

Introduction to Google's TensorFlow

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Ft. Lauderdale Machine Learning Meetup

Sponsors



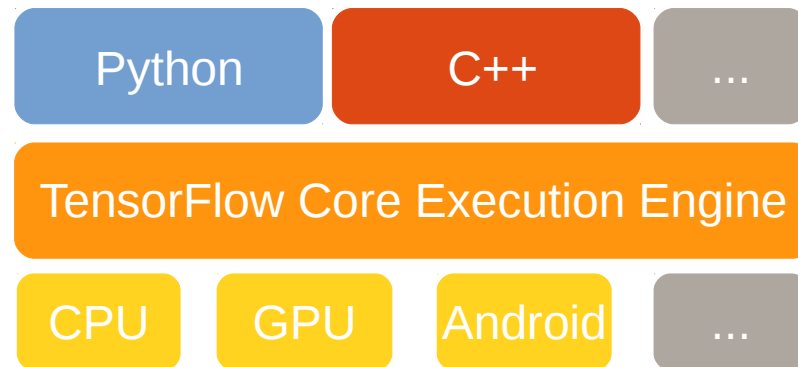
Agenda

- Why Do We Care?
- What is TensorFlow?
- Programming Model
- Underlying Implementations
- Extensions
- Machine Learning in TensorFlow
- TensorBoard
- Resources
- Conclusion

Why Do We Care?

- Has client language support in Python and C++ but many more are popping up in the community (e.g., Java and Rust).
- Has implementations for many platforms from mobile devices to large clusters of high end GPU servers.
- TensorFlow can be used for research and production systems, which greatly simplifies the adoption of machine learning algorithms.
- Support for distributed computation which allows TensorFlow clients to build larger models.
- Google is invested TensorFlow... They have the experience and resources to make it GREAT!

What is TensorFlow?



TensorFlow

/tensər floʊ/

An interface for expressing machine learning algorithms and an implementation for executing such algorithms.

Programming Model

The Data Flow Graph

Data flow graphs describe mathematical computation with a directed acyclic graph of nodes & edges.

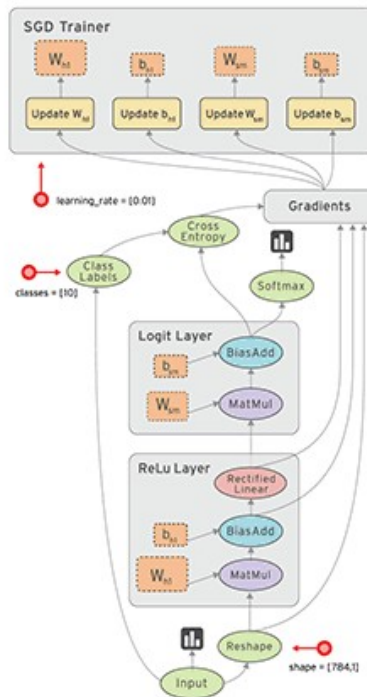


Figure 1: [<https://www.tensorflow.org>] A computation graph

Programming Model

Basic Concepts

- **Operation** – An operation has a name and represents an abstract computation. (e.g., “add” and “dot product”)
- **Kernel** – An implementation of an operation that can run on a particular type of device. (e.g., “CPU” and “GPU”)
- **Session** – Sessions manage and execute TensorFlow computation graphs. Therefore, client programs interact with the TensorFlow system through sessions.
- **Tensor** – A n-dimensional array that flows along the edges of the computation graph. Hence the name *TensorFlow* ;)
- **Variable** – A special kind of operation that returns a handle to a persistent mutable tensor that survives across executions of a graph.

Programming Model

Jupyter Notebook: Hands-on lab

Underlying Implementations

Components

The main components in a TensorFlow system are the client, which uses the Session interface to communicate with the master and one or more worker processes, with each worker process responsible for arbitrating access to one or more computational devices (such as CPU cores and GPU cards).

