

Where we are today with Deep Learning and Kubernetes

江骏 ohmystack @ 饿了么









深度学习平台两大派别



· 继承 Apache 系列,在此之上做调度、研发

• 专注于 TensorFlow、MXNet、pyTorch 这些新的框架,

与 Cloud 结合,提供配套的平台服务









Kubernetes 的帮助下,

个任务"这件事,已经变得非常简单。

这绝不是 Deep Learning Platform

所关注的重点。





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elearn

Preprocess

Hive, Spark, Storm,

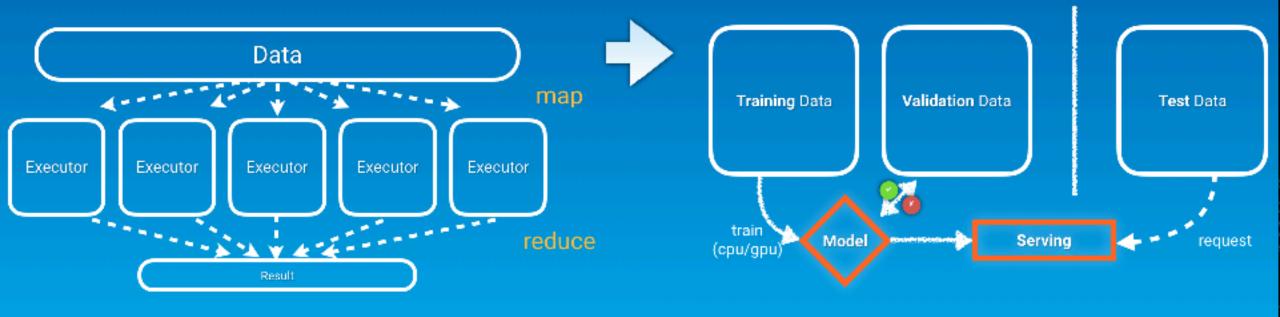
Distributed Storage

NFS, HDFS, S3, ...

Training

TensorFlow

(Distributed Training)



	通常的代码	Deep Learning
开发环境	虚拟机	Where? 数据的展示、分析, 大量的开发数据 & 生产数 据
版本管理	git	How to control the model versions?
发布	各种灰度发布的策略	How to serve the model?
	十分成熟	刚刚起步







Deep Learning Platform 的基本功能



Distributed Training

Model Serving

Model Version
Controlling

Auth

Datastore

Model Serving

Hyper-tuning

Elearn







Kubernetes 中,Deep Learning 任务与普通任务的不同



- 更加需要分布式存储
- 任务的 Restart Policy 往往无法直接满足需求
- Kubernetes Quota 机制无法直接满足需求
- 任务本身的资源需求 (CPU, memory, GPU) 偏大







Papers by Google



Hidden Technical Debt in Machine Learning Systems

https://papers.nips.cc/paper/5656-hidden-technical-debt-in-machine-learning-systems.pdf

TFX: A TensorFlow-Based Production-Scale Machine Learning Platform

http://www.kdd.org/kdd2017/papers/view/tfx-a-tensorflow-based-production-scale-machine-learning-platform

KDD 2017 Applied Data Science Paper

KDD'17, August 13-17, 2017, Halifax, NS, Canada

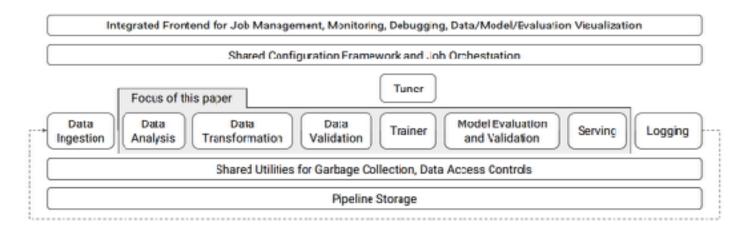


Figure 1: High-level component overview of a machine learning platform.







Cloud Jupyter Notebook



Why we need this

- GPU 卡使用效率低
 - 一个人占用一整台
 - 机器GPU 型任务变成了串行,需要排队等资源
- 任务缺乏管理
- 工程师搭建自己的开发环境麻烦







算法开发环境乱象



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[root@wg-bdi-vision-1 ~]\$ pstree 4048 -a python └-8*[{python}]

