### Deep Learning & AutoML in h2o



### Set up

```
# Load libraries
library(h2o)
library(tidyverse)
library(wesanderson)
library(knitr)
# Disable progress bar in document
h2o.no_progress()
# Start h2o cluster
h2o.init(nthreads = -1,
    max_mem_size = '4G')
## H2O is not running yet, starting it now...
## Note: In case of errors look at the following log files:
      ##
##
      C:\Users\User\AppData\Local\Temp\RtmpaSjdSg\file23f4343678e4/h2o_User_started_from_r.err
##
##
## Starting H2O JVM and connecting: Connection successful!
##
## R is connected to the H2O cluster:
##
      H2O cluster uptime:
                               2 seconds 939 milliseconds
                               Europe/Warsaw
##
      H2O cluster timezone:
##
      H2O data parsing timezone: UTC
##
      H2O cluster version:
                               3.40.0.1
      H2O cluster version age:
##
                               1 month and 6 days
##
                               H2O_started_from_R_User_ltv033
      H2O cluster name:
##
      H2O cluster total nodes:
      H2O cluster total memory: 3.98 GB
##
```

```
##
       H2O cluster total cores:
##
       H2O cluster allowed cores: 16
                                    TRUE
##
       H2O cluster healthy:
##
       H20 Connection ip:
                                    localhost
##
       H20 Connection port:
                                    54321
##
       H2O Connection proxy:
       H20 Internal Security:
                                    FALSE
       R Version:
                                    R version 4.2.2 (2022-10-31 ucrt)
##
# Load file
mushrooms <- read.csv('https://tinyurl.com/hmkhs9au')</pre>
# Transform data set for further analysis
mushrooms <- mushrooms %>%
  # Remove not needed characters
 mutate(across(1:23, ~ substr(.x, 3,3))) %>%
  # Change columns from strings to factors
 mutate(across(everything(), as.factor))
# Load data to h2o cluster
mushrooms_hex <- as.h2o(mushrooms, destination_frame = 'mushrooms_hex')</pre>
# Split data set to train (75%) & test (25%)
mushrooms_split <- h2o.splitFrame(data = mushrooms_hex, ratios = 0.75)</pre>
mushrooms_train <- mushrooms_split[[1]]</pre>
mushrooms_test <- mushrooms_split[[2]]</pre>
```

### Data

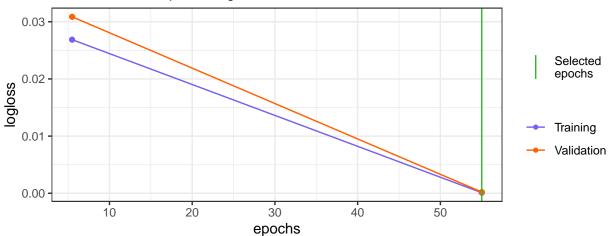
Data set contains 8124 observations of 23 species from Agaricus and Lepiota families. Beside class (e -edible/p - poisonous or uknown edibility) there are 22 physical attributes.

### h2o's Deep Learning

# # Learning curve plot h2o.learning\_curve\_plot(mushrooms\_dl)

## Learning Curve





#### # Confusion matrix

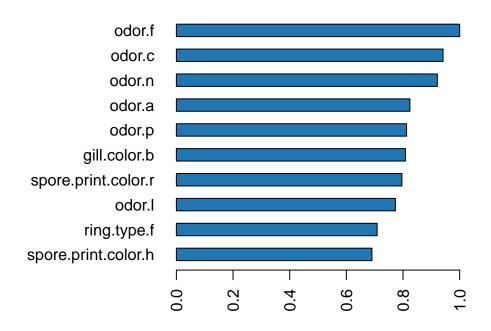
h2o.confusionMatrix(mushrooms\_dl, mushrooms\_test, valid = FALSE, xval = FALSE)

```
## Confusion Matrix (vertical: actual; across: predicted) for max f1 @ threshold = 0.995779564041497:
## e p Error Rate
## e 1050 0 0.000000 =0/1050
## p 0 1005 0.000000 =0/1005
## Totals 1050 1005 0.000000 =0/2055
```

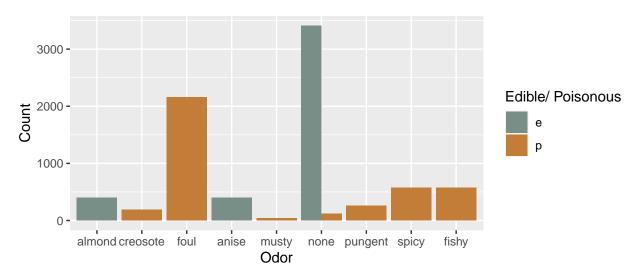
```
# Importance of parameters
```

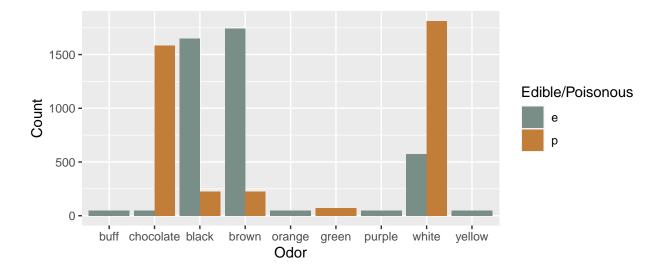
h2o.varimp\_plot(mushrooms\_dl)

