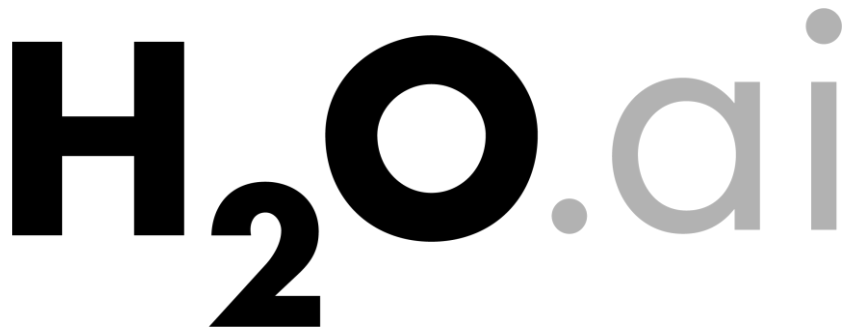
Machine Learning with H_2O



Jo-fai (Joe) Chow

Data Scientist

joe@h2o.ai

@matlabulous

R Addicts Paris Meetup

1st December, 2016

Agenda

- Introduction to H_2O
  - About H_2O.ai
- Our Open Source Products
  - Overview
  - H_2O Platform
  - Steam
  - Live Demo
    - H2O + R + Web + Steam

H_2O.ai

About Me

- Civil (Water) Engineer
  - 2010 – 2015
  - Consultant (UK)
    - Utilities
    - Asset Management
    - Constrained Optimization
  - Industrial PhD (UK)
    - Infrastructure Design Optimization
    - Machine Learning + Water Engineering
    - Discovered H_2O in 2014
- Data Scientist
  - From 2015
  - Virgin Media (UK)
  - Domino Data Lab (Silicon Valley)
  - H_2O.ai (Silicon Valley)

H_2O.ai

I ♥ R

Jo-fai Chow

woobe

Civil Engineer turned Data Scientist

H2O.ai
United Kingdom
jofai.chow@gmail.com
http://www.jofaichow.co.uk/
Joined on Aug 10, 2012

Organizations

Overview | Repositories 41 | Stars 372 | Followers 119 | Following 30

Popular repositories

Customize your pinned repositories

blenditbayes
Code used in my blog "Blend it like a Bayesian!"
R ★ 73 81

deepr
An R package to streamline the training, fine-tuning and predicting processes for deep learning based on 'darch' and 'deepnet'.
R ★ 40 14

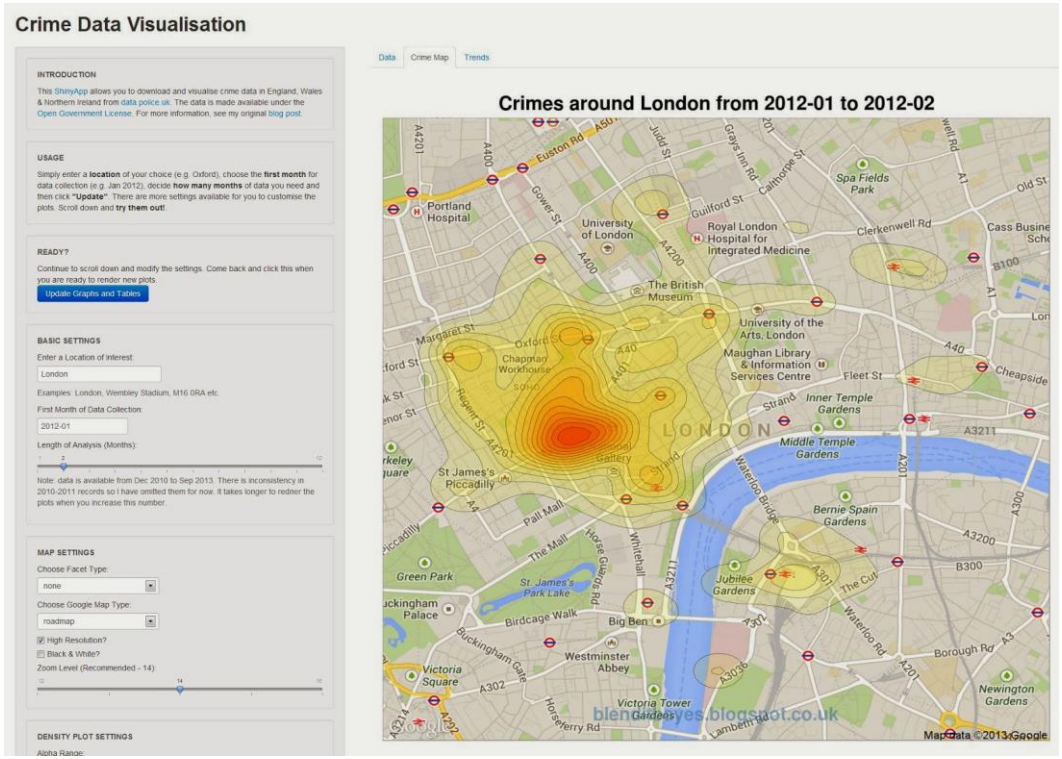
rPlotter
Wrapper functions that make plotting in R a lot easier for beginners.
R ★ 29 4

rCrimemap
This is the next generation of CrimeMap!
R ★ 22 9

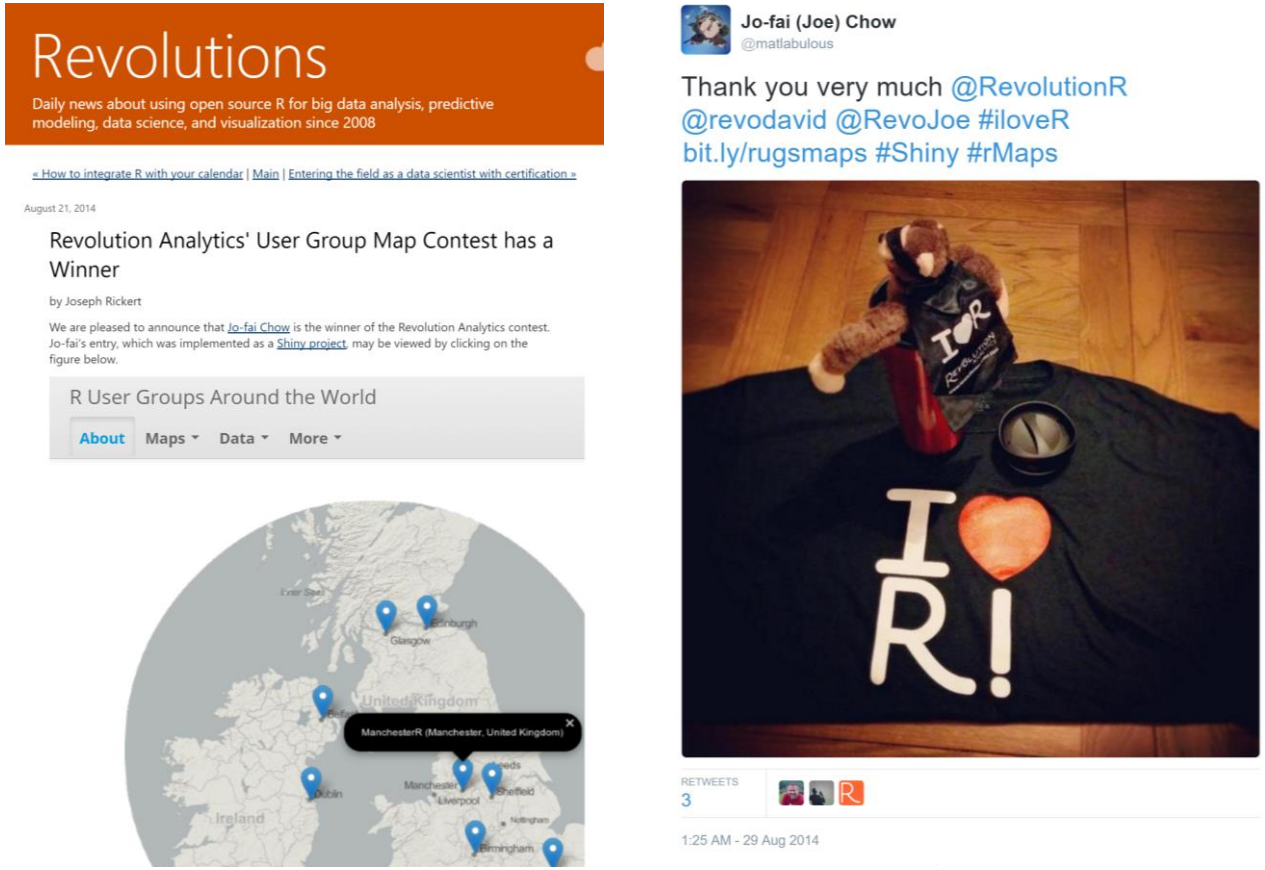
rugsmaps
This app is my submission to the visualization contest held by Revolution Analytics.
R ★ 19 18

rApps
Repository for my R (Shiny) web applications.
R ★ 16 36

I ♥ R

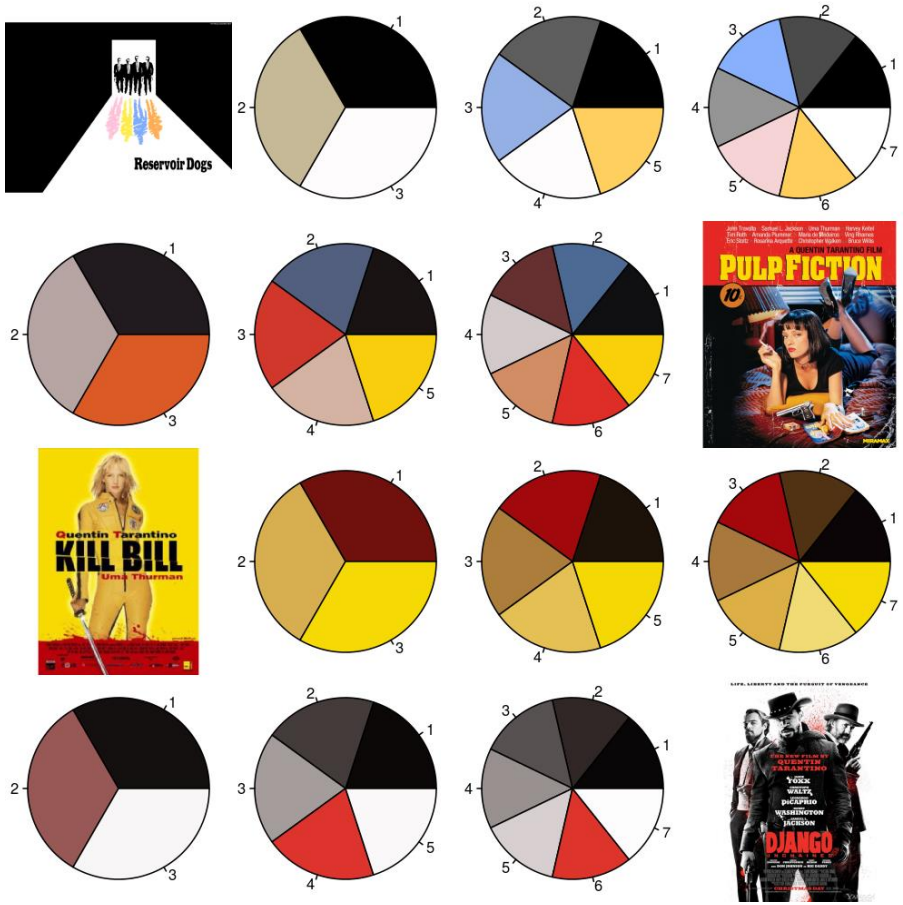


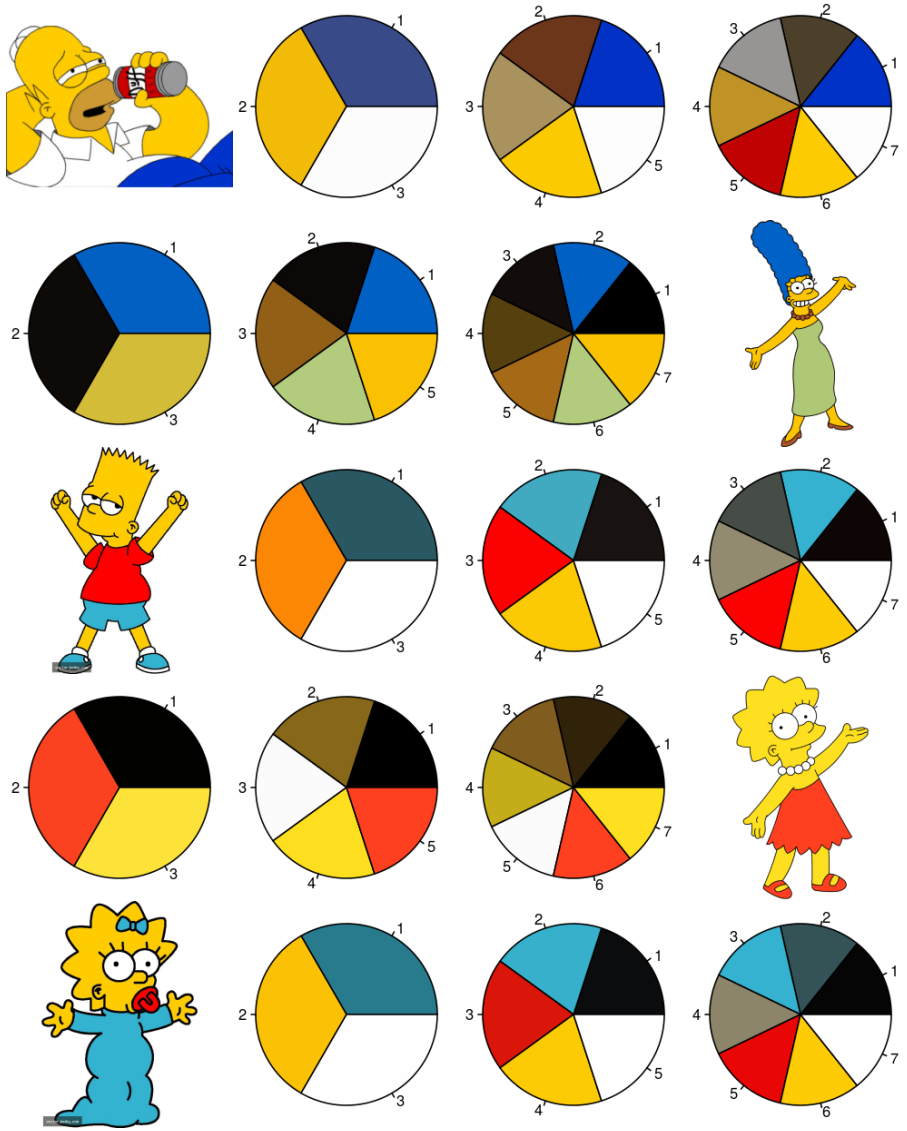
My First Data Viz & Shiny App Experience
CrimeMap (2013)



