LAB2-Linear regression

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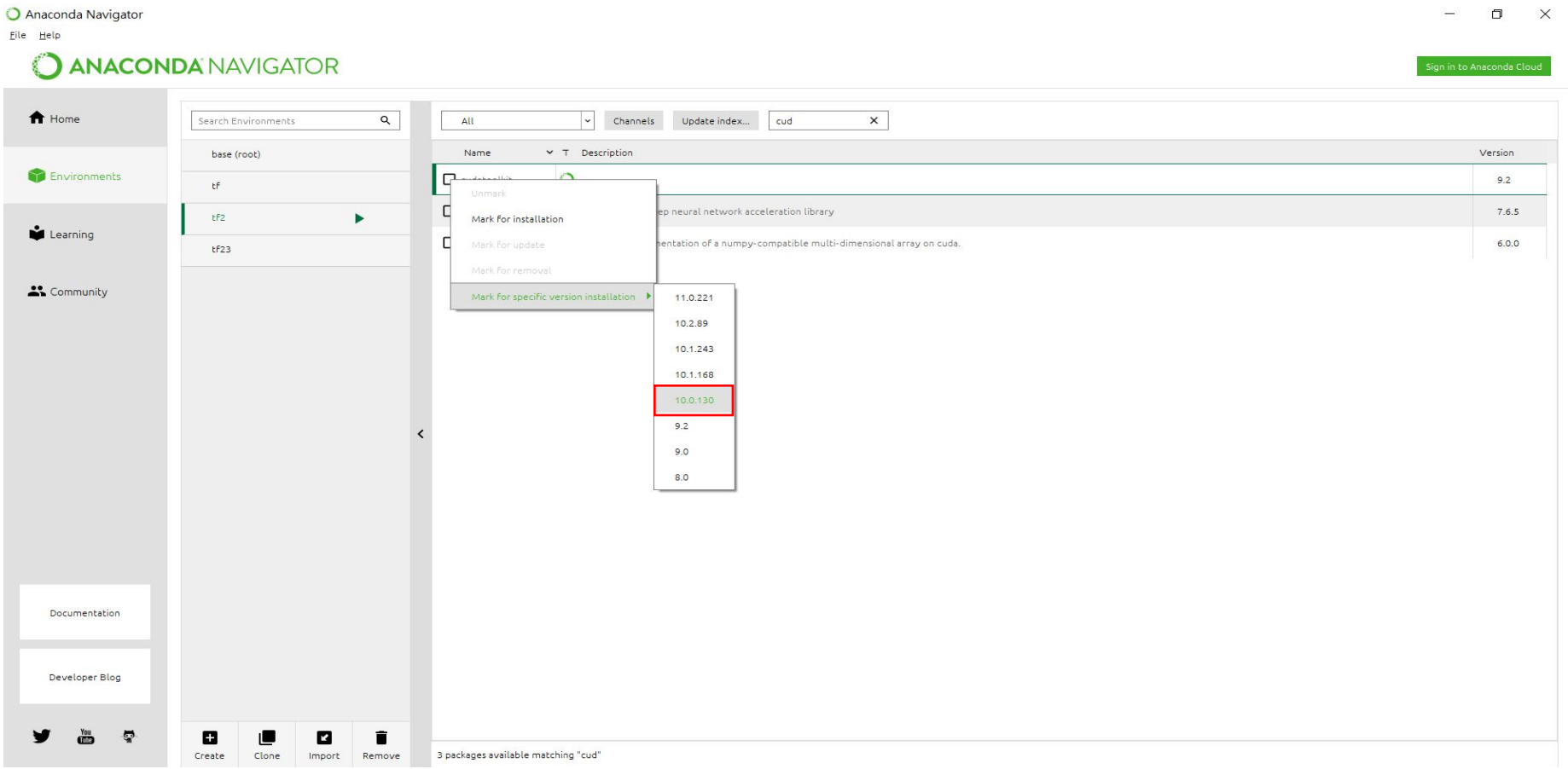
安裝TensorFlow

