

Integrating high-performance machine learning: H2O and KNIME

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Speed - Accuracy

H2O: in-memory machine learning platform designed for speed on distributed systems

H₂O.0

H₂O.αi High Level Architecture



H₂O.ai Distributed Algorithms



Parallel Parse into Distributed Rows



Fine Grain Map Reduce Illustration: Scalable Distributed Histogram Calculation for GBM

Advantageous Foundation

- Foundation for In-Memory Distributed Algorithm Calculation - Distributed Data Frames and columnar compression
- All algorithms are distributed in H₂O: GBM, GLM, DRF, Deep Learning and more. Fine-grained map-reduce iterations.
- Only enterprise-grade, open-source distributed algorithms in the market

User Benefits

- "Out-of-box" functionalities for all algorithms (NO MORE SCRIPTING) and uniform interface across all languages: R, Python, Java
- Designed for all sizes of data sets, especially large data
- Highly optimized Java code for model exports
- In-house expertise for all algorithms



Foundation for Distributed Algorithms

H₂O.ai

Scientific Advisory Council



Dr. Trevor Hastie

- John A. Overdeck Professor of Mathematics, Stanford University
- PhD in Statistics, Stanford University
- · Co-author, The Elements of Statistical Learning: Prediction, Inference and Data Mining
- Co-author with John Chambers, Statistical Models in S
- Co-author, Generalized Additive Models

Dr. Robert Tibshirani

- Professor of Statistics and Health Research and Policy, Stanford University
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- Co-author, The Elements of Statistical Learning: Prediction, Inference and Data Mining
- Author, Regression Shrinkage and Selection via the Lasso
- Co-author, An Introduction to the Bootstrap

Dr. Steven Boyd

- Professor of Electrical Engineering and Computer Science, Stanford University
- PhD in Electrical Engineering and Computer Science, UC Berkeley
- Co-author, Distributed Optimization and Statistical Learning via the Alternating Direction Method of Multipliers
- Co-author, Linear Matrix Inequalities in System and Control Theory
- Co-author, Convex Optimization



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Machine Learning Benchmarks

(https://github.com/szilard/benchm-ml)



Gradient Boosting Machine Benchmark (also available for GLM and Random Forest)

H₂O.ai H2O Algorithms

Supervised Learning

Unsupervised Learning

Statistical Analysis	 Generalized Linear Models: Binomial, Gaussian, Gamma, Poisson and Tweedie Naïve Bayes 	Clustering	 K-means: Partitions observations into k clusters/groups of the same spatial size. Automatically detect optimal k
Ensembles	 Distributed Random Forest: Classification or regression models Gradient Boosting Machine: Produces an ensemble of decision trees with increasing refined approximations 	Dimensionality Reduction	 Principal Component Analysis: Linearly transforms correlated variables to independent components Generalized Low Rank Models: extend the idea of PCA to handle arbitrary data consisting of numerical, Boolean, categorical, and missing data
Deep Neural Networks	• Deep learning : Create multi-layer feed forward neural networks starting with an input layer followed by multiple layers of nonlinear transformations	Anomaly Detection	 Autoencoders: Find outliers using a nonlinear dimensionality reduction using deep learning
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H2O in KNIME

Live Demo



H2O in KNIME

- Offer our users high-performance machine learning algorithms from H2O in KNIME
- Allow to mix & match with other KNIME functionality
 - Data wrangling KNIME Analytics Platform functionality
 - KNIME Big-Data Connectors
 - Text Mining, Image Processing, Cheminformatics, ...
 - and more!



Live Demo



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H2O in KNIME – Cross Validation





H2O in KNIME – Cross Validation

H2O Generalized Linear Model Learner (Classification)



Accuracy statistics. - 2:18 - H2O Binomial Scorer (Score)

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	Statistics#2	0.658	0.497	0.233	0.045	0.483	1,964,234	0.421	0.579	0.99	0.682	
H2O Lo	Statistics#3	0.658	0.497	0.233	0.045	0.483	1,964,604	0.421	0.579	0.99	0.682	
	Statistics#4	0.658	0.496	0.233	0.045	0.483	1,964,437	0.421	0.579	0.989	0.682	
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H2O in KNIME – Parameter Optimization





H2O in KNIME – Parameter Optimization

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	Row0_Stat	Generalized Linear Modeling	0.661081398937558	0.4983155422677105					
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H2O in KNIME – Nodes in KNIME 3.4









H2O in KNIME – What's cooking?





Thank you!



