

实作-2

Lab 2

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声明

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背景

Background

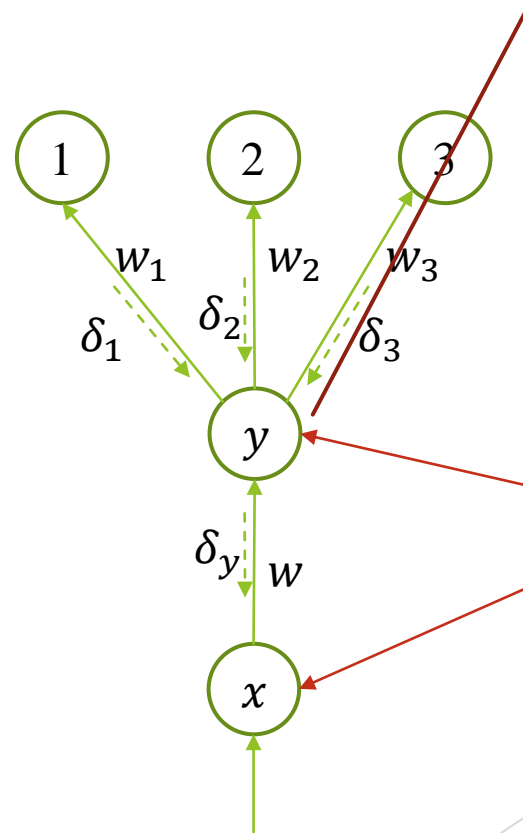
- ▶ 反向传播算法是逐层传播的，如果有中间值，计算起来没有技术难度，但会很繁琐：

Back-propagation takes place in a layer-wise manner. If all intermediate values are maintained, there is no technical difficulties, however, it is time- and resource-consuming:

- ▶ 一般性规则 (General rule)

$$\delta_y \triangleq (\delta_1 w_1 + \delta_2 w_2 + \delta_3 w_3) y'$$

$$w(t+1) = w(t) - \eta \delta_y x$$



$$y = \sigma \left(\sum wx \right)$$

σ 的导数的解析表达均存在，即容易计算 y' 。
The analytical form of the derivative of σ always exists, which makes computation of y' easier.

中间值
(Intermediate Values)

背景

Background

- ▶ 回顾机器学习四要素：数据，模型，损失函数，学习方法，上面的反向传播属于学习方法，同时，我们需要一个库来帮助我们管理其他三个要素。

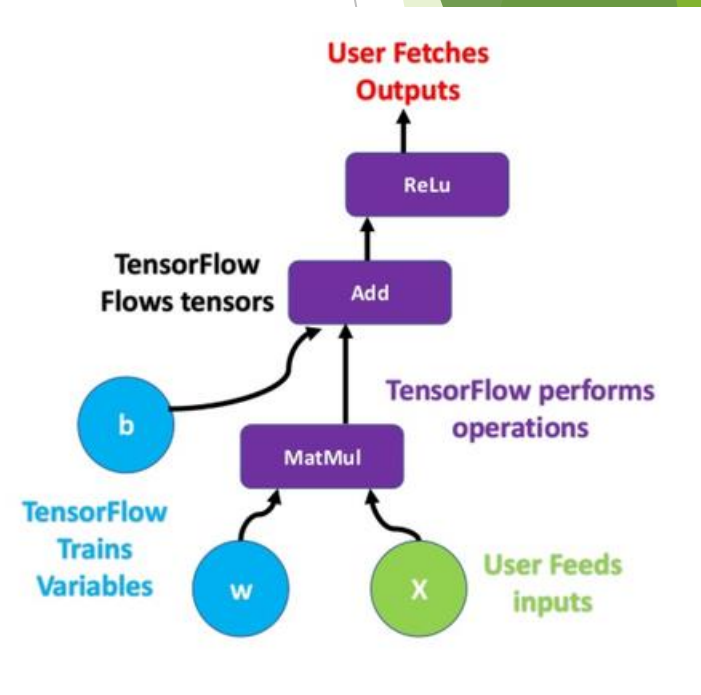
The four elements constituting machine learning include data, model, loss function and learning method. Back-propagation belongs to learning method category; however, a library is still needed to help manage other three elements.

- ▶ 由谷歌开发与开源的TensorFlow便是这样一个开源库，其为研究者及工业界快速实现深度学习模型提供了有力支持。

The library TensorFlow which is developed and open-sourced by Google offers scholars and industries the most powerful tool for conducting deep learning research.

TensorFlow

- TensorFlow从字面意思上讲，可以认为是张量（Tensor）与流（Flow）的加总。这里的张量与数学上的张量是不同的，可以认为是向量和矩阵的推广，即更高维度数组的一般化。TensorFlow在内部将张量表示为基本数据类型的 n 维数组。因为TensorFlow是将对Tensor的操作以单向图的形式表示的，流可以理解为数据流或操作流。
- Superficially speaking, TensorFlow can be treated as the combination of Tensor and Flow. The definition of Tensor here is quite different from the counterpart in mathematics. The tensor in DL context can be regarded as the extension of high dimensional arrays, which is the internal representation of Tensors in TensorFlow. Since TensorFlow arranges tensor in a manner of directed graph, flow is interpreted as data flow or operation flow.