GPU版本

需在tf2環境下安裝 **codatoolkit** 和 **cudnn**

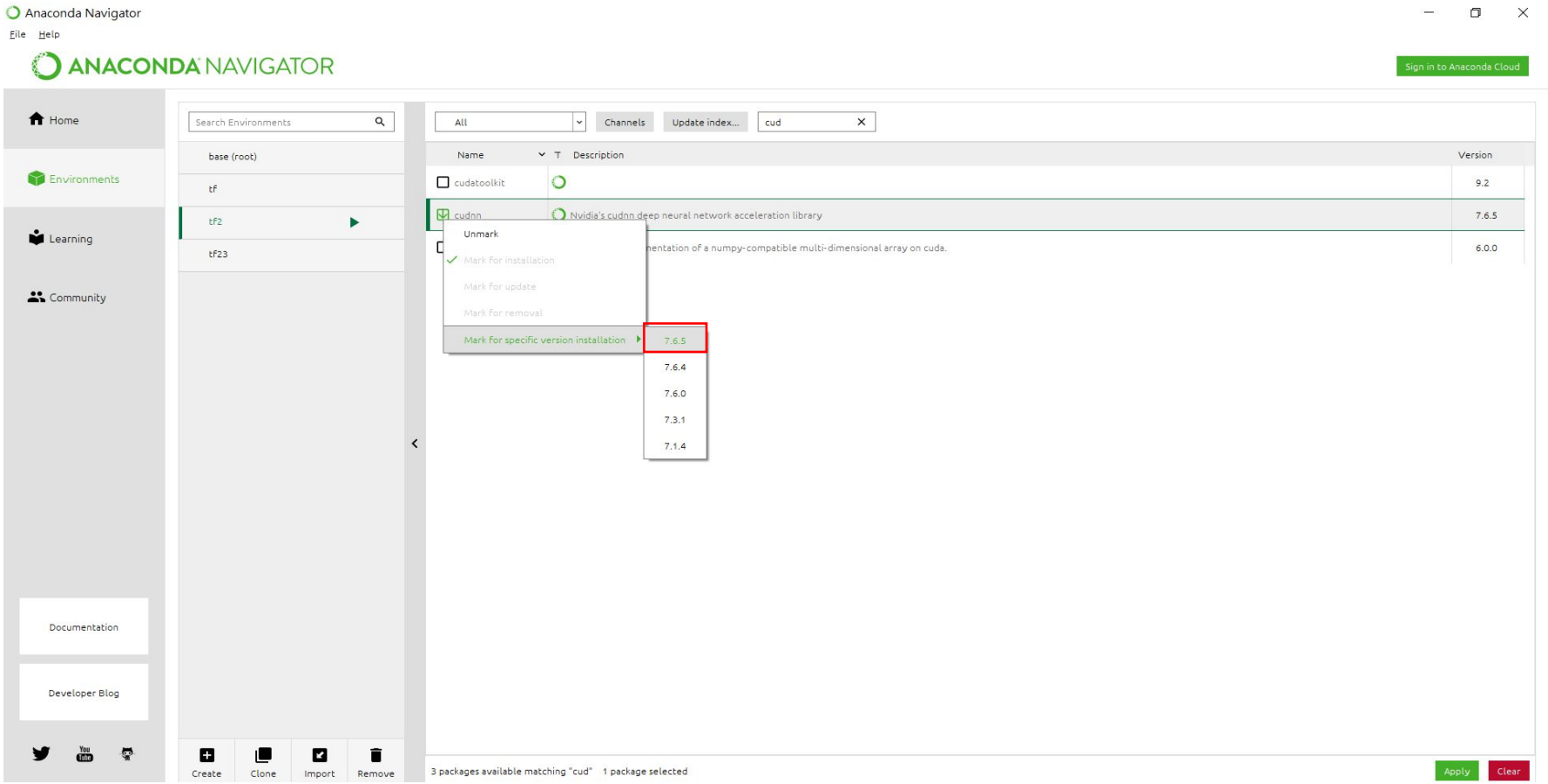
安裝TensorFlow

GPU版本codatoolkit

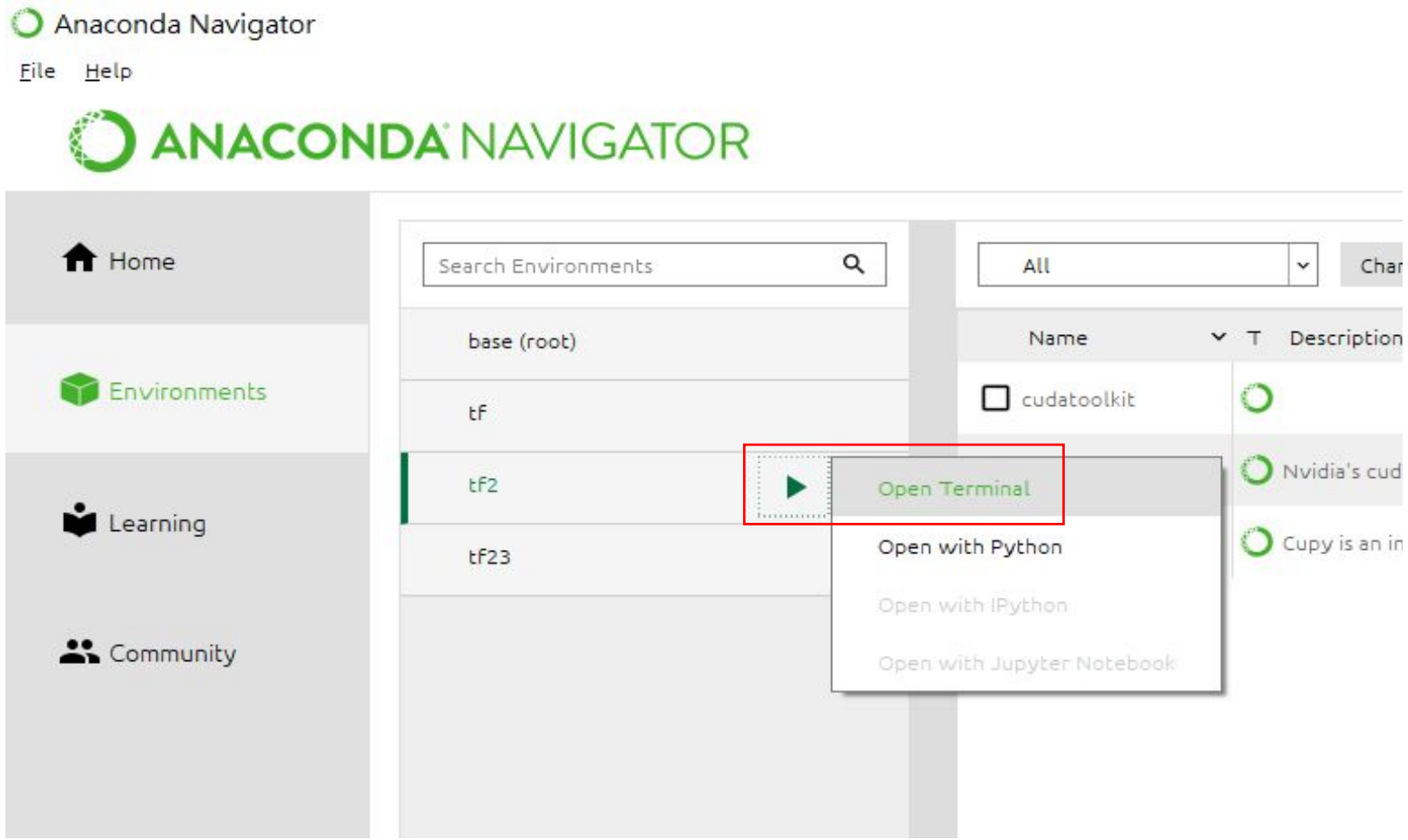


安裝TensorFlow

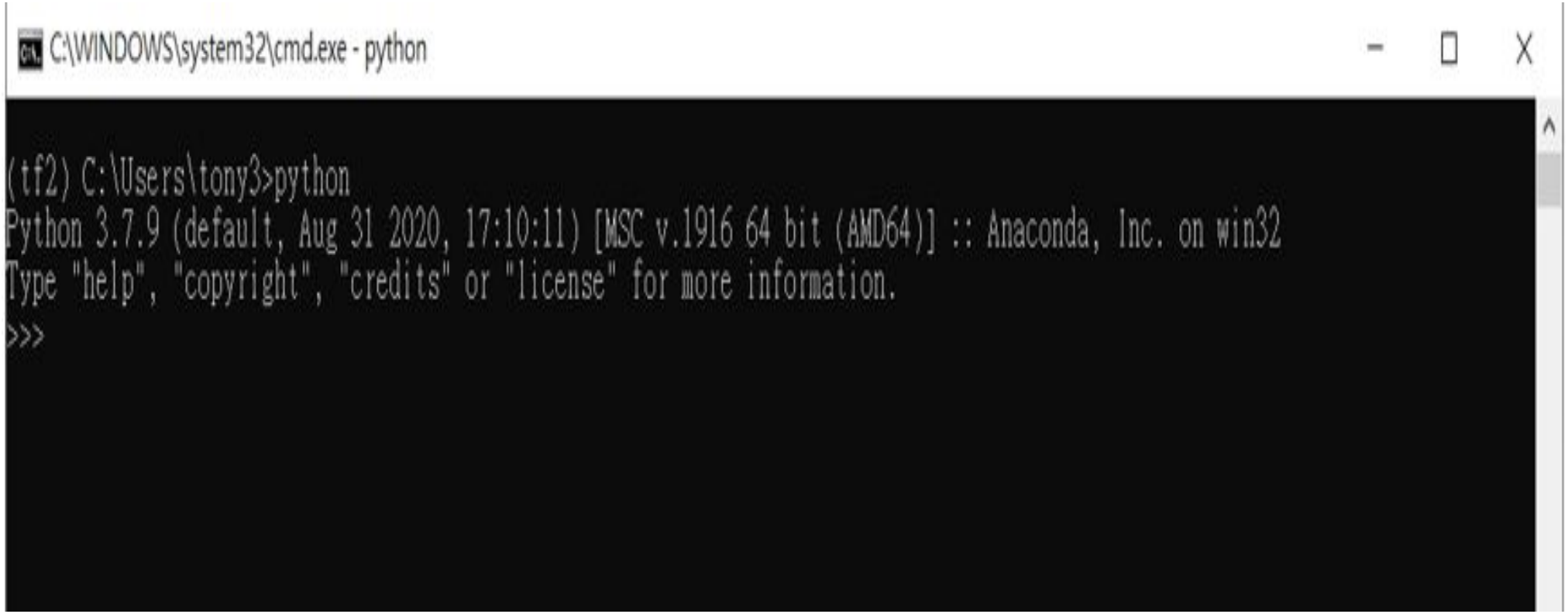
GPU版本cudnn



執行環境

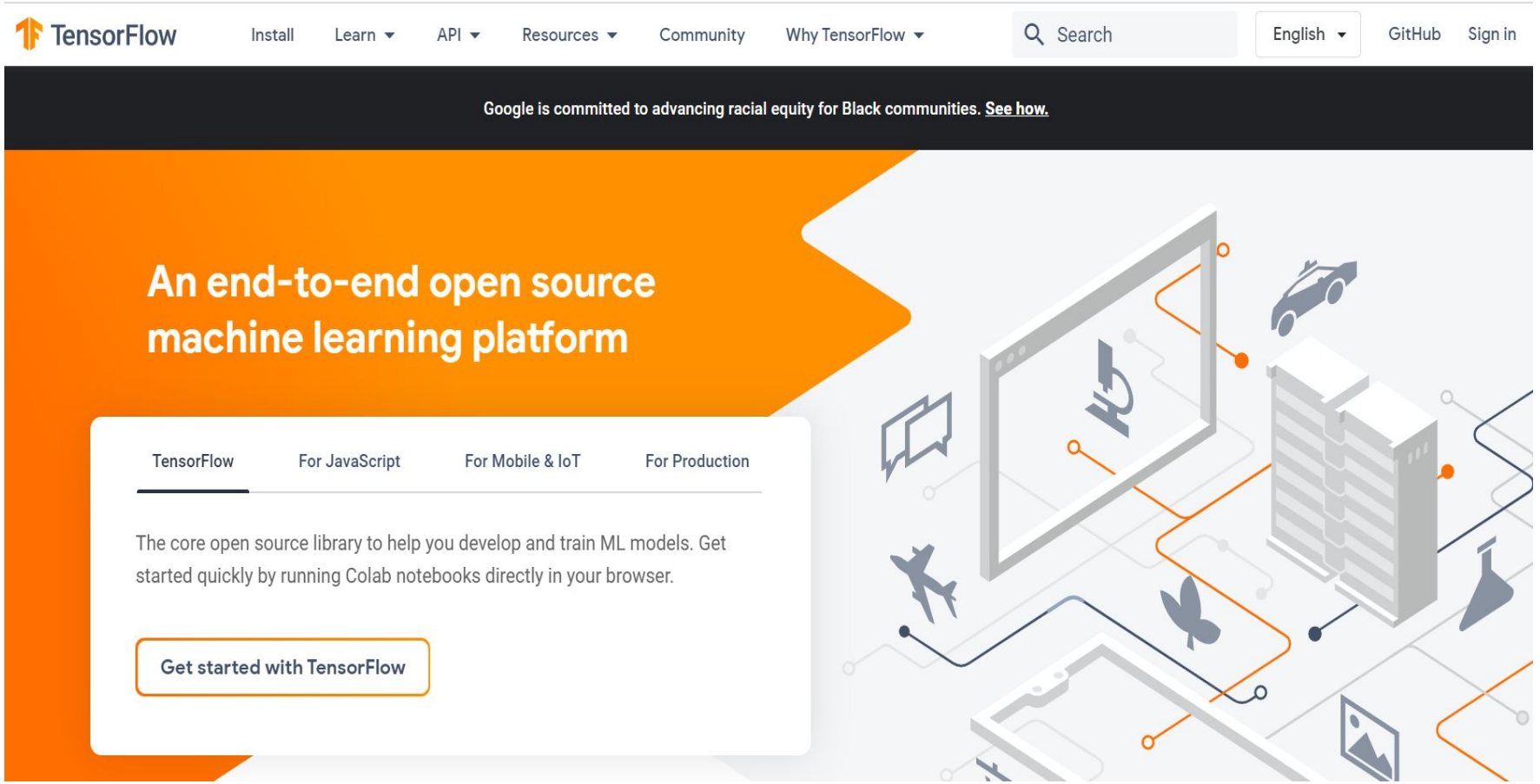

Anaconda Navigator
File Help
ANACONDA NAVIGATOR
Home
Environments
Learning
Community
Search Environments
base (root)
tf
tf2
tf23
All
Name
T
Description
cudatoolkit
Nvidia's cud
Cupy is an in
Open Terminal
Open with Python
Open with IPython
Open with Jupyter Notebook

執行環境


C:\WINDOWS\system32\cmd.exe - python
(tf2) C:\Users\tony3>python
Python 3.7.9 (default, Aug 31 2020, 17:10:11) [MSC v.1916 64 bit (AMD64)] :: Anaconda, Inc. on win32
Type "help", "copyright", "credits" or "license" for more information.
>>>

