## ML and AI project

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```
data <- load("data_activity_recognition.RData")</pre>
tensorflow::tf$random$set_seed(0)
set.seed(19201104)
library(keras)
library(tfruns)
library(reticulate)
library(jsonlite)
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
##
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(Rtsne)
dim(x_test)#checking the dimensions of the variables
## [1] 1520 125
x_org_train <- x_train#storing the original data for CNN
x_org_test <- x_test #storing the original data for CNN</pre>
range(x_test)
## [1] -116.08 120.53
x_train <-array_reshape(x_train,c(nrow(x_train),125*45))</pre>
\#converting \ x \ data \ to \ 2 \ dimensions
x_test <-array_reshape(x_test,c(nrow(x_test),125*45))</pre>
dim(x_test)
## [1] 1520 5625
```

## #checking for new dimensions x\_train <- scale(x\_train)#scaling data</pre> x\_test <- scale(x\_test)</pre> x\_test <- as.matrix(x\_test) #converting data to matrix to perform computations such # as PCA and correlation x\_train <- as.matrix(x\_train)</pre> xval <- sample(1:nrow(x\_train),600)</pre> #separating validation set from original set xntrain <- setdiff(1:nrow(x\_train),xval)</pre> x\_val <- x\_train[xval,]</pre> x\_train <- x\_train[xntrain,]</pre> corr <- cor(x\_train)</pre> #performing correlation, to remove similar variables # this is done to reduce the dimensions and decrease the # number of parameters of the model, which might lead to a higher accuracy corr[upper.tri(corr)] <- 0</pre> #making diagnal as 0, so I don't end up removing # all the variables and upper triangular part as 0 diag(corr)<- 0</pre> $x_{\text{test}} \leftarrow x_{\text{test}}[,!apply(corr,2,function(x) any(x > .70))]$ x\_train <- x\_train[,!apply(corr,2,function(x) any(x > .70))] x\_val <- x\_val[,!apply(corr,2,function(x) any(x > .70))] #removing highly related data with correlation of more than 0.7 dim(x train) ## [1] 7000 1729 #computing pca to reduce dimensions, to improve algorithm accuracy pca <- prcomp(x\_train)</pre> prop <- cumsum(pca\$sdev^2)/sum(pca\$sdev^2)# compute cumulative proportion of variance Q <-length(prop[prop<0.95]) #maintaining atleast 95% of information ## [1] 902 #checking for Q, if large then we need to further decrease the number of dimensions prop <- cumsum(pca\$sdev^2)/sum(pca\$sdev^2)# compute cumulative proportion of variance Q <-length(prop[prop<0.90]) #maintaining atleast 90% of information ## [1] 656

Using 656 dimensions and retaining 90% information

#more than 90 % data can be explained using dimensions, so

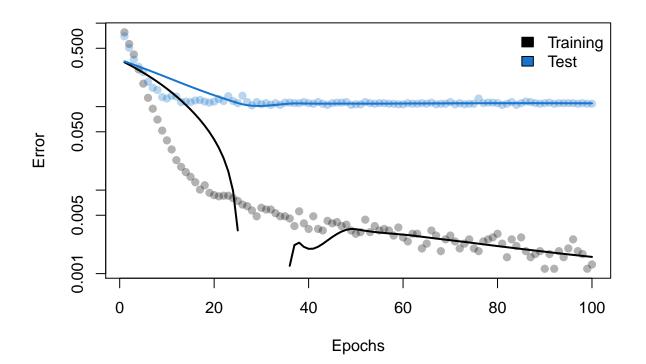
# only a handful is retained