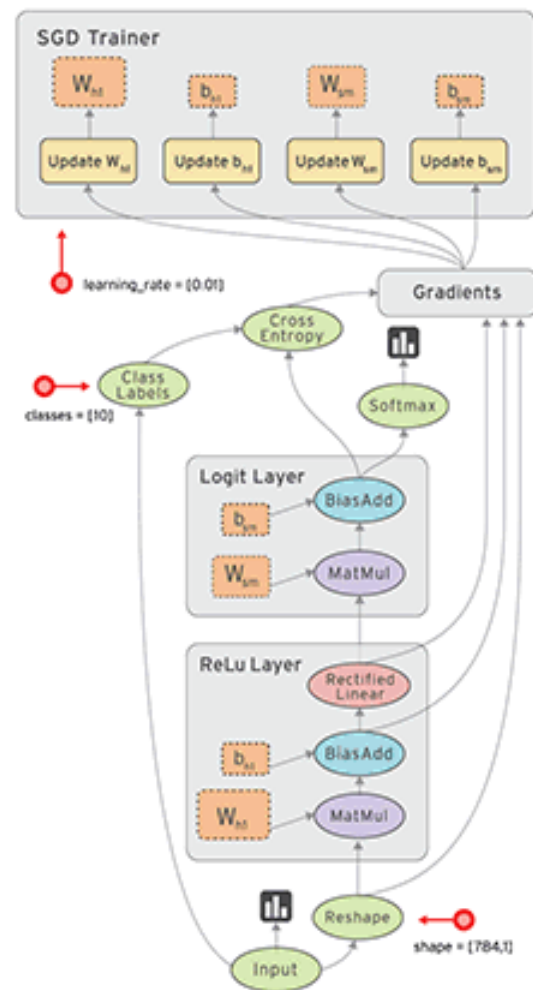
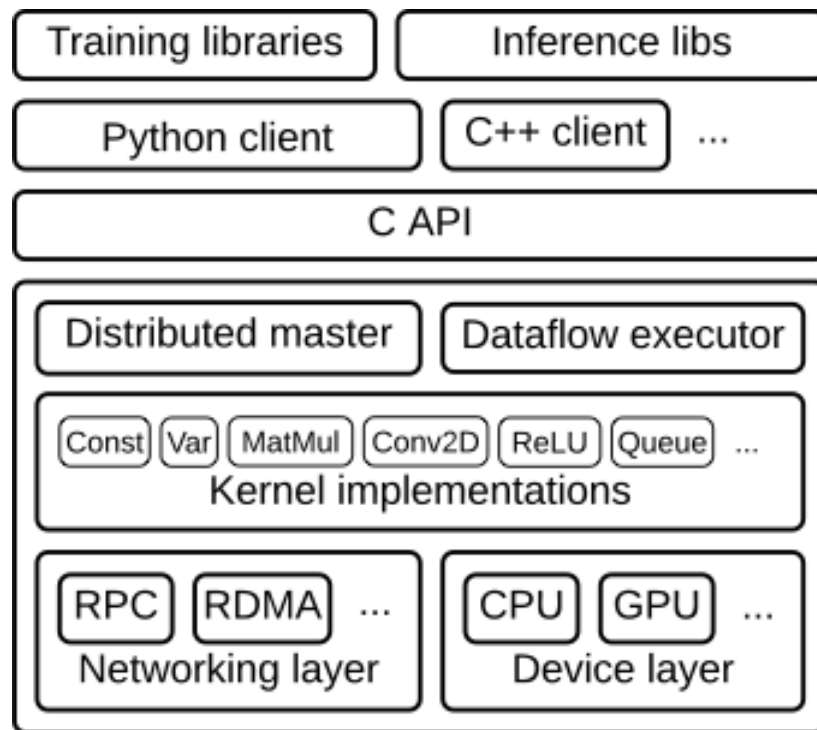


TensorFlow

- TensorFlow 1.x版本基于静态图，即先由Python语言构造好静态计算图，然后交由底层计算引擎执行，返回最后结果；2.x版本基于动态图，即计算图可以交互构建，部分地交由计算引擎执行，返回中间结果，这样有利于开发者调试程序。1.x更偏重于库的形式，即开发者要利用TensorFlow提供的API构建深度学习的整个流程涉及的方方面面；2.x更偏重于框架，即开发者在API中提供适当的回调，即可以完成深度学习的基本流程。

The 1.x version of TensorFlow is based on static computation graph, which is constructed by scripts language such as python, then delivered to the level engine for execution to have the final results return. Conversely, version 2.x is based on the dynamic computation graph, which means the graph can be partially constructed before send to the engine for computation. The intermediate result is also visible to the developer, which can ease the frustration when error occurs. Notably, 1.x is library-oriented, which means the developer has to utilize the provided API to construct each aspect of deep learning, while 2.x is more framework-oriented, which means the process can be automated on condition that proper callbacks are provided.

