



2016 杭州·云栖大会
THE COMPUTING CONFERENCE



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基于 Apache* Spark* 的大规模 分布式机器学习实践

2016
The Computing Conference

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Content

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- Distributed ML on Spark
 - Fraud Detection: End-to-End Solution for Top Payments Company
 - Large-scale, Sparse Logistic Regression for Click-through and Purchase Rate Predictions
 - Deep (Convolutional) neural network
- Infrastructure support for distributed ML
 - Parameter server



20 [飞天·进化] APSARA EVOLUTION Project Overview

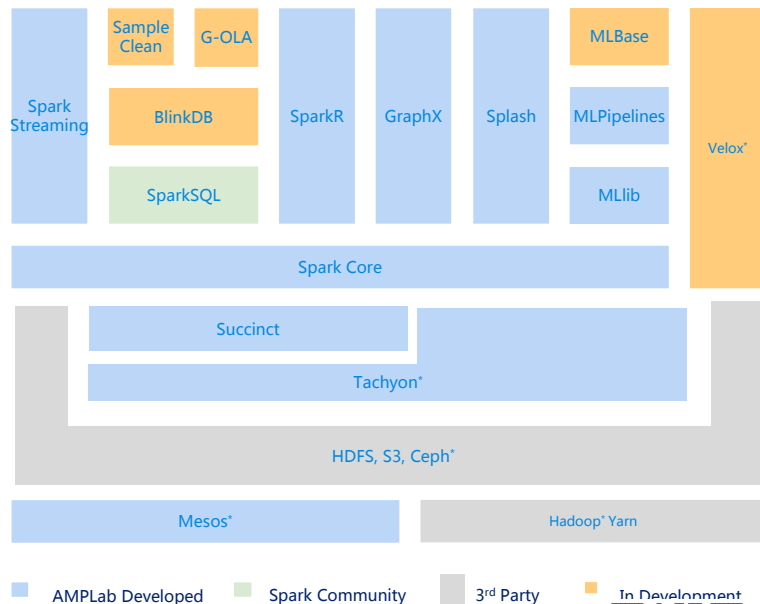


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- Research and open source project initiated by UC Berkeley AMPLab
- Intel is closely collaborating with AMPLab and the community on open source development
 - One of the earliest adopters of Spark* (since 2012)
 - Many key contributions (Netty shuffle, FairScheduler, “yarn-client” mode, ...)
 - Collaborating on other components in BDAS (e.g., Tachyon*, SparkR, ...)
- Intel is partnering with many “web-scale” companies
 - Free! No commercial solution or Consultations
 - Online-LDA, Word2Vec (Merged)
 - SparseML (Separated package)
 - E.g., Tencent, PayPal*, Alibaba*, Baidu*/iQiyi, JD.com, Youku*, etc.

BDAS: Berkeley Data Analytics Stack
(Ref: <https://amplab.cs.berkeley.edu/software/>)



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Large-Scale Distributed ML on Apache Spark

Distributed ML on Spark

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Fraud Detection on Apache Spark

Goal:

- Given transaction details, classify if it's fraud or normal

Evaluation Matrices

- Recall = predicted fraud / all real fraud transaction.
- Precision = predicted fraud correctly / predicted fraud

Fraud can mean:

Buying with stolen credit cards
Abusing promotional programs
Account takeover
Spamming other users



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Intel Customer Story

Problem statement and Pain points

- An old rule-based system that needs significant improvement
- Turn to Spark for data statistics and model training
- Need Neural Network for Fraud Detection on their Spark 1.4 cluster

