

24/07/25

LAB-1

EXPLORING DEEP LEARNING PLATFORMS AND FRAMEWORKS

1. Various Deep Learning Platforms

> Google Colab :

- * Creator /Organization : Google Research

- * Main Features : Cloud-based notebooks, free GPU, auto save to Google Drive.

- * Popular Use-case : Prototyping, student projects, fast testing.

> Jupyter Notebook :

- * Creator /Organization : Project Jupyter.

- * Main Features : Interactive coding, supports multiple languages via kernels.

- * Popular Use case : Data analysis, education, ML/DL coding.

> Kaggle Kernels :

- * Creator /Organization : Kaggle (Google)

- * Main Features : Online notebooks with public dataset & competitions

- * Popular Use-case : ML competitions, code sharing, fast training.

2. Various Deep Learning Frameworks

> TensorFlow :

* Creator / Organization : Google Brain

* Main Features : Static and dynamic graphs,
TensorBoard, TF Lite, TF Serving

* Popular Use-case : NLP, image classification, production
apps.

> PyTorch :

* Creator / Organization : Facebook AI Research (FAIR)

* Main Features : Dynamic graphs, easy debugging

* Popular Use Cases : Research, computer vision, NLP

> Keras :

* Creator / Organization : Google (TF wrapper)

* Main Features : High-level API over TensorFlow

* Popular Use Case : Quick model building, education

⇒ Basic Deep Learning Script

> PyTorch $y = 2x$

import torch

import torch.nn as nn

import torch.optim as optim

model = nn.Linear(1, 1)

criterion = nn.MSELoss()

optimizer = optim.SGD(model.parameters(),
 $lr = 0.01$)

Expected output

Trained weight : tensor([2.0000])

Trained bias : tensor([0.0000])

```
xc = torch.tensor([[1]])  
yc = torch.tensor([[2]])
```

```
for epoch in range(100):
```

```
    y_pred = model(xc)
```

```
    loss = criterion(y_pred, yc)
```

```
    optimizer.zero_grad()
```

```
    loss.backward()
```

```
    optimizer.step()
```

```
print("Trained weight : ", model.state_dict()['linear.weight'])  
print("Trained bias : ", model.state_dict()['linear.bias'])
```

TensorFlow

```
y = 2
```

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

```
x = np.array([[1.0]]),
```

```
y = np.array([[2.0]]),
```

```
model = tf.keras.S
```

```
tf.keras.layers
```

```
])
```

```
model.compile(opt
```

```
model.fit(x, y,
```

```
x = torch.tensor([ [1.0], [2.0], [3.0] ])
```

```
y = torch.tensor([ [2.0], [4.0], [6.0] ])
```

```
for epoch in range(100):
```

```
    y_pred = model(x)
```

```
    loss = criterion(y_pred, y)
```

```
    optimizer.zero_grad()
```

```
    loss.backward()
```

```
    optimizer.step()
```

```
print("Trained weight:", model.weight.data)
```

```
print("Trained bias:", model.bias.data)
```

7 TensorFlow

$$y = 2x$$

~~import TensorFlow as tf~~

~~import numpy as np~~

```
x = np.array([ [1.0], [2.0], [3.0] ], dtype=np.float32)
```

```
y = np.array([ [2.0], [4.0], [6.0] ], dtype=np.float32)
```

```
model = tf.keras.Sequential([
```

```
    tf.keras.layers.Dense(units=1, input_shape=[1])
```

```
])
```

```
model.compile(optimizer='sgd', loss='mean_squared_error')
```

```
model.fit(x,y, epochs=100, verbose=0)
```

Expected Output:

Trained weight : [[2.7]]

Trained bias : [0.]

```
weights = model.layers[0].get_weights()  
print("Trained weight:", weights[0])  
print("Trained bias:", weights[1])
```

Conclusion:

Both models successfully learned the linear relationship between x and y .

