

+This command includes features that are part of [StataNow](#).

Description	Quick start	Menu	Syntax
Options	Remarks and examples	Stored results	Also see

Description

`h2oml gbmiclass` implements gradient boosting multiclass classification for categorical responses. You can validate your model by using validation data or cross-validation, and you can tune hyperparameters and stop early to improve model performance on new data. This command provides only measures of performance. See [\[H2OML\] h2oml postestimation](#) for commands to compute and explain predictions, examine variable importance, and perform other postestimation analyses.

For an introduction to decision trees and the gradient boosting machine (GBM) method, see [\[H2OML\] Intro](#).

Quick start

Before running the `h2oml gbmiclass` command, an H2O cluster must be initialized and data must be imported to an H2O frame; see [\[H2OML\] H2O setup](#) and [Prepare your data for H2O machine learning in Stata](#) in [\[H2OML\] h2oml](#).

Perform gradient boosting multiclass classification of categorical response `y1` on predictors `x1` through `x100`

```
h2oml gbmiclass y1 x1-x100
```

As above, but also report measures of fit for the validation frame named `valid`, and set an H2O random-number seed for reproducibility

```
h2oml gbmiclass y1 x1-x100, validframe(valid) h2orseed(123)
```

As above, but instead of a validation frame, use 3-fold cross-validation

```
h2oml gbmiclass y1 x1-x100, cv(3) h2orseed(123)
```

As above, but set the number of trees to 30, the maximum tree depth to 10, the learning rate to 0.01, and the predictor sampling rate to 0.6

```
h2oml gbmiclass y1 x1-x100, cv(3) h2orseed(123) ntrees(30)           ///
maxdepth(10) lrate(0.01) predsamprate(0.6)
```

As above, but use the default exhaustive grid search to select the optimal number of trees and the maximum tree depth that minimize the log-loss metric

```
h2oml gbmiclass y1 x1-x100, cv(3) h2orseed(123) lrate(0.01)           ///
predsamprate(0.6) ntrees(10(5)100) maxdepth(3(1)10)           ///
tune(metric(logloss))
```

As above, but use a random grid search, set an H2O random-number seed for this search, and limit the maximum search time to 200 seconds

```
h2oml gbmiclass y1 x1-x100, cv(3) h2orseed(123) lrate(0.01)      ///
  predsamprate(0.6) ntrees(10(5)100) maxdepth(3(1)10)           ///
  tune(metric(logloss) grid(random, h2orseed(456)) maxtime(200))
```

As above, but specify a learning-rate decay of 0.9, and tune the number of bins for the categorical and continuous predictors

```
h2oml gbmiclass y1 x1-x100, cv(3) h2orseed(123) lrate(0.01)      ///
  lratedecay(0.9) predsamprate(0.6) ntrees(10(5)100)           ///
  maxdepth(3(1)10) binscont(15(5)50) binscat(500(50)1100)    ///
  tune(metric(logloss) grid(random, h2orseed(456)) maxtime(200))
```

Menu

Statistics > H2O machine learning

Syntax

```
h2oml gbm multiclass response_mult predictors [ , options ]
```

response_mult and *predictors* correspond to column names of the current H2O frame.

<i>options</i>	Description
Model	
<u>validframe</u> (<i>framename</i>)	specify the name of the H2O frame containing the validation dataset that will be used to evaluate the performance of the model
<u>cv</u> [(# [, <i>cvmethod</i>])]	specify the number of folds and method for cross-validation
<u>cv</u> (<i>colname</i>)	specify the name of the variable (H2O column) for cross-validation that identifies the fold to which each observation is assigned
<u>balanceclasses</u>	balance the distribution of classes (categories of the response variable) by oversampling minority classes
<u>h2orseed</u> (#)	set H2O random-number seed for GBM
<u>encode</u> (<i>encode_type</i>)	specify H2O encoding type for categorical predictors; default is <code>encode(enum)</code>
<u>auc</u>	enable potentially time-consuming calculation of the area under the curve and area under the precision–recall curve metrics
<u>stop</u> [(# [, <i>stop_opts</i>])]	specify the number of training iterations and other criteria for stopping GBM training if the stopping metric does not improve
<u>maxtime</u> (#)	specify the maximum run time in seconds for GBM; by default, no time restriction is imposed
<u>scoreevery</u> (#)	specify that metrics be scored after every # trees during training
Hyperparameter	
<u>ntrees</u> (# <i>numlist</i>)	specify the number of trees to build the GBM model; default is <code>ntrees(50)</code>
<u>lrate</u> (# <i>numlist</i>)	specify the learning rate of each tree; default is <code>lrate(0.1)</code>
<u>lratedecay</u> (# <i>numlist</i>)	specify the rate by which the learning rate specified in <code>lrate()</code> is decaying after adding each tree to the GBM; default is <code>lratedecay(1)</code>
<u>maxdepth</u> (# <i>numlist</i>)	specify the maximum depth of each tree; default is <code>maxdepth(5)</code>
<u>minobsleaf</u> (# <i>numlist</i>)	specify the minimum number of observations per child for splitting a leaf node; default is <code>minobsleaf(10)</code>
<u>predsamprate</u> (# <i>numlist</i>)	specify the sampling rate for randomly selecting a fraction of predictors to build a tree; default is <code>predsamprate(1)</code>
<u>samprate</u> (# <i>numlist</i>)	specify the sampling rate for randomly selecting a fraction of observations to build a tree; default is <code>samprate(1)</code>
<u>minsplitthreshold</u> (# <i>numlist</i>)	specify the threshold for the minimum relative improvement needed for a node split; default is <code>minsplitthreshold(1e-05)</code>