Intel Solution

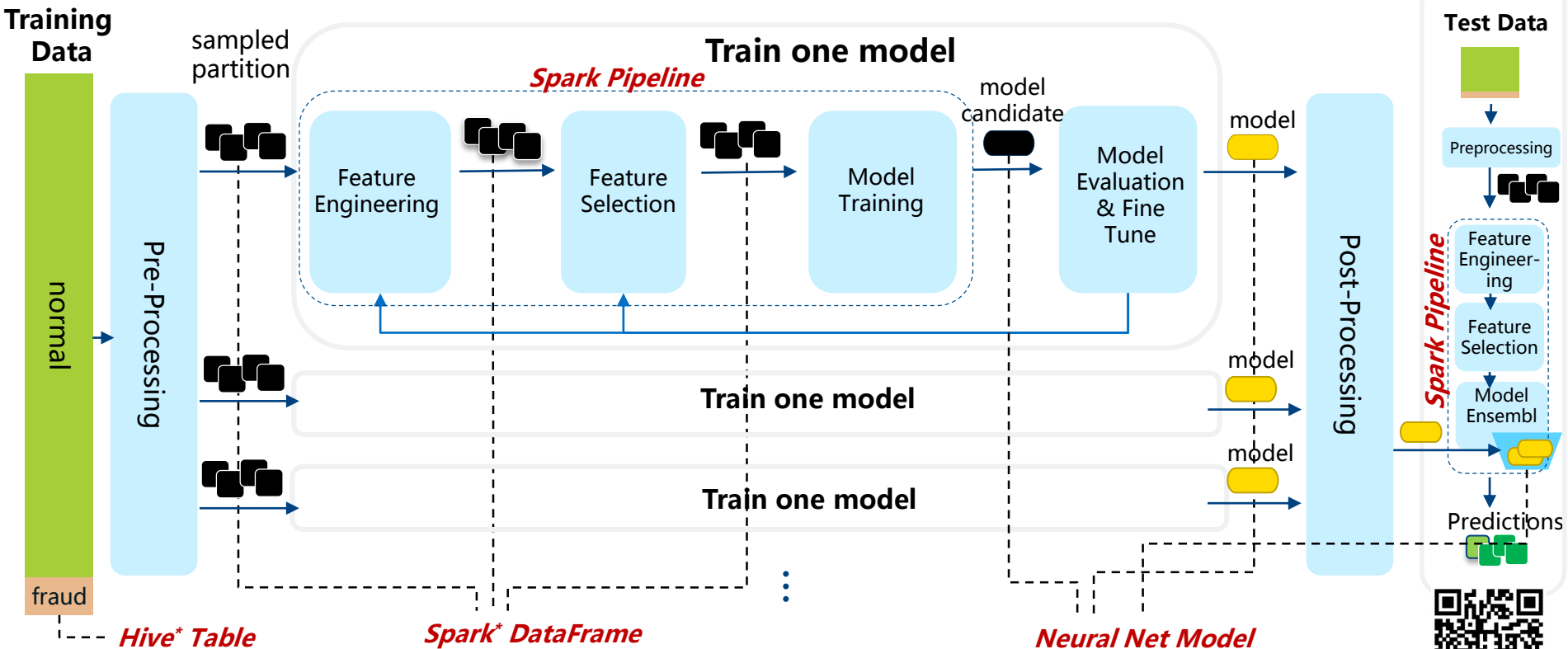
- Implement Neural Network on Spark and help integrate

Business Result

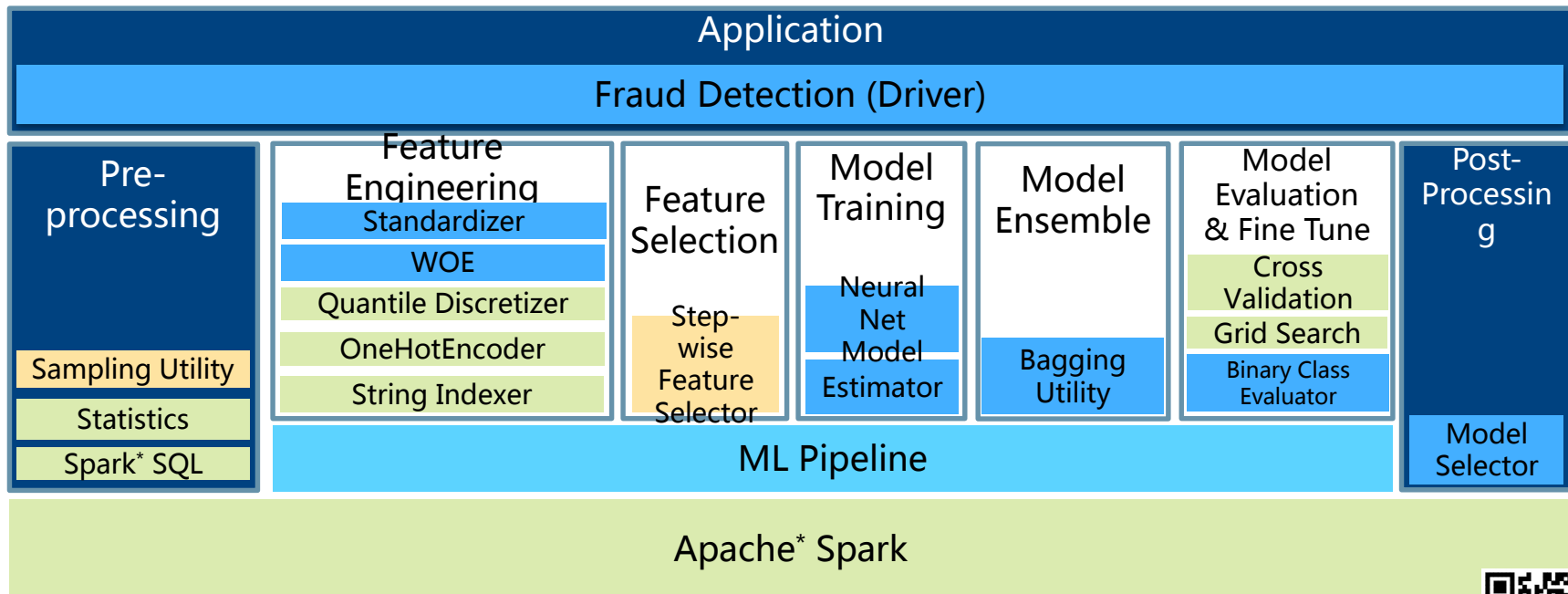
- Neural network model performs better than other algorithm
- Machine Learning system overtakes rule-based system and exceeds expectation
- Improve precision by 15%, improve recall by 30%



