Research Review 2017

Measuring Performance of Big Learning Workloads

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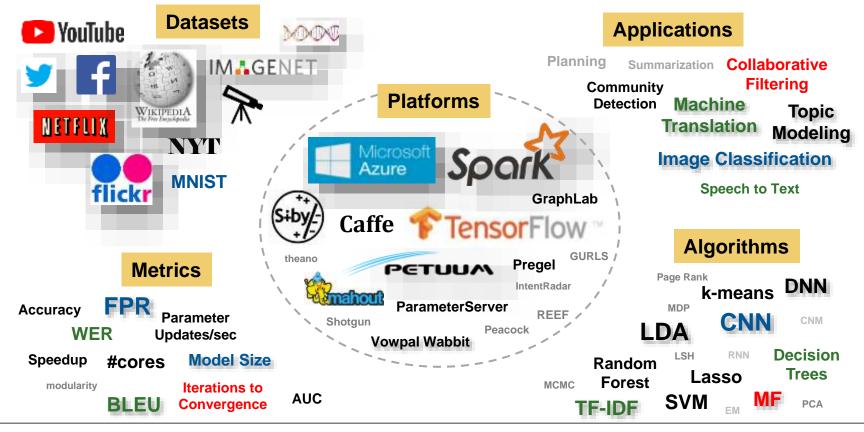
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Project Introduction: Big Learning Landscape



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Big Learning: Large-Scale Machine Learning on Big Data

Problem:

- Over 1,000 papers are published each year in Machine Learning.
- Most are empirical studies and few (if any) provide enough detail to reproduce the results.
- Complexity of systems begets complexity in metrics often partially reported.
- Slows adoption by DoD of new advances in Machine Learning.

Solution:

• Facilitate consistent research comparisons and advancements of Big Learning systems by providing sound *reproducible* ways to measure and report performance.

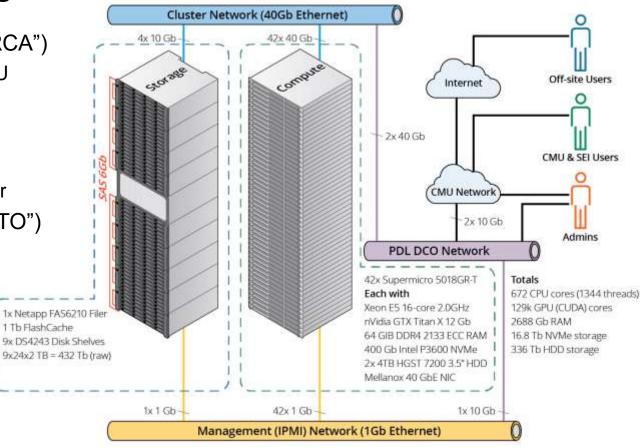
Approach:

- Develop technology platform for consistent (and complete) evaluation.
 - Performance of the computing system
 - Performance of the ML application
- Evaluate relevant Big Learning platforms using this benchmark: Spark+MLlib, Tensorflow, Petuum
- Collaborate with CMU's Big Learning Research Group

¹FACT SHEET: National Strategic Computing Initiative, 29 July 2015.

ORCA: Big Learning Cluster

- 42 compute nodes ("ORCA")
 - 16 core (32 thread) CPU
 - 64GB RAM
 - 400GB NVMe
 - 8TB HDD
 - Titan X GPU accelerator
- Persistent storage ("OTTO")
 - 400+ TB storage
- 40Ge networking



Performance Measurement Workbench (PMW) Architecture

Persistent Services

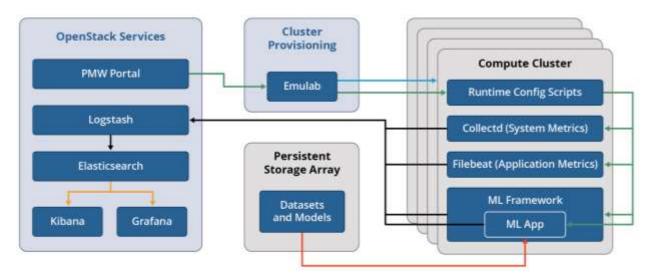
- OpenStack
- Web portal
 - simplifies provisioning
 - coordinates tools
- Data collection/analysis
 - "Elastic Stack"
 - Grafana

