



In-Memory Performance
Durability of Disk





Scalable Machine and Deep Learning with Apache Ignite



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GridGain Director of Product Management

Agenda

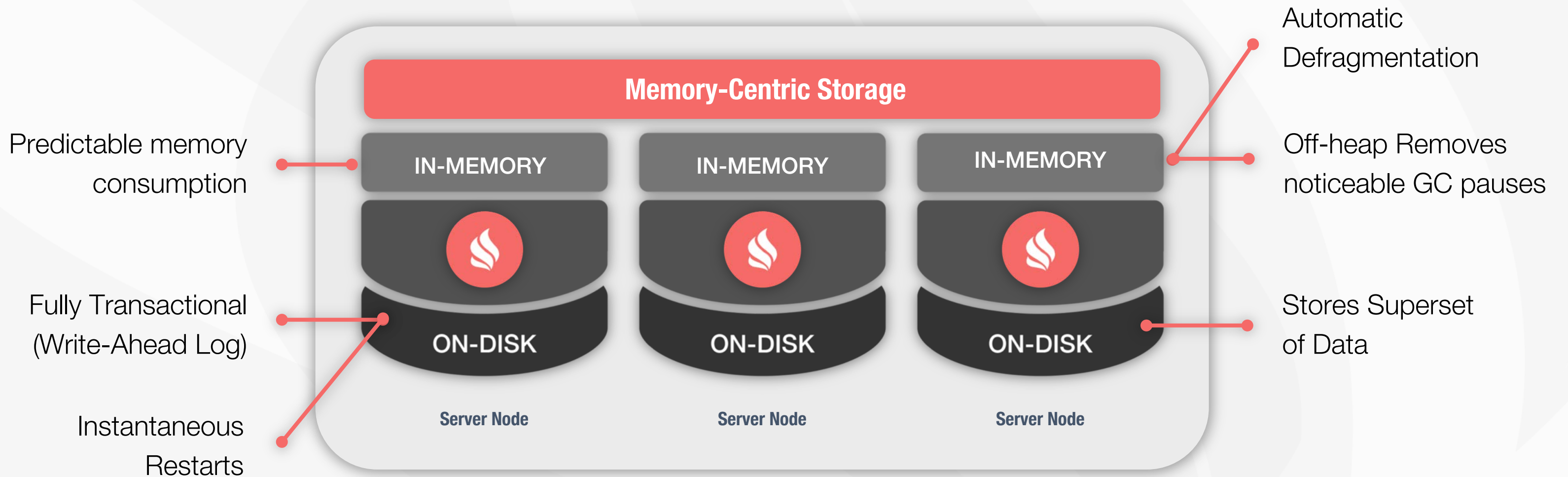
- Why Machine Learning at Scale?
- Ignite Machine Learning
- Genetic Algorithms
- TensorFlow Integration
- Demo
- Q&A

Why Machine Learning at Scale?

1. Models trained and deployed in different systems
 - Move data out for training
 - Wait for training to complete
 - Redeploy models in production
2. Scalability
 - Data exceed capacity of single server
 - Burden for developers