Underlying Implementations

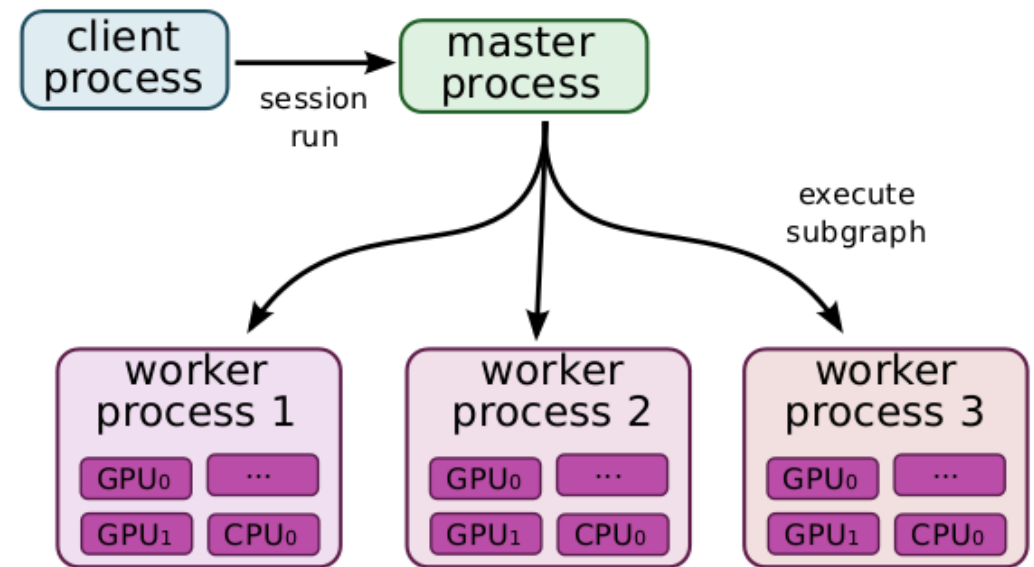
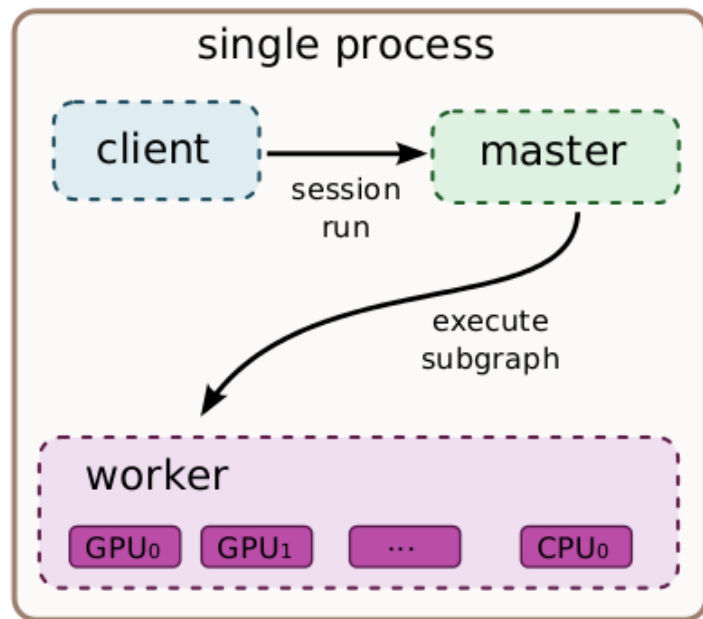


Figure 2: [TensorFlow: Large-scale machine learning on heterogeneous systems.] Single and Distributed System Structure

Extensions

- **Automatic Differentiation** – Automatically computes gradients for data flow graphs.
- **Partial Execution** – Allows TensorFlow clients to execute a subgraph of the entire execution graph.
- **Device Constraints** – Allows TensorFlow clients to control the placement of nodes on a device.
- **Control Flow** – Enables support for conditionals and loops in data flow graphs.
- **Input Operations** – Facilitate efficient loading of data into large scale models from the storage system.
- **Queues** – Allow different portions of the graph to execute asynchronously and to hand off data through Enqueue and Dequeue operation. Enqueue and Dequeue operations are **blocking**.
- **Containers** – The mechanism within TensorFlow for managing longer-lived mutable state.

Extensions

Automatic Differentiation

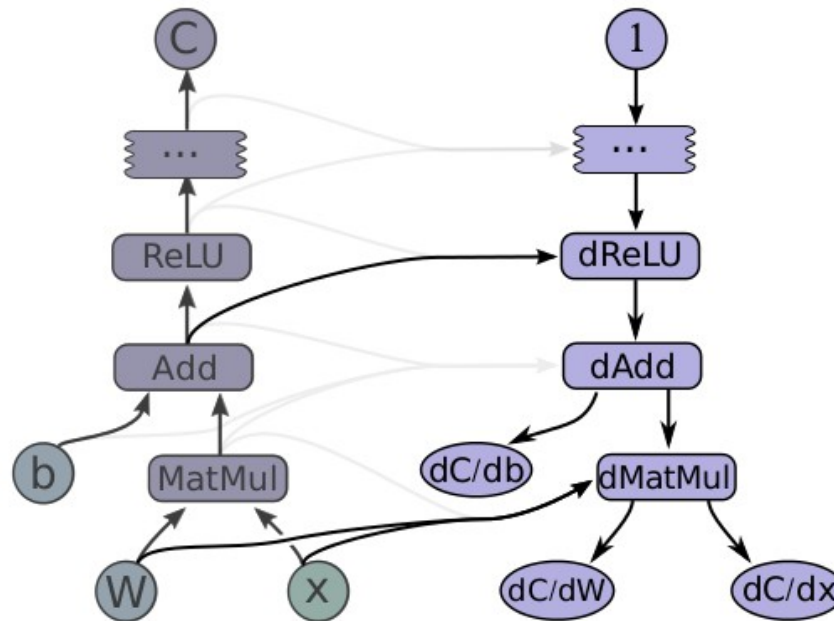


Figure 2: [TensorFlow: Large-scale machine learning on heterogeneous systems.] Gradients computed for graph.

Machine Learning with TensorFlow

Jupyter Notebook: Hands-on lab

Resources

Papers

[1] TensorFlow: Large-scale machine learning on heterogeneous systems. 2015

Websites

TensorFlow

Deep Learning | Udacity

GitHub

Conclusions

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