What is Analytics Zoo



Distributed, High-Performance

Deep Learning Framework

for Apache Spark



https://github.com/intel-analytics/bigdl



Unified Analytics + AI Platform

Distributed TensorFlow, Keras, PyTorch and BigDL on Apache Spark



https://github.com/intel-analytics/analytics-zoo

Accelerating Data Analytics + Al Solutions At Scale



ANALYTICS Cluster Serving



What's Analytics Zoo

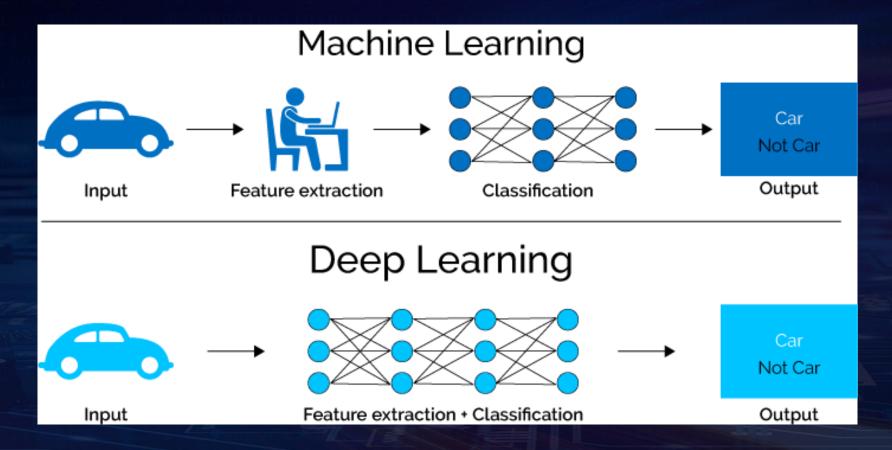


Distributed TensorFlow*, Keras*, PyTorch* and BigDL on Apache Spark*

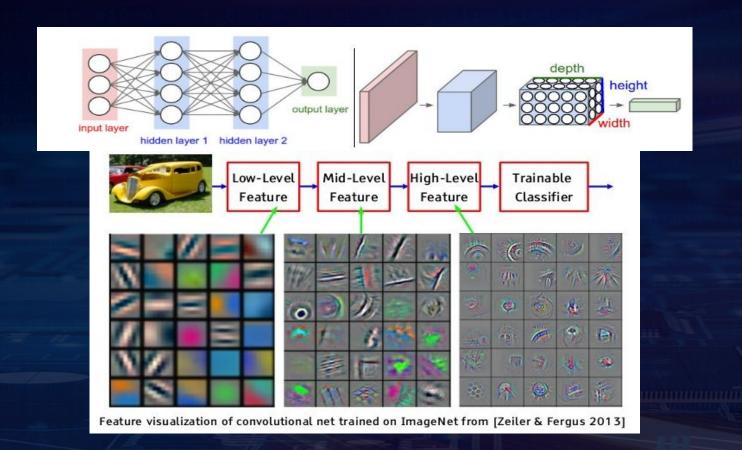
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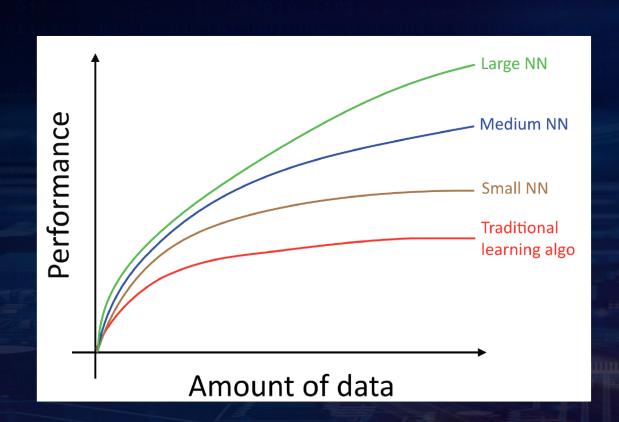
Machine Learning VS Deep Learning