TensorFlow



TensorFlow

- ▶ 由于TensorFlow构建计算图由前端完成，而依据计算图进行计算由后端完成，因此需要有相关机制来请求相关资源进行相关计算，TensorFlow中定义了Session这一概念来完成对应功能。我们可以将计算机科学中进程的概念与TensorFlow中Session的概念进行类比，两者均有在特定的上下文中，进行资源管理与计算执行等动能。

To cater to the case that the graph construction is performed by the front-end whilst computation based on the graph is performed by the backend, it desires a mechanism to allocate the resources and manage the computations. To fulfill these demands, TensorFlow defines session, which can be compared with process in computer science. Both of them manages the resources and execute the computations in specific context.

TensorFlow

source code

```
import tensorflow as tf

x = tf.constant(8)
y = tf.constant(9)
z = tf.mul(x,y)

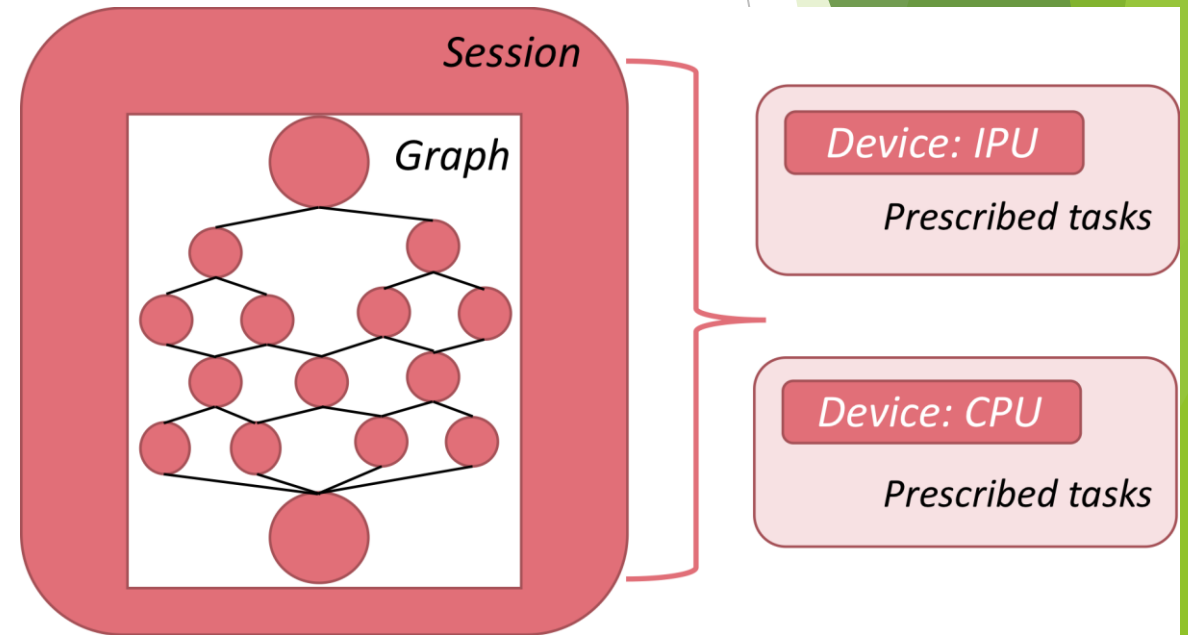
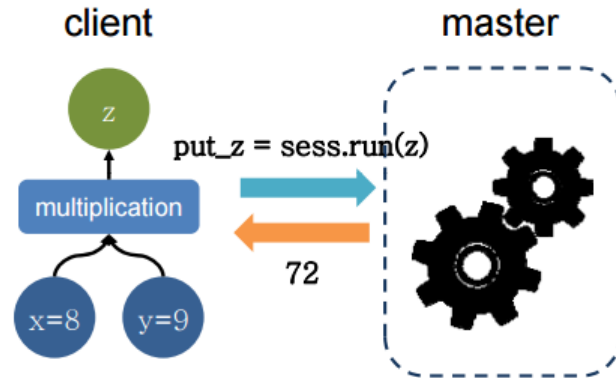
sess = tf.Session()

out_z = sess.run(z)

print('out_z: %d' % out_z)
```

output

out_z: 72



通用逼近定理

Universal proximation theorem

- 在人工神经网络的理论研究中，有一个很著名的定理，即如果网络足够大，足够深，则其可以以任意精度逼近任意性质良好的函数。在数值计算中，此又称为曲线拟合。在本实作中，我们会给出一些简单函数，并尝试构建神经网络逼近这些函数。

During the theoretic research of neural networks, it came up the theorem that by leveraging sufficient deep and wide neural network, any complicated function with well-stated properties can be approximated in any precision. It is also called curve fitting in numeric computations. In this lab, we will construct neural networks to approach some given functions.

习题

Problems

- ▶ 请按照实作-2实验文档，完成各步骤操作并观察结果

Follow the steps in the document of lab-2, finish each procedures and check the corresponding outputs.