Revolution Analytics' Data Viz Contest
RUGSMAPS (2014)

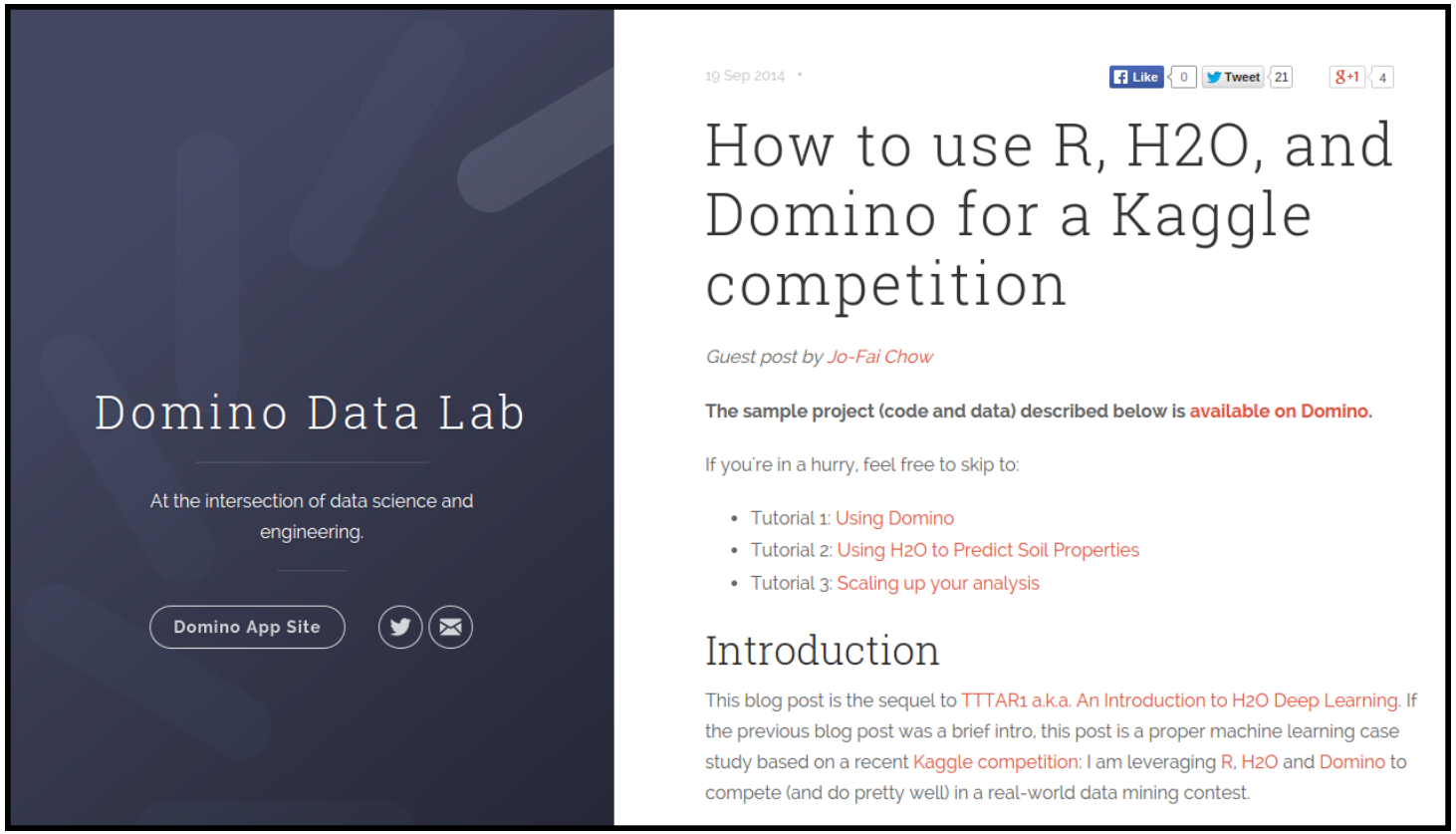
I ♥ R





Developing R Packages for Fun
rPlotter (2014)

I ♥ R



R + H_2O + Domino for Kaggle

Guest Blog Post for Domino & H_2O (2014)

About H_2O.ai

What exactly is H_2O?

Company Overview

Founded	2011 Venture-backed, debuted in 2012
Products	• H_2O Open Source In-Memory AI Prediction Engine • Sparkling Water • Steam
Mission	Operationalize Data Science, and provide a platform for users to build beautiful data products
Team	70 employees • Distributed Systems Engineers doing Machine Learning • World-class visualization designers
Headquarters	Mountain View, CA




H2O.ai

Scientific Advisory Council

Dr. Trevor Hastie

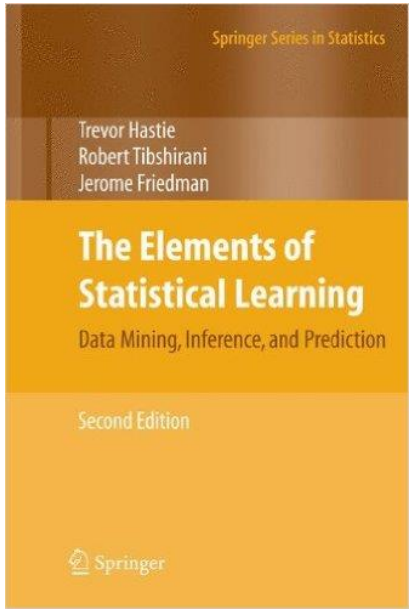
- John A. Overdeck Professor of Mathematics, Stanford University
- PhD in Statistics, Stanford University
- Co-author, *The Elements of Statistical Learning: Prediction, Inference and Data Mining*
- Co-author with John Chambers, *Statistical Models in S*
- Co-author, *Generalized Additive Models*

Dr. Robert Tibshirani

- Professor of Statistics and Health Research and Policy, Stanford University
- PhD in Statistics, Stanford University
- 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*



H_2O is an open source platform empowering business transformation

Bring AI To Business Empower Transformation

Financial Services, Insurance and Healthcare as Our Vertical Focus

Community as Our Foundation

Users In Various Verticals Adore H_2O

H2O.ai

H2O In Action

www.h2o.ai/customers

Capital One

Capital One uses H2O open source machine learning for various use cases.

MarketShare

H2O predictive analytics helps boost the impact and results of digital marketing.

Kaiser

Kaiser uses H2O machine learning to save lives.

Zurich Insurance

Zurich turned to H2O as a strategic differentiator for commercial insurance.

Progressive

Progressive uses H2O predictive analytics for user-based insurance.

Comcast

Comcast uses H2O to improve customer experience.

Hospital Corporation of America

HCA uses H2O to predict patient outcomes in real-time.

McKesson

McKesson discusses the adoption of artificial intelligence in healthcare.

Macy's

Macy's uses H2O for personalized site recommendations.

Transamerica

Transamerica turns to H2O to develop a product recommendation platform for insurance.

Paypal

Paypal turned to H2O Deep Learning for fraud detection and customer churn.

eBay

eBay chose H2O for open source machine learning.

H2O.ai

H_2O.ai Makes A Difference as an AI Platform

Open Source

- 100% open source

Big Data Ecosystem

hadoop

Spark

Flexible Interface

R

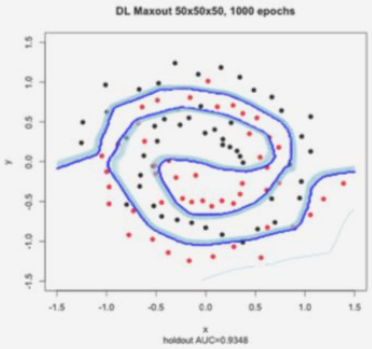
Scala

Java

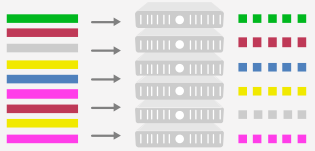
python

H_2O Flow

Smart and Fast Algorithms



Scalability and Performance



- Distributed In-Memory Computing Platform
- Distributed Algorithms
- Fine-Grain MapReduce

Rapid Model Deployment

- Highly portable models deployed in Java (POJO) and Model Object Optimized (MOJO)
- Automated and streamlined scoring service deployment with Rest API

REST API

GPU Enablement

GPU

Cloud Integration

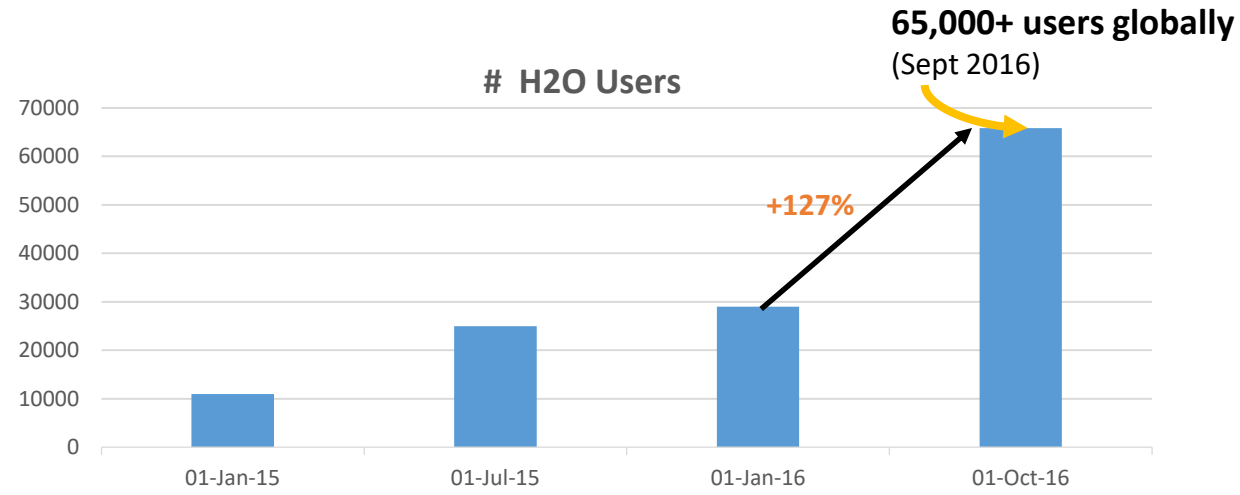
amazon web services™

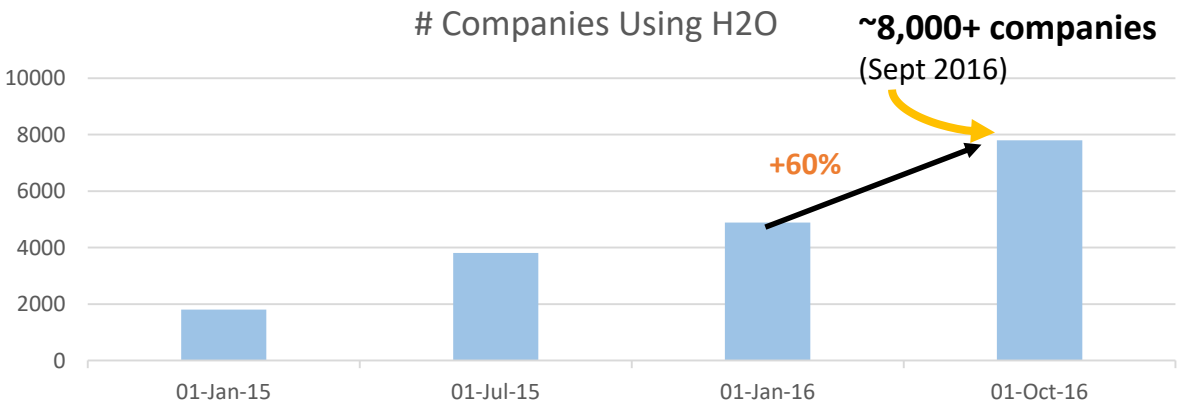
Microsoft Azure

Google

H_2O Community Growth

Tremendous Momentum Globally





* DATA FROM GOOGLE ANALYTICS EMBEDDED IN THE END USER PRODUCT

Large User Circle

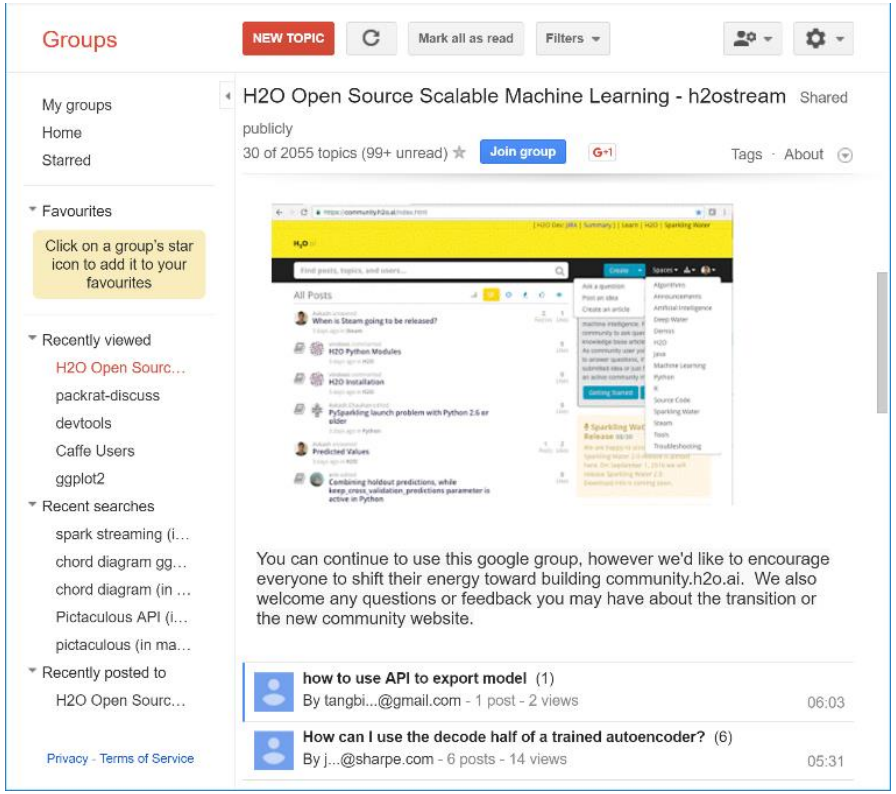
- 65,000+ users from ~8,000 companies in 140 countries. Top 5 from:

