Homework 7

DS Student

04, 14, 2022

```
library('keras')
library('nnet')
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
```

Question 1 Use the Keras library to re-implement the simple neural network discussed during lecture for the mixture data

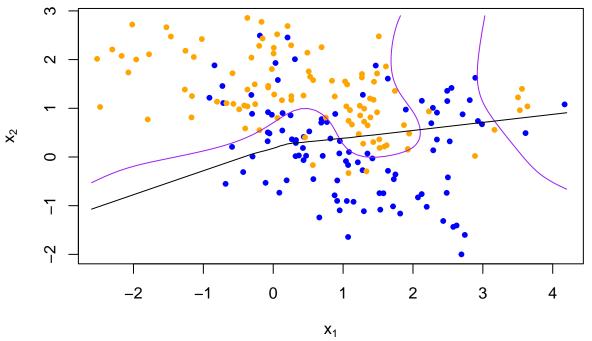
```
# dataset
data <- mixture.example
train_x <- mixture.example$x</pre>
train_y <- mixture.example$y</pre>
test_x <- mixture.example$xnew</pre>
model <- keras_model_sequential()</pre>
## Loaded Tensorflow version 2.6.0
model %>% layer_dense(units=10, activation = "relu", input_shape = c(2))%>% layer_dense(units =2, activ
model %>% compile(optimizer = "rmsprop",
                   loss = "sparse_categorical_crossentropy",
                   metric=c("accuracy"))
model %>% fit(train_x, train_y,
               epochs = 10,
               batch_size = 5)
fit_nnet <- nnet(x = train_x, y= train_y, size=10, entropy=TRUE, decay=0.02)</pre>
## # weights: 41
## initial value 170.298732
## iter 10 value 102.176693
## iter 20 value 93.350233
## iter 30 value 88.802415
```

```
## iter 40 value 87.172780
## iter 50 value 86.752669
## iter 60 value 86.506805
## iter 70 value 86.271908
## iter 80 value 86.212327
## iter 90 value 86.196772
## iter 100 value 86.182970
## final value 86.182970
## stopped after 100 iterations
```

Create a figure to illustrate that the predictions are (or are not) similar using the 'nnet' function versus the Keras model.

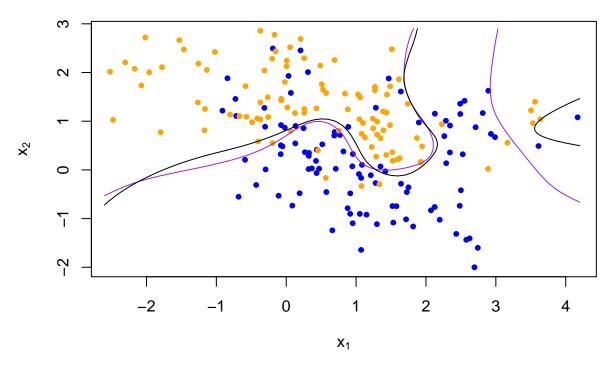
```
prediction_keras <- model%>% predict(test_x)
classes_keras <- rep(0,1)</pre>
for (i in 1:6831){
  if (prediction_keras[i,1] > prediction_keras[i,2]){
    classes_keras[i] <- 1</pre>
    classes_keras[i] <- 2</pre>
  }
}
prediction_nnet <- fit_nnet%>% predict(test_x)
classes_nnet<- rep(0,1)</pre>
for (i in 1:6831){
  if (prediction_nnet[i,1] > 0.5){
    classes nnet[i] <- 1</pre>
  }else{
    classes_nnet[i] <- 2</pre>
  }
}
plot_mixture_data <- expression({</pre>
  plot(data$x[,1], data$x[,2],
       col=ifelse(dat$y==0, 'blue', 'orange'),
       pch=20,
       xlab=expression(x[1]),
       ylab=expression(x[2]))
  ## draw Bayes (True) classification boundary
  prob <- matrix(dat$prob, length(dat$px1), length(dat$px2))</pre>
  cont <- contourLines(dat$px1, dat$px2, prob, levels=0.5)</pre>
  rslt <- sapply(cont, lines, col='purple')</pre>
})
plot_keras_preds <- function(fit, dat=mixture.example) {</pre>
  ## create figure
  eval(plot_mixture_data)
```

```
## compute predictions from nnet
probs <- prediction_keras[,1]
preds <- classes_keras
probm <- matrix(probs, length(data$px1), length(data$px2))
cls <- contourLines(data$px1, data$px2, probm, levels=0.5)
rslt <- sapply(cls, lines, col='black')
}
plot_keras_preds(model)</pre>
```



```
plot_nnet_preds <- function(fit, dat=mixture.example) {
    ## create figure
    eval(plot_mixture_data)

## compute predictions from nnet
    probs <- prediction_nnet
    preds <- prediction_nnet
    probm <- matrix(probs, length(data$px1), length(data$px2))
    cls <- contourLines(data$px1, data$px2, probm, levels=0.5)
    rslt <- sapply(cls, lines, col='black')
}
plot_nnet_preds(model1)</pre>
```



From the plot we can see that the predictions are not similar. The points don't have much overlap.

Convert the neural network into CNN

```
fashion_mnist <- dataset_fashion_mnist()</pre>
c(train_images, train_labels) %<-% fashion_mnist$train</pre>
c(test_images, test_labels) %<-% fashion_mnist$test</pre>
img_rows <- 28</pre>
img_cols <- 28</pre>
train_images<- array_reshape(train_images, c(nrow(train_images), img_rows, img_cols, 1))</pre>
test_images <- array_reshape(test_images, c(nrow(test_images), img_rows, img_cols, 1))</pre>
train_images <- train_images/255</pre>
test_images <- test_images/255</pre>
class_names = c('T-shirt/top',
                  'Trouser',
                  'Pullover',
                  'Dress',
                  'Coat',
                  'Sandal',
                  'Shirt',
                  'Sneaker',
                  'Bag',
                  'Ankle boot')
train_labels <- to_categorical(train_labels, 10)</pre>
test_labels <- to_categorical(test_labels, 10)</pre>
```

```
model1 <- keras_model_sequential() %>%
  layer_conv_2d(filters = 32, kernel_size = c(3,3), activation = 'relu', input_shape = c(28,28,1)) %>%
  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_dropout(rate = 0.25) %>%
  layer_flatten() %>%
  layer_dense(units = 128, activation = 'relu') %>%
  layer_dropout(rate = 0.5) %>%
  layer_dense(units = 10, activation = 'softmax')
model1 %>% compile(
  loss = loss_categorical_crossentropy,
  optimizer = optimizer_adadelta(),
  metrics = c('accuracy')
)
history <- model1 %>%
  fit(
    x = train_images, y = train_labels,
    epochs = 10
plot(history)
0.55
       0
                                                                             o loss
0.45
               0
0.35
                                0
                                        0
                                                0
0.25
                                                        0
                                                                 0
                                                                         0
                                                                                 0
                                                                                 0
                                                                         0
                                                                 0
                                                        0
                                                0
                                        0
0.88
                                0
                        0
               0
0.84

    accuracy

               2
                                4
                                                6
                                                                 8
                                                                                 10
```