TensorFlow

https://www.tensorflow.org/

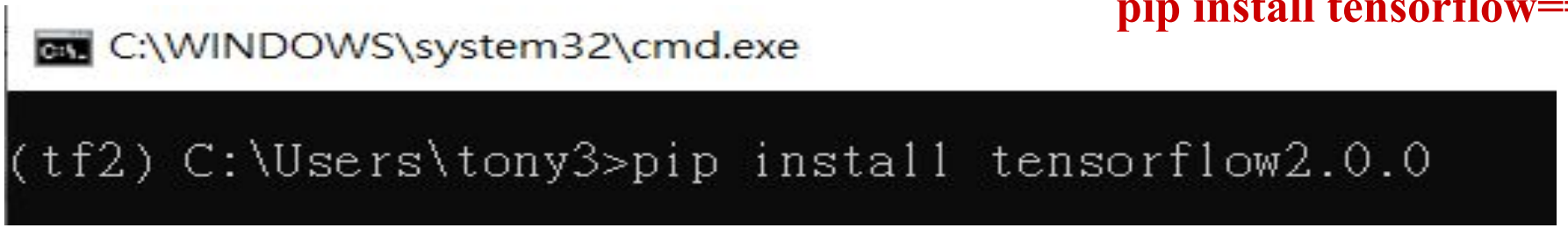


TensorFlow

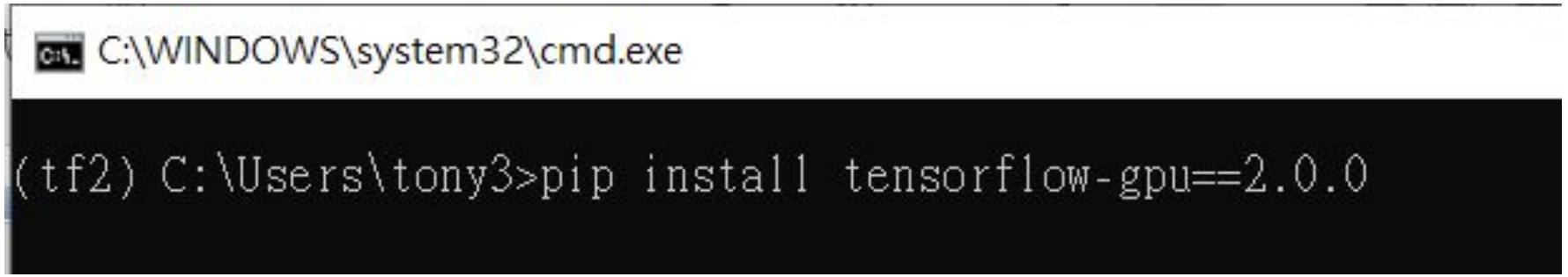
CPU版本

在 ES302 有些電腦請 tensorflow安裝2.5.0

pip install tensorflow==2.5.0



GPU版本



TensorFlow

安裝後可以在python中import tensorflow。
若沒有出現Error代表安裝成功。

```
(tf2) C:\Users\tony3>python
Python 3.7.9 (default, Aug 31 2020, 17:10:11) [MSC v.1916 64 bit (AMD64)] :: Anaconda, Inc. on win32
Type "help", "copyright", "credits" or "license" for more information.
>>> import tensorflow as tf
>>> tf.__version__
'2.0.0'
>>>
```