<code>binscat(# numlist)</code>	specify the number of bins to build the histogram for node splits for categorical predictors (enum columns in H2O); default is <code>binscat(1024)</code>
<code>binsroot(# numlist)</code>	specify the number of bins to build the histogram for root node splits for continuous predictors (real and int columns in H2O); default is <code>binsroot(1024)</code>
<code>binscont(# numlist)</code>	specify the number of bins to build the histogram for node splits for continuous predictors (real and int columns in H2O); default is <code>binscont(20)</code>

Tuning

<code>tune(tune_opts)</code>	specify hyperparameter tuning options for selecting the best-performing model
------------------------------	---

Only one of `validframe()` or `cv()` is allowed.

If neither `validframe()` nor `cv()` is specified, the evaluation metrics are reported for the training dataset.

When `numlist` is specified in one or more hyperparameter options, tuning is performed for those hyperparameters.

`collect` is allowed; see [\[U\] 11.1.10 Prefix commands](#).

See [\[U\] 20 Estimation and postestimation commands](#) for more capabilities of estimation commands.

<code>cvmethod</code>	Description
<code>random</code>	randomly split the training dataset into folds; the default
<code>modulo</code>	evenly split the training dataset into folds using the modulo operation
<code>stratify</code>	evenly distribute observations from the different classes of the response to all folds

<code>stop_opts</code>	Description
<code>metric(metric_option)</code>	specify the stopping metric for training or grid search
<code>tolerance(#)</code>	specify the tolerance value by which a model must improve before the training or grid search stops; default is <code>tolerance(1e-3)</code>

<code>tune_opts</code>	Description
<code>metric(metric_option)</code>	specify the metric for selecting the best-performing model
<code>grid(gridspec)</code>	specify whether to perform an exhaustive or random search for all hyperparameter combinations
<code>maxmodels(#)</code>	specify the maximum number of models considered in the grid search; default is all configurations
<code>maxtime(#)</code>	specify the maximum run time for the grid search in seconds; default is no time limit
<code>stop([# [, stop_opts]])</code>	specify the number of iterations and other criteria for stopping GBM training if the stopping metric does not improve in the grid search
<code>parallel(#)</code>	specify the number of models to build in parallel during the grid search; default is <code>parallel(1)</code> , sequential model building
<code>nooutput</code>	suppress the table summarizing hyperparameter tuning

If any of `maxmodels()`, `maxtime()`, or `stop[()]` is specified, then `grid(random)` is implied.

Options

Model

`validframe()`, `cv[()]`, `balanceclasses`, `h2orseed()`, `encode()`, `auc`, `stop[()]`, `maxtime()`, and `scoreevery()`; see [H2OML] ***h2oml gbm***.

Hyperparameter

`ntrees()`, `lrate()`, `lratedecay()`, `maxdepth()`, `minobsleaf()`, `predsamprate()`, `samprate()`, `minsplithreshold()`, `binscat()`, `binsroot()`, and `binscont()`; see [H2OML] ***h2oml gbm***.

Tuning

`tune()`; see [H2OML] ***h2oml gbm***.

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Remarks and examples

For examples, see *Remarks and examples* in [H2OML] ***h2oml gbm***.