```
#using pca to to finally convert the main data to reduced number of dimensions
x1_train <- predict(pca,x_train)[,1:Q]</pre>
x1_test <- predict(pca,x_test)[,1:Q]</pre>
x1_val <- predict(pca,x_val)[,1:Q]</pre>
#converting data to factor, followed by one-hot encoding for analysis
#separating y into training testing and validation
y1_train <- as.numeric(factor(y_train))</pre>
y1_test <- as.numeric(factor(y_test))</pre>
yc_train <- to_categorical(y1_train)</pre>
yc_test <- to_categorical(y1_test)</pre>
yc_train <- yc_train</pre>
yc_test <- yc_test</pre>
yf_train <- as.factor(y_train)</pre>
yf_test <- as.factor(y_test)</pre>
yc_test <- yc_test[,-1]</pre>
yc train <- yc train[,-1]
yc_val <- yc_train[xval,]</pre>
yc_train <- yc_train[xntrain,]</pre>
V <- ncol(x1_train)</pre>
#trying out a basic neural network with 4 layers to check out how neural network performs
model2 <- keras_model_sequential() %>%
layer_dense(units = 128, activation = "relu", input_shape = V) %>%# first hidden
layer_dense(units = 128, activation = "relu") %>%
                                                        #second layer
layer_dense(units = 64, activation = "relu") %>%
                                                        #third layer
layer dense(units = 64, activation = "relu") %>%
                                                       #fourth layer
layer_dense(units = 19, activation = "softmax") %>% #outputlayer
compile(
loss = "categorical_crossentropy", metrics = "accuracy",
optimizer = optimizer_sgd(),
)
# count parameters
count_params(model2)
## [1] 114259
fit <- model2 %>% fit(
x = x1_{train}, y = yc_{train},
validation_data = list(x1_val, yc_val),
epochs = 100,
verbose = 0,
# store accuracy on test set for each run
score <- model2 %>% evaluate(
  x1_test, yc_test,
  verbose = 0
)
score
```

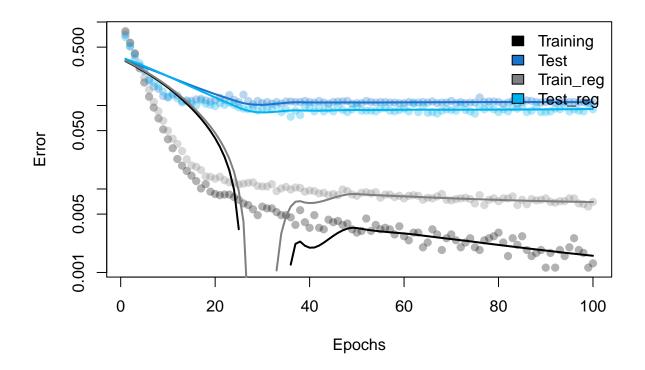
## \$loss

```
## [1] 0.6068159
##
## $accuracy
## [1] 0.8855263
```

```
# to add a smooth line to points
smooth_line <- function(y) {</pre>
x <- 1:length(y)
out <- predict( loess(y ~ x) )</pre>
return(out)
}
# some colors will be used later
cols <- c("black", "dodgerblue3", "gray50", "deepskyblue2")</pre>
# check performance ---> error
out <- 1 - cbind(fit$metrics$accuracy,fit$metrics$val_accuracy)</pre>
matplot(out, pch = 19, ylab = "Error", xlab = "Epochs",
col = adjustcolor(cols[1:2], 0.3),log = "y")
# on log scale to visualize better differences
matlines(apply(out, 2, smooth_line), lty = 1, col = cols[1:2], lwd = 2)
legend("topright", legend = c("Training", "Test"),
fill = cols[1:2], bty = "n")
```



```
#trying the same model as above, but with regularization to prevent overfitting
model_reg <- keras_model_sequential() %>%
layer_dense(units = 128, activation = "relu", input_shape = V,
kernel_regularizer = regularizer_12(1 = 0.009)) %>%
layer dense(units = 128, activation = "relu",
kernel_regularizer = regularizer_12(1 = 0.009)) %>%
layer_dense(units = 64, activation = "relu",
kernel_regularizer = regularizer_12(1 = 0.009)) %>%
layer dense(units = 32, activation = "relu",
kernel_regularizer = regularizer_12(1 = 0.009)) %>%
layer_dense(units = 19, activation = "softmax") %>%
compile(
loss = "categorical_crossentropy",
optimizer = optimizer_sgd(),
metrics = "accuracy"
)
# train and evaluate on test data at each epoch
fit_reg <- model_reg %>% fit(
x = x1_{train}, y = yc_{train},
validation_data = list(x1_val, yc_val),
epochs = 100,
verbose = 0
)
# store accuracy on test set for each run
score <- model_reg %>% evaluate(
 x1_test, yc_test,
  verbose = 0
)
score
## $loss
## [1] 0.6161871
## $accuracy
## [1] 0.9131579
out <- 1 - cbind(fit$metrics$accuracy,
fit$metrics$val_accuracy,
fit reg$metrics$accuracy,
fit_reg$metrics$val_accuracy)
# check performance
matplot(out, pch = 19, ylab = "Error", xlab = "Epochs",
col = adjustcolor(cols, 0.3),
log = "y")
matlines(apply(out, 2, smooth_line), lty = 1, col = cols, lwd = 2)
legend("topright", legend = c("Training", "Test", "Train_reg", "Test_reg"),
fill = cols, bty = "n")
```



```
# get all weights
w_all <- get_weights(model2)</pre>
w_all_reg <- get_weights(model_reg)</pre>
# weights of first hidden layer
# one input --> 64 units
w <- w_all[[3]][1,]
w_reg <- w_all_reg[[3]][1,]</pre>
# compare visually the magnitudes
par(mfrow = c(2,1), mar = c(2,2,0.5,0.5))
r <- range(w)
n <- length(w)
plot(w, ylim = r, pch = 19, col = adjustcolor(1, 0.5))
abline(h = 0, lty = 2, col = "red")
segments(1:n, 0, 1:n, w)
plot(w_reg, ylim = r, pch = 19, col = adjustcolor(1, 0.5))
abline(h = 0, lty = 2, col = "red")
segments(1:n, 0, 1:n, w_reg)
```