> cpu 100%, 但根本看不出进程是在跑什么内容

别人的进程被强行 OOM kill



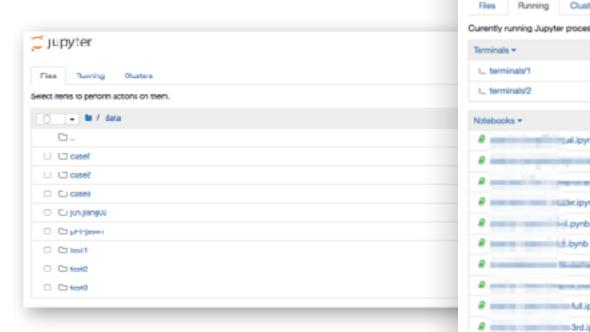


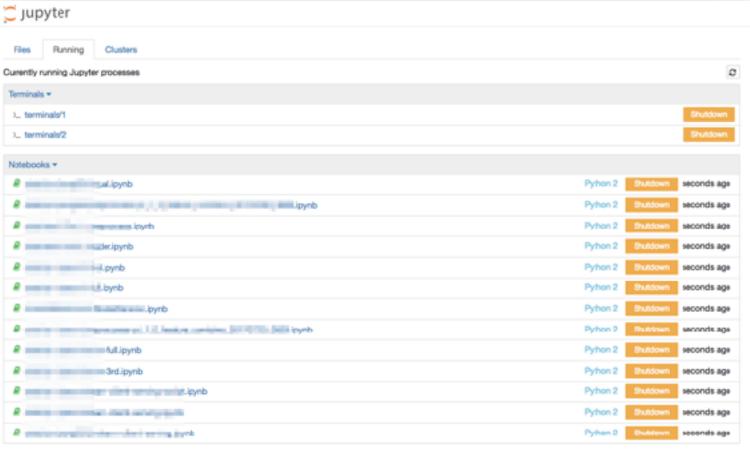


Cloud Jupyter Notebook



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子台,我们真正看的是它的附加值。

而不是封装,或者是界面的改变。



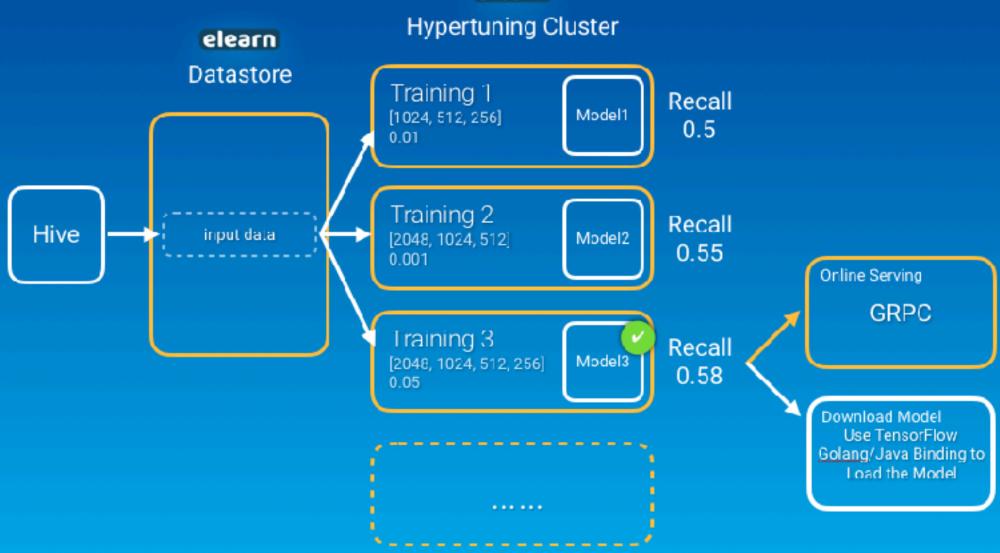
主办





Hypertuning

elearn





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Google Vizier

https://research.google.com/pubs/pub46180.html

Google Vizier: A Service for Black-Box Optimization

Venue

ACM (2017)

Publication Year

2017

Authors

Daniel Golovin, Benjamin Solnik, Subhodeep Moitra, Greg Kochanski, John Elliot Karro, D. Sculley

Abstract





Any sufficiently complex system acts as a black box when it becomes easier to experiment with than to understand. Hence, black-box optimization has become increasingly important as systems have become more complex. In this paper we describe **Google Vizier**, a Google-internal service for performing black-box optimization that has become the defacto parameter tuning engine at Google. Google Vizier is used to optimize many of our machine learning models and other systems, and also provides core capabilities to Google's Cloud Machine Learning **HyperTune** subsystem. We discuss our requirements, infrastructure design, underlying algorithms, and advanced features such as transfer learning and automated early stopping that the service provides.





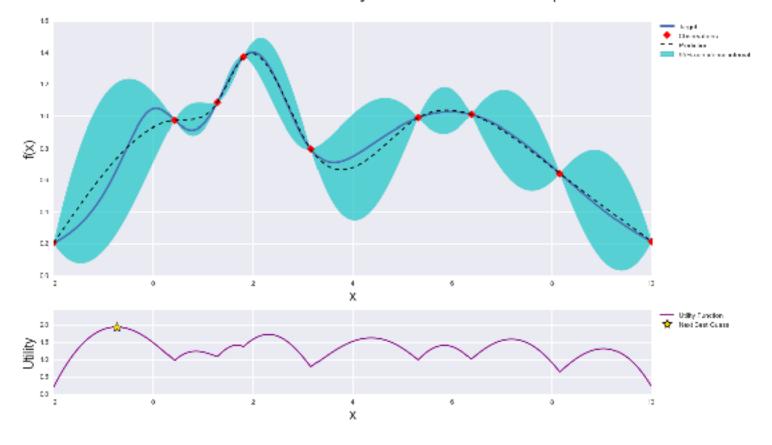




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https://github.com/fmfn/BayesianOptimization