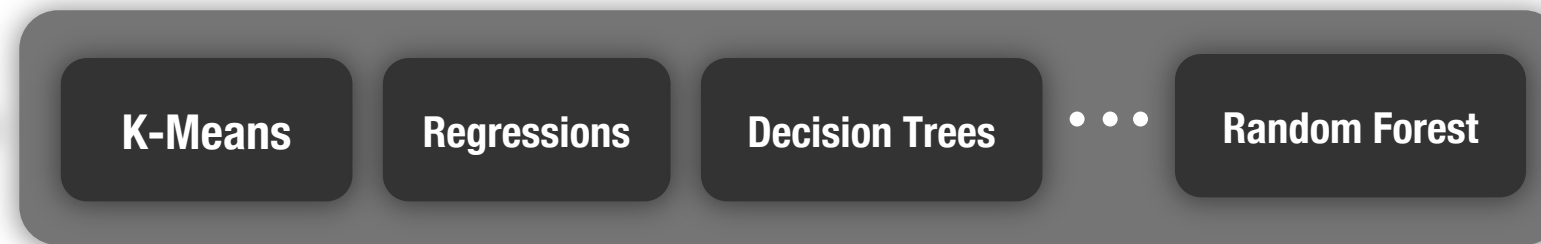
Memory-Centric Storage



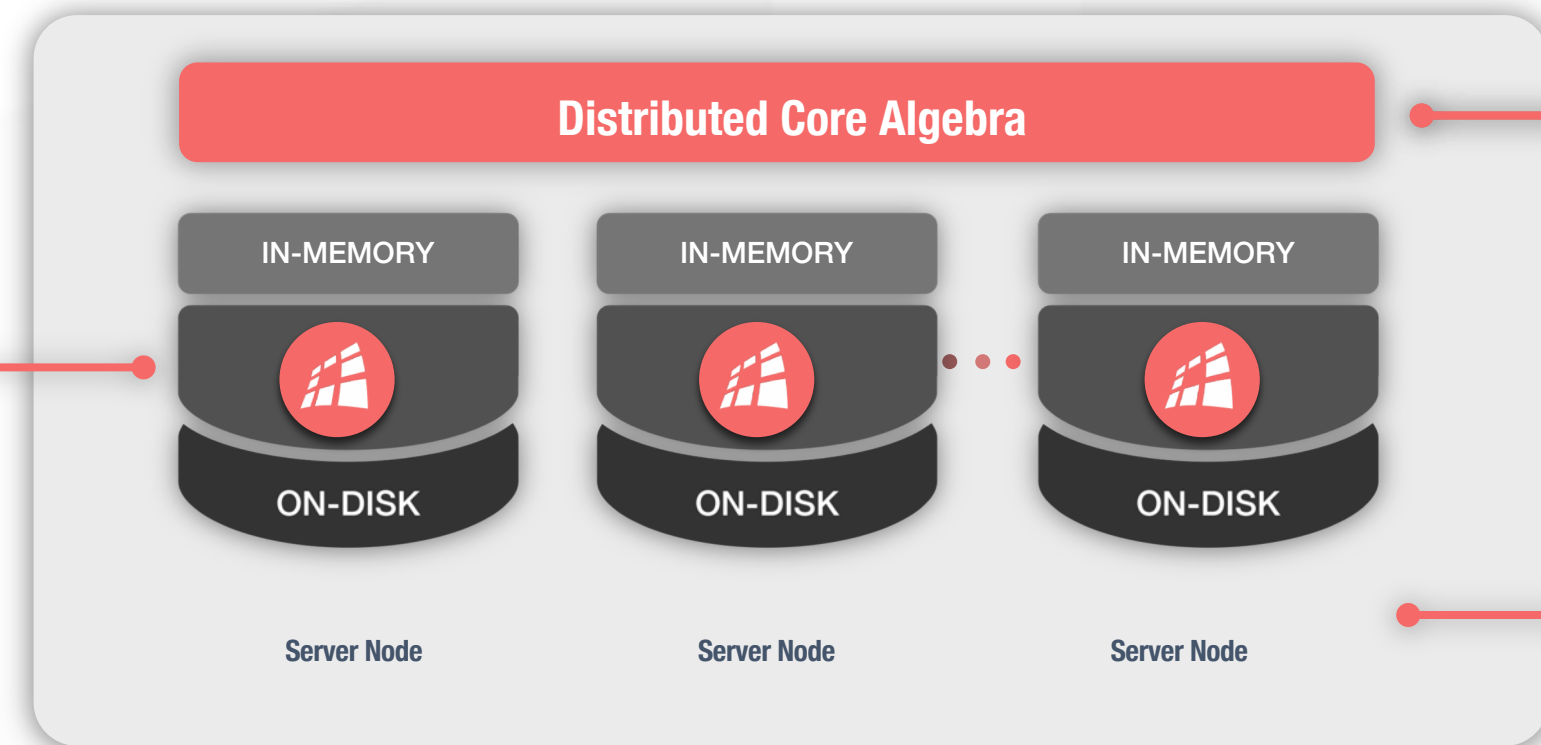
Machine Learning



Multi-Language Support



Distributed Algorithms



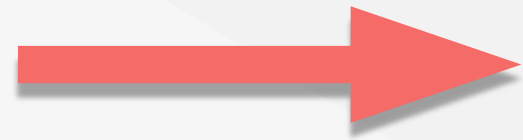
Large Scale Parallelization

Dense and Sparse Algebra

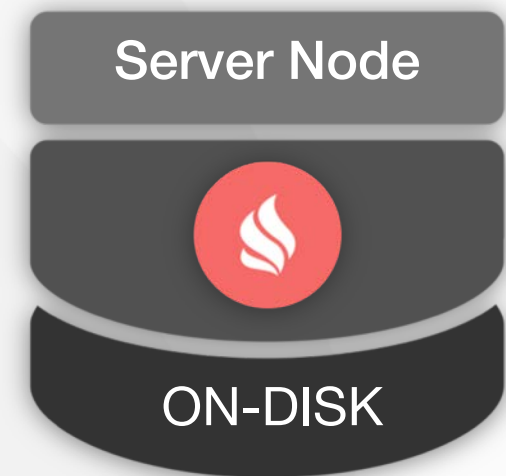
No ETL

Record to Node Mapping

Key

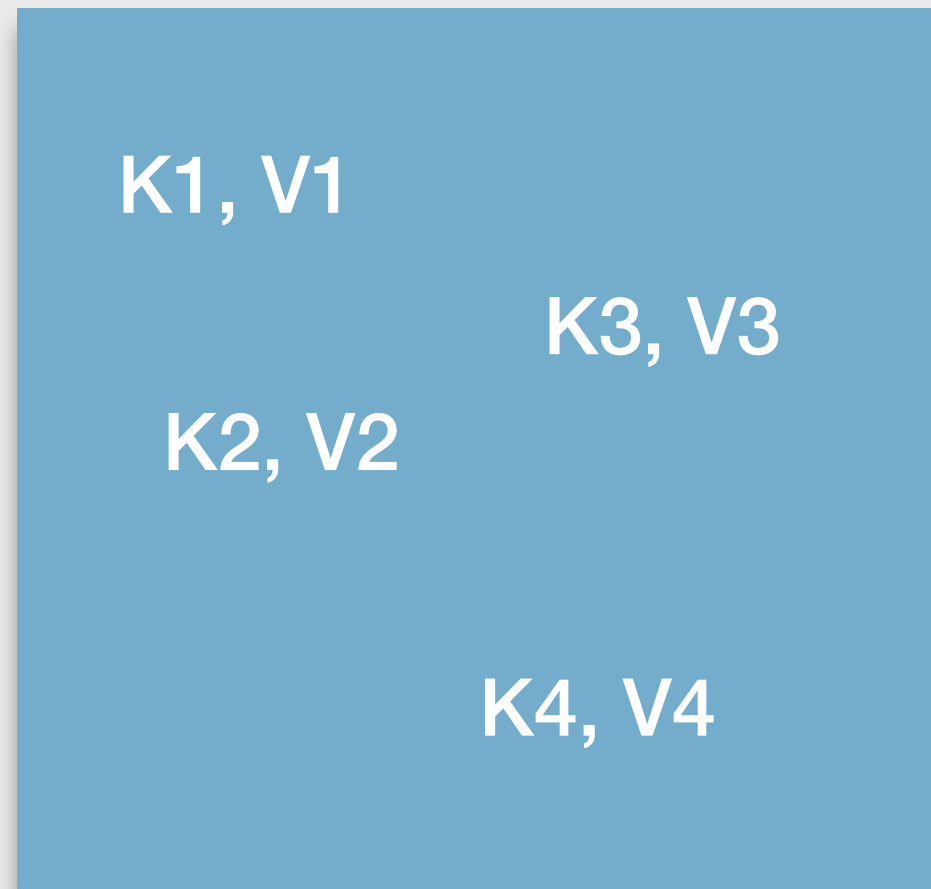


Partition



Caches and Partitions

Cache

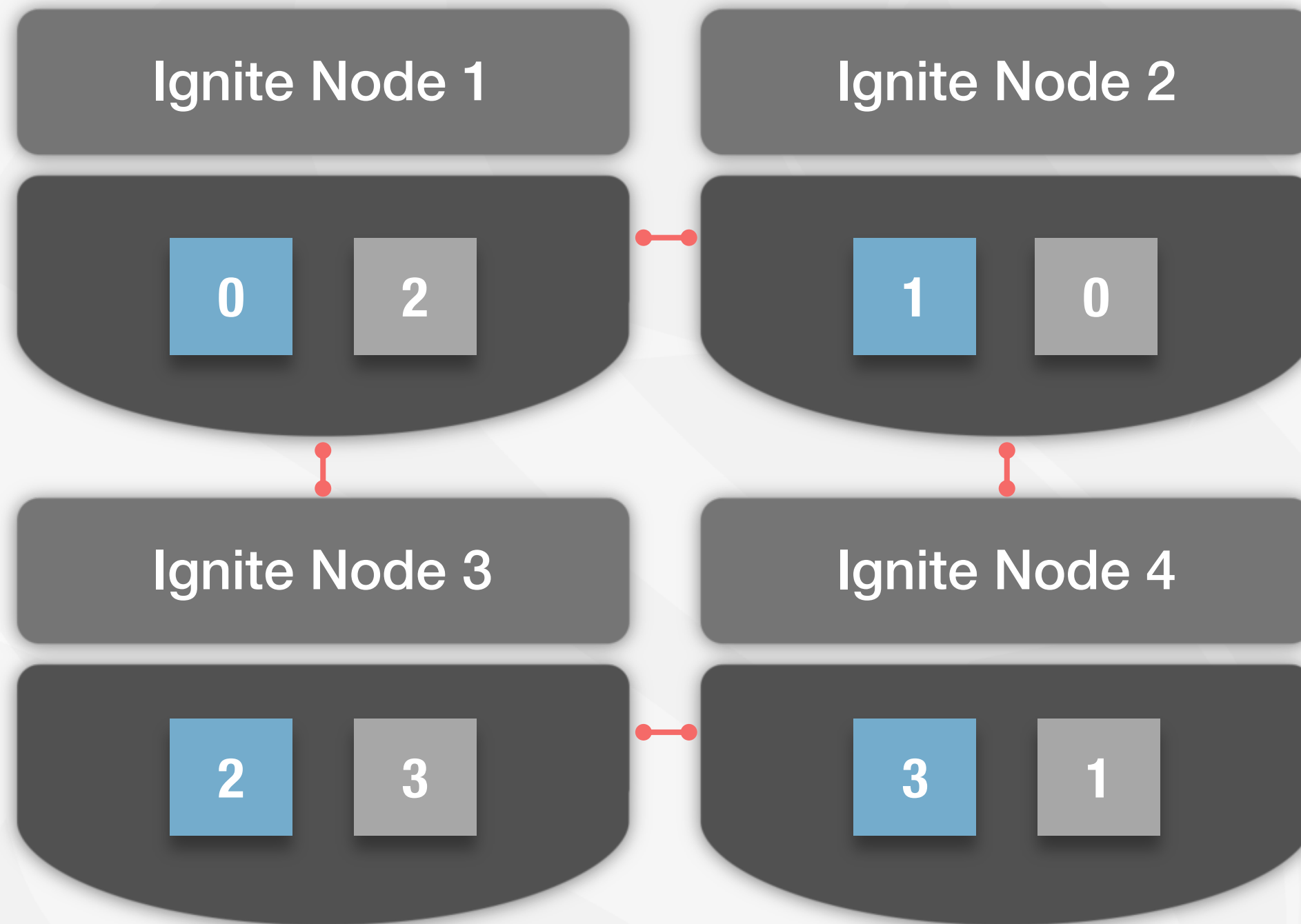


