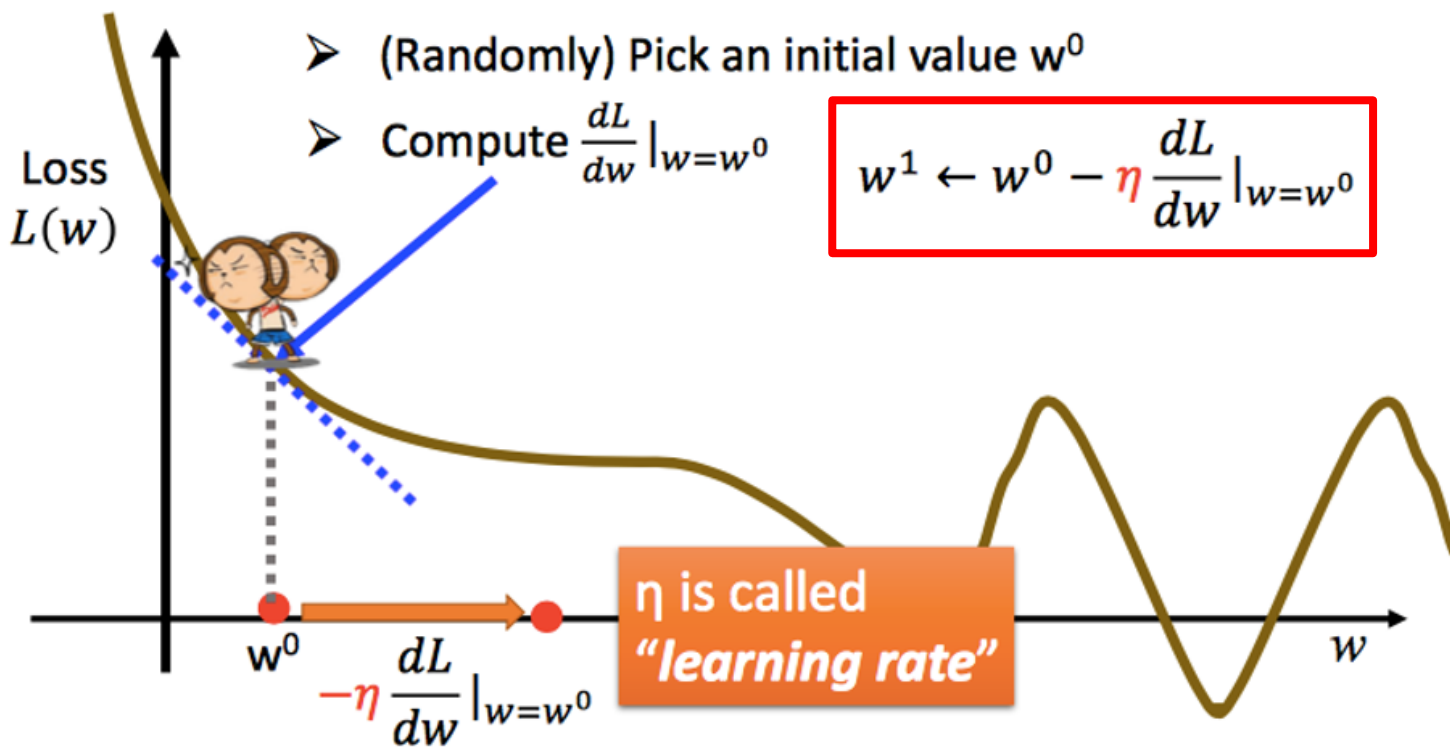
Linear Regression (6/7)

<http://chico386.pixnet.net/album/photo/171572850>

Step 3: Gradient Descent

$$w^* = \arg \min_w L(w)$$

- Consider loss function $L(w)$ with one parameter w :