Solution Architecture Overview



Tool Stack Overview



Spark Community
 Intel Developed
 Intel Improved
 In Development



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Large-Scale Distributed ML on Apache Spark

Distributed ML on Spark

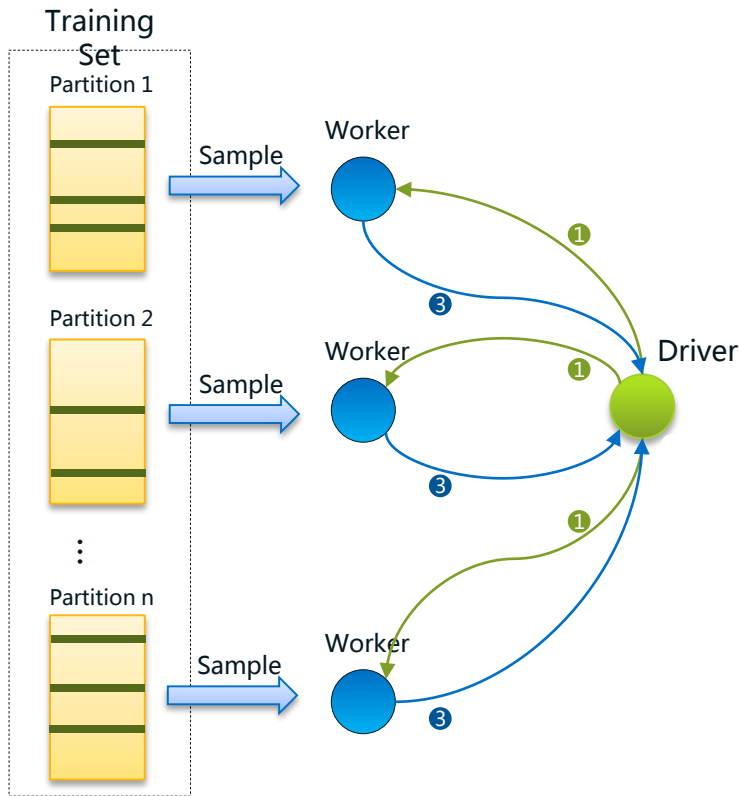
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Logistic Regression on Spark* with Mini-Batch SGD



“Canonical” implementation

Repeat {

① Driver broadcasts W to each worker

Workers compute gradient for the next batch of B records from the training set

Each task (running on workers) samples records from its data

partition

③ Each task computes local gradient

Aggregates gradient (possibly through tree aggregation)