```
0.1
0.0
-0.1
        0
                    20
                                 40
                                                                       100
                                                                                    120
                                              60
                                                           80
0.1
0.0
-0.1
                    20
        0
                                 40
                                              60
                                                           80
                                                                       100
                                                                                    120
#CNN data preprocessing
# we need to convert this data to 3 dimensions, so that we can use convolution neural network
dim(x_org_train)
## [1] 7600 125
                    45
dim(x_org_test)
## [1] 1520 125
                    45
xcnn_train <-array_reshape(x_org_train,c(nrow(x_org_train),125,45,1))</pre>
xcnn_test <-array_reshape(x_org_test,c(nrow(x_org_test),125,45,1))</pre>
xcnn_val <- xcnn_train[xval,1:125,1:45,1]</pre>
xcnn_train <- xcnn_train[xntrain,1:125,1:45,1]</pre>
xcnn_train <- array_reshape(xcnn_train,c(nrow(xcnn_train),125,45,1))</pre>
xcnn_val <- array_reshape(xcnn_val,c(nrow(xcnn_val),125,45,1))</pre>
#cnn model
model <- keras_model_sequential() %>%
# convolutional layers
layer_conv_2d(filters = 32, kernel_size = c(2,2), activation = "relu",input_shape = c(125,45,1)) %>%
```

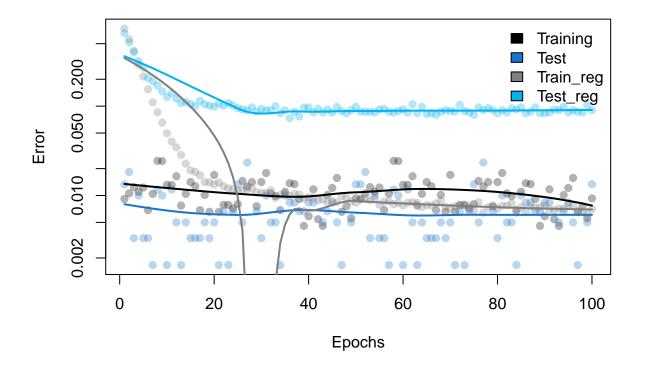
0.2

```
layer_max_pooling_2d(pool_size = c(2,2)) %>%
layer conv 2d(filters = 64, kernel size = c(3,3), activation = "relu") %%
layer_max_pooling_2d(pool_size = c(2,2)) %>%
layer_conv_2d(filters = 64, kernel_size = c(3,3), activation = "relu") %>%
layer_max_pooling_2d(pool_size = c(2,2)) %>%
# fully connected layers
layer_flatten() %>%
layer_dense(units = 64, activation = "relu", kernel_regularizer = regularizer_12(0.1)) %>%
layer_dropout(0.1) %>%
layer_dense(units = 19, activation = "softmax") %>%
# compile
compile(
loss = "categorical_crossentropy",
metrics = "accuracy",
optimizer = optimizer_adam()
)
# model training
# NOTE : this will require some time
fit <- model %>% fit(
x = xcnn_train, y = yc_train,
validation_data = list(xcnn_val, yc_val),
epochs = 50,
batch_size = 40, # roughly 0.5% of training observations
verbose = 0
)
score <- model %>% evaluate(
 xcnn_test, yc_test,
  verbose = 0
)
score
## $loss
## [1] 0.1585414
## $accuracy
## [1] 0.9868421
# to add a smooth line to points
smooth_line <- function(y) {</pre>
x \leftarrow 1:length(y)
out <- predict( loess(y ~ x) )</pre>
return(out)
# check performance
cols <- c("black", "dodgerblue3")</pre>
out <- cbind(fit$metrics$accuracy,</pre>
```

```
fit$metrics$val_accuracy)
matplot(out, pch = 19, ylab = "Accuracy", xlab = "Epochs",
col = adjustcolor(cols[1:2], 0.3),
log = "v")
matlines(apply(out, 2, smooth_line), lty = 1, col = cols[1:2], lwd = 2)
legend("bottomright", legend = c("Training", "Test"),
fill = cols[1:2], bty = "n")
model_reg <- keras_model_sequential() %>%
# convolutional layers
layer_conv_2d(filters = 32, kernel_size = c(2,2), activation = "relu",input_shape = c(125,45,1),kernel_
layer_max_pooling_2d(pool_size = c(2,2)) %>%
layer_conv_2d(filters = 64, kernel_size = c(3,3), activation = "relu", kernel_regularizer = regularizer_
layer_max_pooling_2d(pool_size = c(2,2)) %>%
layer_conv_2d(filters = 64, kernel_size = c(3,3), activation = "relu", kernel_regularizer = regularizer_
layer_max_pooling_2d(pool_size = c(2,2)) %>%
# fully connected layers
layer flatten() %>%
layer_dense(units = 64, activation = "relu", kernel_regularizer = regularizer_12(0.1)) %>%
layer_dropout(0.1) %>%
layer_dense(units = 19, activation = "softmax") %>%
# compile
compile(
loss = "categorical_crossentropy",
metrics = "accuracy",
optimizer = optimizer_adam()
# model training
# NOTE : this will require some time
fit <- model %>% fit(
x = xcnn_train, y = yc_train,
validation_data = list(xcnn_val, yc_val),
epochs = 50,
batch_size = 40, # roughly 0.5% of training observations
verbose = 0
)
score <- model %>% evaluate(
  xcnn_test, yc_test,
  verbose = 0
)
score
```

## \$loss