1. United States
2. India
3. Japan
4. Germany
5. United Kingdom

H2O.ai

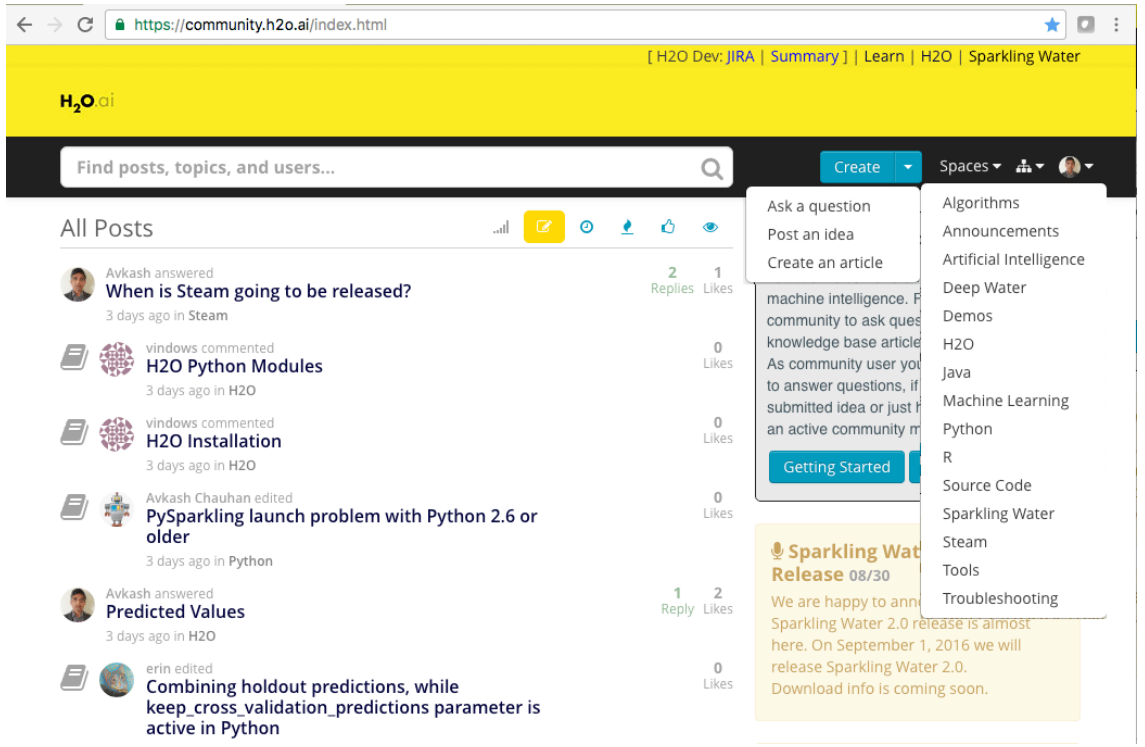
H_2O Community Support

Google forum – h2osteam



Please try

community.h2o.ai



H_2O for Kaggle Competitions

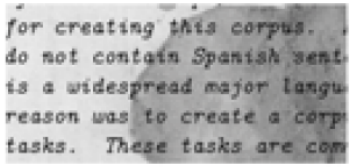
CIFAR-10 Competition Winners: Interviews with Dr. Ben Graham, Phil Culliton, & Zygmunt Zając

Triskelion | 01.02.2015

READ MORE

"I did really like H2O's deep learning implementation in R, though - the interface was great, the back end extremely easy to understand, and it was scalable and flexible. Definitely a tool I'll be going back to."

Kaggle challenge 2nd place winner Colin Priest



Completed • Knowledge • 161 teams

Denoising Dirty Documents

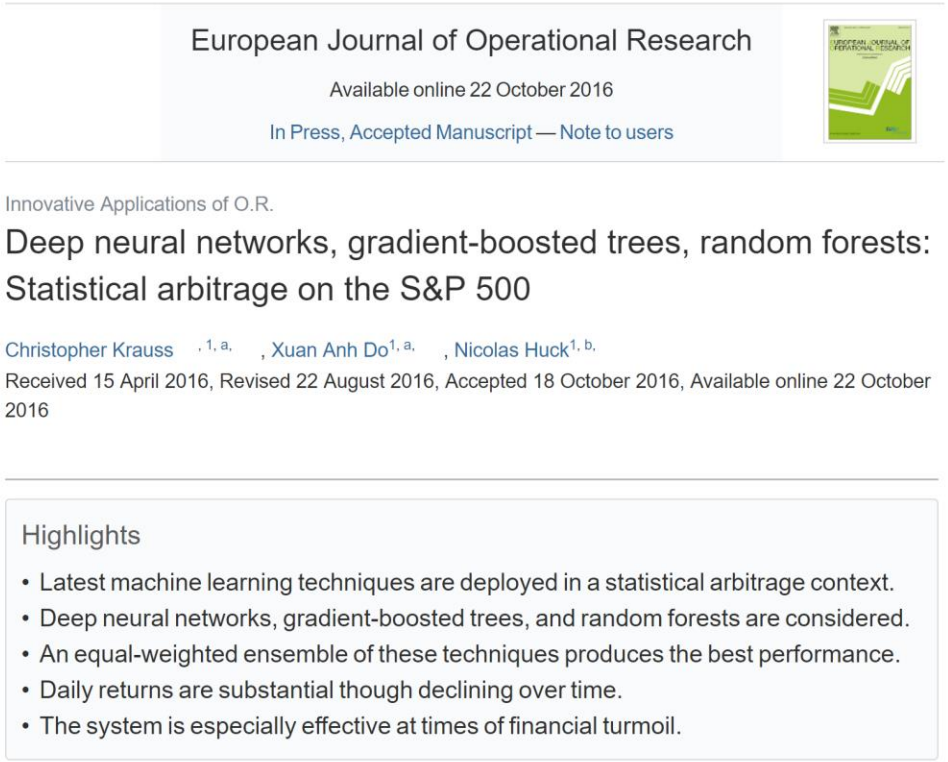
Mon 1 Jun 2015 – Mon 5 Oct 2015 (3 months ago)

READ MORE

"For my final competition submission I used an ensemble of models, including 3 deep learning models built with R and h2o."

H_2O.ai

H_2O for Academic Research



European Journal of Operational Research

Available online 22 October 2016

In Press, Accepted Manuscript — Note to users

Innovative Applications of O.R.

Deep neural networks, gradient-boosted trees, random forests: Statistical arbitrage on the S&P 500

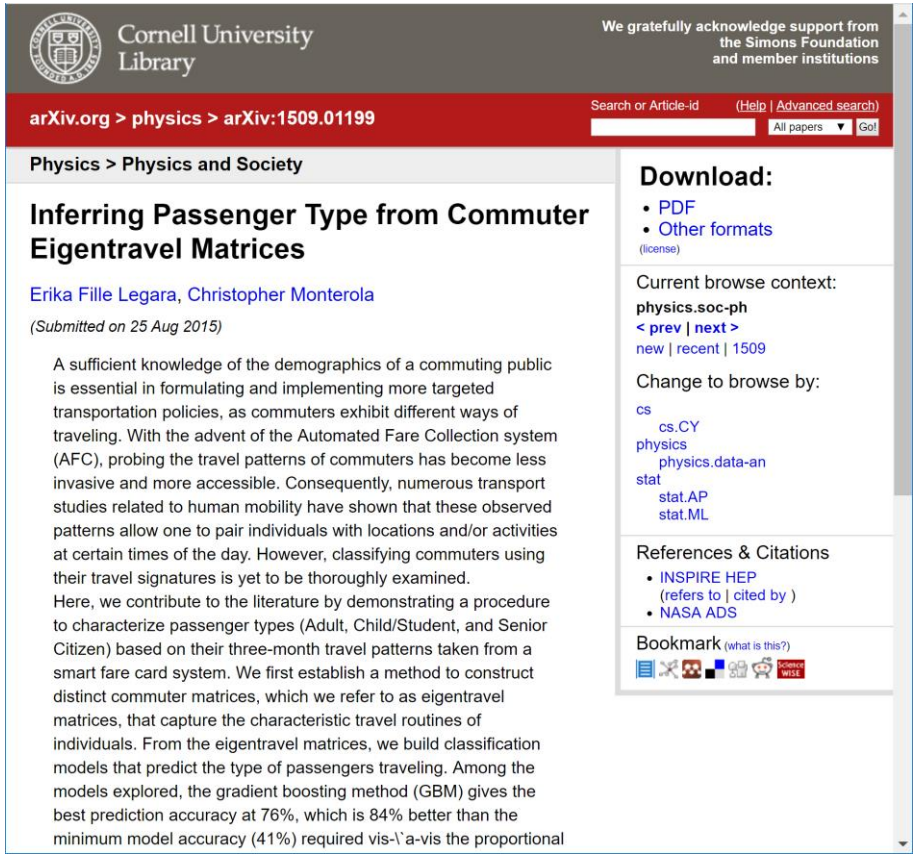
Christopher Krauss [1, a], Xuan Anh Do [1, a], Nicolas Huck [1, b]

Received 15 April 2016, Revised 22 August 2016, Accepted 18 October 2016, Available online 22 October 2016

Highlights

- Latest machine learning techniques are deployed in a statistical arbitrage context.
- Deep neural networks, gradient-boosted trees, and random forests are considered.
- An equal-weighted ensemble of these techniques produces the best performance.
- Daily returns are substantial though declining over time.
- The system is especially effective at times of financial turmoil.

http://www.sciencedirect.com/science/article/pii/S0377221716308657



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arXiv.org > physics > arXiv:1509.01199

Search or Article-id (Help | Advanced search) All papers Go!

Physics > Physics and Society

Inferring Passenger Type from Commuter Eigentravel Matrices

Erika Fille Legara, Christopher Monterola

(Submitted on 25 Aug 2015)

A sufficient knowledge of the demographics of a commuting public is essential in formulating and implementing more targeted transportation policies, as commuters exhibit different ways of traveling. With the advent of the Automated Fare Collection system (AFC), probing the travel patterns of commuters has become less invasive and more accessible. Consequently, numerous transport studies related to human mobility have shown that these observed patterns allow one to pair individuals with locations and/or activities at certain times of the day. However, classifying commuters using their travel signatures is yet to be thoroughly examined. Here, we contribute to the literature by demonstrating a procedure to characterize passenger types (Adult, Child/Student, and Senior Citizen) based on their three-month travel patterns taken from a smart fare card system. We first establish a method to construct distinct commuter matrices, which we refer to as eigentravel matrices, that capture the characteristic travel routines of individuals. From the eigentravel matrices, we build classification models that predict the type of passengers traveling. Among the models explored, the gradient boosting method (GBM) gives the best prediction accuracy at 76%, which is 84% better than the minimum model accuracy (41%) required vis-\`a-vis the proportional

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cs
cs.CY
physics
physics.data-an
stat
stat.AP
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References & Citations

- INSPIRE HEP (refers to | cited by)
- NASA ADS

Bookmark (what is this?)

https://arxiv.org/abs/1509.01199

H_2O

democratizes

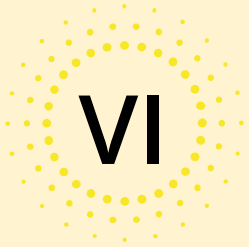
artificial intelligence & big data science

Our Open Source Products

100% Open Source. Big Data Science for Everyone!

H_2O.ai Offers AI Open Source Platform

Product Suite to Operationalize Data Science with Visual Intelligence



Visual Intelligence and UX Framework For Data Interpretation and Story Telling on top of Beautiful Data Products

100% Open Source

H_2O.ai

In-Memory, Distributed Machine Learning Algorithms with Speed and Accuracy

Deep Water

State-of-the-art Deep Learning on GPUs with TensorFlow, MXNet or Caffe with the ease of use of H2O

Spark + H_2O

SPARKLING WATER

H2O Integration with Spark. Best Machine Learning on Spark.

Steam

Operationalize and Streamline Model Building, Training and Deployment Automatically and Elastically

H_2O.ai Offers AI Open Source Platform

Product Suite to Opera… Visual Intelligence

This Talk + Live Demos

… Framework For Data Interpretation and
…autiful Data Products

100% Open Source

H_2O.ai

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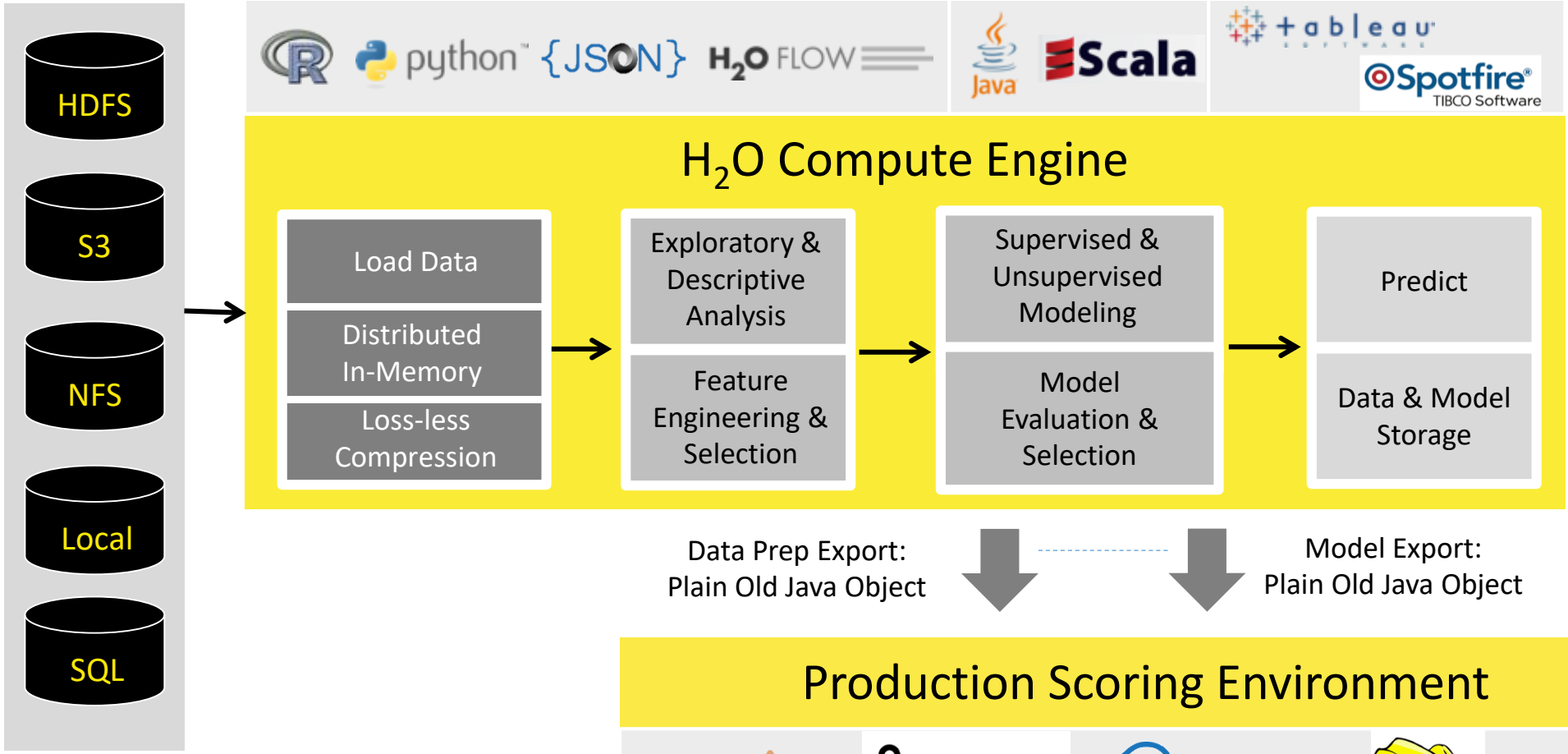
Spark + H_2O SPARKLING WATER

H2O Integration with Spark. Best Machine Learning on Spark.