Stored results

`h2oml gbm` stores the following in `e()`:

Scalars

<code>e(N_train)</code>	number of observations in the training frame
<code>e(N_valid)</code>	number of observations in the validation frame (with option <code>validframe()</code>)
<code>e(N_cv)</code>	number of observations in the cross-validation (with option <code>cv()</code>)
<code>e(n_cvfolds)</code>	number of cross-validation folds (with option <code>cv()</code>)
<code>e(k_predictors)</code>	number of predictors
<code>e(n_class)</code>	number of classes
<code>e(n_trees)</code>	number of trees
<code>e(n_trees_a)</code>	actual number of trees used in GBM
<code>e(maxdepth)</code>	maximum specified tree depth
<code>e(depth_min_a)</code>	achieved minimum tree depth
<code>e(depth_avg_a)</code>	achieved average depth among trees
<code>e(depth_max_a)</code>	achieved maximum tree depth
<code>e(minobsleaf)</code>	minimum specified number of observations for a child leaf
<code>e(lrate)</code>	learning rate
<code>e(lratedecay)</code>	learning rate decay
<code>e(samprate)</code>	observation sampling rate
<code>e(predsamprate)</code>	predictor sampling rate
<code>e(minsplithr)</code>	minimum split improvement threshold
<code>e(binscat)</code>	number of bins for categorical predictors
<code>e(binsroot)</code>	number of bins for root node
<code>e(binscont)</code>	number of bins for continuous predictors
<code>e(h2orseed)</code>	H2O random-number seed
<code>e(auc)</code>	1 if <code>auc</code> ; 0 otherwise
<code>e(maxtime)</code>	maximum run time
<code>e(balanceclass)</code>	1 if classes are balanced; 0 otherwise
<code>e(stop_iter)</code>	maximum iterations before stopping training without metric improvement
<code>e(stop_tol)</code>	tolerance for metric improvement before training stops

e(scoreevery)	number of trees before scoring metrics during training
e(tune_h2orseed)	random-number seed for tuning (with option <code>tune()</code>)
e(tune_stop_iter)	maximum iterations before stopping tuning without metric improvement (with option <code>tune()</code>)
e(tune_stop_tol)	tolerance for metric improvement before tuning stops (with option <code>tune()</code>)
e(tune_maxtime)	maximum run time for tuning grid search (with option <code>tune()</code>)
e(tune_maxmodels)	maximum number of models considered in tuning grid search (with option <code>tune()</code>)
Macros	
e(cmd)	<code>h2oml gbm</code>
e(cmdline)	command as typed
e(subcmd)	<code>gbm</code>
e(method)	<code>classification</code>
e(method_type)	<code>multiclass</code>
e(class_type)	<code>Gradient boosting multiclass classification</code>
e(method_full_name)	name of response
e(response)	names of predictors
e(predictors)	title in estimation output
e(title)	name of the loss function
e(loss)	name of the training frame
e(train_frame)	name of the validation frame (with option <code>validframe()</code>)
e(valid_frame)	fold assignment method (with option <code>cv()</code>)
e(cv_method)	name of variable identifying cross-validation folds (with option <code>cv()</code>)
e(cv_varname)	encoding type for categorical predictors
e(stop_metric)	stopping metric for training
e(tune_grid)	grid search method used for tuning (with option <code>tune()</code>)
e(tune_metric)	name of the tuning metric (with option <code>tune()</code>)
e(tune_stop_metric)	stopping metric for tuning (with option <code>tune()</code>)
e(properties)	<code>nob noV</code>
e(estat_cmd)	program used to implement <code>h2omlestat</code>
e(predict)	program used to implement <code>h2omlpredict</code>
e(marginsnotok)	predictions disallowed by <code>margins</code>
Matrices	
e(metrics)	training, validation, and cross-validation metrics
e(hyperparam_table)	minimum, maximum, and selected hyperparameter values

Also see

[H2OML] **h2oml postestimation** — Postestimation tools for `h2oml gbm` and `h2oml rf`⁺

[H2OML] **h2oml** — Introduction to commands for Stata integration with H2O machine learning⁺

[H2OML] **h2oml gbm** — Gradient boosting machine for regression and classification⁺

[H2OML] **h2oml gbbinclass** — Gradient boosting binary classification⁺

[H2OML] **h2oml gbregress** — Gradient boosting regression⁺

[H2OML] **h2oml rfmulticlass** — Random forest multiclass classification⁺

[U] 20 Estimation and postestimation commands

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