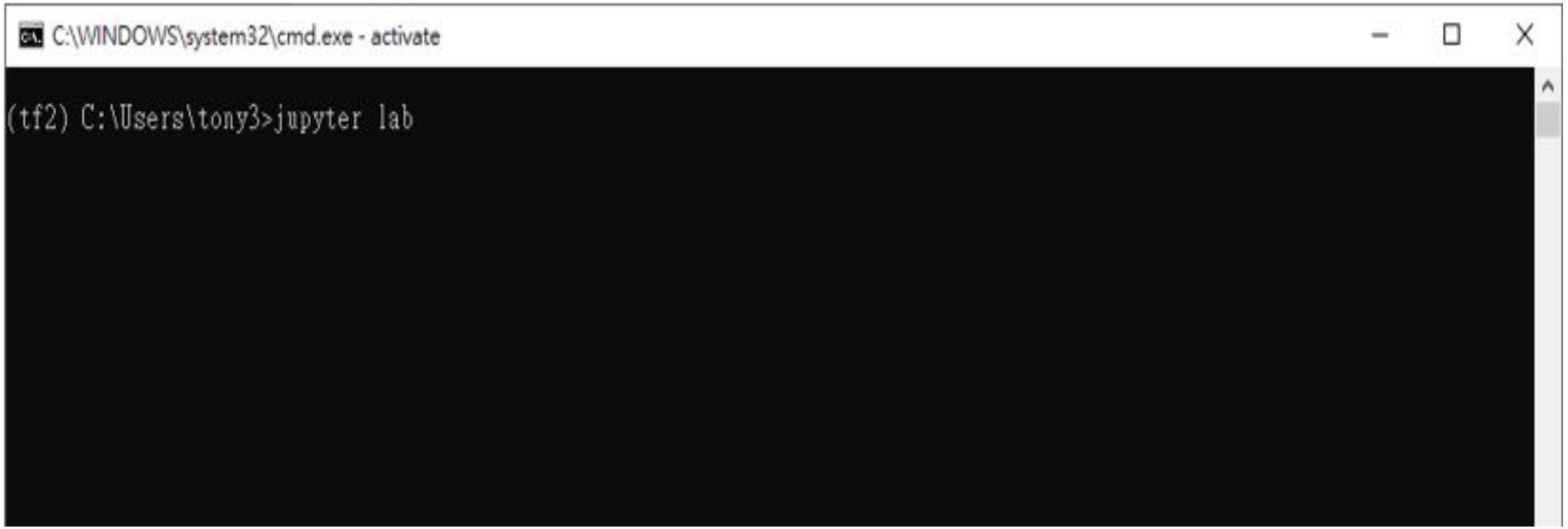
JUPYTER LAB

建議可安裝jupyter lab

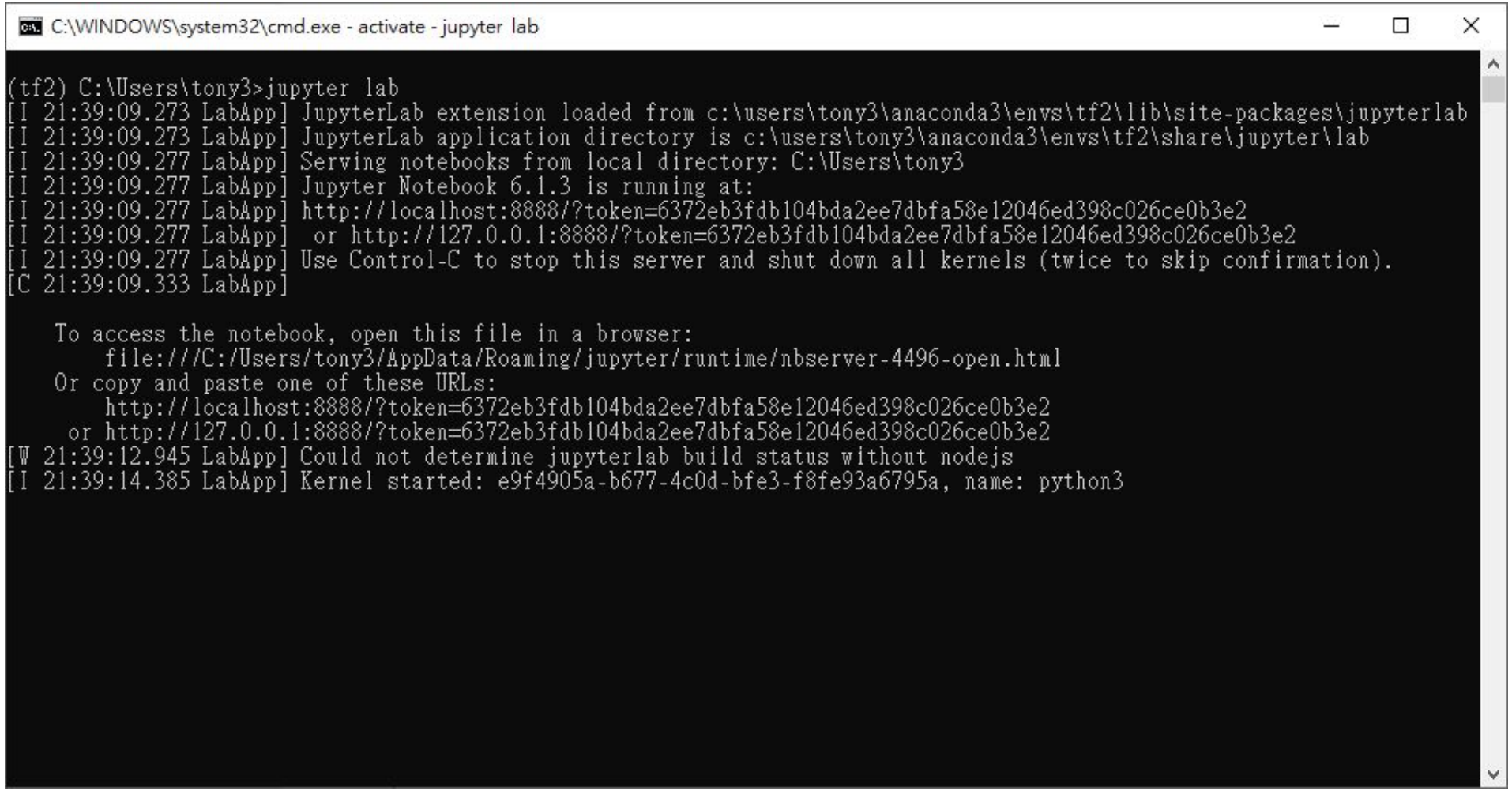
& pip install jupyterlab

執行JUPYTER LAB

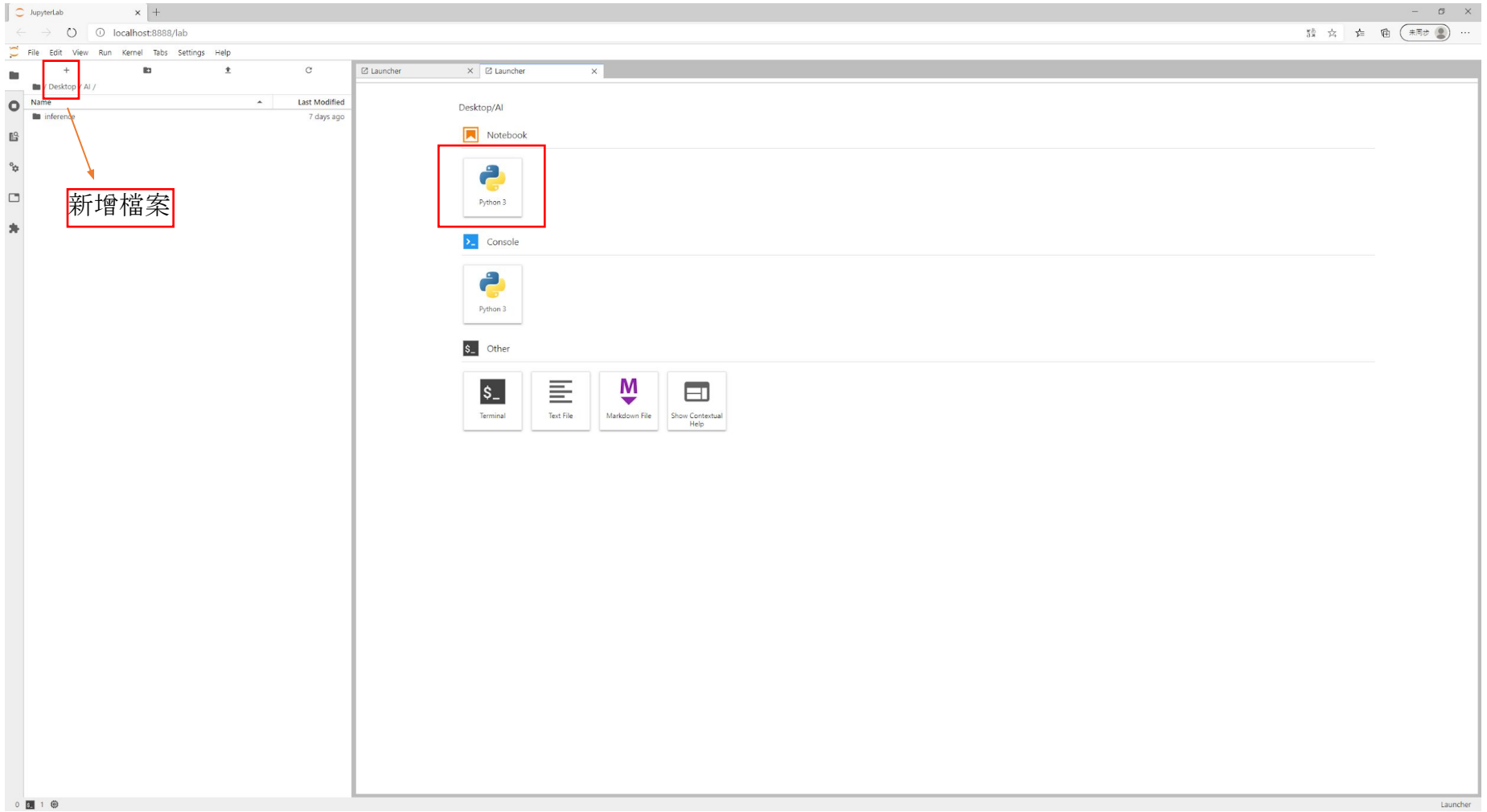
& jupyter lab



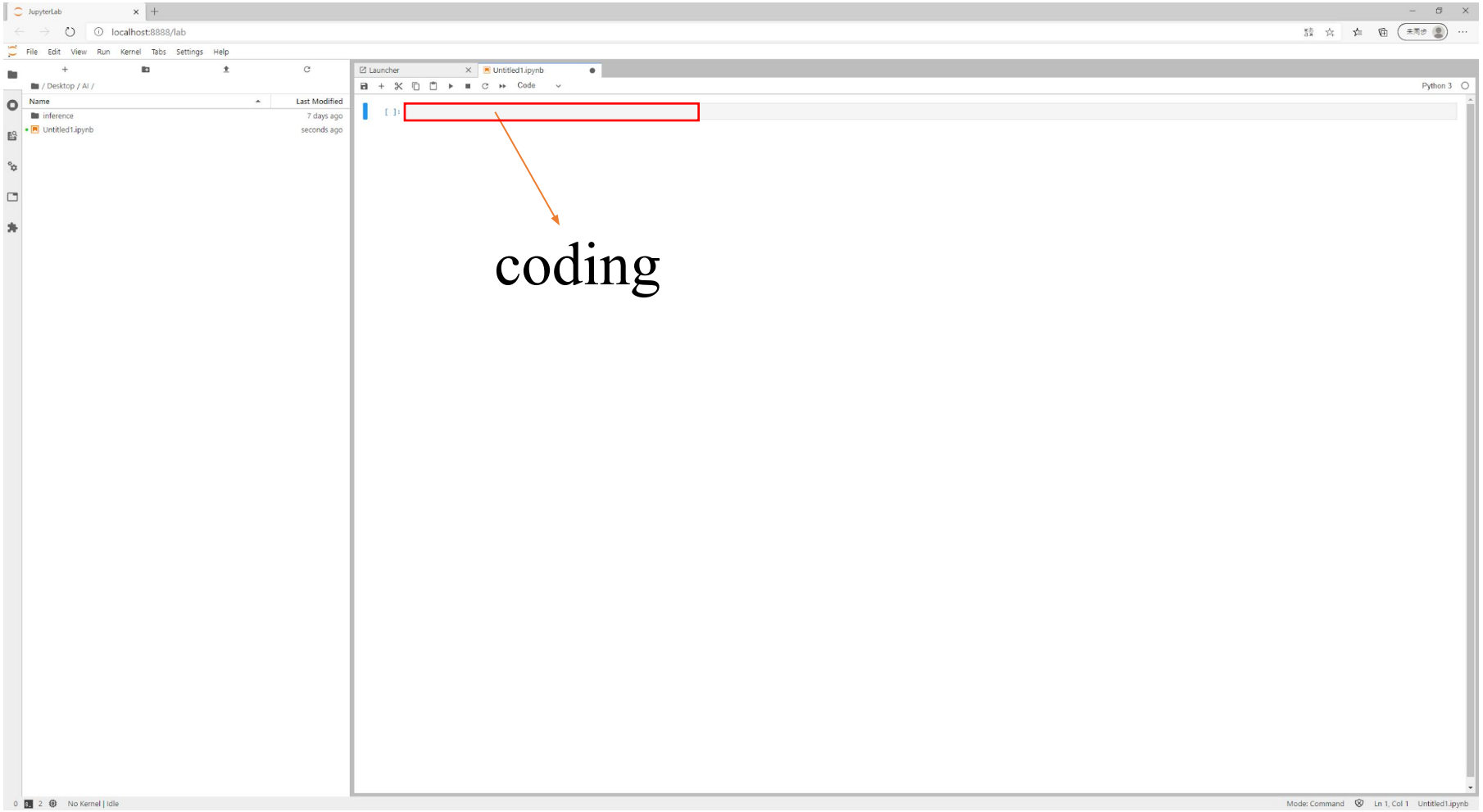
執行JUPYTER LAB



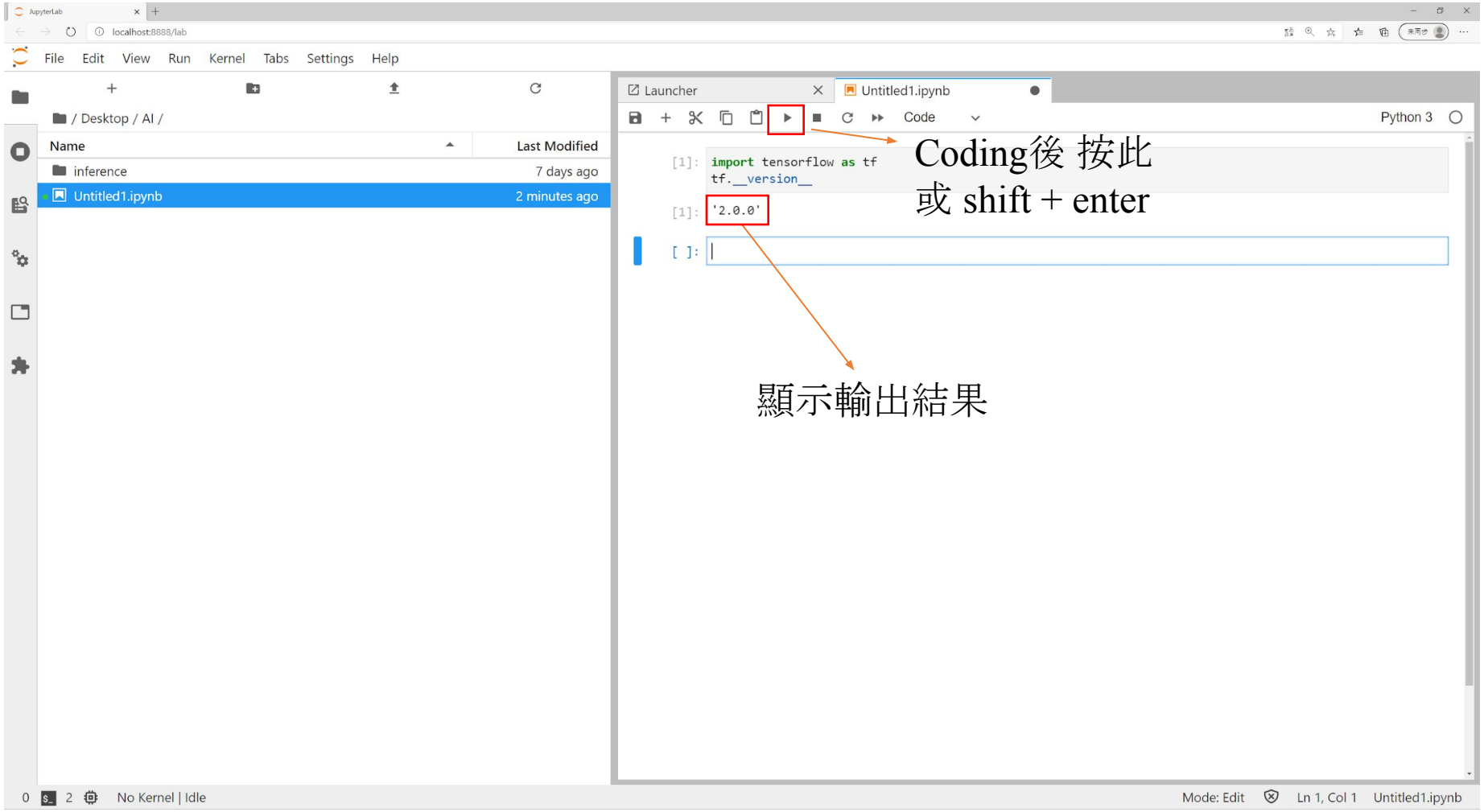
執行JUPYTER LAB



執行JUPYTER LAB



執行JUPYTER LAB



START!!!

Import library

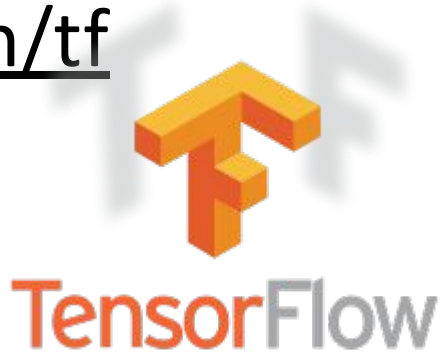
```
import tensorflow as tf
import numpy as np
```

Import library

```
import tensorflow as tf
```

TensorFlow是一個開源軟體庫, 用於各種感知和語言理解任務的機器學習。

https://www.tensorflow.org/api_docs/python/tf



Import library

```
import numpy as np
```

Numpy 是 Python 的一個重要模組，主要用於資料處理上。

Import library

安裝numpy

& pip install numpy

```
(tf2.3) C:\Users\user>pip install numpy
```

dataset

```
X = np.random.rand(1000).astype(np.float32)
print('X=',X)
n = X.shape[0]
Y = X * 9 + 5
print('Y=',Y)
```

X為訓練資料, 亂數產生1000筆

Y為對應X的答案

dataset

numpy.random.rand

numpy.random.rand

`numpy.random.` **`rand`** (*d0, d1, ..., dn*)¶

Random values in a given shape.

Create an array of the given shape and populate it with random samples from a uniform distribution over `[0, 1)`.

Parameters: **d0, d1, ..., dn** : *int, optional*

The dimensions of the returned array, should all be positive. If no argument is given a single Python float is returned.

Returns: **out** : *ndarray, shape* `(d0, d1, ..., dn)`

Random values.

權重初始化

```
W = tf.Variable(tf.random.normal([1]))
print(W)
b = tf.Variable(tf.random.normal([1]))
print(b)
```

```
<tf.Variable 'Variable:0' shape=(1,) dtype=float32, numpy=array([0.17674959], dtype=float32)>
<tf.Variable 'Variable:0' shape=(1,) dtype=float32, numpy=array([0.16365877], dtype=float32)>
```

權重初始化

tf.Variable()

The `Variable()` constructor requires an initial value for the variable, which can be a `Tensor` of any type and shape. This initial value defines the type and shape of the variable. After construction, the type and shape of the variable are fixed. The value can be changed using one of the assign methods.