Feature Visualization



Data & Performance Relationship



"Machine Learning Yearning",
Andrew Ng, 2016

Real-World ML/DL Applications Are Complex Data Analytics Pipelines

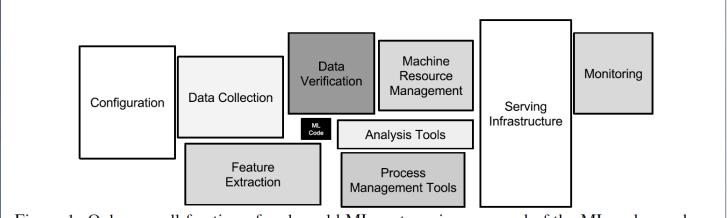
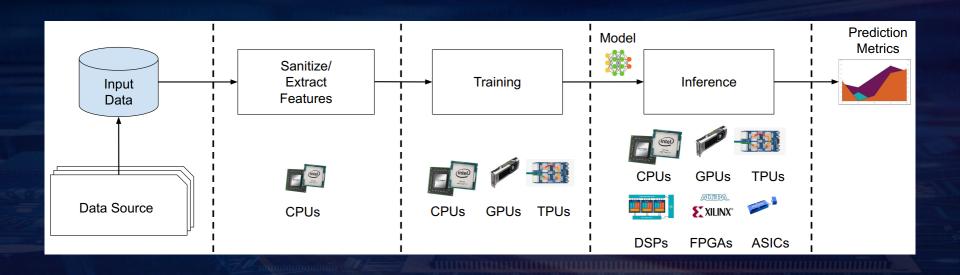


Figure 1: Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small black box in the middle. The required surrounding infrastructure is vast and complex.

"Hidden Technical Debt in Machine Learning Systems", Sculley et al., Google, NIPS 2015 Paper

Data Analytics Pipeline from production perspective



End-to-End Big Data Analytics and AI Pipeline

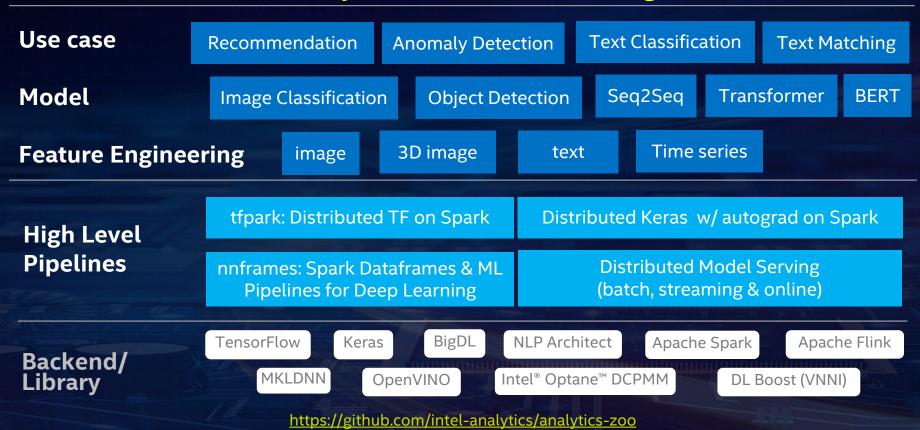
Seamless Scaling from Laptop to Production with



- "Zero" code change from laptop to distributed cluster
- Directly access production data (Hadoop/Hive/HBase) without data copy
- Easily prototype the end-to-end pipeline
- Seamlessly deployed on production big data clusters

