## HW9 GUO

Qing Guo 11/23/2019

```
setwd( "/Users/cheshibin/STAT_5014_2019_-qguo0701-")
#install.packages("devtools")
library(devtools)
## Loading required package: usethis
#devtools::install_github("rstudio/keras")
library(keras)
install_keras()
## Installation complete.
mnist<-dataset_mnist()</pre>
x train<-mnist$train$x</pre>
y train<-mnist$train$y</pre>
x_test<-mnist$test$x</pre>
y_test<-mnist$test$y</pre>
x_train<-array_reshape(x_train,c(nrow(x_train),784))</pre>
x_test<-array_reshape(x_test,c(nrow(x_test),784))</pre>
x_train<-x_train/255
x_test<-x_test/255
y_train<-to_categorical(y_train,10)</pre>
y_test<-to_categorical(y_test,10)</pre>
model<-keras_model_sequential()</pre>
model%>%
 layer_dense(units=256,activation="relu",input_shap=c(784))%>%
 layer dropout(rate=0.4)%>%
 layer_dense(units=128,activation="relu")%>%
 layer_dropout(rate=0.3)%>%
 layer_dense(units=10,activation="softmax")
summary(model)
## Model: "sequential"
## Layer (type)
                          Output Shape
                                                         Param #
## dense (Dense)
                                (None, 256)
                                                          200960
                               (None, 256)
## dropout (Dropout)
## dense_1 (Dense)
                               (None, 128)
                                                         32896
## dropout_1 (Dropout)
                             (None, 128)
## dense_2 (Dense) (None, 10)
                                            1290
## Total params: 235,146
## Trainable params: 235,146
```

```
## Non-trainable params: 0
model%>%compile(
   loss="categorical_crossentropy",
   optimizer=optimizer_rmsprop(),
   metrics=c("accuracy")
)
   history<-model%>%fit(
      x_train,y_train,
      epochs=30,batch_size=128,
      validation_split=0.2
plot(history)
       0.4 -
       0.3
       0.2 -
       0.1
                                                                                                                        data
                                                                                                                          training
     0.99 -
                                                                                                                               validation
     0.96
accuracy
     0.93
     0.90 -
     0.87 -
                             5
                                            10
                                                                            20
             0
                                                            15
                                                                                             25
                                                                                                             30
                                                           epoch
model %>% evaluate(x_test, y_test)
## $loss
## [1] 0.1142836
##
## $accuracy
## [1] 0.9796
model %>% predict_classes(x_test)
##
           [1] \ 7 \ 2 \ 1 \ 0 \ 4 \ 1 \ 4 \ 9 \ 5 \ 9 \ 0 \ 6 \ 9 \ 0 \ 1 \ 5 \ 9 \ 7 \ 3 \ 4 \ 9 \ 6 \ 6 \ 5 \ 4 \ 0 \ 7 \ 4 \ 0 \ 1 \ 3 \ 1 \ 3 \ 4
##
          [35] 7 2 7 1 2 1 1 7 4 2 3 5 1 2 4 4 6 3 5 5 6 0 4 1 9 5 7 8 9 3 7 4 6 4
           [69] \  \  \, 3\  \  \, 0\  \  \, 7\  \  \, 0\  \  \, 2\  \  \, 9\  \  \, 1\  \  \, 7\  \  \, 3\  \  \, 2\  \  \, 9\  \  \, 7\  \  \, 7\  \  \, 6\  \  \, 1\  \  \, 3\  \  \, 6\  \  \, 9\  \  \, 3\  \  \, 1\  \  \, 4\  \  \, 1\  \  \, 7\  \  \, 6\  \  \, 9\  \  \, 6\  \  \, 0