Steam

Operationalize and Streamline Model Building, Training and Deployment Automatically and Elastically

High Level Architecture



Algorithms Overview

Supervised Learning

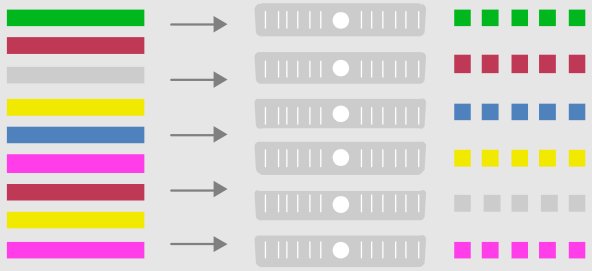
Statistical Analysis	• **Generalized Linear Models**: Binomial, Gaussian, Gamma, Poisson and Tweedie • **Naïve Bayes**
Ensembles	• **Distributed Random Forest**: Classification or regression models • **Gradient Boosting Machine**: Produces an ensemble of decision trees with increasing refined approximations
Deep Neural Networks	• **Deep learning**: Create multi-layer feed forward neural networks starting with an input layer followed by multiple layers of nonlinear transformations

Unsupervised Learning

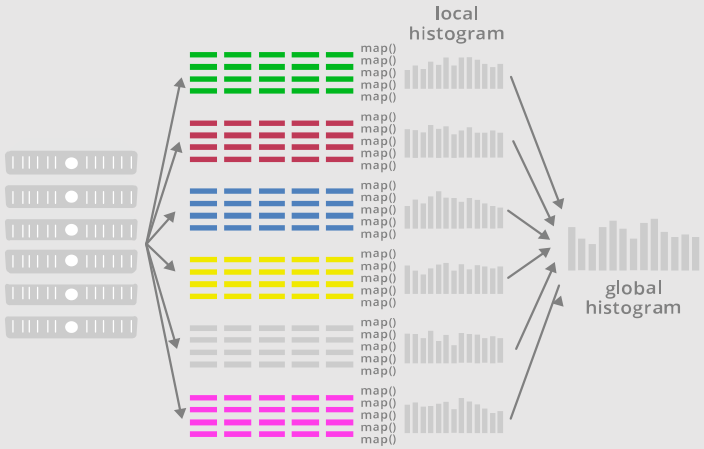
Clustering	• **K-means**: Partitions observations into k clusters/groups of the same spatial size. Automatically detect optimal k
Dimensionality Reduction	• **Principal Component Analysis**: Linearly transforms correlated variables to independent components • **Generalized Low Rank Models:** extend the idea of PCA to handle arbitrary data consisting of numerical, Boolean, categorical, and missing data
Anomaly Detection	• **Autoencoders**: Find outliers using a nonlinear dimensionality reduction using deep learning

Distributed Algorithms

Foundation for 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 H_2O: 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**

H_2O Deep Learning in Action

116M rows, 6GB CSV file
800+ predictors (numeric + categorical)

airlines_all_selected_cols.hex

Actions: View Data | Split... | Build Model... | Predict | Download | Export

Rows	Columns	Compressed Size
116695259	12	2GB

Job

Run Time 00:00:36.712
Remaining Time 00:00:17.188
Type Model
Key deeplearning-dd2f42f7-81f7-42e8-9d98-e34437309828
Description DeepLearning
Status RUNNING
Progress 69%

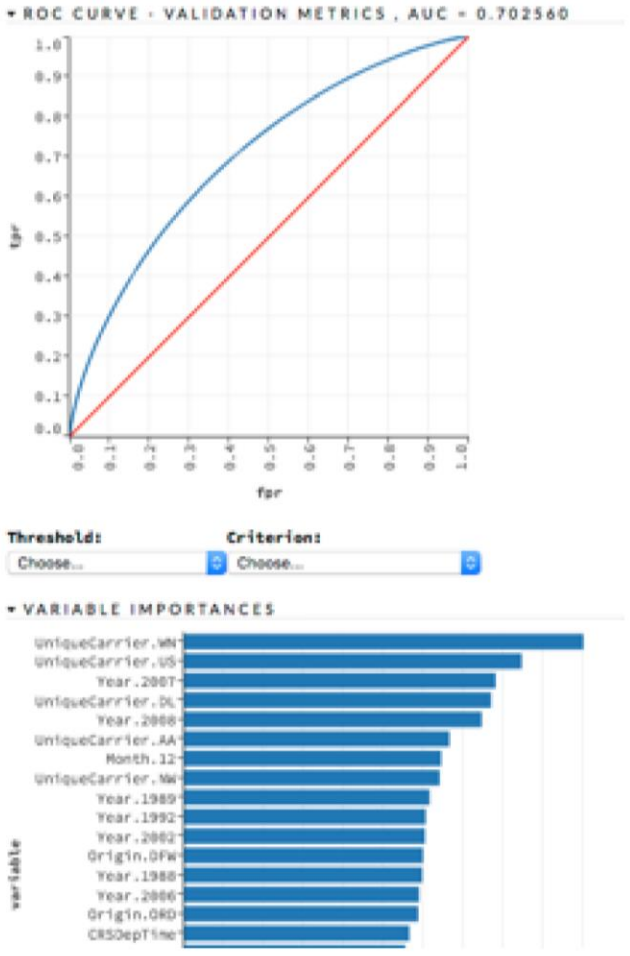
Iterations: 12. Epochs: 0.628821. Speed: 2,243,735 samples/sec. Estimated time left: 21.849 sec

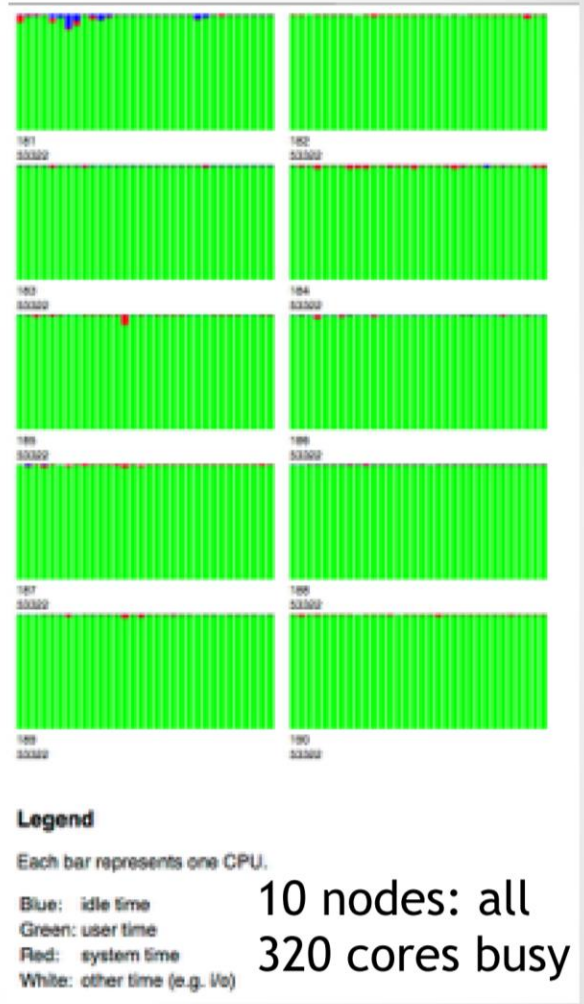
Actions: View | Cancel Job

model trained in <1 min:
2M+ samples/second

▾ OUTPUT - STATUS OF NEURON LAYERS (PREDICTING ISDEPDELAYED, 2-CLASS CLASSIFICATION, BERNOULLI DISTRIBUTION, CROSSENTROPY LOSS, 17,462 WEIGHTS/BIASES, 221.3 KB, 106,585,365 TRAINING SAMPLES, MINI-BATCH SIZE 1)

layer	units	type	dropout	l1	l2	mean_rate	rate_RMS	momentum	mean_weight	weight_RMS	mean_bias	bias_RMS
1	807	Input	0									
2	20	Rectifier	0	0	0	0.0493	0.2020	0	-0.0021	0.2111	-0.9139	1.0036
3	20	Rectifier	0	0	0	0.0197	0.0227	0	-0.1033	0.5362	-1.3988	1.5259
4	20	Rectifier	0	0	0	0.0517	0.0446	0	-0.1575	0.3068	-0.6846	0.6846
5	20	Rectifier	0	0	0	0.0761	0.0844	0	-0.0374	0.2275	-0.2647	0.2481
6	2	Softmax		0	0	0.0161	0.0083	0	0.0741	0.7260	0.4269	0.2056





10 nodes: all
320 cores busy

real-time, interactive
model inspection in Flow

H_2O.ai **Deep Learning Model**

Steam

DATA

BUSINESS INSIGHTS

Steam

Automated Platform to Build and Scale Smart Data Products

Smart Data Product

Smart Data Product

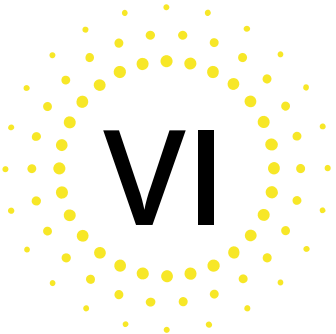
Smart Data Product

Smart Data Product

AI – Machine Learning

Automation

Scalability



Visual Intelligence

H2O.ai

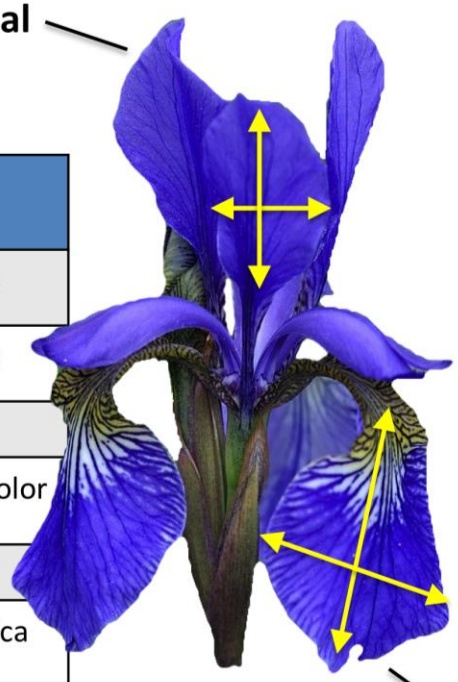
H_2O + R + Web + Steam

Quick Demo (5 mins)

Simple Demo – Iris

Samples
(instances, observations)

	Sepal length	Sepal width	Petal length	Petal width	Class label
1	5.1	3.5	1.4	0.2	Setosa
2	4.9	3.0	1.4	0.2	Setosa
			...		
50	6.4	3.5	4.5	1.2	Versicolor
			...		
150	5.9	3.0	5.0	1.8	Virginica



Class labels
(targets)

Features
(attributes, measurements, dimensions)

H_2O + R

h2o_iris_demo.R

Source on Save | Run | Source

```r
# ------------------------------------------------------------------------------
# Build a simple classification model using iris dataset
# ------------------------------------------------------------------------------

# Start and connect to a local H2O cluster
library(h2o)
h2o.init(nthreads = -1)

# Import data from a R data frame
data(iris)
d_iris <- as.h2o(iris)

# Define Targets and Features
target <- "Species"
features <- setdiff(colnames(d_iris), c("Species"))

# ------------------------------------------------------------------------------
# Train a H2O Model
# ------------------------------------------------------------------------------

# Train three basic H2O models
model_drf <- h2o.randomForest(x = features,
                              y = target,
                              model_id = "iris_random_forest",
                              training_frame = d_iris)

model_gbm <- h2o.gbm(x = features,
                     y = target,
                     model_id = "iris_gbm",
                     training_frame = d_iris)

model_dnn <- h2o.deeplearning(x = features,
                              y = target,
                              model_id = "iris_deep_learning",
                              training_frame = d_iris)
```

