Machine Learning with H2O for R Users

Professor Matthew Lanham Academic Director, MS BAIM Program



Tech Talk Disclaimer

- I am not affiliated with H2O, nor receive any compensation or benefits for giving this talk.
- I'm a professor that teaches Data Mining, Predictive Analytics, and R courses at Purdue University.
- I wanted to share one lecture on the topic that I teach to my M.S. in Business Analytics & Information Management (BAIM) students that introduces the platform.
- If H2O would like to provide me competition or benefits for this talk or future talks, I'm happy to have that discussion ☺

Materials

You can get the slides and codes for this presentation here:

https://github.com/MatthewALanham/InformsBA2019

Agenda

- About H2O Platform
- h2o R Library/API demo
- AutoML Automated Machine Learning demo
- H2O Flow demo





- https://www.h2o.ai/
- H2O was founded in 2012
- Provide scalable architecture + distributed machine learning algorithms to tackle small and big data problems
- They claim to be focusing on automated AI and making AI more accessible to everyone
- Have 5000+ customers and there are several interesting customer stories on their website (https://www.h2o.ai/customer-stories/)

H20 Pros

- Open source (Apache 2.0 licensed)
- Well-documented and commercially supported
 - Bookmark: http://docs.h2o.ai/h2o/latest-stable/h2o-docs/welcome.html
- Easy to use (technical to sort-of-technical person)
- Scalable to big data
- Has mature architecture
- OS: Windows, Mac, Ubuntu, RHEL/CentOS
- APIs R, Python, Scala, or a Web GUI (no programming)
- Automated Model Building functionality via AutoML
- Can train and tune a deep learning model in one line of code

H2O Platform

High performance learning

- Distributed (multi-core + multi-node) implementations of ML algorithms
- Core algorithms written in high performance Java

Use favorite language, environment, and easily deploy into action

- APIs available that make it easy to work with big data from you laptop using your favorite analytics language
- Meant to work anywhere Your laptop, Hadoop, Spark, EC2, etc.
- Easily deploy models to production as pure Java code OR if you can just save your models to disk as R/Python

H2O Distributed Computing

H2O Cluster

- Multi-node cluster with shared memory model
- All computations are in memory
- Each node only sees some rows of the data
- No limit on cluster size

Distributed Key Value Store

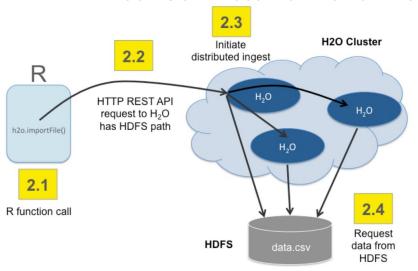
- Objects in the H2O cluster such as data frames, models, and results are all referenced by key
- Any node in the cluster can access any object in the cluster by key

H2O Frame

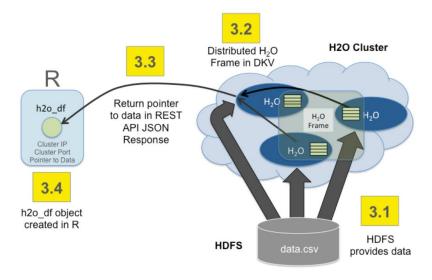
- Distributed data frames (collection of vectors)
- Columns are distributed (across nodes) arrays
- Works just like R's data.frame or Python Pandas DataFrame

H2O Distributed Computing

The R client tells the cluster to read the data



The data is returned from HDFS into a distributed H2O Frame



H2O Algorithms + Common Workflow Tasks

Supervised

Unsupervised

Common

Miscellaneous

- Cox Proportional Hazards (CoxPH)
 Aggregator
- Deep Learning (Neural Networks)
- Distributed Random Forest (DRF)
- Generalized Linear Model (GLM)
- Naïve Baves Classifier
- Stacked Ensembles
- XGBoost

- Generalized Low Rank Models (GLRM)
 Early Stopping
- Isolation Forest
- K-Means Clustering
- Gradient Boosting Machine (GBM)
 Principal Component Analysis (PCA)

Quantiles

- Word2vec

Generic Models

Generic Models

Common workflow tasks:

- Imputation, normalization, auto one-hot encoding
- Cross-validation, grid or random search
- Variable importance, model evaluation metrics, plots

Source: http://docs.h2o.ai/h2o/latest-stable/h2o-docs/data-science.html#

H2O Installation

You obtain h2o for R just like any other R package:

install.packages("h2o")

- The library is really just an API
- All the data is stored on the cluster (the server), not on our client. Even when the client and cluster are the same machine.

Thus, when we want to train a model or make predictions, we first have to get the data into the H2O cluster.

 You will need java, but most likely you already have that on your machine. If not, go to: http://www.oracle.com/technetwork/java/javase/downloads/index.html

H2O Demo in RStudio



Demo

See *h2o.R* script

AutoML



AutoML

AutoML is the path of least resistance for finding a competitive predictive model.

Data Preparation

- Imputation
- One-hot encoding
- Standardization
- Label/Target encoding
- Feature selection

Model Generation

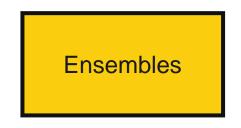
- Logistic regression?
- Tree?
- Neural network?
- Hyperparameter tuning
- Generalizability/early stopping

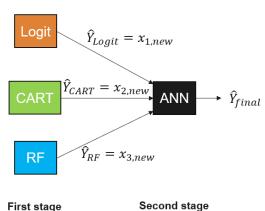
Ensembles

- Focus on predictive performance?
- Stacking/metamodeling?
- Ensemble selection

AutoML

- The goal of AutoML is to achieve competitive prediction.
- Ensembling
 - Combining multiple learners together.
 - Many different ways to do this.
 - Can perform well if you have decent base learners and the models have uncorrelated errors
- Stacking tries to find the optimal combination of base learners via a meta-model.
- AutoML uses Random Grid Search (GBMs, GLMs, etc.) with Stacked Ensembles
- Provides a Leaderboard output of the top models.
 Demo...





H2O Flow



H2O Flow

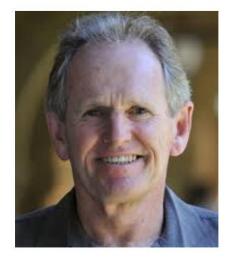
- Flow is the name of the web GUI that is part of H2O. It is just another client, making the same web service calls to the H2O backend that the R client is making.
- You can do the following:
 - View data you have uploaded through your client
 - Upload data directly
 - View models you have created through your R client (and those currently being created)
 - Create models directly
 - View and run predictions you have generated through your client
- If using your RStudio Desktop, open a browser and go here to see your flow
 - http://127.0.0.1:54321/flow/index.html
- If using RStudio Server, the link will be slightly modified (example):
 - http://rstudio.scholar.rcac.purdue.edu:54321/flow/index.html

Future investigations

- Testing performance on large scale datasets
- Importing non-native h2o models into the workflow
- Integrating models into applications outside of H2O

H2O.ai provides impressively scalable implementations of many of the important machine learning tools in a user-friendly environment. Allowing for free academic use sets a generous example for commercial software developers — it is also the way forward in the era of open-source software.

Trevor J. Hastie
John A. Overdeck Professor of Mathematical Sciences
Professor of Statistics
Professor of Biomedical Data Science
Department of Statistics
Stanford University
USA



Additional Resources/Examples

- https://github.com/DarrenCook/h2o
- https://github.com/h2oai/h2omeetups/tree/master/2018_09_05_SF_Meetup_AutoML