```
## [1] 0.1581469
##
## $accuracy
## [1] 0.9868421
out <- 1 - cbind(fit$metrics$accuracy,</pre>
fit$metrics$val_accuracy,
fit_reg$metrics$accuracy,
fit_reg$metrics$val_accuracy)
# check performance
matplot(out, pch = 19, ylab = "Error", xlab = "Epochs",
col = adjustcolor(cols, 0.3),
log = "y")
## Warning in xy.coords(x, y, xlabel, ylabel, log = log): 6 y values <= 0</pre>
## omitted from logarithmic plot
matlines(apply(out, 2, smooth_line), lty = 1, col = cols, lwd = 2)
legend("topright", legend = c("Training", "Test", "Train_reg", "Test_reg"),
fill = cols, bty = "n")
```



```
#setting a grid of values for the flags/hyperparameters of interest:
hdlayer1 <- c(128,64,32)
dropout1 <- c(0,0.1,0.3)</pre>
```

```
hdlayer2 <- c(64,16)
dropout2 \leftarrow c(0,0.2)
hdlayer3 <- c(64,32,16)
dropout3 <- c(0,0.1)
# total combinations 3 x 3 x 2 x 2 x 3 x 2 = 216
runs <- tuning_run("mlai_1.R", #creating runs to simulate output
                     runs_dir = "nn",
                     flags = list(
                    hdlayer_1 = hdlayer1,
                    dropout_1 = dropout1,
                    hdlayer_2 = hdlayer2,
                    dropout_2 = dropout2,
                    hdlayer_3 = hdlayer3,
                    dropout_3 = dropout3
                       ),
                   sample = 0.1)
## 216 total combinations of flags (sampled to 21 combinations)
## Training run 1/21 (flags = list(32, 0.3, 16, 0.2, 16, 0.1))
## Using run directory nn/2020-04-28T12-06-00Z
##
## > library(keras)
##
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
## +
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-06-00Z
## Training run 2/21 (flags = list(32, 0, 16, 0.2, 32, 0.1))
## Using run directory nn/2020-04-28T12-06-55Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-06-55Z
## Training run 3/21 (flags = list(32, 0.1, 64, 0, 16, 0.1))
## Using run directory nn/2020-04-28T12-07-52Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-07-52Z
## Training run 4/21 (flags = list(64, 0, 16, 0, 64, 0))
## Using run directory nn/2020-04-28T12-08-33Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-08-33Z
## Training run 5/21 (flags = list(32, 0, 16, 0, 16, 0.1))
## Using run directory nn/2020-04-28T12-09-04Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-09-04Z
## Training run 6/21 (flags = list(64, 0, 64, 0, 16, 0))
## Using run directory nn/2020-04-28T12-09-54Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-09-54Z
## Training run 7/21 (flags = list(32, 0, 64, 0, 32, 0.1))
## Using run directory nn/2020-04-28T12-10-20Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-10-20Z
## Training run 8/21 (flags = list(32, 0.3, 16, 0, 64, 0))
## Using run directory nn/2020-04-28T12-11-07Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-11-07Z
## Training run 9/21 (flags = list(32, 0.1, 16, 0.2, 64, 0.1))
## Using run directory nn/2020-04-28T12-11-46Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-11-46Z
## Training run 10/21 (flags = list(64, 0.1, 64, 0, 16, 0))
## Using run directory nn/2020-04-28T12-12-39Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-12-39Z
## Training run 11/21 (flags = list(128, 0.3, 64, 0, 16, 0.1))
## Using run directory nn/2020-04-28T12-13-20Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-13-20Z
## Training run 12/21 (flags = list(128, 0, 64, 0, 16, 0.1))
## Using run directory nn/2020-04-28T12-14-02Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-14-02Z
## Training run 13/21 (flags = list(64, 0.1, 64, 0.2, 64, 0))
## Using run directory nn/2020-04-28T12-14-51Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-14-51Z
## Training run 14/21 (flags = list(64, 0.3, 16, 0, 16, 0))
## Using run directory nn/2020-04-28T12-15-24Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-15-24Z
## Training run 15/21 (flags = list(32, 0, 16, 0.2, 16, 0.1))
## Using run directory nn/2020-04-28T12-16-15Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-16-15Z
## Training run 16/21 (flags = list(64, 0.1, 16, 0.2, 16, 0.1))
## Using run directory nn/2020-04-28T12-16-58Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-16-58Z
## Training run 17/21 (flags = list(128, 0.3, 64, 0, 32, 0.1))
## Using run directory nn/2020-04-28T12-17-54Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-17-54Z
## Training run 18/21 (flags = list(128, 0.1, 16, 0, 16, 0.1))
## Using run directory nn/2020-04-28T12-18-36Z
```

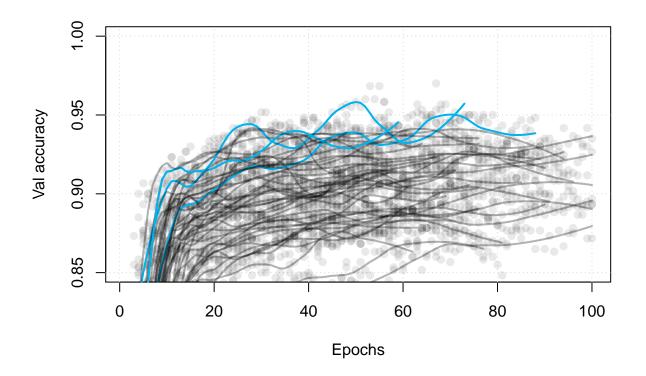
```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-18-36Z
## Training run 19/21 (flags = list(32, 0.1, 16, 0, 16, 0.1))
## Using run directory nn/2020-04-28T12-19-15Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-19-15Z
## Training run 20/21 (flags = list(32, 0, 16, 0, 64, 0))
## Using run directory nn/2020-04-28T12-19-50Z
```