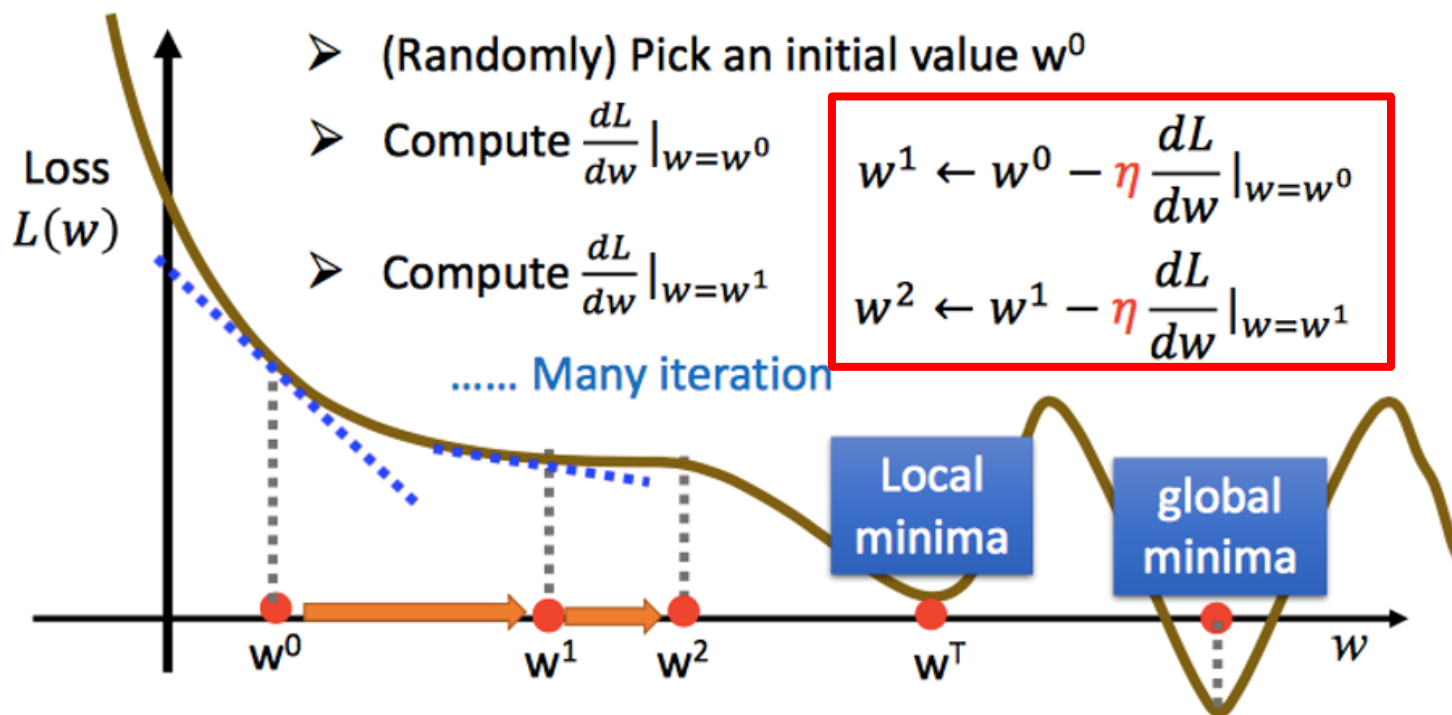


Linear Regression (7/7)

Step 3: Gradient Descent

$$w^* = \arg \min_w L(w)$$

- Consider loss function $L(w)$ with one parameter w :



Autograd: Automatic Differentiation (1/3)

```
import torch
x = torch.ones((2,2), requires_grad = True)
z = 4*x*x
y = z.norm() #平方和開根號
y.backward()
print(z.requires_grad)
print(z)
print(y)
print(x.grad)
```

```
True
tensor([[4., 4.],
        [4., 4.]], grad_fn=<MulBackward0>)
tensor(8., grad_fn=<NormBackward0>)
tensor([[4., 4.],
        [4., 4.]])
```



Autograd: Automatic Differentiation (2/3)

- backward():
 - 如果呼叫 backward() 的 tensor 不是 scalar 會報錯

```
import torch
x = torch.ones((2,2), requires_grad = True)
z = 4*x*x
z.backward()
```

```
RuntimeError: grad can be implicitly created only for scalar outputs
```

Autograd: Automatic Differentiation (3/3)

- with torch.autograd.no_grad() ➡ 此參數之後的參數不需要再算gradient, 可降低內存、加速計算
- A.grad.data.zero_() ➡ 將 gradient 歸零
 - 在 nn.Module裡面,被包裝成.zero_grad()

```
import torch
x=torch.ones((2,2),requires_grad=True)
with torch.no_grad():
    z=4*x*x
y=z.norm()
print(x.requires_grad)
print(z.requires_grad)
print(y)
```

```
True
False
tensor(8.)
```

Torch.nn

- `import torch.nn as nn`
 - pytorch 針對類神經網路包成一包模組
- torch.nn 提供的function：
 - Convolution
 - Pooling
 - Linear
 - Dropout
 - Activation function
 - Loss function
 - ...

