

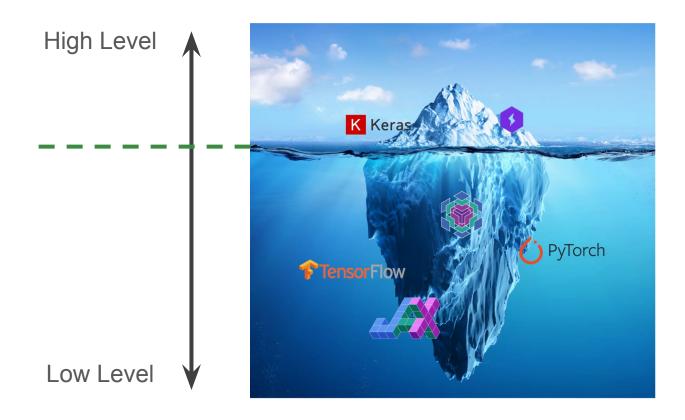


Introduction to

ML Frameworks

Together for a Better Algerian Economy

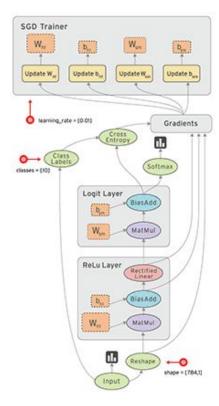
Overview





An open source Deep Learning Library

- Created in 2012
- C++, Python, CUDA
- Develop ML and DL models
- Released by Google Brain in 2015







An open source Deep Learning Library

- Supports many machine learning tasks (linear regression to complex deep learning architectures)
- Allows for low-level control over the model architecture.
- Higher-level control can be done via Keras, which, with the 2nd version, became a core part of Tensorflow
- Keras 3.0 is again moving away from a Tensorflow-only approach to support a variety of backends: JAX, Tensorflow, and Pytorch.
- Wide selection of platforms available with mobile devices (Tensorflow Lite) or large-scale distributed systems (Tensorflow Serving), production-ready deployment (TFX - Tensorflow Extended) or web deployment (Tensorflow.js for running models in Node.js).