```
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
         0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
        .... [TRUNCATED]
## +
## > model <- keras_model_sequential() %>% layer_dense(units = FLAGS$hdlayer_1,
## +
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
##
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn/2020-04-28T12-19-50Z
## Training run 21/21 (flags = list(128, 0.1, 64, 0, 64, 0.1))
## Using run directory nn/2020-04-28T12-20-29Z
##
## > library(keras)
## > FLAGS <- flags(flag_integer("hdlayer_1", 256), flag_numeric("dropout_1",</pre>
        0.2), flag_integer("hdlayer_2", 64), flag_numeric("dropout_2",
## +
        .... [TRUNCATED]
##
## > model <- keras_model_sequential() %% layer_dense(units = FLAGS$hdlayer_1,
         input_shape = ncol(x1_train), activation = "relu", name = "layer_1 ..." ... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = 64, verbose = 0, callback .... [TRUNCATED]
## +
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn/2020-04-28T12-20-29Z
#sampling 21 models
#Determing the optimal configuration for the data
#Extracting values from the stored runs
```

```
path <- pasteO(path, "/")</pre>
if(is.null(files)) files <- list.files(path)</pre>
n <- length(files)</pre>
out <- vector("list", n)</pre>
for(i in 1:n) {
dir <- paste0(path, files[i], "/tfruns.d/")</pre>
out[[i]] <- jsonlite::fromJSON(pasteO(dir, "metrics.json"))</pre>
out[[i]]$flags <- jsonlite::fromJSON(paste0(dir, "flags.json"))</pre>
out[[i]]$evaluation <- jsonlite::fromJSON(paste0(dir,"evaluation.json"))</pre>
}
return(out)
}
#Plotting the corresponding validation learning curves
plot_learning_curve <- function(x, ylab = NULL, cols = NULL, top = 3,</pre>
span = 0.4, ...)
smooth_line <- function(y) {</pre>
x \leftarrow 1:length(y)
out <- predict(loess(y~x, span = span))</pre>
return(out)
}
matplot(x, ylab = ylab, xlab = "Epochs", type = "n", ...)
matplot(x, pch = 19, col = adjustcolor(cols, 0.3), add = TRUE)
tmp <- apply(x, 2, smooth_line)</pre>
tmp <- sapply(tmp, "length<-", max(lengths(tmp)))</pre>
set <- order(apply(tmp, 2, max, na.rm = TRUE), decreasing = TRUE)[1:top]</pre>
cl <- rep(cols, ncol(tmp))</pre>
cl[set] <- "deepskyblue2"</pre>
matlines(tmp, lty = 1, col = cl, lwd = 2)
}
# extract results
out <- read_metrics("nn")</pre>
# extract validation accuracy and plot learning curve
acc <- sapply(out, "[[", "val_accuracy")</pre>
```

plot\_learning\_curve(acc, col = adjustcolor("black", 0.3), ylim = c(0.85, 1), ylab = "Val accuracy", top

read\_metrics <- function(path, files =NULL)</pre>



res1<- ls\_runs(metric\_val\_accuracy > 0.8, runs\_dir = "nn", order = metric\_val\_accuracy)
res1