Please try

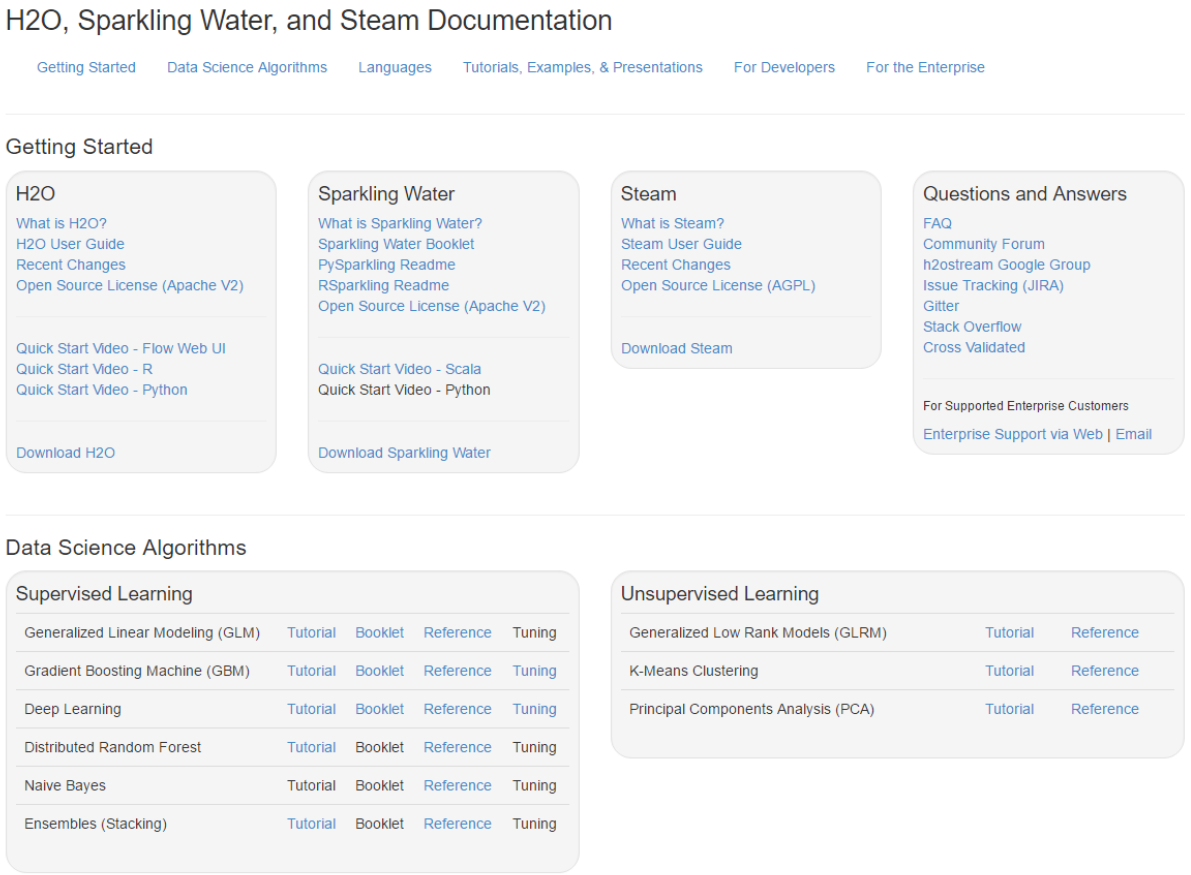
H_2O.ai

Demo Time

R, Flow and then Steam

Key Learning Resources

- Help Documentations
  - docs.h2o.ai
- Meetups
  - bit.ly/h2o_meetups
- YouTube Channel
  - bit.ly/h2o_youtube



H2O, Sparkling Water, and Steam Documentation

Getting Started | Data Science Algorithms | Languages | Tutorials, Examples, & Presentations | For Developers | For the Enterprise

Getting Started

H2O
What is H2O?
H2O User Guide
Recent Changes
Open Source License (Apache V2)

Quick Start Video - Flow Web UI
Quick Start Video - R
Quick Start Video - Python

Download H2O

Sparkling Water
What is Sparkling Water?
Sparkling Water Booklet
PySparkling Readme
RSparkling Readme
Open Source License (Apache V2)

Quick Start Video - Scala
Quick Start Video - Python

Download Sparkling Water

Steam
What is Steam?
Steam User Guide
Recent Changes
Open Source License (AGPL)

Download Steam

Questions and Answers
FAQ
Community Forum
h2ostream Google Group
Issue Tracking (JIRA)
Gitter
Stack Overflow
Cross Validated

For Supported Enterprise Customers
Enterprise Support via Web | Email

Data Science Algorithms

Supervised Learning

Generalized Linear Modeling (GLM)	Tutorial	Booklet	Reference	Tuning
Gradient Boosting Machine (GBM)	Tutorial	Booklet	Reference	Tuning
Deep Learning	Tutorial	Booklet	Reference	Tuning
Distributed Random Forest	Tutorial	Booklet	Reference	Tuning
Naive Bayes	Tutorial	Booklet	Reference	Tuning
Ensembles (Stacking)	Tutorial	Booklet	Reference	Tuning

Unsupervised Learning

Generalized Low Rank Models (GLRM)	Tutorial	Reference
K-Means Clustering	Tutorial	Reference
Principal Components Analysis (PCA)	Tutorial	Reference

TensorFlow

PARIS

rs AI Open Source Platform

Operationalize Data Science with Visual Intelligence

Visual Intelligence and U… nd
Story Telling on top of Be…

Meetup Talk
Yesterday
bit.ly/h2o_paris_2_slides

3rd talk this evening
by Jakub (Kuba)

100% Open Source

H2O.ai

In-Memory, Distributed Machine Learning Algorithms with Speed and Accuracy

Deep Water

State-of-the-art Deep Learning on GPUs with TensorFlow, MXNet or Caffe with the ease of use of H2O

Spark + H2O

SPARKLING WATER

H2O Integration with Spark. Best Machine Learning on Spark.

Steam

Operationalize and Streamline Model Building, Training and Deployment Automatically and Elastically

H_2O's Mission

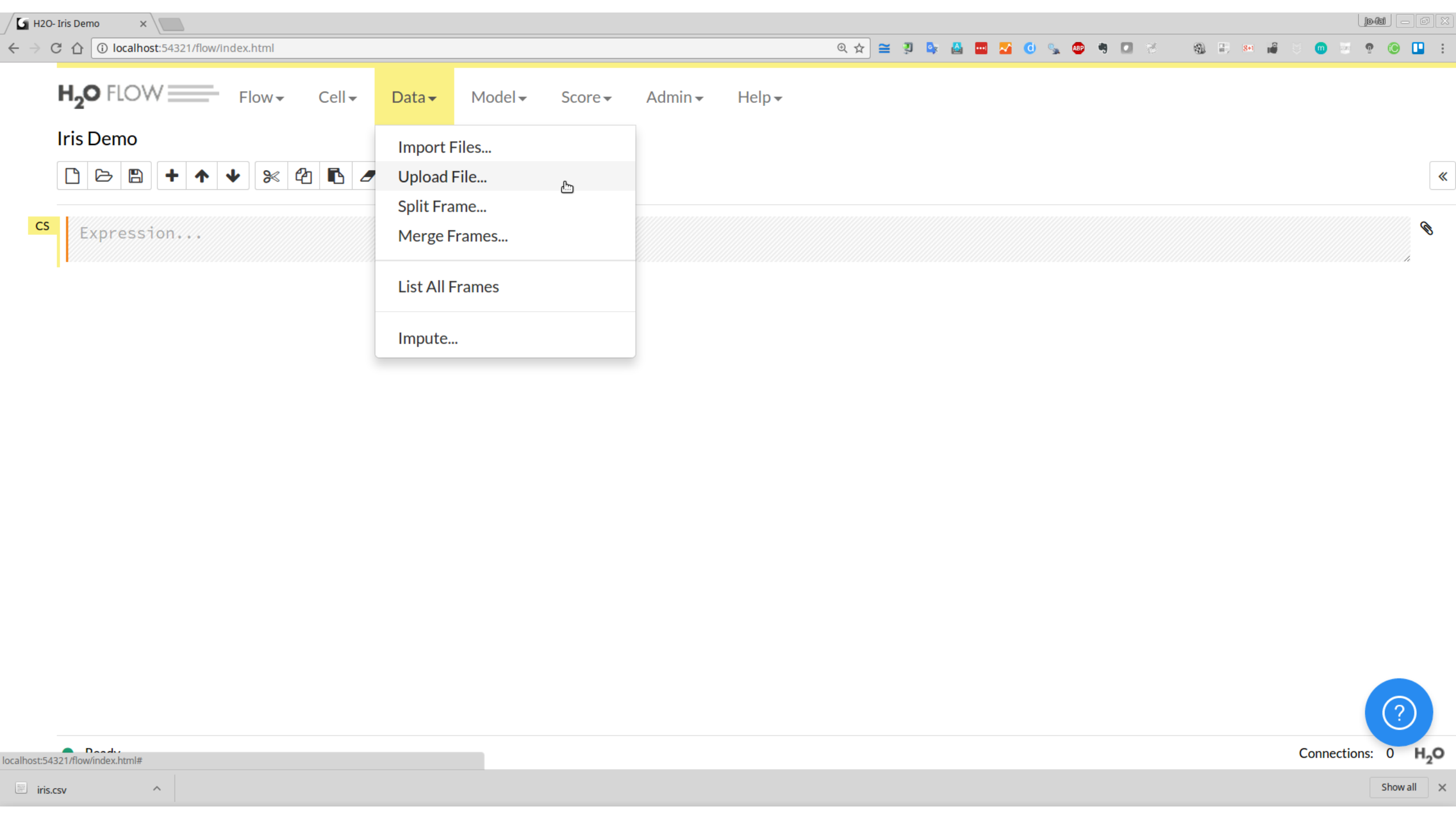
Making Machine Learning Accessible to Everyone

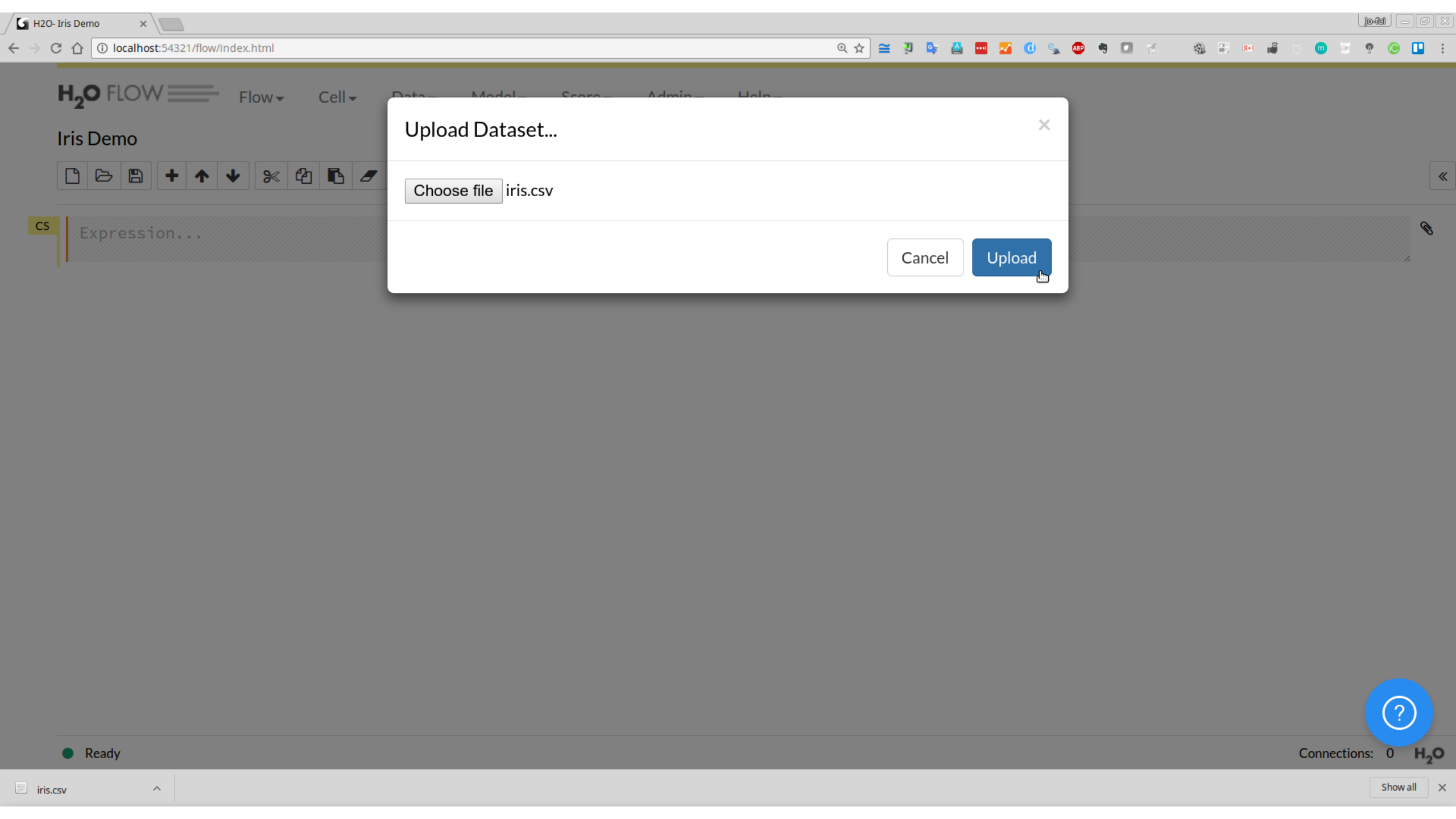
Photo credit: Virgin Media

Merci beaucoup!

- Organizers & Sponsors
  - Diane, Vincent, Barthelemy, François, Julie and Timeri
  - NUMA
- Code, Slides & Documents
  - bit.ly/h2o_meetups
  - docs.h2o.ai
- Contact
  - joe@h2o.ai
  - @matlabulous
  - github.com/woobe

Extra Slides (Iris Demo)


H2O- Iris Demo
localhost:54321/flow/index.html
H2O FLOW
Flow
Cell
Data
Model
Score
Admin
Help
Iris Demo
Import Files...
Upload File...
Split Frame...
Merge Frames...
List All Frames
Impute...
CS
Expression...
Connections: 0
localhost:54321/flow/index.html#
iris.csv
Show all


H2O- Iris Demo
localhost:54321/flow/index.html
H2O FLOW
Flow
Cell
Iris Demo
CS
Expression...
Upload Dataset...
Choose file
iris.csv
Cancel
Upload
Ready
Connections: 0
iris.csv
Show all

Iris Demo

Setup Parse

PARSE CONFIGURATION

Sources iris.csv

ID iris_from_csv

Parser CSV

Separator ,: '44'

Column Headers Auto
First row contains column names
First row contains data

Options Enable single quotes as a field quotation character
Delete on done

EDIT COLUMN NAMES AND TYPES

Search by column name...