Creating Models

Discovering

TF Hub ML Kit **Data**

KYD TF Data **Tooling**

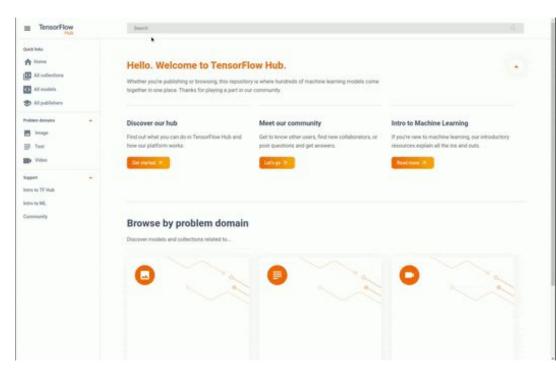
TFLMM TF Cloud COLAB





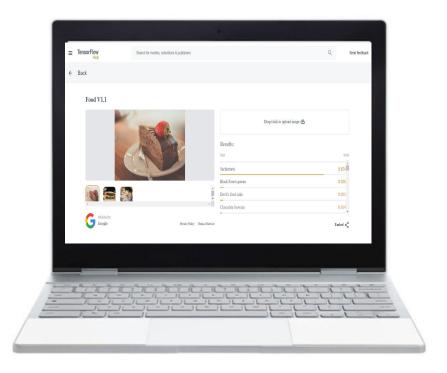
TF Hub

- +10000 pretrained models
- Test models on browser



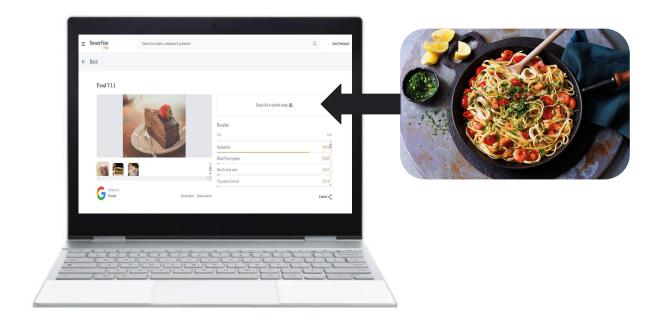
















Food V1.1



Privacy Policy Terms of Service

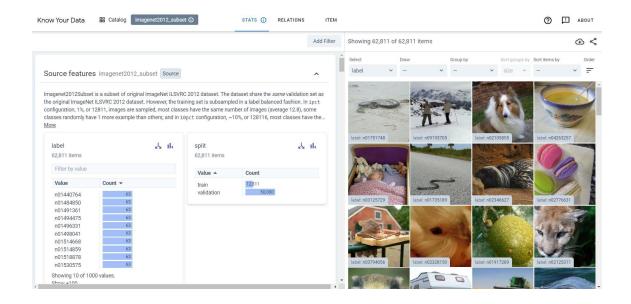






Embed <

Know Your Data



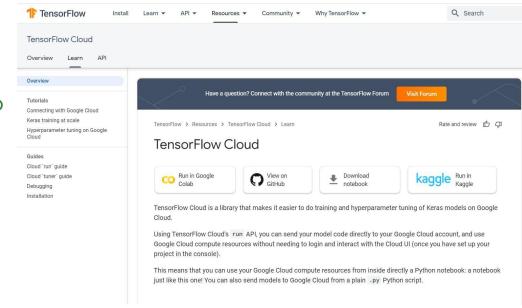




TensorFlow Cloud Easily train with the power of GCP

 Connect your local environment to Google Cloud

Get started: github.com/tensorflow/cloud







Analysing and Optimizing Models

Analysing your Model

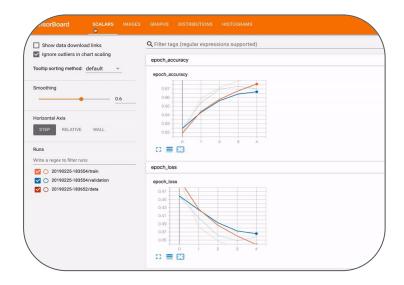
TensorBoard TF Profiler

Optimizing your Model

Model Optimization Toolkit Systrace Perfetto







Trace View overview_page · /device:GPU:0 (pid 1) Stream #19(Compute) input pipeline analyzer Stream #20(MemcpyH2D) Stream #21(MemcpyD2H) kernel stats ▼ TensorFlow Name Scope memory_profile tensorflow_stats TensorFlow Ops host:CPU (pid 501) trace viewer train_function Hosts (1) EagerExecute:__inference_train_function_664 51194053e9b6 EaperLocalExecute: Inference train function 894 tf Compute/159544737 Nothing selected. Tap stuff.

TensorBoard

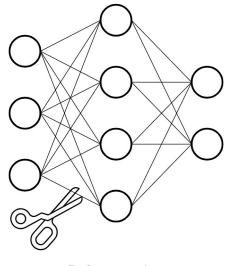
TensorFlow Profiler



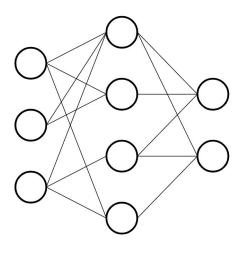


Model Optimization Toolkit

- Quantization
- Pruning



Before pruning

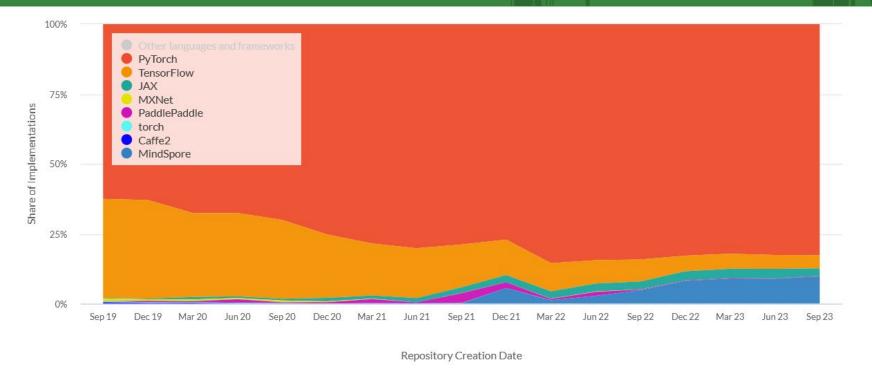


After pruning





PyTorch



The popularity of frameworks in new Machine Learning papers, source: visio.ai

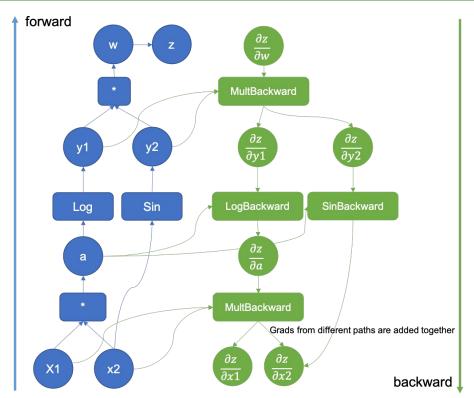


PyTorch

An open source DL Library

- Released in 2016 by Facebook's Al Research Lab (FAIR)
- <u>Dynamic Computation Graph</u>

 (i.e. Graph built on the fly) and ease of use
- Easier Net debugging and runtime modification
- Auto grad computation
- Growth of specialized tools and libraries (MONAI, Diffusers, Pytorch 3D, fast.ai, etc.)
- Dethroned TF's stable position





PyTorch Lightning

An open source DL Library

- High Level Framework built on top of PyTorch
- Abstracts complicated operations that had to be written (i.e., training loops and validation steps) and allows researchers to focus on the core aspects of model development.
- Easy scalability with multiple GPUs and nodes.
- Automatic handling of common tasks like checkpointing, logging or early stopping.
- Even more modular structure, making code easier to read and maintain.