##
         \begin{smallmatrix} 103 \end{smallmatrix} \begin{smallmatrix} 5 & 4 & 9 & 9 & 2 & 1 & 9 & 4 & 8 & 7 & 3 & 9 & 7 & 4 & 4 & 4 & 9 & 2 & 5 & 4 & 7 & 6 & 7 & 9 & 0 & 5 & 8 & 5 & 6 & 6 & 5 & 7 & 8 & 1 \\ \end{smallmatrix}
##
```

```
[9317] 8 6 2 9 4 8 8 7 1 0 8 7 7 5 8 5 3 4 6 1 1 5 5 0 7 2 3 6 4 1 2 4 1 5
   [9351] 4 2 0 4 8 6 1 9 0 2 5 6 9 3 6 3 6 0 1 2 3 4 5 6 7 8 9 0 1 2 3 4 5 6
##
## [9385] 7 8 9 0 1 2 3 5 6 7 8 1 0 9 5 7 5 1 8 6 9 0 4 1 9 3 8 4 4 7 0 1 9 2
## [9419] 8 7 8 2 3 9 6 0 6 5 5 3 3 3 9 8 1 1 0 6 1 0 0 6 2 1 1 3 2 7 7 8 8 7
    [9453] 8 4 6 0 2 0 7 0 3 6 8 7 1 5 9 9 3 7 2 4 9 4 3 6 2 2 5 3 2 5 5 9 4 1
## [9487] 7 2 0 1 2 3 4 5 6 7 8 9 0 1 2 3 4 5 6 7 8 9 0 1 2 3 4 5 6 7 8 9 1 0
## [9521] 1 2 7 5 3 4 4 0 0 6 9 6 6 5 7 2 3 4 4 9 1 4 0 7 9 5 7 2 3 1 4 4 0 9
## [9555] 9 6 1 8 3 3 7 3 9 8 8 4 7 7 6 2 1 9 8 7 8 8 7 2 2 3 9 3 3 5 5 0 7 4
    [9589] 5 6 5 1 4 1 1 2 8 2 6 1 5 0 1 2 3 4 5 6 7 8 9 0 1 2 3 4 5 6 7 8 9 0
## [9623] 1 2 3 4 5 6 7 8 8 0 6 0 1 2 3 7 9 4 7 1 9 1 7 1 4 0 0 1 7 5 7 1 3 3
## [9657] 3 1 6 9 7 1 3 0 7 6 0 8 9 4 3 5 4 8 1 5 9 0 6 3 3 8 1 4 7 5 2 0 0 1
## [9691] 7 8 9 6 8 8 2 3 6 1 8 9 5 2 0 1 2 3 4 5 6 7 8 9 0 1 2 3 4 5 6 7 8 9
   [9725] 0 1 2 3 4 6 6 7 8 9 7 4 6 1 4 0 9 9 3 7 8 2 7 5 8 6 3 2 2 0 5 8 6 0
## [9759] 3 8 1 0 3 0 4 7 4 9 0 9 0 7 1 7 1 6 6 5 6 2 8 7 6 4 9 9 5 3 7 4 3 0
## [9793] 4 6 6 1 1 3 2 1 0 0 1 2 3 4 7 8 9 0 1 2 3 4 5 6 7 8 0 1 2 3 4 7 8 9
##
   [9827] \ 0 \ 8 \ 3 \ 9 \ 5 \ 5 \ 2 \ 6 \ 8 \ 4 \ 1 \ 7 \ 1 \ 7 \ 3 \ 5 \ 6 \ 9 \ 1 \ 1 \ 1 \ 2 \ 1 \ 2 \ 0 \ 7 \ 7 \ 5 \ 8 \ 2 \ 9 \ 8 \ 6 \ 7
## [9861] 3 4 6 8 7 0 4 2 7 7 5 4 3 4 2 8 1 5 1 0 2 3 3 5 7 0 6 8 6 3 9 9 8 2
## [9895] 7 7 1 0 1 7 8 9 0 1 2 3 4 5 6 7 8 0 1 2 3 4 7 8 9 7 8 6 4 1 9 3 8 4
## [9929] 4 7 0 1 9 2 8 7 8 2 6 0 6 5 3 3 9 9 1 4 0 6 1 0 0 6 2 1 1 7 7 8 4 6
   [9963] 0 7 0 3 6 8 7 1 5 2 4 9 4 3 6 4 1 7 2 6 6 0 1 2 3 4 5 6 7 8 9 0 1 2
## [9997] 3 4 5 6
fashion_mnist <- dataset_fashion_mnist()</pre>
c(train_images, train_labels) %<-% fashion_mnist$train
c(test_images, test_labels) %<-% fashion_mnist$test
class names = c('T-shirt/top',
                 'Trouser',
                 'Pullover'.
                 'Dress',
                 'Coat',
                'Sandal',
                'Shirt',
                'Sneaker',
                 'Bag',
                 'Ankle boot')
dim(train_images)
## [1] 60000
                       28
dim(train_labels)
## [1] 60000
library(tidyr)
library(ggplot2)
image_1 <- as.data.frame(train_images[1, , ])</pre>
colnames(image_1) <- seq_len(ncol(image_1))</pre>
image_1$y <- seq_len(nrow(image_1))</pre>
image_1 <- gather(image_1, "x", "value", -y)</pre>
image_1$x <- as.integer(image_1$x)</pre>
ggplot(image_1, aes(x = x, y = y, fill = value)) +
  geom_tile() +
  scale_fill_gradient(low = "white", high = "black", na.value = NA) +
  scale y reverse() +
  theme_minimal() +
  theme(panel.grid = element_blank())
```

```
theme(aspect.ratio = 1) +
xlab("") +
ylab("")
```

0



```
train_images <- train_images / 255
test_images <- test_images / 255
par(mfcol=c(5,5))
par(mar=c(0, 0, 1.5, 0), xaxs='i', yaxs='i')
for (i in 1:25) {
   img <- train_images[i, , ]
   img <- t(apply(img, 2, rev))
   image(1:28, 1:28, img, col = gray((0:255)/255), xaxt = 'n', yaxt = 'n',
        main = paste(class_names[train_labels[i] + 1]))
}</pre>
```