1	Sepal.Length	Numeric	5.1	4.9	4.7	4.6	5	5.4	4.6	5	4.4
2	Sepal.Width	Numeric	3.5	3	3.2	3.1	3.6	3.9	3.4	3.4	2.9
3	Petal.Length	Numeric	1.4	1.4	1.3	1.5	1.4	1.7	1.4	1.5	1.4
4	Petal.Width	Numeric	0.2	0.2	0.2	0.2	0.2	0.4	0.3	0.2	0.2
5	Species	Enum	setosa	setosa	setosa	setosa	setosa	setosa	setosa	setosa	setosa

Previous page Next page

Parse

Iris Demo

47ms

iris_from_csv

Actions: View Data | Split... | Build Model... | Predict | Download | Export | Delete

Rows	*Columns*	*Compressed Size*
150	5	1KB

COLUMN SUMMARIES

label	*type*	*Missing*	*Zeros*	*+Inf*	*-Inf*	*min*	*max*	*mean*	*sigma*	*cardinality*	*Actions*
Sepal.Length	real	0	0	0	0	4.3000	7.9000	5.8433	0.8281	·	·
Sepal.Width	real	0	0	0	0	2.0	4.4000	3.0573	0.4359	·	·
Petal.Length	real	0	0	0	0	1.0	6.9000	3.7580	1.7653	·	·
Petal.Width	real	0	0	0	0	0.1000	2.5000	1.1993	0.7622	·	·
Species	enum	0	50	0	0	0	2.0	·	·	3	Convert to numeric

Previous 20 Columns | Next 20 Columns

CHUNK COMPRESSION SUMMARY

FRAME DISTRIBUTION SUMMARY

H2O- Iris Demo

localhost:54321/flow/index.html#

H2O FLOW Flow Cell Data Model Score Admin Help

Iris Demo

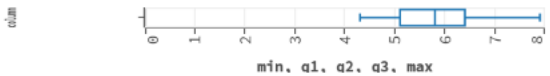
84ms

Summary: Sepal.Length

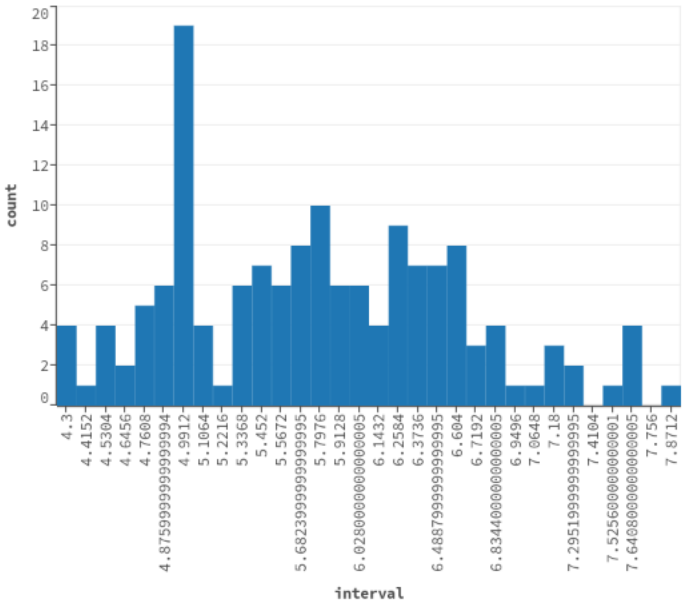
Actions Impute Inspect

CHARACTERISTICS

SUMMARY

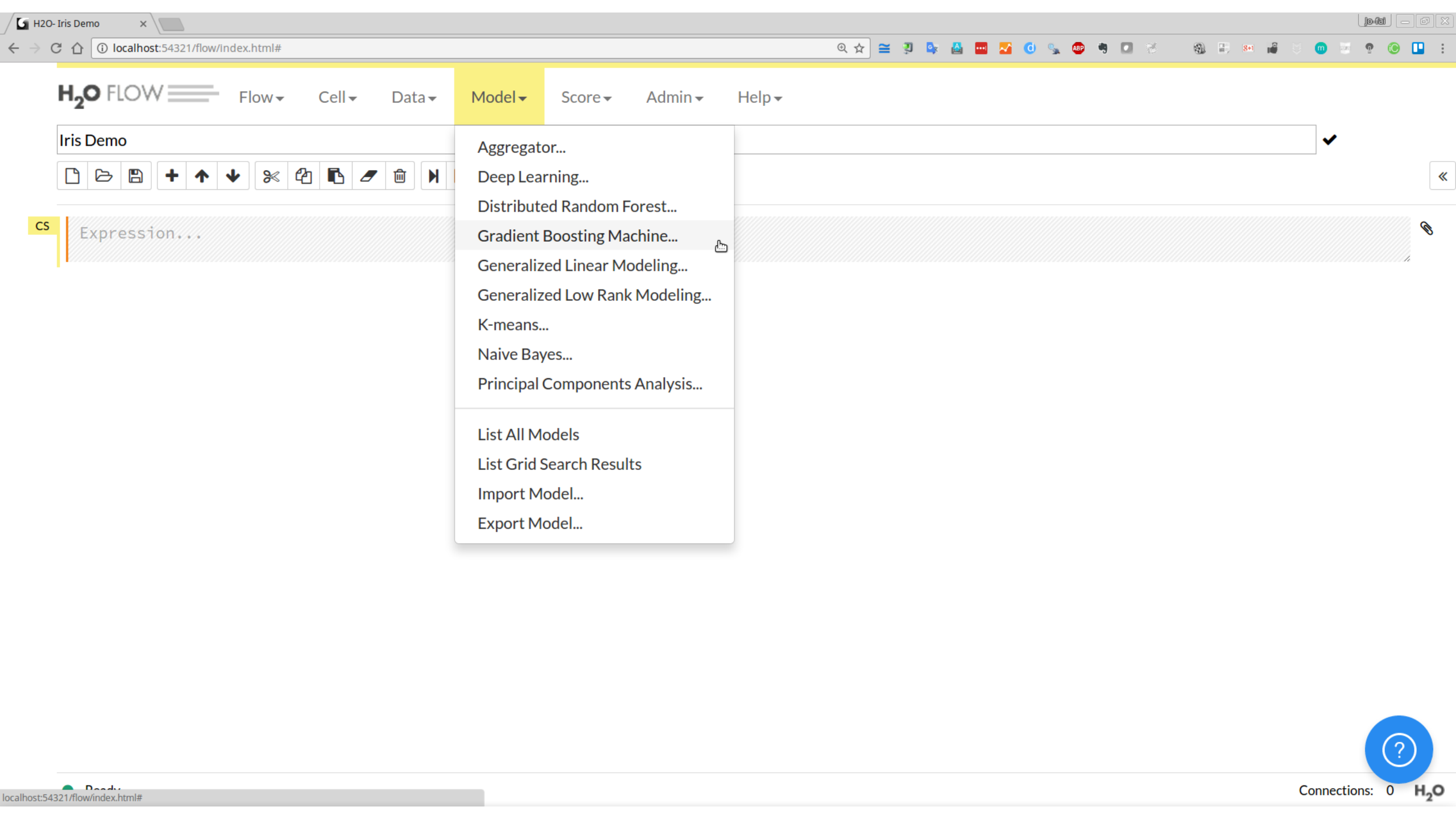


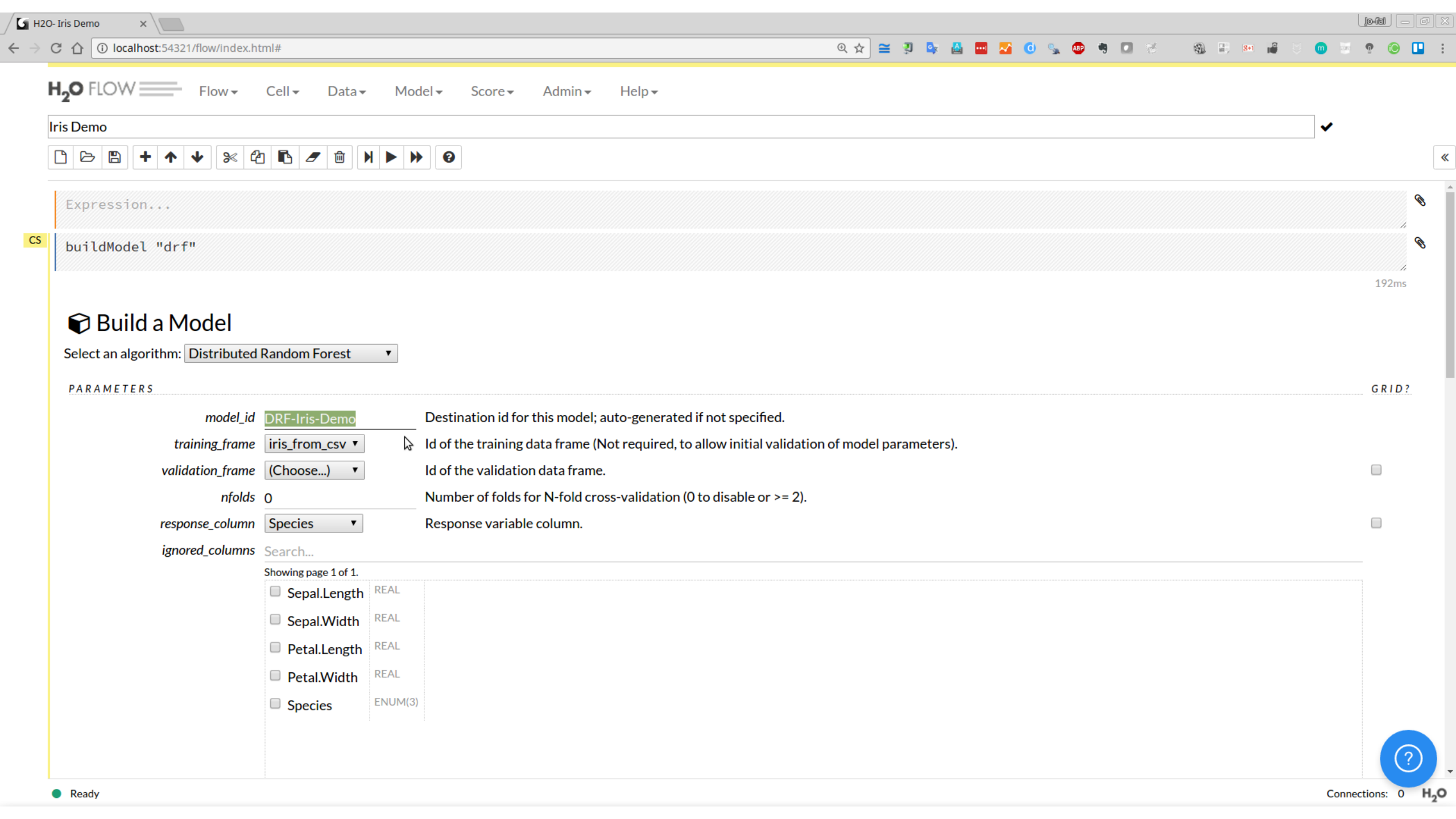
DISTRIBUTION

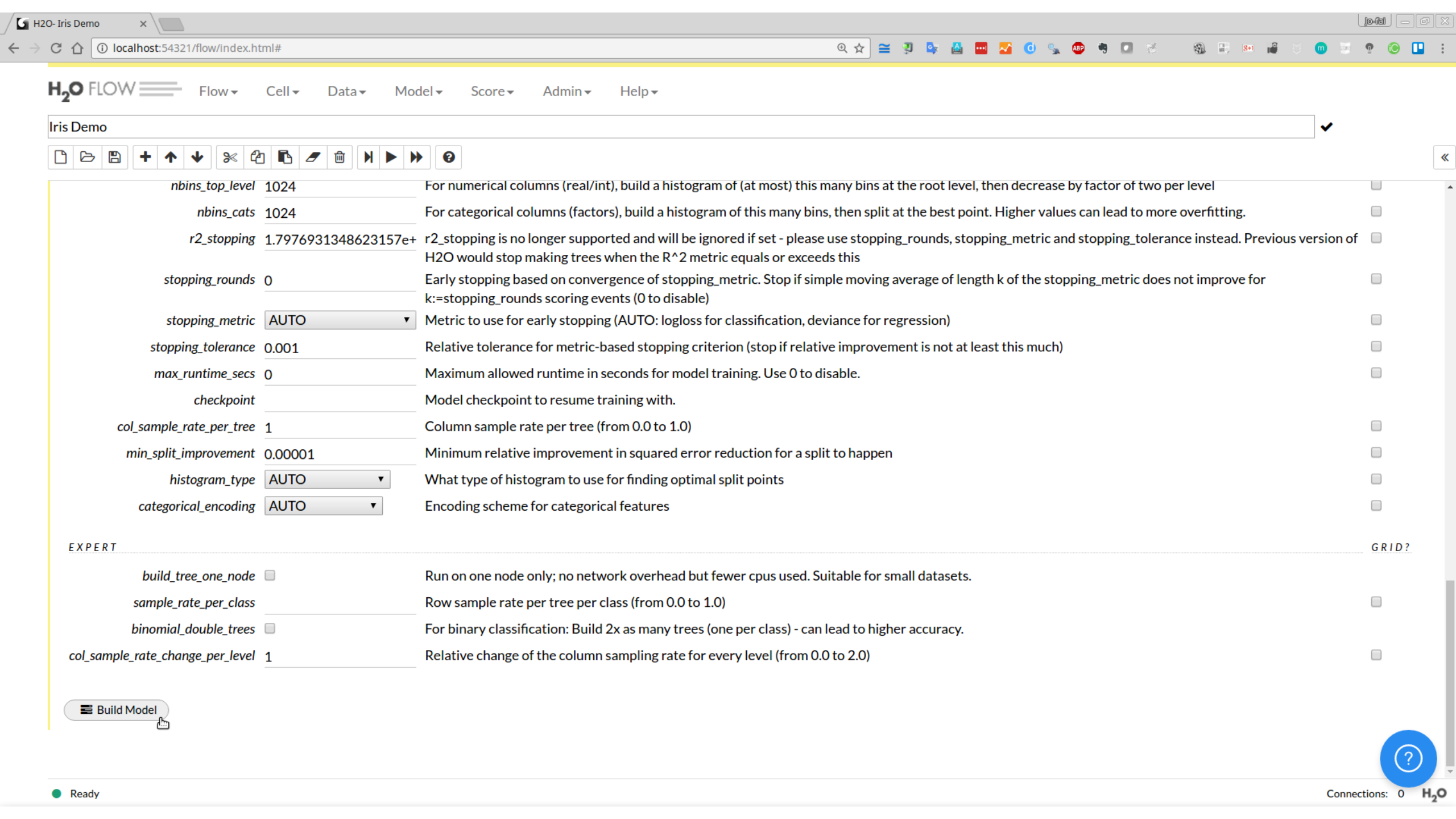


Ready

Connections: 0 H2O


H2O- Iris Demo
localhost:54321/flow/index.html#
H2O FLOW
Flow
Cell
Data
Model
Score
Admin
Help
Iris Demo
CS
Expression...
Aggregator...
Deep Learning...
Distributed Random Forest...
Gradient Boosting Machine...
Generalized Linear Modeling...
Generalized Low Rank Modeling...
K-means...
Naive Bayes...
Principal Components Analysis...
List All Models
List Grid Search Results
Import Model...
Export Model...
Connections: 0
localhost:54321/flow/index.html#