Partition 1



Partition 2

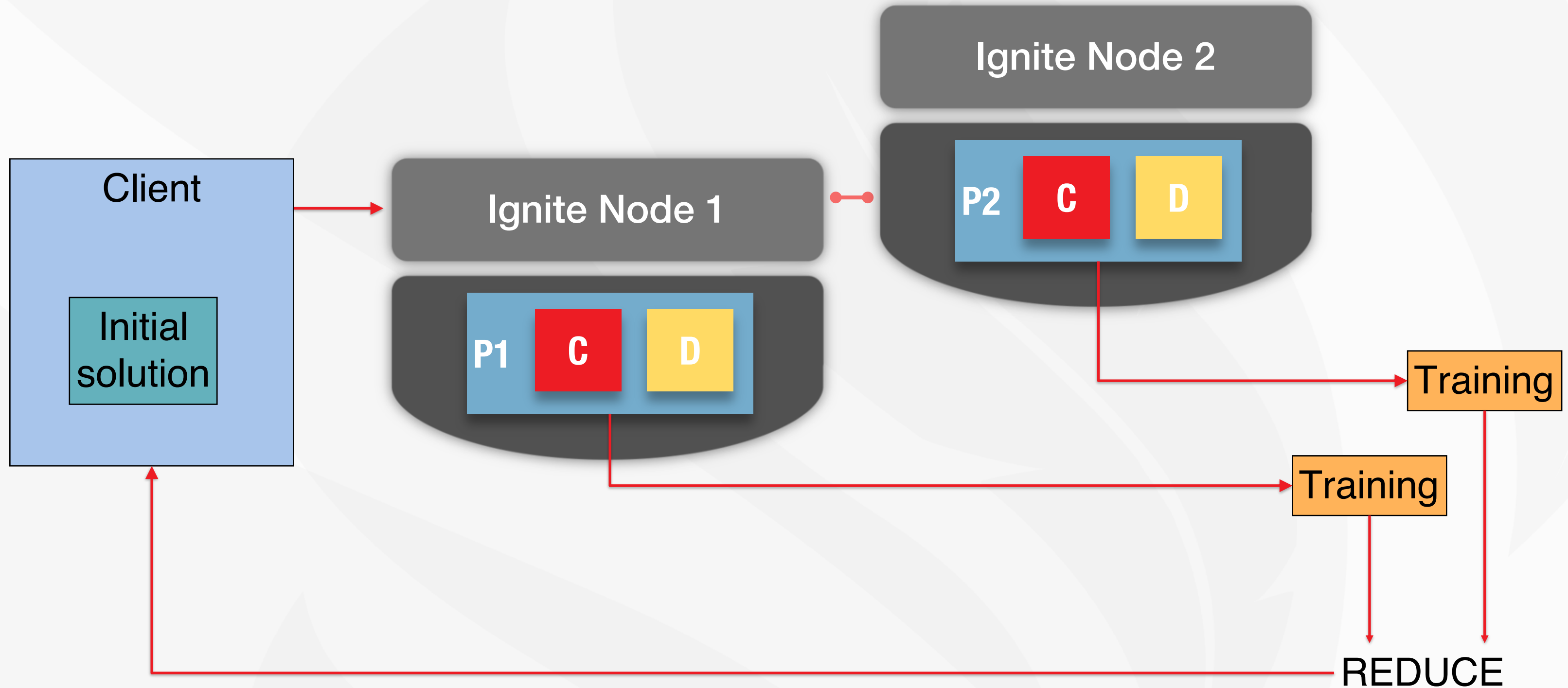
Partitions Distribution



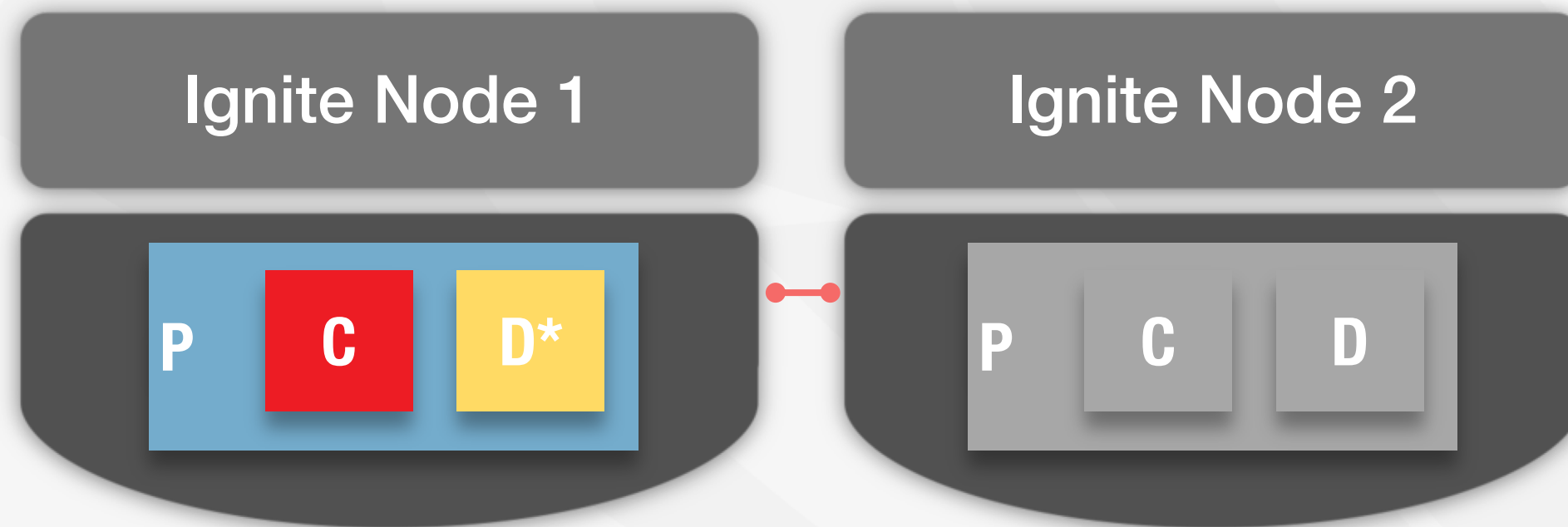
Primary

Backup

Partition-Based Dataset



Training Failover



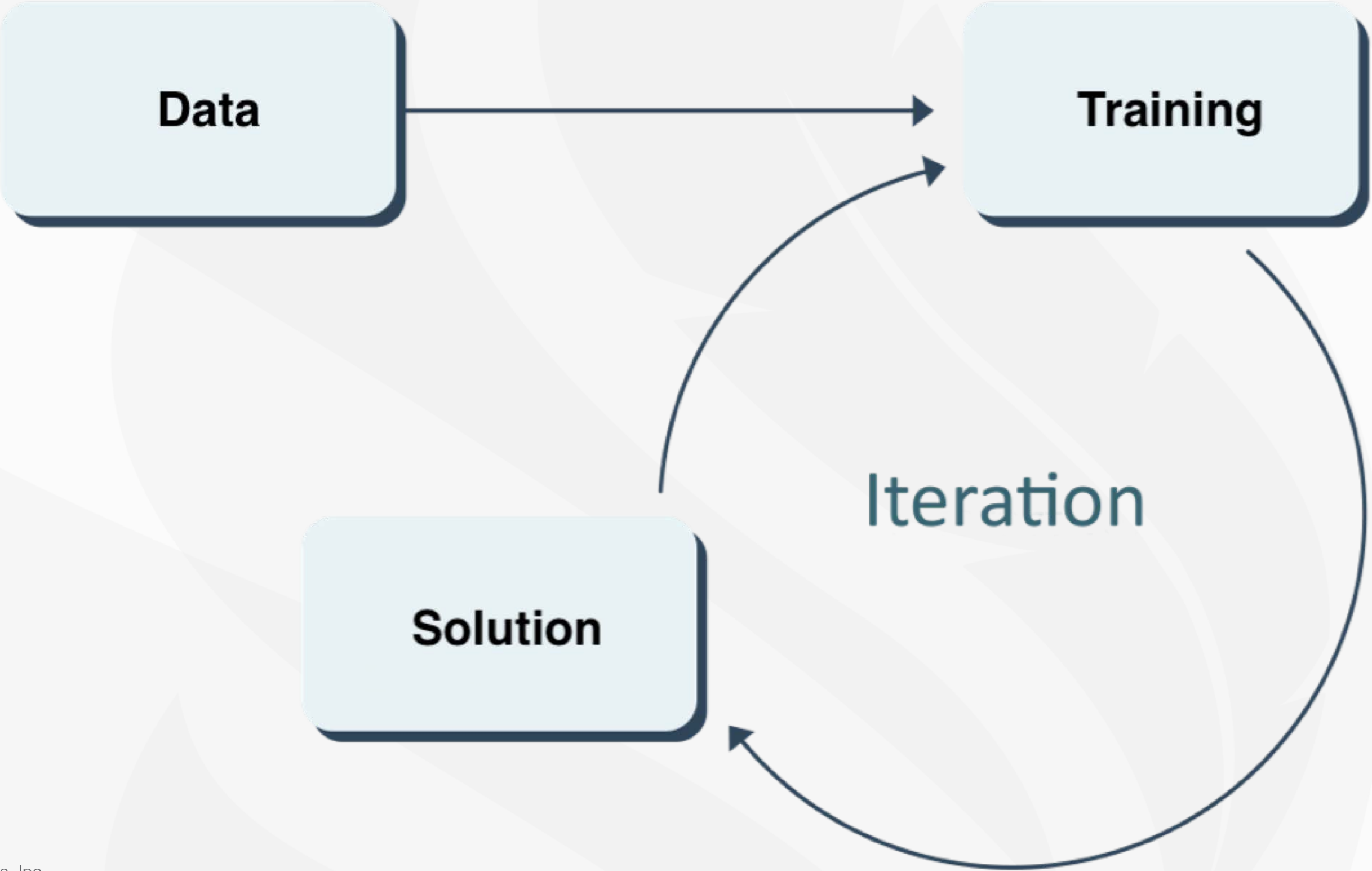
P = Partition

C = Partition Context

D = Partition Data

D* = Local ETL

Continuous Learning



Algorithms and Applicability

	Classification	Regression
Description	Identify to which category a new observation belongs, on the basis