```
## Data frame: 42 x 30
##
                       run_dir metric_val_accuracy eval_loss eval_accuracy
      nn/2020-04-28T09-01-45Z
                                             0.9500
                                                        0.2112
                                                                      0.9566
      nn/2020-04-28T09-03-43Z
                                                        0.2681
                                             0.9483
                                                                      0.9480
## 3
      nn/2020-04-28T12-20-29Z
                                             0.9450
                                                        0.3519
                                                                      0.9362
## 4
      nn/2020-04-28T12-17-54Z
                                             0.9433
                                                        0.3944
                                                                      0.9447
      nn/2020-04-28T09-03-05Z
                                             0.9383
                                                        0.3586
                                                                      0.9388
## 6
      nn/2020-04-28T12-16-58Z
                                             0.9350
                                                        0.5153
                                                                      0.9092
      nn/2020-04-28T12-13-20Z
## 7
                                             0.9350
                                                        0.3743
                                                                      0.9447
## 8
      nn/2020-04-28T12-10-20Z
                                             0.9300
                                                        0.3194
                                                                      0.9276
      nn/2020-04-28T12-09-54Z
                                             0.9283
                                                        0.3819
                                                                      0.9296
## 10 nn/2020-04-28T12-18-36Z
                                             0.9267
                                                        0.4442
                                                                      0.9237
##
      metric_loss metric_accuracy metric_val_loss
## 1
           0.0787
                            0.9977
                                             0.2378
## 2
           0.1171
                            0.9987
                                             0.2840
## 3
           0.1673
                            0.9923
                                             0.3488
## 4
           0.2484
                            0.9809
                                             0.3871
## 5
           0.1844
                            0.9884
                                             0.3809
## 6
           0.3509
                            0.9444
                                             0.4131
## 7
           0.2567
                            0.9756
                                             0.3664
## 8
           0.0926
                            0.9953
                                             0.3363
## 9
           0.1453
                            0.9970
                                             0.4122
## 10
           0.2276
                            0.9801
                                             0.4434
```

```
## # ... with 32 more rows
## # ... with 23 more columns:
       flag_hdlayer_1, flag_dropout_1, flag_hdlayer_2, flag_dropout_2,
       flag_hdlayer_3, flag_dropout_3, samples, batch_size, epochs,
## #
## #
       epochs_completed, metrics, model, loss_function, optimizer,
## #
       learning_rate, script, start, end, completed, output, source_code,
## #
       context, type
res1 <- res1[,c(2,4,8:13)]
res1[1:10,]
## Data frame: 10 x 8
      metric_val_accuracy eval_accuracy flag_hdlayer_1 flag_dropout_1
## 1
                   0.9500
                                  0.9566
                                                     128
## 2
                   0.9483
                                  0.9480
                                                     128
                                                                     0.0
## 3
                   0.9450
                                  0.9362
                                                     128
                                                                     0.1
## 4
                   0.9433
                                  0.9447
                                                     128
                                                                     0.3
## 5
                   0.9383
                                  0.9388
                                                     128
                                                                     0.1
## 6
                   0.9350
                                  0.9092
                                                      64
                                                                     0.1
## 7
                   0.9350
                                  0.9447
                                                     128
                                                                     0.3
## 8
                   0.9300
                                  0.9276
                                                      32
                                                                     0.0
## 9
                   0.9283
                                  0.9296
                                                      64
                                                                     0.0
## 10
                   0.9267
                                  0.9237
                                                     128
                                                                     0.1
      flag_hdlayer_2 flag_dropout_2 flag_hdlayer_3 flag_dropout_3
##
## 1
                                 0.0
                  64
                                                  32
                                                                 0.0
## 2
                  64
                                 0.0
                                                  16
                                                                 0.0
## 3
                                 0.0
                                                                 0.1
                  64
                                                  64
## 4
                  64
                                 0.0
                                                  32
                                                                 0.1
## 5
                  64
                                 0.2
                                                  32
                                                                 0.0
                                 0.2
## 6
                  16
                                                  16
                                                                 0.1
## 7
                                 0.0
                                                                 0.1
                  64
                                                  16
## 8
                  64
                                 0.0
                                                  32
                                                                 0.1
## 9
                  64
                                 0.0
                                                  16
                                                                 0.0
## 10
                  16
                                 0.0
                                                  16
                                                                 0.1
dropout_set \leftarrow c(0, 0.3, 0.4, 0.5)
lambda_set <- c(0, exp(seq(-6, -4, length = 9)))
lr_set \leftarrow c(0.001, 0.002, 0.005, 0.01)
bs_set \leftarrow c(0.005, 0.01, 0.02, 0.03)*nrow(x1_train)
runs <- tuning_run("mlai_2.R",
                   runs_dir = "nn2",
                    flags = list(
                      dropout = dropout_set,
                      lambda = lambda_set,
                      lr = lr_set,
                      bs = bs_set
                   ),
```

## 640 total combinations of flags (sampled to 12 combinations)

sample = 0.02)