Provisioning

- Emulab
- "bare-metal" provisioning

Hardware Resources

- Compute cluster
- Data storage





Performance Measurement Workbench (PMW) Architecture

Persistent Services

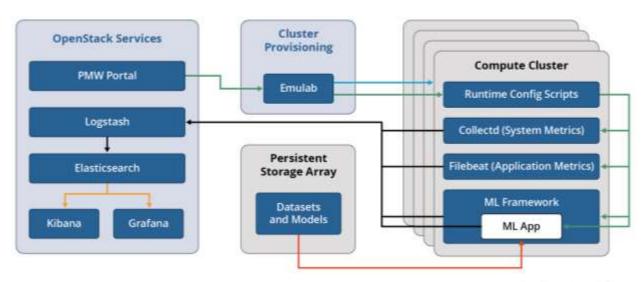
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Provisioning

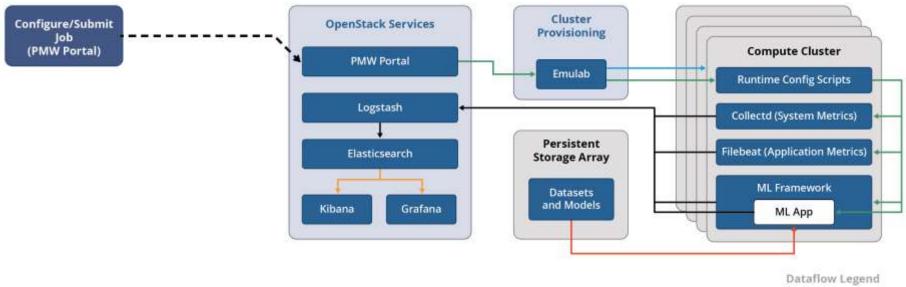
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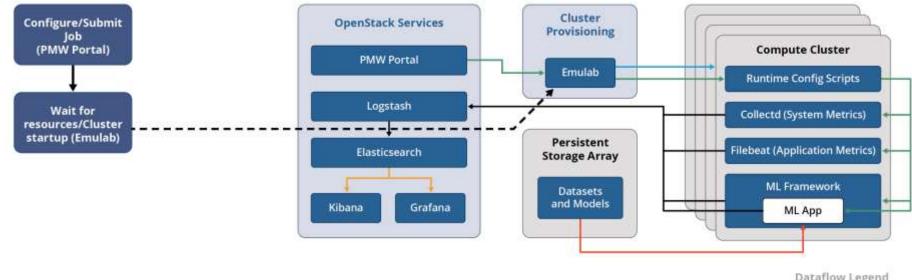
Hardware Resources