of a training set of data	Modeling the relationship between a scalar dependent variable y and one or more explanatory variables x
Applicability	spam detection, image recognition, credit scoring, disease identification	drug response, stock prices, supermarket revenue
Algorithms	nearest neighbor, decision tree classification, neural network	linear regression, decision tree regression, nearest neighbor, neural network

Algorithms and Applicability

	Clustering	Preprocessing
Description	Grouping a set of objects in such a way that objects in the same group are more similar to each other than to those in other groups	Feature extraction and normalization
Applicability	customer segmentation, grouping experiment outcomes, grouping shopping items	transform input data, such as text, for use with machine learning algorithms
Algorithms	k-means	Normalization preprocessor

Linear Regression

- Ordinary Least Squares
- Linear Regression Trainer
 - QR Decomposition
 - Gradient Descent

```
// y = bx + a
LinearRegressionModel model = trainer.train(trainSet);
double prediction = model.predict(sampleObject);

// Prepare trainSet
...

// QR Decomposition
LinearRegressionQRTrainer trainer = new LinearRegressionQRTrainer();
LinearRegressionModel mdl = trainer.train(trainSet);

// Gradient Descent
LinearRegressionSGDTrainer trainer = new LinearRegressionSGDTrainer(
    1000, 1e-6);
LinearRegressionModel mdl = trainer.train(trainSet);
```