H2O- Iris Demo
localhost:54321/flow/index.html#
H2O FLOW
Flow
Cell
Data
Model
Score
Admin
Help
Iris Demo
Expression...
CS
buildModel "drf"
192ms
Build a Model
Select an algorithm: Distributed Random Forest
PARAMETERS
GRID?
model_id
DRF-Iris-Demo
Destination id for this model; auto-generated if not specified.
training_frame
iris_from_csv
Id of the training data frame (Not required, to allow initial validation of model parameters).
validation_frame
(Choose...)
Id of the validation data frame.
nfolds
0
Number of folds for N-fold cross-validation (0 to disable or >= 2).
response_column
Species
Response variable column.
ignored_columns
Search...
Showing page 1 of 1.
Sepal.Length REAL
Sepal.Width REAL
Petal.Length REAL
Petal.Width REAL
Species ENUM(3)
Ready
Connections: 0


H2O- Iris Demo
localhost:54321/flow/index.html#
H2O FLOW
Flow
Cell
Data
Model
Score
Admin
Help
Iris Demo
nbins_top_level 1024 For numerical columns (real/int), build a histogram of (at most) this many bins at the root level, then decrease by factor of two per level
nbins_cats 1024 For categorical columns (factors), build a histogram of this many bins, then split at the best point. Higher values can lead to more overfitting.
r2_stopping 1.7976931348623157e+ r2_stopping is no longer supported and will be ignored if set - please use stopping_rounds, stopping_metric and stopping_tolerance instead. Previous version of H2O would stop making trees when the R^2 metric equals or exceeds this
stopping_rounds 0 Early stopping based on convergence of stopping_metric. Stop if simple moving average of length k of the stopping_metric does not improve for k:=stopping_rounds scoring events (0 to disable)
stopping_metric AUTO Metric to use for early stopping (AUTO: logloss for classification, deviance for regression)
stopping_tolerance 0.001 Relative tolerance for metric-based stopping criterion (stop if relative improvement is not at least this much)
max_runtime_secs 0 Maximum allowed runtime in seconds for model training. Use 0 to disable.
checkpoint Model checkpoint to resume training with.
col_sample_rate_per_tree 1 Column sample rate per tree (from 0.0 to 1.0)
min_split_improvement 0.00001 Minimum relative improvement in squared error reduction for a split to happen
histogram_type AUTO What type of histogram to use for finding optimal split points
categorical_encoding AUTO Encoding scheme for categorical features
EXPERT
GRID?
build_tree_one_node Run on one node only; no network overhead but fewer cpus used. Suitable for small datasets.
sample_rate_per_class Row sample rate per tree per class (from 0.0 to 1.0)
binomial_double_trees For binary classification: Build 2x as many trees (one per class) - can lead to higher accuracy.
col_sample_rate_change_per_level 1 Relative change of the column sampling rate for every level (from 0.0 to 2.0)
Build Model
Ready
Connections: 0

Iris Demo

col_sample_rate_change_per_level 1 Relative change of the column sampling rate for every level (from 0.0 to 2.0)

Build Model

CS

```
buildModel 'drf', {"model_id":"DRF-Iris-Demo","training_frame":"iris_from_csv","nfolds":0,"response_column":"Species","ignored_columns":
[],"ignore_const_cols":true,"ntrees":50,"max_depth":20,"min_rows":1,"nbins":20,"seed":-1,"mtries":-1,"sample_rate":0.6320000290870667,"score_each_iteration":false
,"score_tree_interval":0,"balance_classes":false,"max_confusion_matrix_size":20,"max_hit_ratio_k":0,"nbins_top_level":1024,"nbins_cats":1024,"r2_stopping":1.79769
31348623157e+308,"stopping_rounds":0,"stopping_metric":"AUTO","stopping_tolerance":0.001,"max_runtime_secs":0,"checkpoint":"","col_sample_rate_per_tree":1,"min_sp
lit_improvement":0.00001,"histogram_type":"AUTO","categorical_encoding":"AUTO","build_tree_one_node":false,"sample_rate_per_class":
[],"binomial_double_trees":false,"col_sample_rate_change_per_level":1}
```

1.1s

Job

Run Time 00:00:00.183
Remaining Time 00:00:00.0
Type Model
Key DRF-Iris-Demo
Description DRF
Status DONE
Progress 100%

Done.

Actions View

Iris Demo

161ms

Model

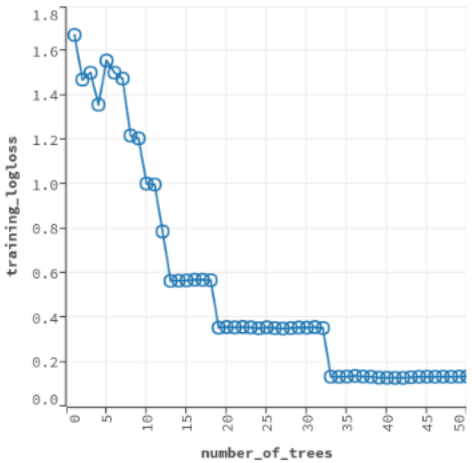
Model ID: DRF-Iris-Demo

Algorithm: Distributed Random Forest

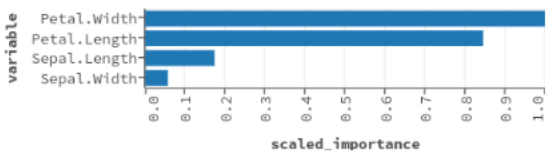
Actions: Refresh | Predict... | Download POJO | Download Model Deployment Package | Export | Inspect | Delete

MODEL PARAMETERS

SCORING HISTORY - LOGLOSS



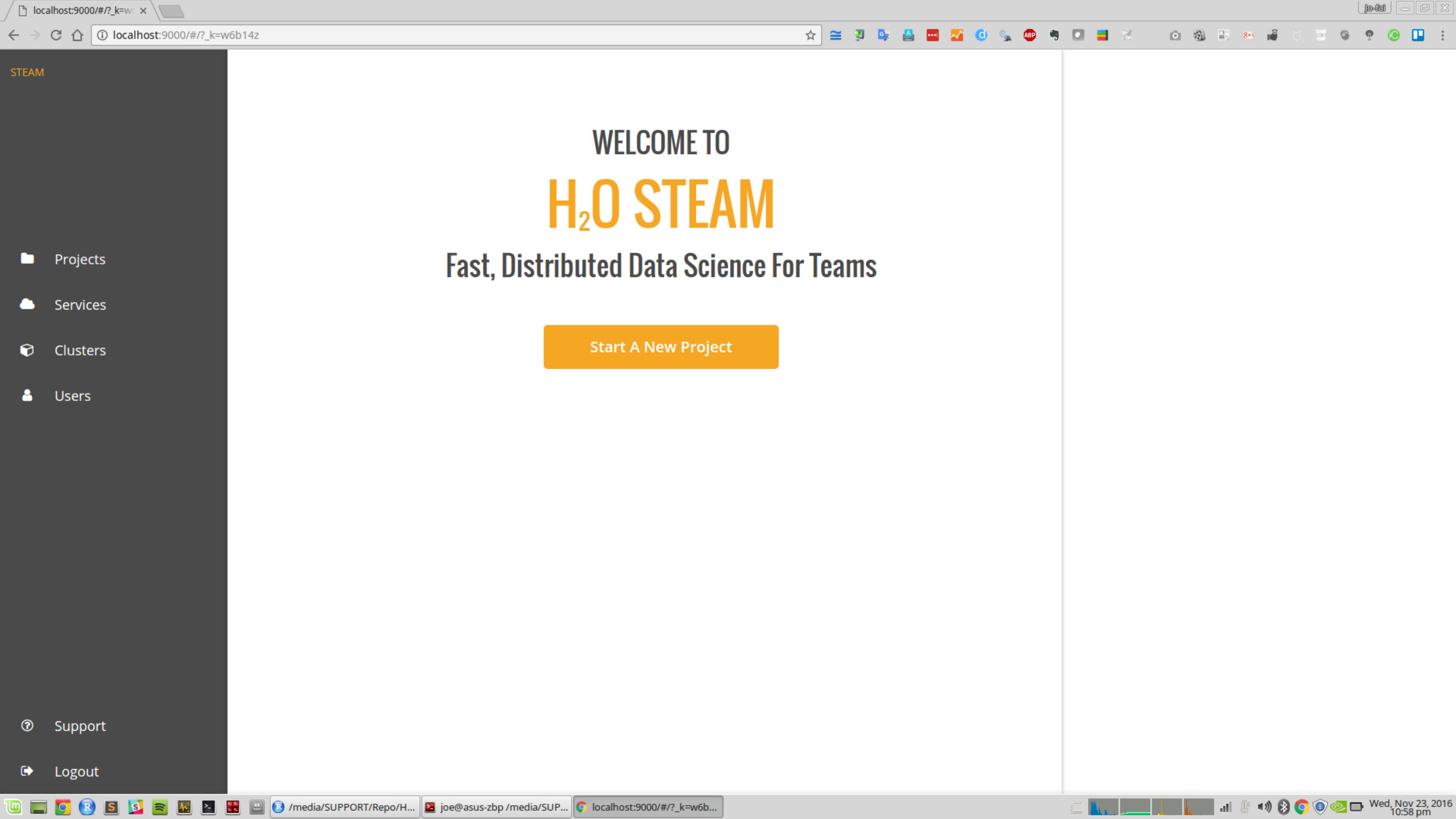
VARIABLE IMPORTANCES

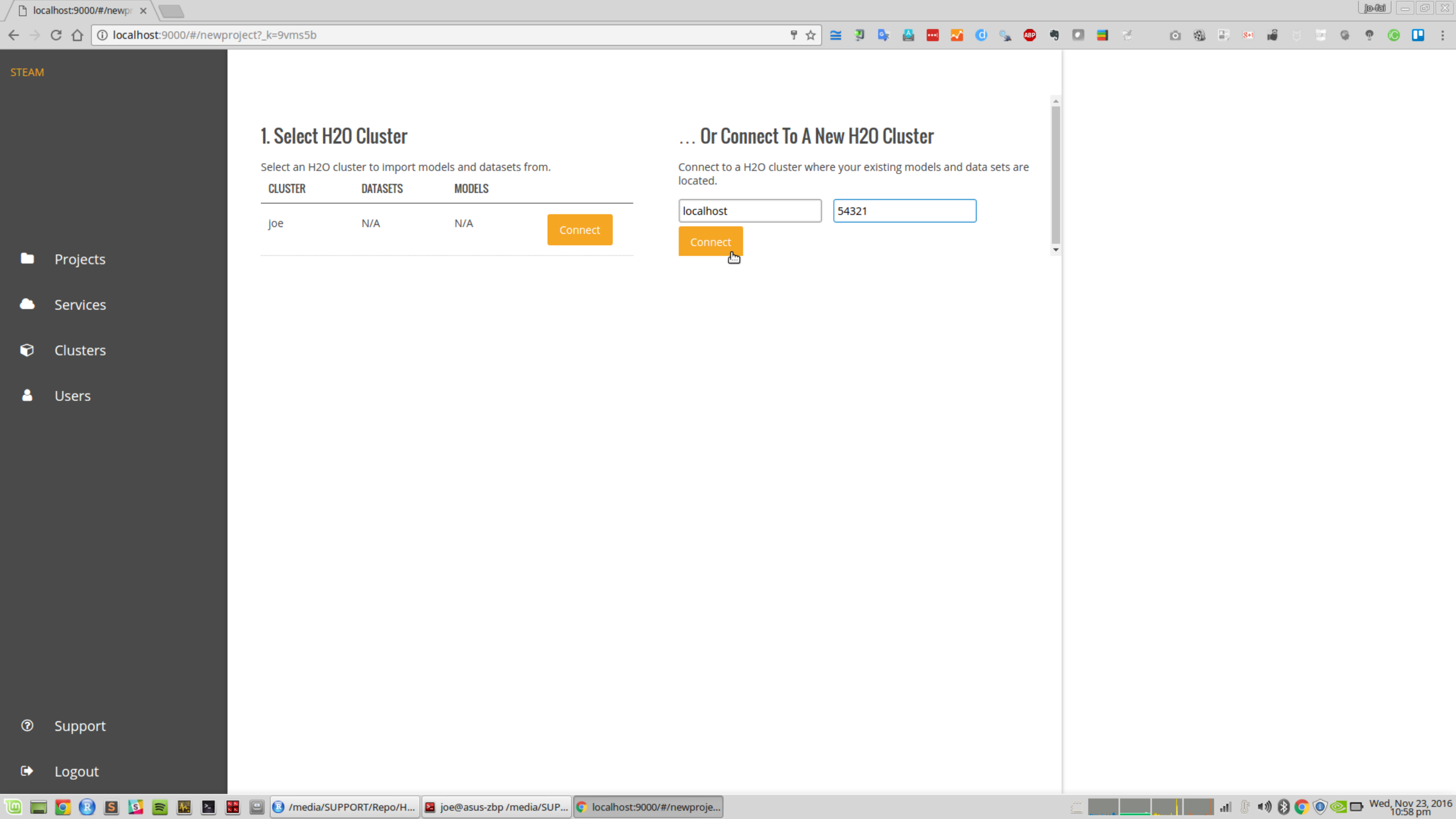


TRAINING METRICS - CONFUSION MATRIX VERTICAL: ACTUAL; ACROSS: PREDICTED

	setosa	versicolor	virginica	Error	Rate
setosa	50	0	0	**0**	**0 / 50**
versicolor	0	46	4	**0.0800**	**4 / 50**
virginica	0	4	46	**0.0800**	**4 / 50**
Total	**50**	**50**	**50**	**0.0533**	**8 / 150**