An open source Numerical Computation Library

- Developed by Google.
- Combines Python flexibility with hardware acceleration.
- Automatic differentiation (autograd) native with Python and NumPy.
- <u>Just-In-Time Compilation</u> (jit), allowing for speedups of Python function into optimised machine code, able to run on CPU, GPU and TPU.
- Extensive vectorisation and parallelism (vmap) enabling efficient code execution across different hardware platforms.



Flax

A High Level Neural Network Library

- Developed by Google.
- Built on top of JAX.
- Provide modular design with ease of customisation.
- Easy manage model parameters, state, and randomness.
- Separate model and training logic, making code easier to debug.

Combining <u>JAX and Flax</u> is the primary weapon of choice by both Deepmind and Google AI.



Popularity











More than 75% of new deep learning papers use PyTorch



Differences between Frameworks

- Different Initializers of Weights
 - Tensorflow uses <u>Xavier Uniform initialisation</u>
 - Pytorch uses <u>Kaiming Uniform initialisation</u>
 - Flax uses <u>LeCun Normal initialisation</u>

Different initializers might greatly <u>influence how the models start the</u>
<u>training</u> and in which <u>direction</u> they might go



Differences between Frameworks

- Different Processing Backends
 - Images and data augmentation handling:
 - Pytorch uses a PIL-based approach
 - Tensorflow/JAX use more native solutions.

Python Libraries For Image Processing SimpleTK OpenCV Scikit-Image Pgmagick Pillow/PIL Matplotlib SciPy Mahotas

96



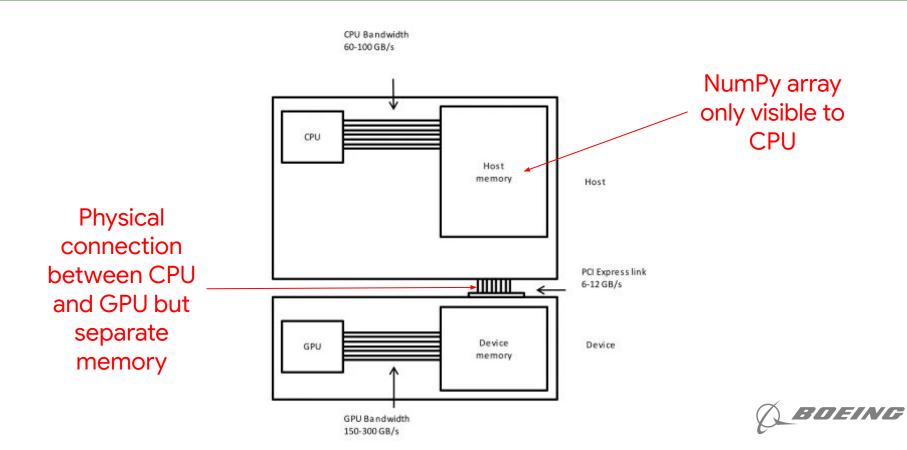
Differences between Frameworks

- Different GPU Memory Allocation Algorithms
 - Pytorch and Pytorch Lightning incrementally allocate memory, allocating more when needed.
 - Tensorflow and JAX operate in a greedy fashion, which might cause strange errors when used in the same scope.

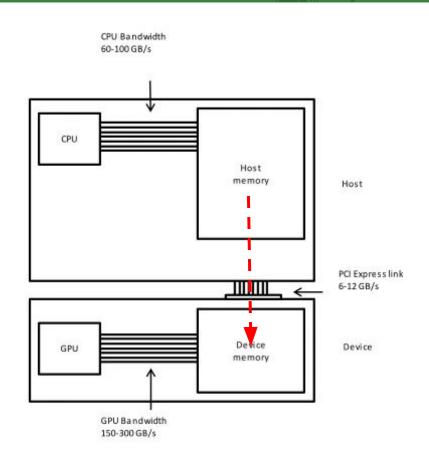
Different strategies influence <u>processing time</u> but also <u>allocate more</u> <u>memory</u> when needed, blocking the possibility of using different components that might require GPU.



Processing Units



Processing Units



Data Transfer required from CPU to GPU



Conclusion

Which is best?

- Pytorch is well established.
- Tensorflow has technical debt from previous solutions.
- JAX + Flax still need to show their great potential as their community grows by the day.



Which Framework would you like to kickstart with?



References

What's new in TensorFlow? Google I/O Extended 2021 by Ihssene Brahimi.

https://softwaremill.com/ml-engineer-comparison-of-pytorch-tensorflow-jax-and-flax/
Introduction to GPU programming with cuda and python by Ihssene Brahimi.







World Learning EDUCATION | DEVELOPMENT | EXCHANGE