Decision Trees

- Data stored by features
- Related data on same node
- Features
 - Continuous
 - Categorical

```
// Train the model
DecisionTreeModel mdl = trainer.train(
    new BiIndexedCacheColumnDecisionTreeTrainerInput(
        cache, new HashMap<>(), ptsCnt, featCnt));

// Estimate the model on the test set
IgniteTriFunction<Model<Vector, Double>,
    Stream<IgniteBiTuple<Vector, Double>>,
    Function<Double, Double>,
    Double> mse = Estimators.errorsPercentage();

Double accuracy = mse.apply(mdl, testMnistStream.map(
    v -> new IgniteBiTuple<>(v.viewPart(0, featCnt), v.getX(featCnt))),
    Function.identity());

System.out.println(">>> Errs percentage: " + accuracy);
```

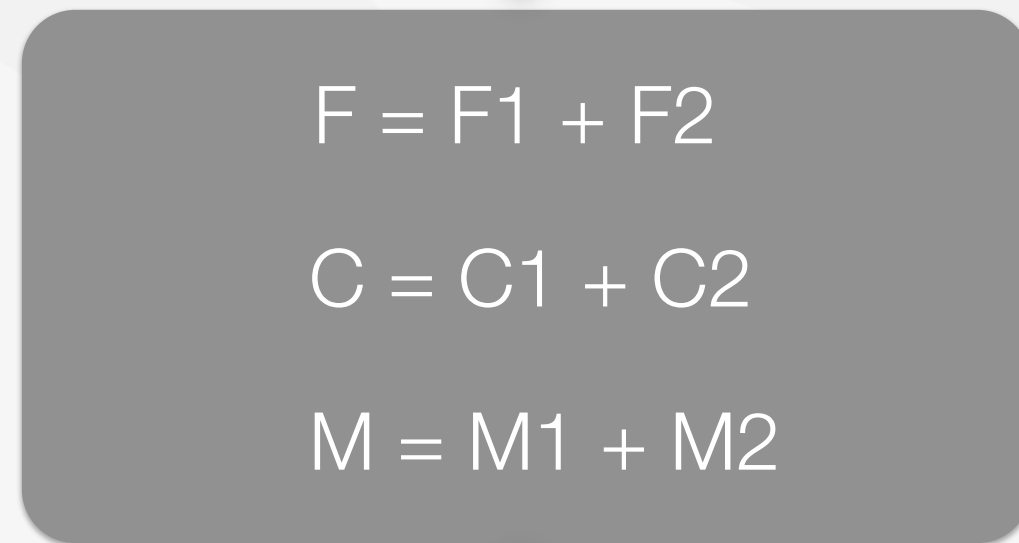



Demo: Fraud Detection



Genetic Algorithms

Biological Evolution
Simulation



Collocated Computation

F1, C1, M1

F2, C2, M2

Chromosome and Genes Cluster

IN-MEMORY

ON-DISK

IN-MEMORY

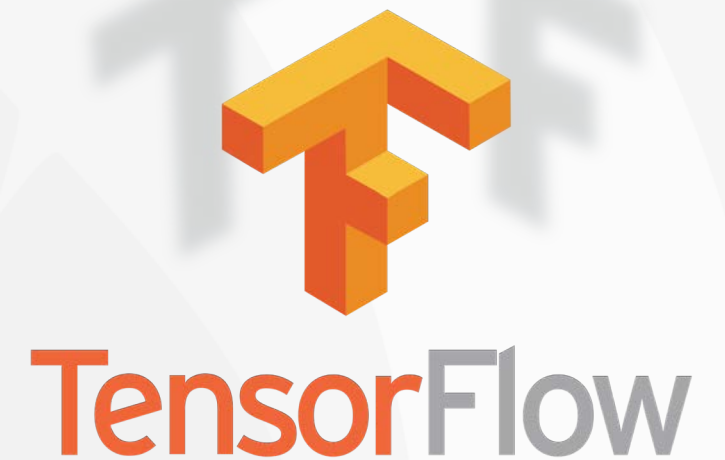
ON-DISK

Ignite Cluster

F = Fitness Calculation
C = Crossover
M = Mutation

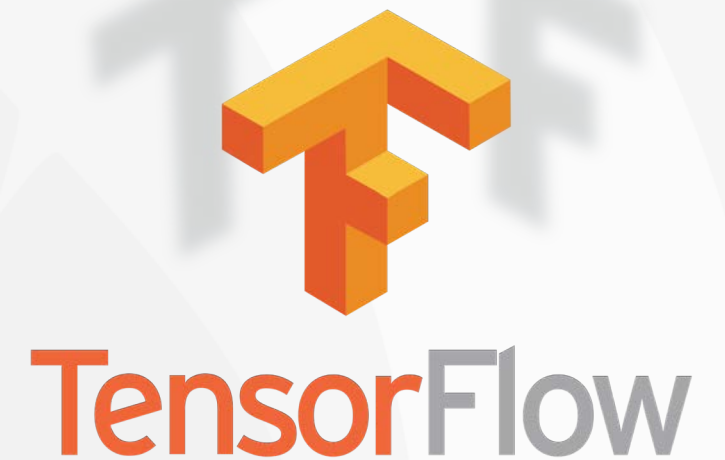
TensorFlow Integration: Benefits

- Ignite as distributed data source
 - Perfect fit for distributed TF training
- Less ETL
 - TF nodes deployed together with Ignite nodes
 - In-machine data movement only
- TF tasks execution in-place in Ignite
 - Roadmap



TensorFlow Integration: Main Features

- Distribution of user tasks written in Python