Gaussian Process and Utility Function After 9 Steps



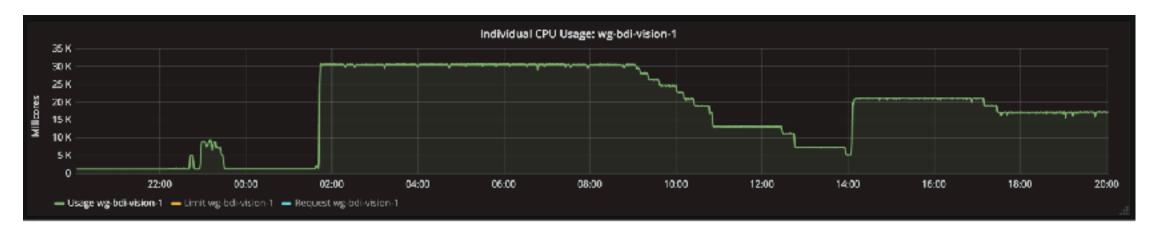


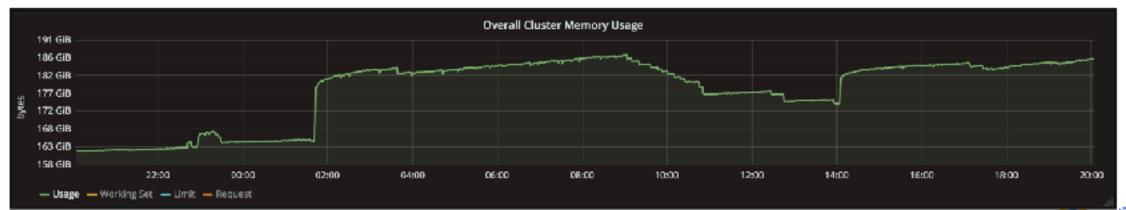






Hypertuning 时,集群中某单台物理机 metrics。充分发挥 Cloud 的计算能力。







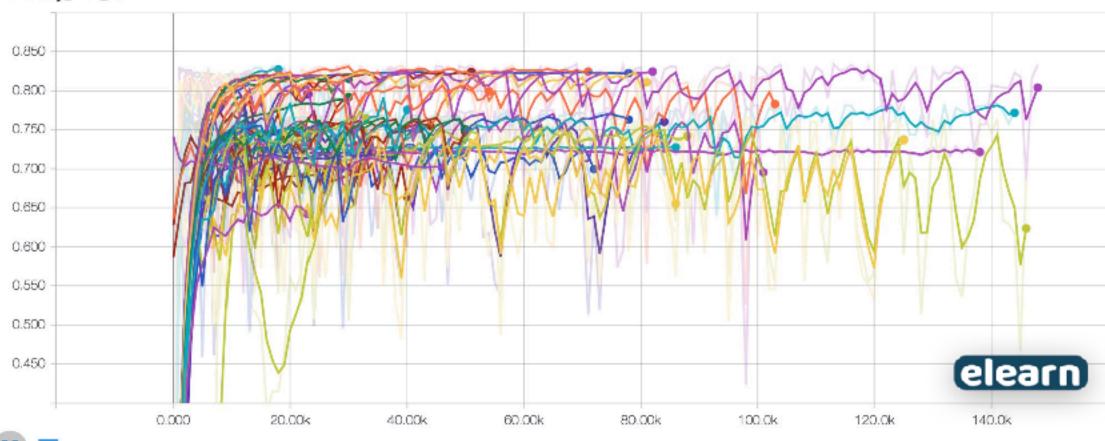


Hyper Parameter Tuning in a Real Model



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accuracy_with_10













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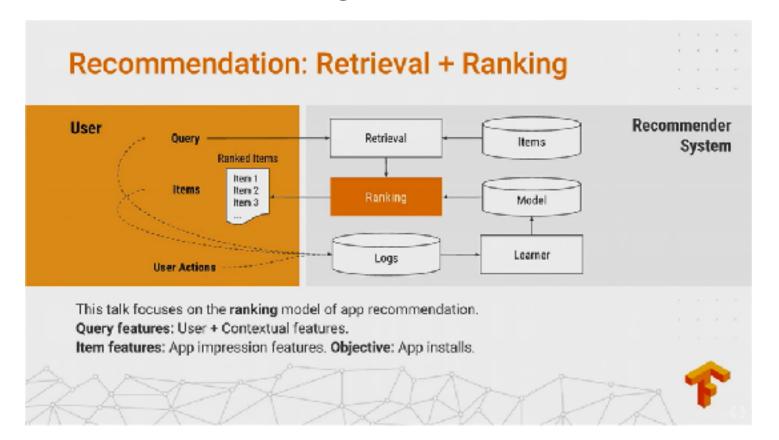




Much more things to do ...



Streaming Live Model Fine-tuning









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Q&A

江骏 @ 饿了么

https://github.com/ohmystack http://weibo.com/jiangjun1990