Torch.nn

```
[12] import torch
      import torch.nn as nn

      layer = nn.Linear(2, 3)  #nn.Linear也是一個繼承nn.module來的class

      print(layer)             #print出module資訊
      print(layer.weight)      #print出這個module的weight參數
```

```
↳ Linear(in_features=2, out_features=3, bias=True)
   Parameter containing:
   tensor([[ -0.0398, -0.3860],
          [ 0.4566, -0.6518],
          [-0.3717,  0.4265]], requires_grad=True)
```

Torch.nn.Module

```
import torch.nn as nn
import torch.nn.function as F

class Model(nn.Module): # 繼承 nn.module
    def __init__(self): # Define what modules you need
        super(Model, self).__init__()
        self.fc1 = nn.Linear(2,3)

    def forward(self, x): # Define how to pass data
        out = self.fc1(x)
        out = F.relu(out)
        return out
```

- 核心功能：
 - `add_module(name,module)` ➡ 將子模塊加到當前的模塊
 - `forward(*input)` ➡ 前向傳播
 - `state_dict()` ➡ 保存module的參數資訊
 - `load_state_dict()` ➡ 用來加載模型參數

Torch.nn.Module

- 查看模塊
 - parameters() ➡ “Model” 的所有參數
 - modules() ➡ “Model” 這個 module 和 “Model” 裡的所有子 module
- 設置模式:
 - train() ➡ 將 module 設置為 training mode
 - zero_grad() ➡ 將 module 所有梯度設為 0

Torch.nn.Sequential

- modules依順序添加到容器中

```
class Model(nn.Module):  
    def __init__(self):  
        super(Model, self).__init__()  
        self.layers = nn.Sequential(  
            nn.Linear(2,3),  
            F.relu()  
        )  
  
    def forward(self, x):  
        out = self.layers(x)  
        return out
```



Saving & Loading a Model (1/2)

- Save entire model
 - `torch.save(model, PATH)`
 - 將 model 儲存到 PATH 的位置
- Load entire model
 - `model = torch.load(PATH)`

Saving & Loading a Model (2/2)

- Saving & Loading a General Checkpoint for Inference and/or Resuming Training

```
torch.save({ "epoch": epoch,
            "model_state_dict": model.state_dict(),
            "optimizer_state_dict": optimizer.state_dict(),
            "loss": loss,
            ...
            }, PATH)

model = LeNet()
optimizer = optimizer = torch.optim.SGD()

checkpoint = torch.load(PATH)
model.load_state_dict(checkpoint["model_state_dict"])
optimizer.load_state_dict(checkpoint["optimizer_state_dict"])
epoch = checkpoint["epoch"]
loss = checkpoint["loss"]

model.eval()
```

Torch.nn.functional

- `import torch.nn.functional as F`
- `torch.nn` vs. `torch.nn.functional`
 - `torch.nn.functional`提供的是純函數
 - `torch.nn`提供的是包括完整的`nn.Module`(包含參數資訊)
- 需要維持參數狀態的，主要是`convolution layer` 和 `linear layer`，所以用`torch.nn`所提供的`module`
- 而在計算時，`relu`、`dropout`、`pooling`不需要保存狀態的可以直接使用`torch.nn.functional`

Example Code

```
class LeNet(nn.Module):
    def __init__(self):
        # nn.Module的子類函數必須在構造函數中執行父類的構造函數
        super(LeNet, self).__init__() # 等價於nn.Module.__init__()

        # nn.Conv2d返回的是一個Conv2d class的一個對象，該類中包含forward函數的實現
        # 當調用self.conv1(input)的時候，就會調用該類的forward函數
        self.conv1 = nn.Conv2d(1, 6, (5, 5)) # output (N, C_{out}, H_{out}, W_{out})
        self.conv2 = nn.Conv2d(6, 16, (5, 5))
        self.fc1 = nn.Linear(256, 120)
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)

    def forward(self, x):
        x = F.max_pool2d(F.relu(self.conv1(x)), (2, 2))
        x = F.max_pool2d(F.relu(self.conv2(x)), (2, 2))
        x = x.view(x.size()[0], -1)
        x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = F.relu(self.fc3(x))
        return x

model = LeNet() #為我們定義的這個class生出一個object
output = model(input) #會自動調用forward函數
```



References

- 台大李宏毅老師Youtube
 - <https://www.youtube.com/channel/UC2ggjtuuWvxrHHHiaDH1dIQ>
- 莫凡pytorch
 - <https://morvanzhou.github.io/tutorials/machine-learning/torch/>

Thank you