訓練

```
for step in range(0,1001):
    with tf.GradientTape() as g:
        pred = W*X+b
        loss = tf.reduce_sum(tf.pow(pred-Y,2))/n

    gradient = g.gradient(loss,[W,b])

    tf.optimizers.Adam(.1).apply_gradients(zip(gradient, [W, b]))

    if step % 100 == 0:
        pred = W*X+b
        loss = tf.reduce_sum(tf.pow(pred-Y,2))/n
        print("step: %i, loss: %f, W: %f, b: %f" % (step, loss, W.numpy(), b.numpy()))
```

預測為W*X+b

Loss採用mean square

Optimizer採用Adam

$$\frac{1}{N}\sum_{i=1}^{N}(y_i-\hat{y}_i)^2$$

訓練

loss = tf.reduce_sum(tf.pow(pred-Y,2))/n

$$\frac{1}{N}\sum_{i=1}^{N}(y_i - \hat{y}_i)^2$$

訓練

```
step: 0, loss: 87.877869, W: 0.276753, b: 0.263662
step: 100, loss: 0.020041, W: 8.676969, b: 5.263853
step: 200, loss: 0.020030, W: 8.677159, b: 5.263757
step: 300, loss: 0.020019, W: 8.677350, b: 5.263662
step: 400, loss: 0.020008, W: 8.677541, b: 5.263566
step: 500, loss: 0.019997, W: 8.677732, b: 5.263471
step: 600, loss: 0.019986, W: 8.677922, b: 5.263376
step: 700, loss: 0.019976, W: 8.678113, b: 5.263280
step: 800, loss: 0.019965, W: 8.678304, b: 5.263185
step: 900, loss: 0.019954, W: 8.678494, b: 5.263090
step: 1000, loss: 0.019943, W: 8.678685, b: 5.262994
```

結果

```
import matplotlib.pyplot as plt
plt.plot(X, Y, 'ro', label='Original data')
plt.plot(X, np.array( W * X + b), label='pred line')
plt.legend()
plt.show()
```

結果

import matplotlib.pyplot as plt

matplotlib是Python程式語言及其數值數學擴展包 NumPy的可視化操作界面。

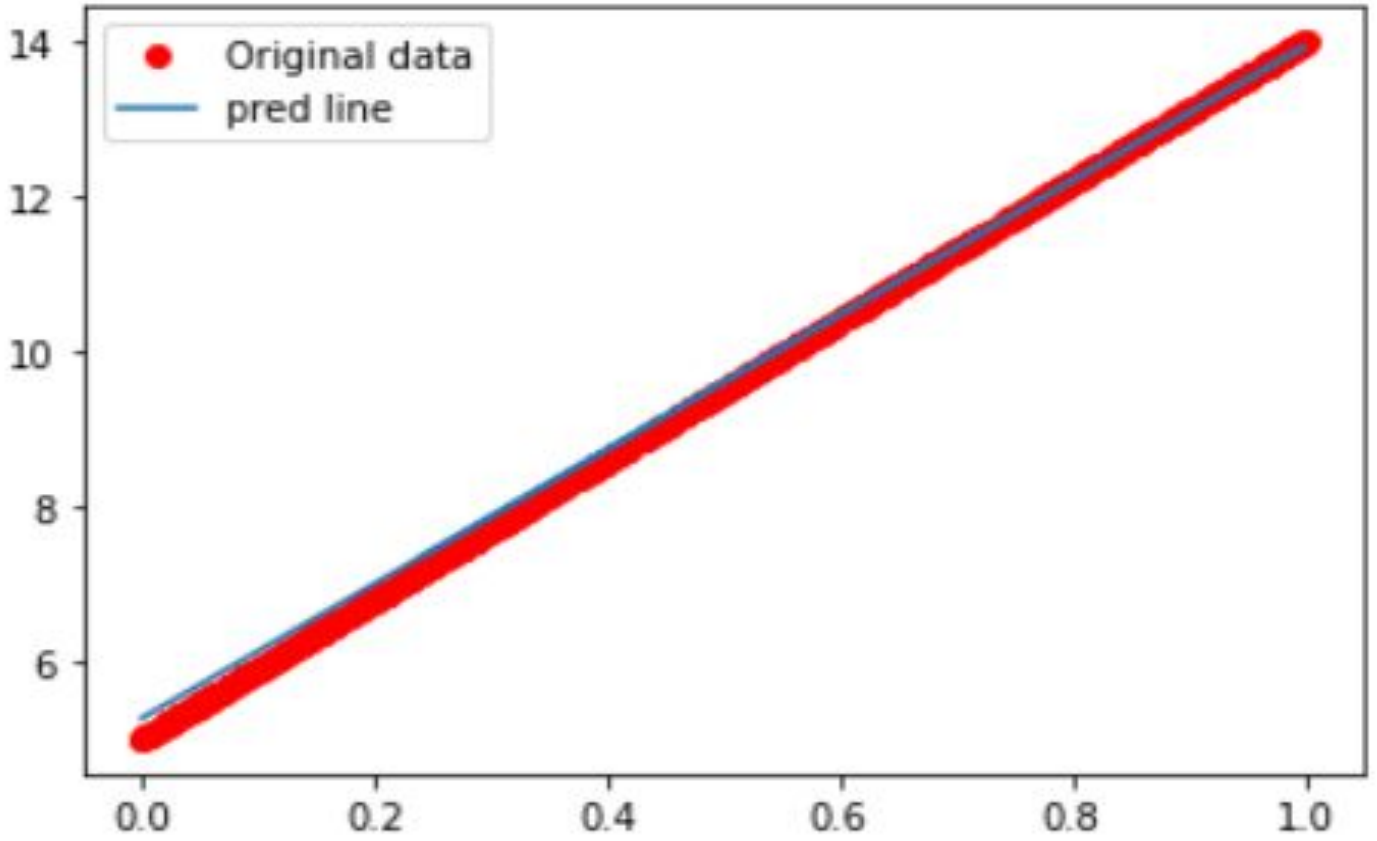
結果

安裝matplotlib

& pip install matplotlib

```
(tf2.3) C:\Users\user>pip install matplotlib
```

結果



訓練房價與面積關係

LotArea
8450
9600
11250
9550
14260
14115
10084
10382
6120
7420
11200
11924
12968
10652
10920
6120
11241
10791

SalePrice
208500
181500
223500
140000
250000
143000
307000
200000
129900
118000
129500
345000
144000
279500
157000
132000
149000
90000

Import library

```
import tensorflow as tf
import numpy as np
import pandas as pd
```

Import library

import pandas as pd

pandas是Python程式語言的用於數據操縱和分析的軟體庫。

dataset

```
train = pd.read_csv("train.csv")
train = train[train['LotArea'] < 5000]
train_X = train['LotArea'].values.reshape(-1,1)
train_Y = train['SalePrice'].values.reshape(-1,1)
n_sample = train_X.shape[0]
```

dataset

pd.read_csv("train.csv")