```
## Training run 1/12 (flags = list(0.4, 0.00247875217666636, 0.01, 140))
## Using run directory nn2/2020-04-28T12-20-58Z
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
        0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1 test, yc test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn2/2020-04-28T12-20-58Z
## Training run 2/12 (flags = list(0.3, 0.0111089965382423, 0.005, 35))
## Using run directory nn2/2020-04-28T12-21-21Z
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras_model_sequential() %% layer_dense(units = 128,</pre>
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
## +
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn2/2020-04-28T12-21-21Z
## Training run 3/12 (flags = list(0.5, 0.00408677143846407, 0.002, 35))
## Using run directory nn2/2020-04-28T12-22-05Z
```

```
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag numeric("lr", 0.01), flag numeric("bs", 100))
##
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input shape = V, activation = "relu", name = "layer 1", kernel regularizer = r .... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn2/2020-04-28T12-22-05Z
## Training run 4/12 (flags = list(0.5, 0.00865169520312063, 0.002, 70))
## Using run directory nn2/2020-04-28T12-23-05Z
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc val), epochs = 100, batch size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
##
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn2/2020-04-28T12-23-05Z
## Training run 5/12 (flags = list(0.3, 0.0183156388887342, 0.002, 140))
## Using run directory nn2/2020-04-28T12-23-41Z
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## +
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
```

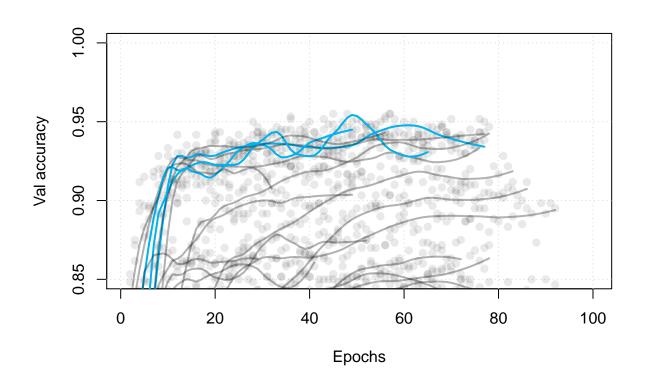
```
input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
##
## > score <- model %>% evaluate(x1 test, yc test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn2/2020-04-28T12-23-41Z
## Training run 6/12 (flags = list(0.3, 0.00408677143846407, 0.005, 35))
## Using run directory nn2/2020-04-28T12-23-58Z
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
        0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn2/2020-04-28T12-23-58Z
## Training run 7/12 (flags = list(0.5, 0.0142642339089993, 0.005, 35))
## Using run directory nn2/2020-04-28T12-24-35Z
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
##
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
```

```
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn2/2020-04-28T12-24-35Z
## Training run 8/12 (flags = list(0.4, 0.00408677143846407, 0.002, 210))
## Using run directory nn2/2020-04-28T12-25-23Z
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras model sequential() %>% layer dense(units = 128,
## +
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn2/2020-04-28T12-25-23Z
## Training run 9/12 (flags = list(0.5, 0.00247875217666636, 0.001, 70))
## Using run directory nn2/2020-04-28T12-25-43Z
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
        0.01), flag numeric("lr", 0.01), flag numeric("bs", 100))
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED
## +
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn2/2020-04-28T12-25-43Z
```

```
## Training run 10/12 (flags = list(0.5, 0.0142642339089993, 0.005, 70))
## Using run directory nn2/2020-04-28T12-26-23Z
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
        0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
##
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1 test, yc test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn2/2020-04-28T12-26-23Z
## Training run 11/12 (flags = list(0.4, 0.00865169520312063, 0.001, 210))
## Using run directory nn2/2020-04-28T12-26-51Z
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
## > model <- keras_model_sequential() %% layer_dense(units = 128,</pre>
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
## +
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
        yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
## Run completed: nn2/2020-04-28T12-26-51Z
## Training run 12/12 (flags = list(0.3, 0.00247875217666636, 0.002, 140))
## Using run directory nn2/2020-04-28T12-27-12Z
```