- Automatic creation and maintenance of TF cluster
- Minimization of ETL costs
- Fault tolerance for both Ignite and TF instances





Demo: TensorFlow and Ignite



Summary: Apache Ignite Benefits

- Massive scalability
 - Horizontal + Vertical
 - RAM + Disk
- Zero-ETL
 - Train models and run algorithms in place
- Fault tolerance and continuous learning
 - Partition-based dataset



Resources

- Apache Ignite ML Documentation:
 - <https://apacheignite.readme.io/docs>
- ML Blogging Series:
 - [Genetic Algorithms with Apache Ignite](#)
 - [Introduction to Machine Learning with Apache Ignite](#)
 - [Using Linear Regression with Apache Ignite](#)
 - [Using k-NN Classification with Apache Ignite](#)
 - [Using K-Means Clustering with Apache Ignite](#)
 - [Using Apache Ignite's Machine Learning for Fraud Detection at Scale](#)





Among Top 5 Apache Projects

Over 1M downloads per year

Top 5 Developer Mailing Lists

1. Ignite 
2. Kafka
3. Tomcat
4. Beam
5. James

Top 5 User Mailing Lists

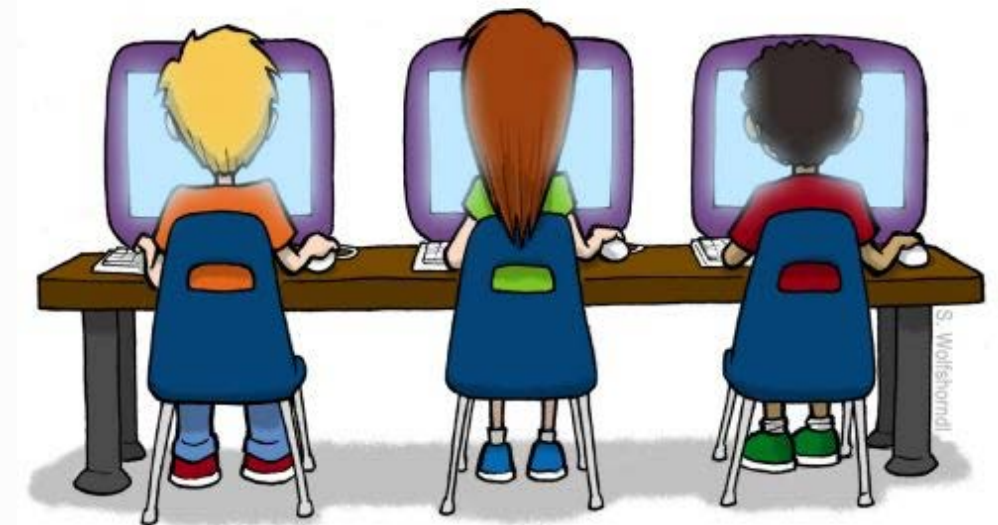
1. Lucene/Solr
2. Ignite 
3. Flink
4. Kafka
5. Cassandra

Top 5 by Commits

1. Hadoop
2. Ambari
3. Camel
4. Ignite 
5. Beam

Apache Ignite – We’re Hiring ;)

- Very Active Community
- Great Way to Learn Distributed Computing
- How To Contribute:
 - <https://ignite.apache.org/>





Any Questions?

Thank you for joining us. Follow the conversation.
<http://ignite.apache.org>



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