->讀取csv資料

train = train[train['LotArea'] < 5000]

->只讀取‘LotArea’中小於5000的資料

dataset

train_X = train['LotArea'].values.reshape(-1,1)

-> LotArea資料擺入train_X

train_Y = train['SalePrice'].values.reshape(-1,1)

-> SalePrice資料擺入train_Y

n_sample = train_X.shape[0]

-> n_sample 為train_X有幾筆資料

權重初始化

```
W = tf.Variable(tf.random.normal([1]))
print(W)
b = tf.Variable(tf.random.normal([1]))
print(b)
```

```
<tf.Variable 'Variable:0' shape=(1,) dtype=float32, numpy=array([1.2125201], dtype=float32)>
<tf.Variable 'Variable:0' shape=(1,) dtype=float32, numpy=array([-0.7893576], dtype=float32)>
```

訓練

```
for step in range(0,10001):
    with tf.GradientTape() as g:
        pred = W*train_X+b
        loss = tf.reduce_sum(tf.pow(pred-train_Y,2))/(n_sample)

    gradient = g.gradient(loss,[W,b])

    tf.optimizers.Adam(2).apply_gradients(zip(gradient, [W, b]))

    if step % 100 == 0:
        pred = W*train_X+b
        loss = tf.reduce_sum(tf.pow(pred-train_Y,2))/(n_sample)
        print("step: %i, loss: %f, W: %f, b: %f" % (step, loss, W.numpy(), b.numpy()))
```

預測為W*train_X+b

Loss採用mean square

Optimizer採用Adam

$$\frac{1}{N}\sum_{i=1}^{N}(y_i-\hat{y}_i)^2$$

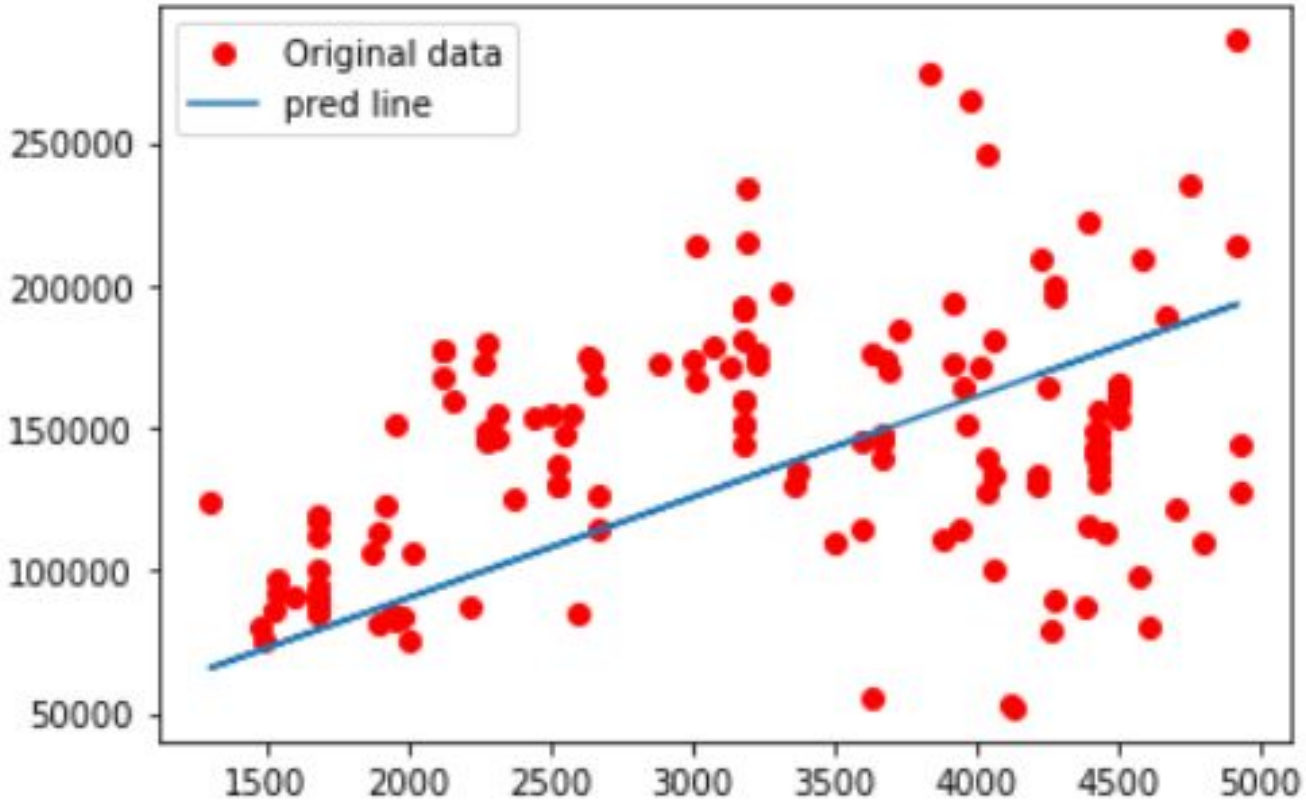
訓練

```
step: 8000, loss: 2411719936.000000, W: 39.213654, b: 16001.242188
step: 8100, loss: 2411141888.000000, W: 39.213654, b: 16201.242188
step: 8200, loss: 2409086720.000000, W: 35.213531, b: 16401.242188
step: 8300, loss: 2403479040.000000, W: 35.213531, b: 16601.242188
step: 8400, loss: 2397951488.000000, W: 35.213531, b: 16801.242188
step: 8500, loss: 2392503296.000000, W: 35.213531, b: 17001.242188
step: 8600, loss: 2387135744.000000, W: 35.213531, b: 17201.242188
step: 8700, loss: 2381848064.000000, W: 35.213531, b: 17401.242188
step: 8800, loss: 2376640256.000000, W: 35.213531, b: 17601.242188
step: 8900, loss: 2371512320.000000, W: 35.213531, b: 17801.242188
step: 9000, loss: 2366464512.000000, W: 35.213531, b: 18001.242188
step: 9100, loss: 2361496832.000000, W: 35.213531, b: 18201.242188
step: 9200, loss: 2356609024.000000, W: 35.213531, b: 18401.242188
step: 9300, loss: 2351801344.000000, W: 35.213531, b: 18601.242188
step: 9400, loss: 2347073536.000000, W: 35.213531, b: 18801.242188
step: 9500, loss: 2342425600.000000, W: 35.213531, b: 19001.242188
step: 9600, loss: 2337857792.000000, W: 35.213531, b: 19201.242188
step: 9700, loss: 2333370368.000000, W: 35.213531, b: 19401.242188
step: 9800, loss: 2328962560.000000, W: 35.213531, b: 19601.242188
step: 9900, loss: 2324634880.000000, W: 35.213531, b: 19801.242188
step: 10000, loss: 2320387072.000000, W: 35.213531, b: 20001.242188
```

結果

```
import matplotlib.pyplot as plt
plt.plot(train_X, train_Y, 'ro', label='Original data')
plt.plot(train_X, np.array( W * train_X + b), label='pred line')
plt.legend()
plt.show()
```

結果


Original data
pred line
250000
200000
150000
100000
50000
1500
2000
2500
3000
3500
4000
4500
5000

END!!!