- Compute cluster
- Data storage

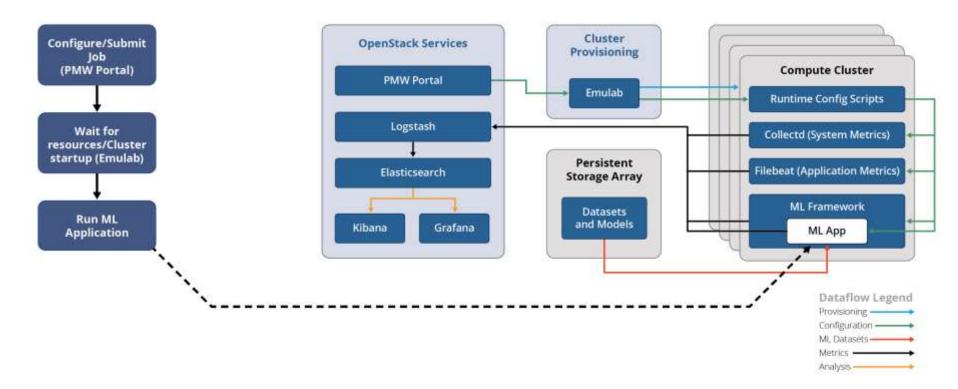


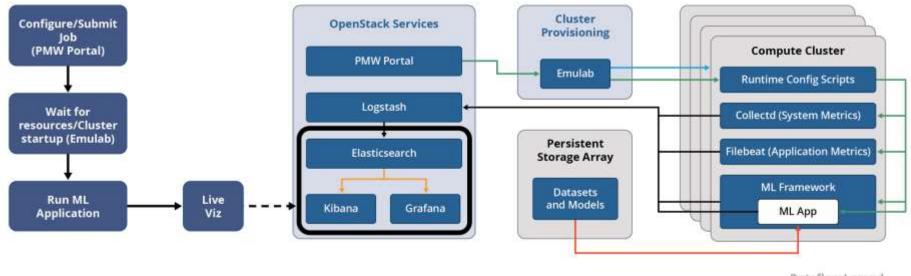


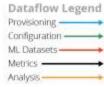


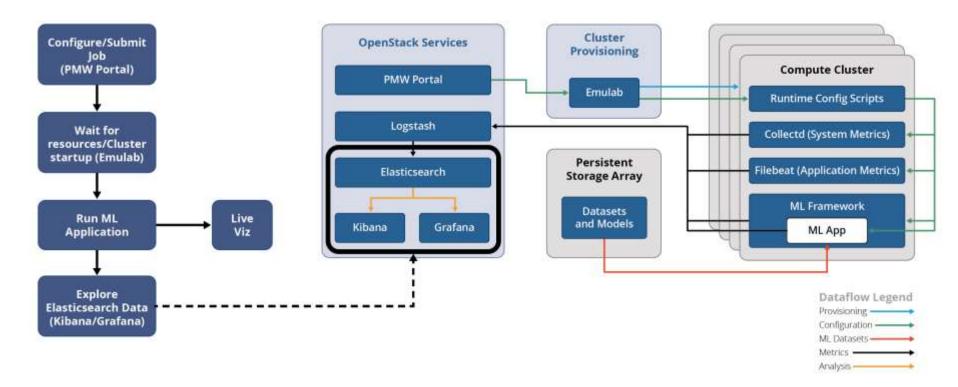








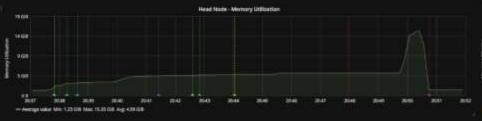


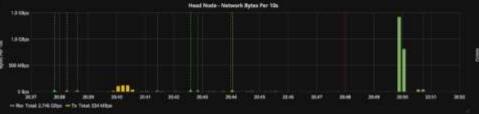


😚 - 🏙 Eight Node Cluster (8 + 1) - 03-13-2017 - 👥 🛤 🚯 🛛

Dashboard (Live Display and Historic)*



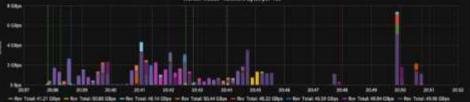




Heart Node - Disk Dytes Per 10c TO Miles 102408-0 THE VALUE 41 10 10 IN MICH 10 10 10 10 10 4/01 wate on a set 2.814 10.11 2014 100.00 20.00 ant 10042 1000 - Read Total 10.6 Million - Write Total 181.0 Million

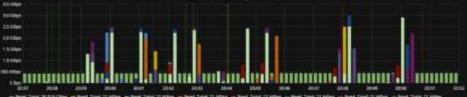


Worker Nodes - Network Bytes per 10s



- Ren Transf 47.21 Gilge - Ren Transf 40.84 Gilge - New Transf 49.14 Gilge - Ren Transf 49.14 Gilge - Te Transf 49.14 Gilg

Worker Nodes - Disk Bytes Per 10c

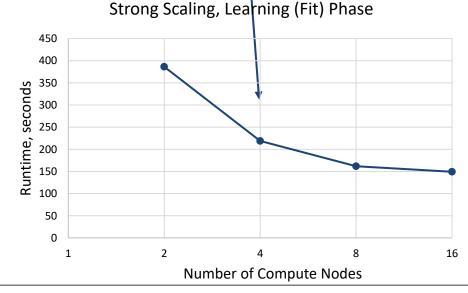


- Point Theor 2019 Cites - Intel Tool 21 Million - Book Tool 21 Million - Anna Tool 21 Million - Noval Tool 21 Million -

In-Depth Analysis

framework	algorithm	data_set	nodes_count ~	load_training_phase_time	fit_phase_time	load_testing_phase_time	validate_phase_time
spark	linear_regression	5000_by_500000	3	68.713	386.479	18.952	321.56
spark	linear_regression	5000_by_500000	5	35.767	218.848	16,905	199.124
spark	linear_regression	5000_by_500000	9	26.86	161.769	14.948	164.846
spark	linear_regression	5000_by_500000	17	25.815	149.232	14.173	147.584

- Supports arbitrary queries
- Tabular data format
- Database format allows for queries across experiments
- Example: producing scaling plots



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Approaching Complete Reproducibility

Configuration tracking

- Complete OS Image(s) used
 - Distribution version
 - Installed package versions
- ML Platform configuration/tuning parameters (Spark, Tensorflow, etc.)
- The ML application code command line parameters
- Dataset(s) used
- Caveats (future work)
 - Not tracking hardware firmware levels.
 - Application code not yet integrated with revision control system
 - Datasets assumed to be tracked independently