### Variable Importance: Deep Learning



Most important parameters are odor and spore print color:





#### AutoML

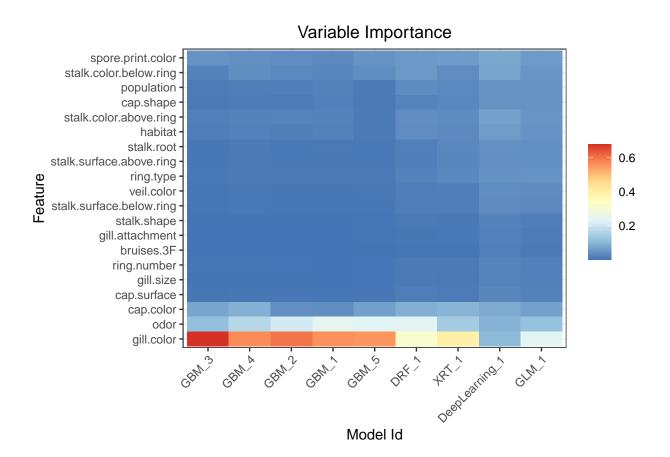
```
##
## 00:04:11.44: AutoML: XGBoost is not available; skipping it.
## 00:04:11.72: _train param, Dropping bad and constant columns: [veil.type]
## 00:04:14.402: _train param, Dropping bad and constant columns: [veil.type]
## 00:04:31.965: _train param, Dropping bad and constant columns: [veil.type]
## 00:04:32.977: _train param, Dropping bad and constant columns: [veil.type]
## 00:04:41.334: _train param, Dropping bad and constant columns: [veil.type]
## 00:04:50.140: _train param, Dropping bad and constant columns: [veil.type]
## 00:05:02.23: _train param, Dropping bad and constant columns: [veil.type]
## 00:05:02.932: _train param, Dropping bad and constant columns: [veil.type]
## 00:05:09.915: _train param, Dropping bad and constant columns: [veil.type]
```

```
# Leader board
df <- h2o.get_leaderboard(object = mushrooms_auoml)
df <- as.data.frame(df)
df$model_id <- substr(df$model_id,1,5)
kable(df)</pre>
```

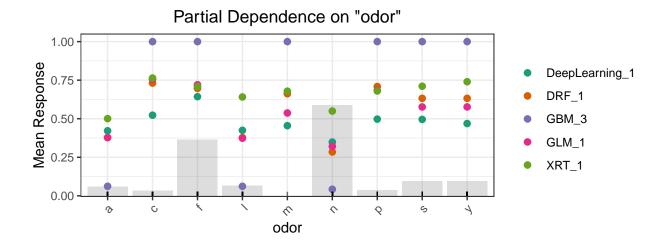
model_id	auc	logloss	aucpr	mean_per_class_error	rmse	mse
$\mathrm{DeepL}$	1.0000000	0.0002825	1.0000000	0.0000000	0.0074360	0.0000553
$GBM_3$	1.0000000	0.0000000	1.0000000	0.0000000	0.0000000	0.0000000
$GBM\_4$	1.0000000	0.0000000	1.0000000	0.0000000	0.0000000	0.0000000

model_id	auc	logloss	aucpr	mean_per_class_error	rmse	mse
GBM_1	1.0000000	0.0000000	1.0000000	0.0000000	0.0000000	0.0000000
$DRF\_1$	1.0000000	0.0020105	1.0000000	0.0000000	0.0120680	0.0001456
$GBM\_5$	1.0000000	0.0014990	1.0000000	0.0000000	0.0207927	0.0004323
$GBM_2$	1.0000000	0.0000000	1.0000000	0.0000000	0.0000000	0.0000000
$GLM\_1$	1.0000000	0.0016588	1.0000000	0.0000000	0.0123610	0.0001528
XRT_1	0.9999735	0.2961661	0.9999712	0.0022703	0.2830245	0.0801029

# Variables importance heatmap for different AutoML models
h2o.varimp\_heatmap(mushrooms\_auoml)



# Effect of odor variable for each model
h2o.pd\_multi\_plot(mushrooms\_auoml, mushrooms\_test, "odor")



### References

- 1. Data set: https://www.kaggle.com/datasets/ulrikthygepedersen/mushroom-attributes
- 2. Agaricus family graphic: https://en.wikipedia.org/wiki/Agaricus
- 3. Mushrooms graphics: <a href="https://biolwww.usask.ca/fungi/glossary.html">https://biolwww.usask.ca/fungi/glossary.html</a>
- $\begin{array}{lll} \textbf{4. h2o Deep Learning:} & \text{https://docs.h2o.ai/h2o/latest-stable/h2o-docs/data-science/deep-learning.} \\ & \text{html} \end{array}$
- 5. **h2o AutoML**: https://docs.h2o.ai/h2o/latest-stable/h2o-docs/automl.html