Extra Slides (Steam Demo)


localhost:9000/#/?_k=w6b14z
STEAM
Projects
Services
Clusters
Users
Support
Logout
WELCOME TO
H2O STEAM
Fast, Distributed Data Science For Teams
Start A New Project
Wed, Nov 23, 2016
10:58 pm


localhost:9000/#/newpr
localhost:9000/#/newproject?_k=9vms5b
STEAM
Projects
Services
Clusters
Users
Support
Logout
1. Select H2O Cluster
Select an H2O cluster to import models and datasets from.
CLUSTER
DATASETS
MODELS
joe
N/A
N/A
Connect
… Or Connect To A New H2O Cluster
Connect to a H2O cluster where your existing models and data sets are located.
localhost
54321
Connect
/media/SUPPORT/Repo/H...
joe@asus-zbp /media/SUP...
localhost:9000/#/newproje...
Wed, Nov 23, 2016
10:58 pm

STEAM

- Projects
- Services
- Clusters
- Users
- Support
- Logout

1. Select H2O Cluster

H2O_started_from_R_joe_eon283
localhost:54321
✖ use a different cluster

2. Select Dataframe

iris

3. Select Model Category

Multinomial

4. Pick Models To Import

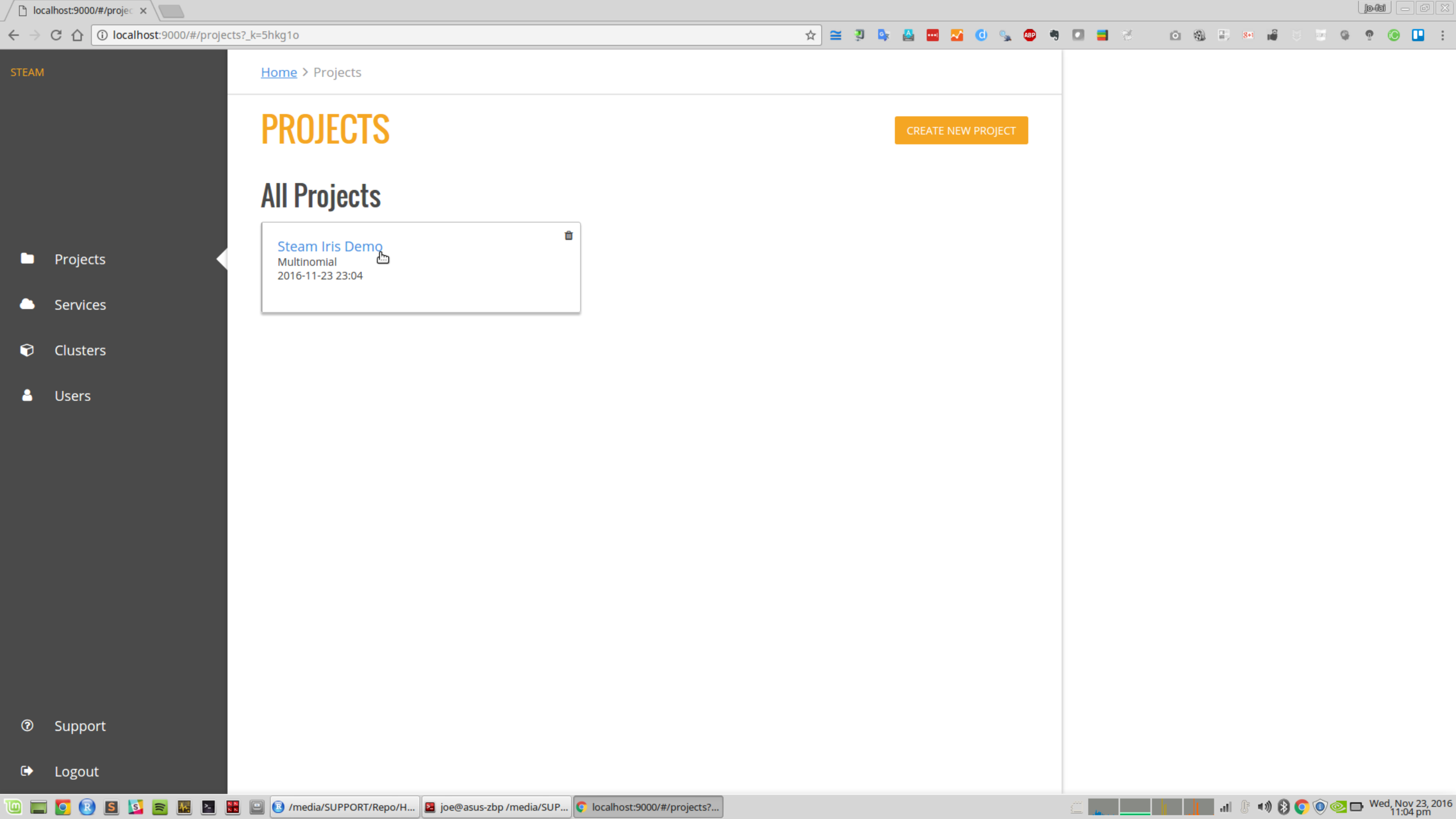
Models in a project must share the same feature set and response column to enable comparison. By default, Steam picks the most optimized model format for you to import. Advanced users can choose your own model type here.

MODEL	RESPONSE COLUMN	CATEGORICAL	
iris_deep_learning	Species	Multinomial	☑ Select for Import
iris_random_forest	Species	Multinomial	☑ Select for Import
iris_gbm	Species	Multinomial	☑ Select for Import

5. Name Project

Steam Iris Demo

Create Project


localhost:9000/#/projec
localhost:9000/#/projects?_k=5hkg1o
STEAM
Home > Projects
PROJECTS
CREATE NEW PROJECT
All Projects
Steam Iris Demo
Multinomial
2016-11-23 23:04
Projects
Services
Clusters
Users
Support
Logout
/media/SUPPORT/Repo/H...
joe@asus-zbp /media/SUP...
localhost:9000/#/projects?...
Wed, Nov 23, 2016
11:04 pm

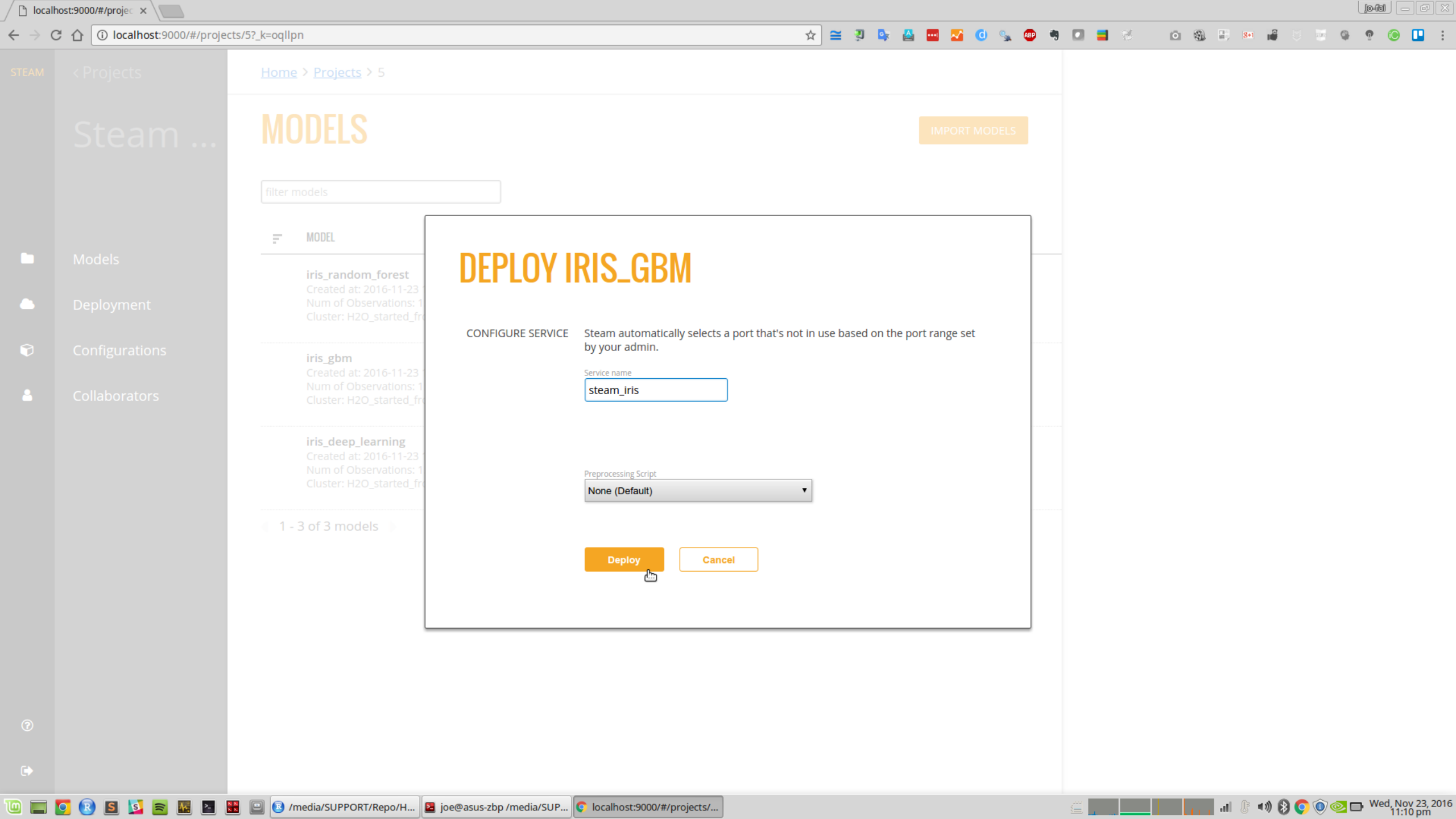
MODELS

IMPORT MODELS

filter models

MODEL	MSE	LOGLOSS	R^2	ACTIONS
iris_random_forest Created at: 2016-11-23 11:04:38 Num of Observations: 150 Cluster: H2O_started_from_R_joe_eon283	0.034834	0.122869	0.947749	view model details label as deploy model delete model
iris_gbm Created at: 2016-11-23 11:04:40 Num of Observations: 150 Cluster: H2O_started_from_R_joe_eon283	0.002838	0.018819	0.995744	view model details label as deploy model delete model
iris_deep_learning Created at: 2016-11-23 11:04:37 Num of Observations: 150 Cluster: H2O_started_from_R_joe_eon283	0.127301	0.577885	0.809048	view model details label as deploy model delete model

1 - 3 of 3 models


localhost:9000/#/projec
localhost:9000/#/projects/5?_k=oqllpn
STEAM
Projects
Steam ...
Models
Deployment
Configurations
Collaborators
Home > Projects > 5
MODELS
IMPORT MODELS
filter models
MODEL
iris_random_forest
Created at: 2016-11-23
Num of Observations: 1
Cluster: H2O_started_fr
iris_gbm
Created at: 2016-11-23
Num of Observations: 1
Cluster: H2O_started_fr
iris_deep_learning
Created at: 2016-11-23
Num of Observations: 1
Cluster: H2O_started_fr
1 - 3 of 3 models
DEPLOY IRIS_GBM
CONFIGURE SERVICE
Steam automatically selects a port that's not in use based on the port range set by your admin.
Service name
steam_iris
Preprocessing Script
None (Default)
Deploy
Cancel
/media/SUPPORT/Repo/H...
joe@asus-zbp /media/SUP...
localhost:9000/#/projects/...
Wed, Nov 23, 2016
11:10 pm

DEPLOYMENT

UPLOAD NEW PACKAGE

DEPLOYED SERVICES | PACKAGING

steam_iris @ 192.168.1.80:41788
started

Model 7
Status OK

✖ Stop Service

Prediction Service

Steam

Select input parameters, OR enter your own custom query string to predict

MODEL INPUT PARAMETERS

Parameters

1. Sepal.Length	1.2
2. Sepal.Width	0.6
3. Petal.Length	0.8
4. Petal.Width	1.1

Query String

The parameters above gets automatically built into a REST API query string. You can also input your own string if that's easier for you.

http://192.168.1.80:41788/predict? Sepal.Length=1.2&Sepal.Width=0.6&Petal.Length=0.8&Petal.Width=1.1

PREDICT CLEAR

BATCH PREDICTION *OPTIONAL

Select a Batch JSON file

PREDICTION RESULTS

Model Predictions

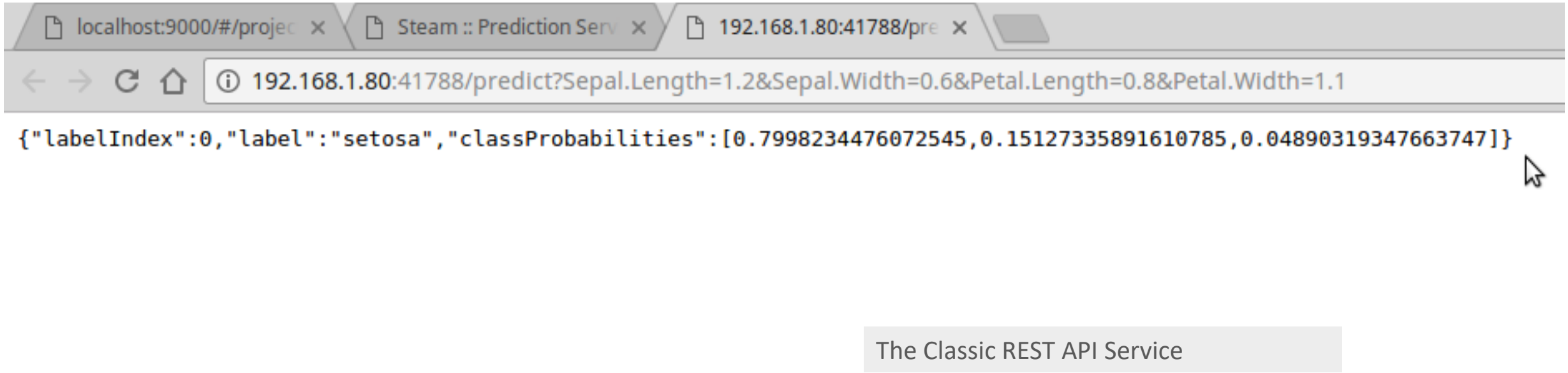
Predicting setosa

Index	Labels	Probability
0	setosa	0.7998
1	versicolor	0.1513
2	virginica	0.0489

Model Runtime Stats

Service started	2016-11-23 23:10:20 UTC
Uptime	28 s

MORE STATS



The Classic REST API Service