Driver updates weight

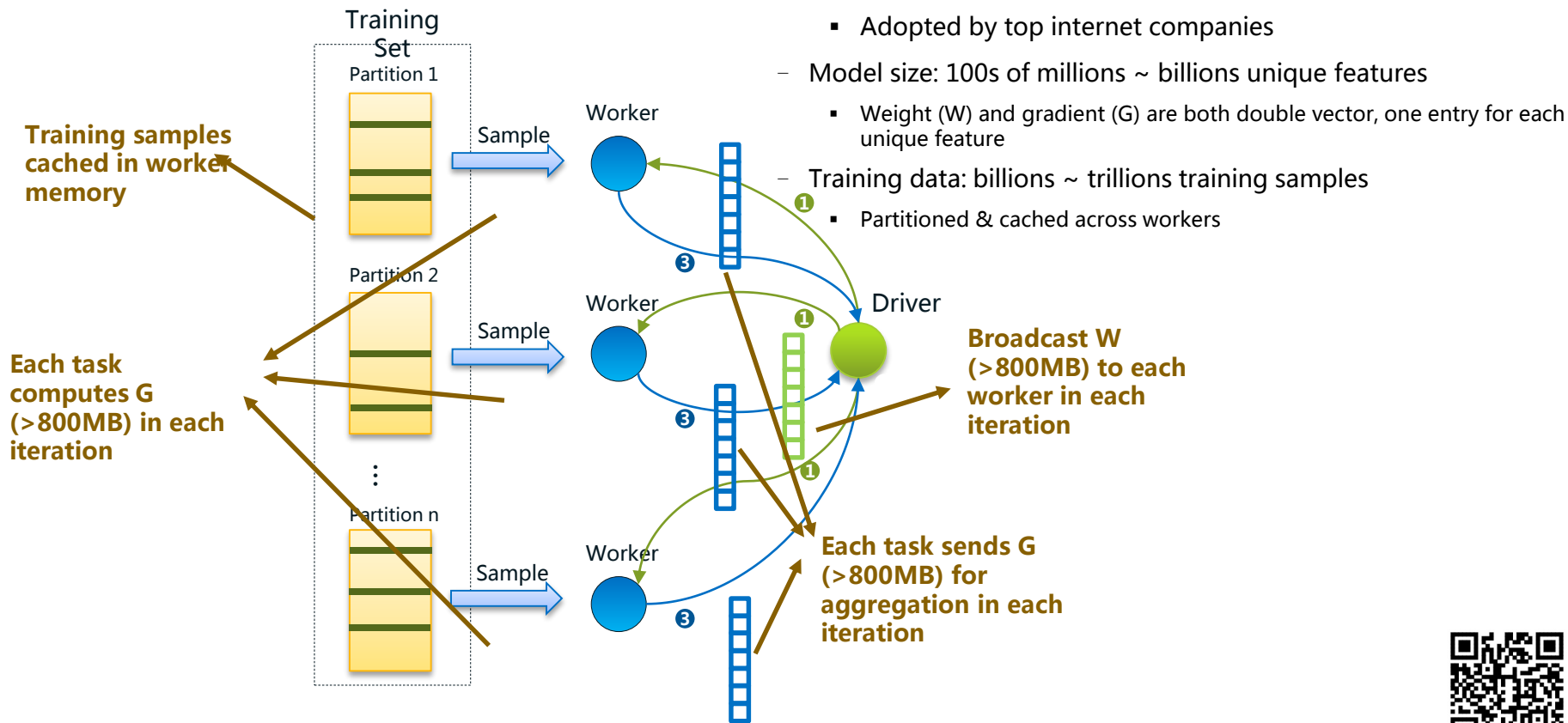
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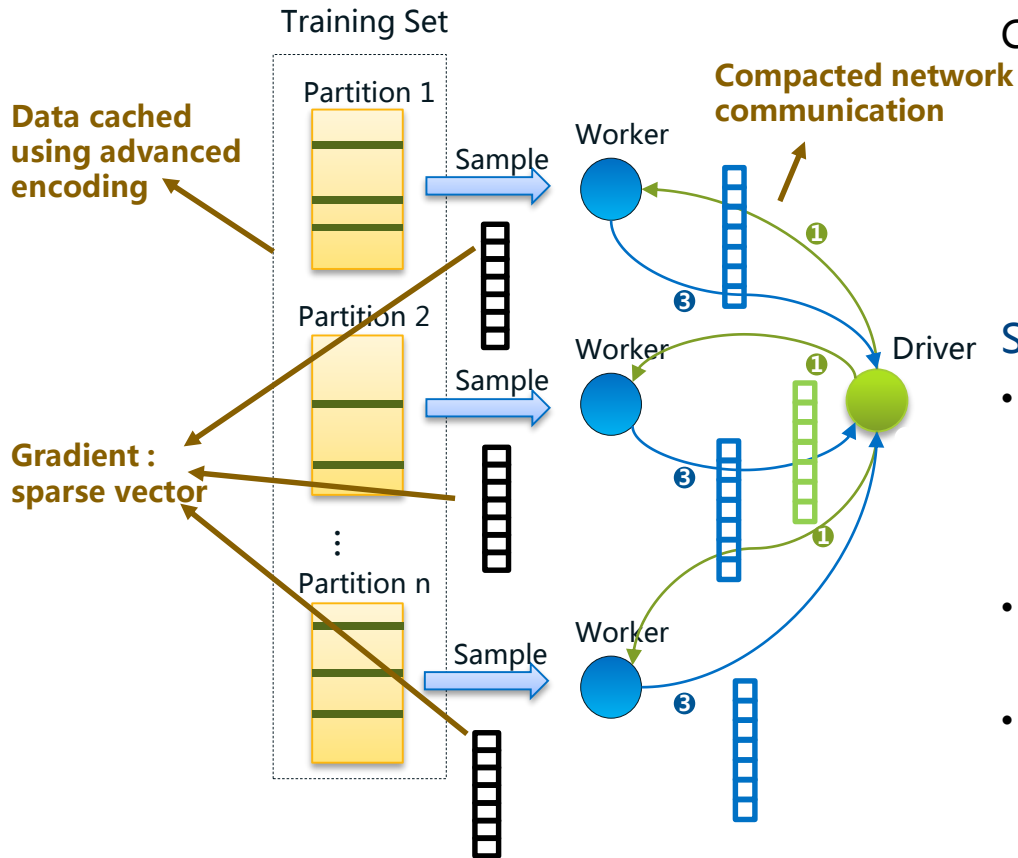