Analytics Zoo

Unified Analytics + AI Platform for Big Data



What's Analytics Zoo

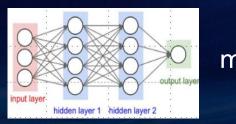


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What's Serving



model

Input Data

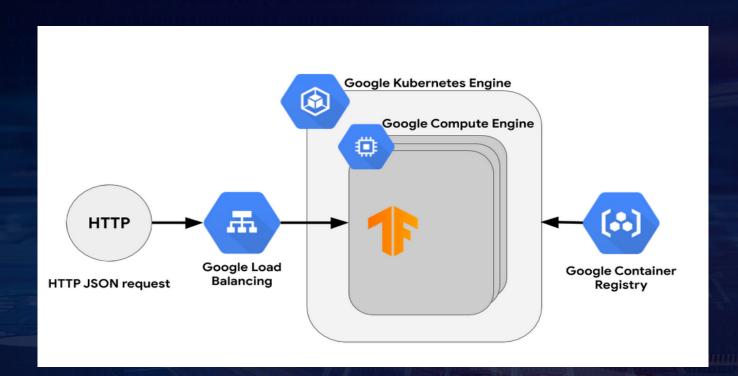
Preprocessing

Predict

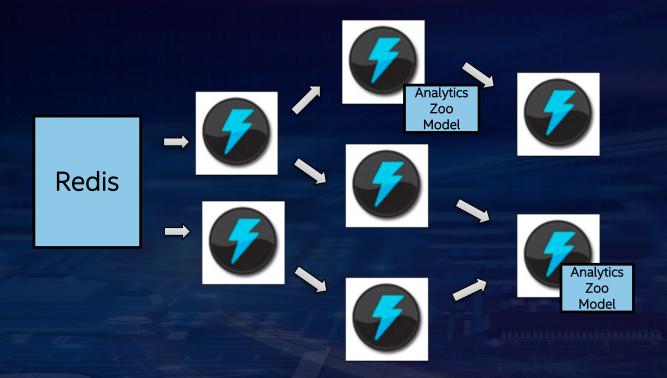
Postprocessing

Result

Example of TFServing



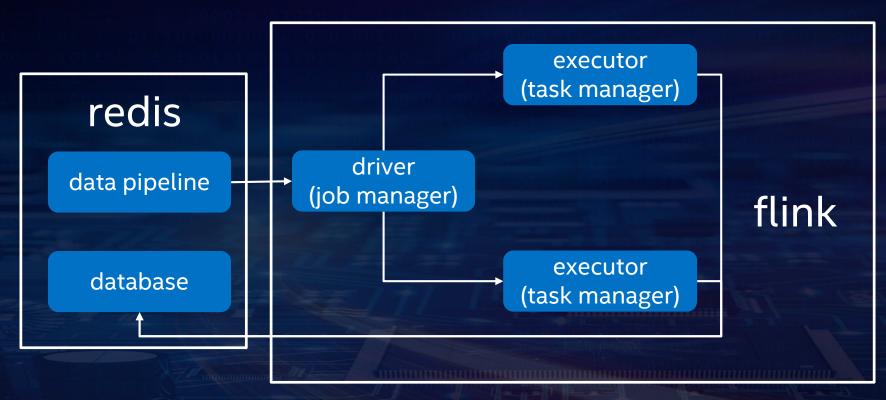
Distributed Model Serving



Distributed model serving in Web Service, Flink, Kafka, Storm, etc.

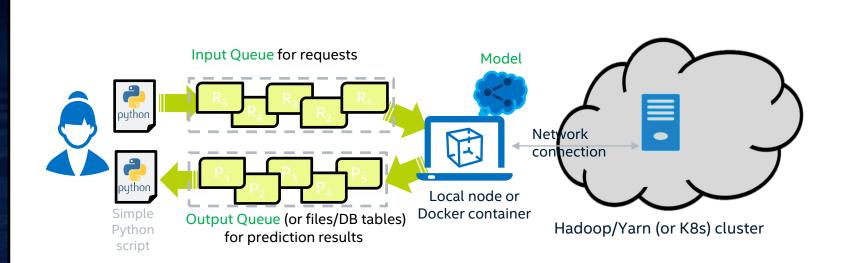
Plain Java or Python API, with OpenVINO and DL Boost (VNNI) support

Main Version of Cluster Serving



Version based on Spark Streaming is also supported.

Data pipeline User Perspective



Deploy Your Own Cluster Serving

One command to pull docker image, and customize your config, then call cluster-serving-start to start your serving

example of config:

```
## Analytics-zoo Cluster Serving
model:
    # model path must be set
    path: resources
data:
    # redis address
    src: XXXXXXX:6379
...
```

API Introductions

http API

data are represented by json format, and call http post method to enqueue your data into pipeline (http API is compatible with TFServing)

python API

data are represented by ndarray, and call python method to enqueue your data into pipeline

Use Case – Medical Imaging Analysis



Consider a Very large medical image of patient, the mission is to determine if tumor exists

End-to-end pipeline would contain image preprocessing, predict, all reduce analysis

tumor



if num > 1% the condition is bad

Advantages

Wide Range Deep Learning model support

Tensorflow, Caffe, OpenVINO, Pytorch, BigDL

Low Latency

Continuous Streaming pipeline is supported by Apache Flink, also Spark version is supported for users who are more familiar.

High Throughput & Scalability

Optimization of multithread control, and could easily scale out to clusters.

Very Quick Start

docker run -itd --name cluster-serving --net=host intelanalytics/zoo-cluster-serving:0.7.0

Log into the container using docker exec -it cluster-serving bash.

We already prepared analytics-zoo and opency-python with pip in this container. And prepared model in model directory with following structure.

Start Cluster Serving using cluster-serving-start.

Run python program python quick_start.py to push data into queue and get inference result.

Then you can see the inference output in console.

```
image: fish1.jpeg, classification-result:class: 5's prob: 0.18204997
image: dog1.jpeg, classification-result:class: 267's prob: 0.27166227
image: cat1.jpeg, classification-result:class: 292's prob: 0.32633427
```

End-to-End Big Data and AI Pipelines

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https://github.com/intel-analytics/analyticszoo/blob/master/docs/docs/ClusterServingGuide/ProgrammingGuide.md



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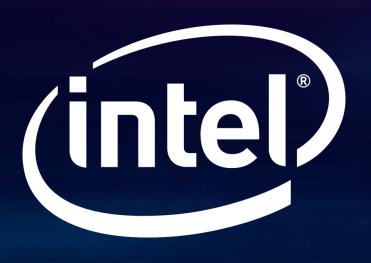
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