```
##
## > FLAGS <- flags(flag_numeric("dropout", 0.4), flag_numeric("lambda",
         0.01), flag_numeric("lr", 0.01), flag_numeric("bs", 100))
##
## > model <- keras_model_sequential() %>% layer_dense(units = 128,
         input_shape = V, activation = "relu", name = "layer_1", kernel_regularizer = r .... [TRUNCATED]
## > fit <- model %>% fit(x = x1_train, y = yc_train, validation_data = list(x1_val,
         yc_val), epochs = 100, batch_size = FLAGS$bs, verbose = 0,
## + .... [TRUNCATED]
## > score <- model %>% evaluate(x1_test, yc_test, verbose = 0)
## Warning in value[[3L]](cond): Error occurred resetting tf graph:
## AttributeError: module 'tensorflow' has no attribute 'reset_default_graph'
##
## Run completed: nn2/2020-04-28T12-27-12Z
# extract results
out <- read metrics("nn2")</pre>
# extract validation accuracy and plot learning curve
```

plot\_learning\_curve(acc, col = adjustcolor("black", 0.3), ylim = c(0.85, 1), ylab = "Val accuracy", top



acc <- sapply(out, "[[", "val\_accuracy")</pre>

```
res2<- ls_runs(metric_val_accuracy > 0.8, runs_dir = "nn2", order = metric_val_accuracy)
res2
## Data frame: 20 x 28
                        run_dir metric_val_accuracy eval_loss eval_accuracy
##
      nn2/2020-04-28T10-44-20Z
                                              0.9500
                                                        0.4423
                                                                       0.9467
## 2 nn2/2020-04-28T12-25-23Z
                                                        0.6642
                                                                       0.9211
                                              0.9467
## 3 nn2/2020-04-28T10-48-08Z
                                              0.9467
                                                        0.3771
                                                                       0.9395
## 4 nn2/2020-04-28T10-47-28Z
                                                        0.7194
                                                                       0.9112
                                              0.9450
      nn2/2020-04-28T12-27-12Z
                                              0.9433
                                                                       0.9329
                                                        0.5596
## 6 nn2/2020-04-28T10-45-43Z
                                              0.9433
                                                        0.4117
                                                                       0.9368
## 7 nn2/2020-04-28T10-44-39Z
                                             0.9383
                                                        0.6286
                                                                       0.9322
## 8 nn2/2020-04-28T12-26-51Z
                                              0.9117
                                                        0.7487
                                                                       0.9046
     nn2/2020-04-28T12-25-43Z
                                             0.9117
                                                        0.6977
                                                                       0.8934
## 10 nn2/2020-04-28T12-23-41Z
                                              0.9083
                                                        1.0036
                                                                       0.8980
##
      metric_loss metric_accuracy metric_val_loss
## 1
                            0.9889
           0.3020
                                             0.4346
## 2
           0.7645
                            0.8974
                                             0.6423
## 3
           0.2359
                            0.9943
                                             0.3668
## 4
           0.6889
                                             0.6639
                            0.9326
## 5
           0.5571
                            0.9383
                                             0.5441
## 6
           0.2766
                            0.9927
                                             0.4302
## 7
           0.6164
                            0.9404
                                             0.6079
## 8
           0.8320
                            0.8814
                                             0.7140
## 9
           0.8503
                            0.8397
                                             0.6558
## 10
           1.1218
                            0.8631
                                             1.0004
## # ... with 10 more rows
## # ... with 21 more columns:
## #
       flag_dropout, flag_lambda, flag_lr, flag_bs, samples, batch_size,
## #
       epochs, epochs_completed, metrics, model, loss_function,
## #
       optimizer, learning_rate, script, start, end, completed, output,
## #
       source_code, context, type
res2 \leftarrow res2[,c(2,4,8:13)]
res2[1:10,]
## Data frame: 10 x 8
      metric_val_accuracy eval_accuracy flag_dropout flag_lambda flag_lr
## 1
                    0.9500
                                  0.9467
                                                   0.0
                                                            0.0052
                                                                      0.001
## 2
                                                            0.0041
                    0.9467
                                  0.9211
                                                   0.4
                                                                      0.002
## 3
                   0.9467
                                  0.9395
                                                   0.0
                                                            0.0032
                                                                      0.002
## 4
                                  0.9112
                                                   0.3
                                                            0.0032
                                                                      0.002
                    0.9450
## 5
                   0.9433
                                  0.9329
                                                   0.3
                                                            0.0025
                                                                      0.002
## 6
                                                   0.0
                    0.9433
                                  0.9368
                                                            0.0111
                                                                      0.001
## 7
                    0.9383
                                  0.9322
                                                   0.3
                                                            0.0052
                                                                      0.002
## 8
                    0.9117
                                  0.9046
                                                   0.4
                                                            0.0087
                                                                      0.001
## 9
                                                   0.5
                                                            0.0025
                    0.9117
                                  0.8934
                                                                      0.001
## 10
                    0.9083
                                  0.8980
                                                   0.3
                                                            0.0183
                                                                      0.002
      flag_bs samples batch_size
##
## 1
          210
                 7000
                              210
## 2
          210
                 7000
                              210
## 3
          210
                 7000
                              210
## 4
                 7000
           70
                               70
```

##	5	140	7000	140
##	6	35	7000	35
##	7	210	7000	210
##	8	210	7000	210
##	9	70	7000	70
##	10	140	7000	140