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Network and Memory Bottlenecks



Sparse Logistic Regression



Click-through and purchase rate predictions

- Adopted by top internet companies
- Model size: 100s of millions ~ billions unique features
- Training data: billions ~ trillions training samples

Solution

- Cached using sparse format
 - Using float16 (instead of double values)
 - Extra Support for binary (0 or 1) values
- Only Calc & sync gradient with non-zero data
- Better Communication



Large-Scale Distributed ML on Apache Spark

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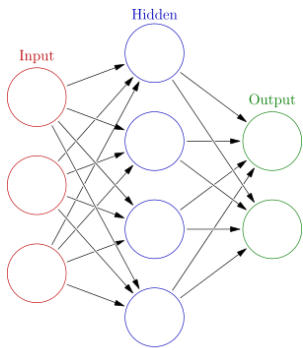
Distributed Neural Network

Multi-Layer Perceptron (MLP)

- Fully connected, feed-forward

Deep learning

- CNN, autoencoder, RBM, etc.

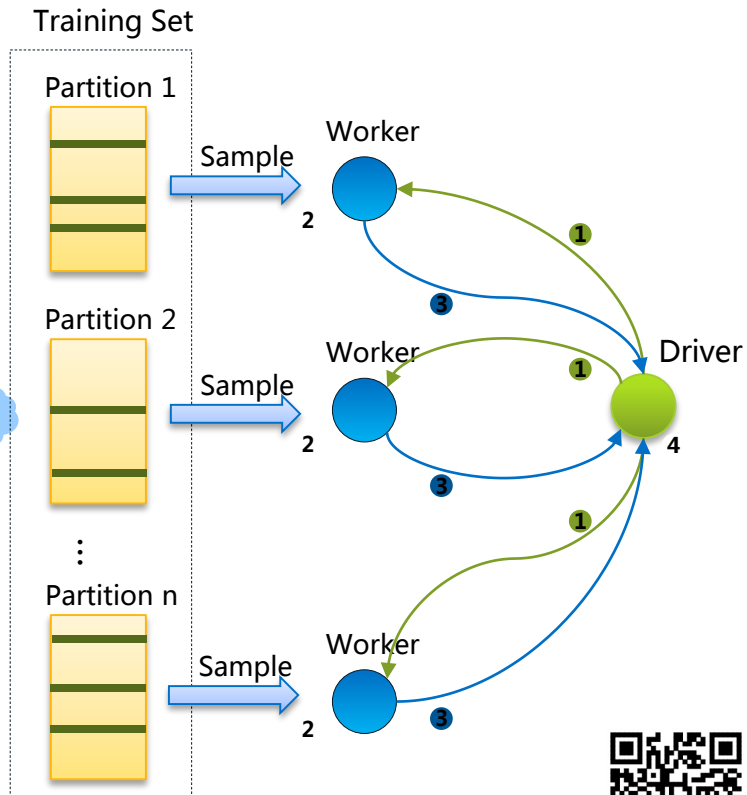


Training A Neural Network

Repeat {

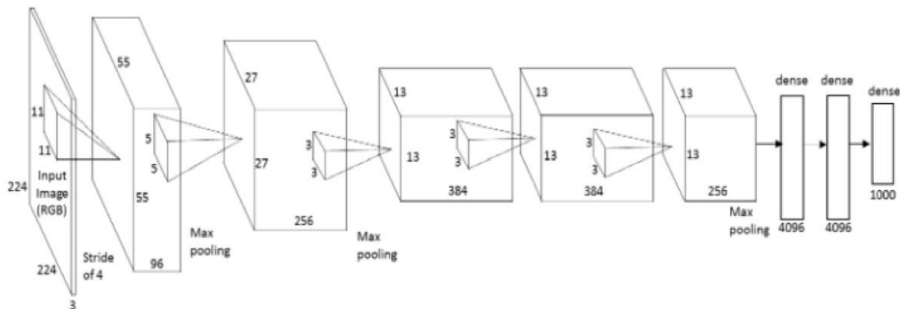
- 1 Driver broadcasts parameters (weights & biases) to each worker
Workers process the next batch of B records from the training set
- 2 Each task (running on workers) samples records from its data partition
- 3 Each task computes the *forward* and *backpropagation* pass
- 4 Driver aggregates gradient
Driver updates parameters (weights & biases)

}



Deep (Convolutional) Neural Network

Intuitive API with layer-based interface



```
val trainData = loadData()
val model = new Sequential(...)
model += new Convolution(...)
model += new maxPooling(...)
...
val criterion = new ClassNLLCriterion()
val optimizer = new ParallelOptimizer(model, new SGD)
optimizer.setCrossValidation(evaluator.accuracy)
optimizer.setPath("./model_save.obj")
optimizer.optimize(trainData)
```

Built on top of standard Big Data platforms

- Easily utilize your existing clusters

Engaging industry users and community early

- Evolving with feedback from real-world use cases
- Community version compatible with Spark* MLP

Targeting Full function coverage:

- Auto Encoder, Sparse Encoder
- Convolution with max and avg pooling
- RBM and DBN

Benchmark with popular dataset / models

- GoogleNet, AlexNet on ImageNet

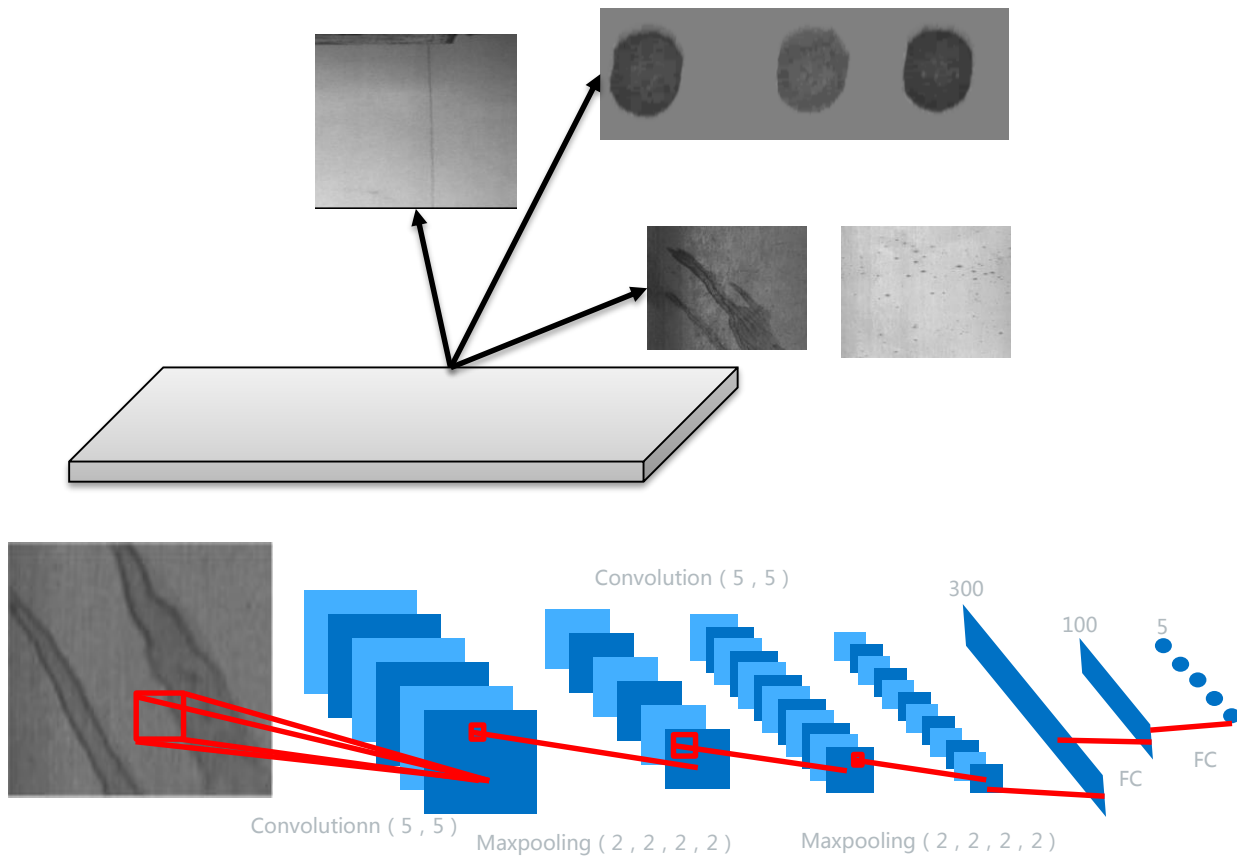
Easy MKL[†] integration for Intel® Architecture acceleration

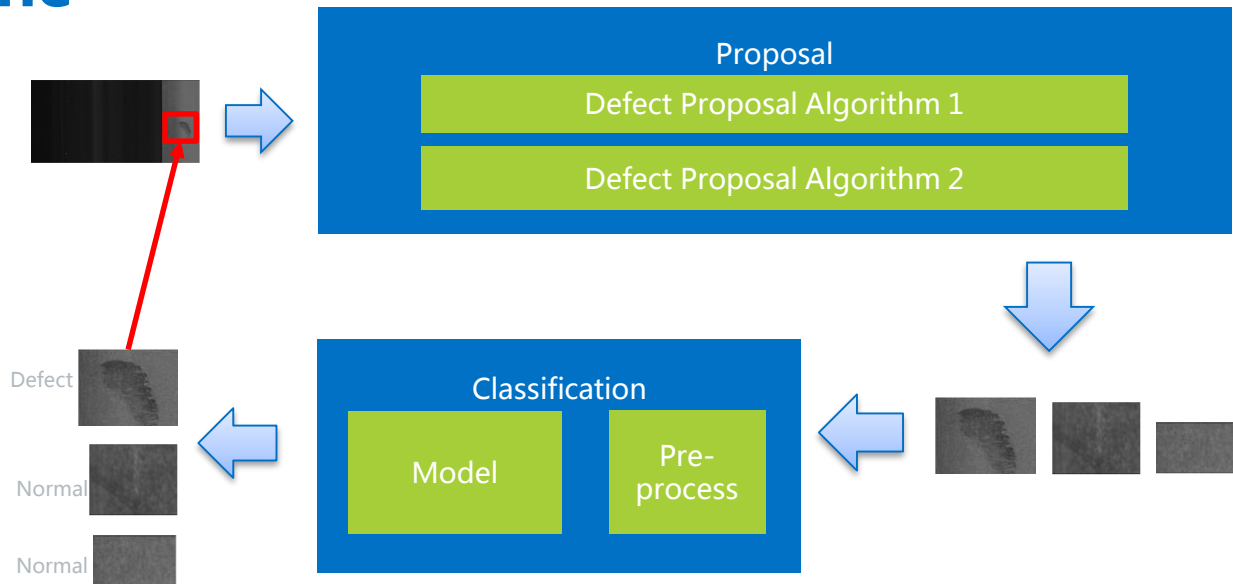
Better communication: All-to-one, All-reduce on spark(CaffeOnSpark), ParameterS

[†]Free community license (https://software.intel.com/en-us/articles/free_mkl)



Flaw detection in steel product





Large-Scale Distributed ML on Apache Spark

Distributed ML on Spark

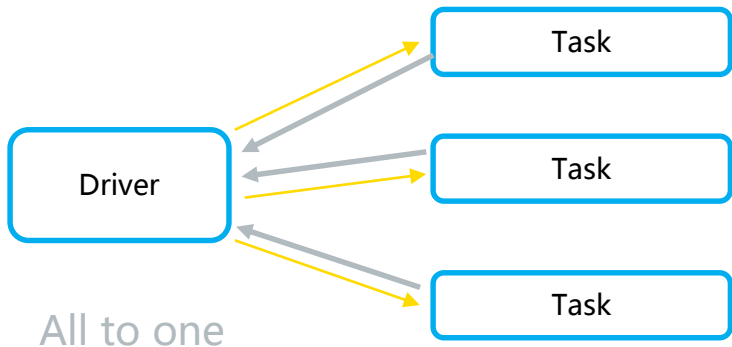
- Fraud Detection: End-to-End Solution for Top Payments Company
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Infrastructure support for distributed ML

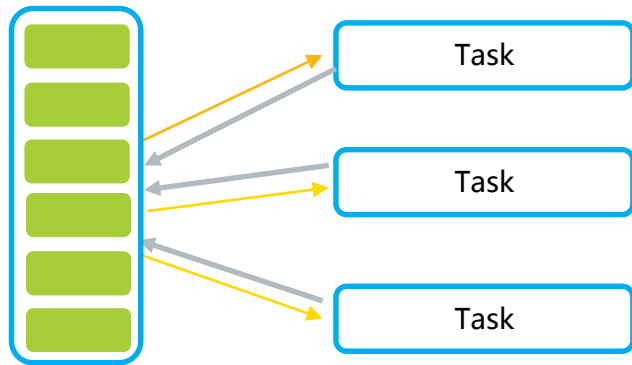
- Parameter server



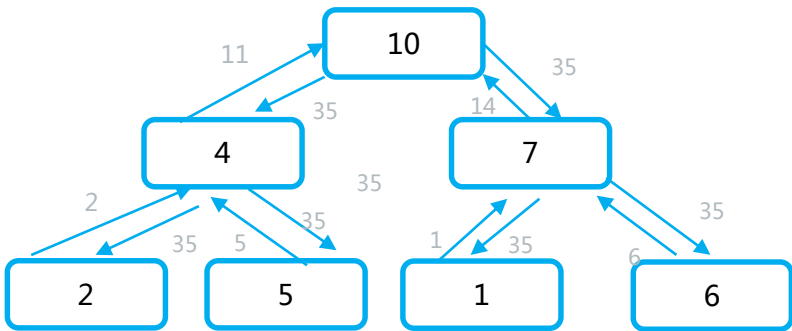
Communication Model



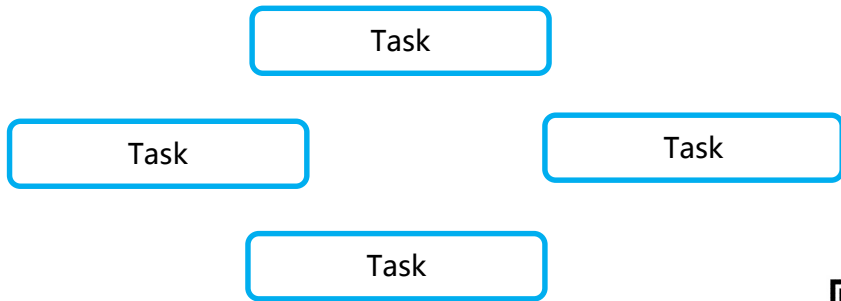
All to one



Parameter Server



All reduce (tree aggregation)

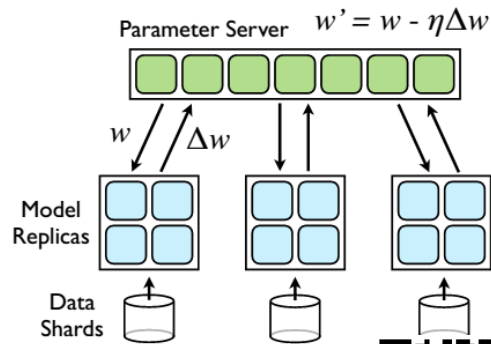


All reduce



“Parameter Server” support?

- Very large scale model/graph (billions of unique features)
- Leveraging further data sparsity in each worker (only a subset of weight vector needed)
- Possible weakly-synchronized model (BSP vs. SSP vs.ASP, etc.)
- Distributed parameter aggregation & update in Parameter Server
- Easily integration with Apache Spark^{*}.
- Fault Tolerance
- Co-partitioning



Source: Dean J, Corrado G, Monga R, et al. Large scale distributed deep networks[C]//Advances in neural information processing systems. 2012: 1223

^{*}Other names and brands may be claimed as the property of others.



Reference & Resources

Intel packages

- <https://github.com/intel-analytics/SparseML>
- <https://github.com/intel-analytics/FraudDetection>

Intel Analytics:

- <https://github.com/intel-analytics>

Contact

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