## HW9\_Zhengzhi

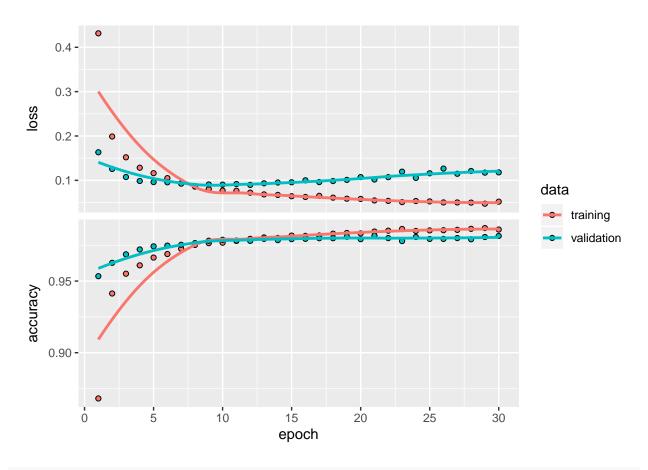
zhengzhi lin 2019.11.17

```
library(keras)
mnist <- dataset mnist()</pre>
x train <- mnist$train$x
y_train <- mnist$train$y</pre>
x_test <- mnist$test$x</pre>
y_test <- mnist$test$y</pre>
# reshape
x_train <- array_reshape(x_train, c(nrow(x_train), 784))</pre>
x_test <- array_reshape(x_test, c(nrow(x_test), 784))</pre>
# rescale
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_shape = 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
## dropout (Dropout)
                             (None, 256)
## 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)
```



```
model %>% evaluate(x_test, y_test)
```

```
## $loss
## [1] 0.1089825
##
## $accuracy
## [1] 0.9816
```

## model %>% predict\_classes(x\_test)

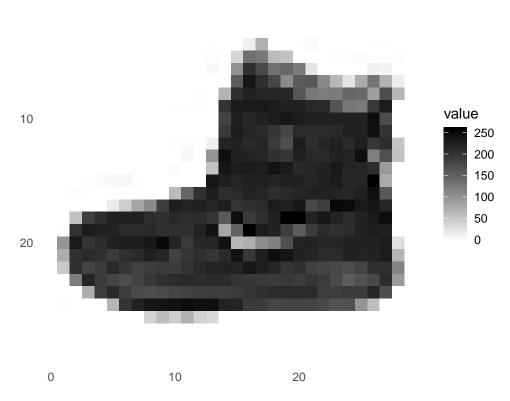
```
## [1] 7 2 1 0 4 1 4 9 5 9 0 6 9 0 1 5 9 7 3 4 9 6 6 5 4 0 7 4 0 1 3 1 3 4 ## [35] 7 2 7 1 2 1 1 7 4 2 3 5 1 2 4 4 6 3 5 5 6 0 4 1 9 5 7 8 9 3 7 4 6 4 ## [69] 3 0 7 0 2 9 1 7 3 2 9 7 7 6 2 7 8 4 7 3 6 1 3 6 9 3 1 4 1 7 6 9 6 0
```

```
[9283] 1 3 4 7 8 9 7 5 5 1 9 9 7 1 0 0 5 9 7 1 7 2 2 3 6 8 3 2 0 0 6 1 7 5
##
   [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 2 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 0 7 5 8 5 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] 7 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 \ 8 \ 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 3 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
                       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$y <- seq_len(nrow(image_1))
image_1 <- gather(image_1, "x", "value", -y)
image_1$x <- as.integer(image_1$x)

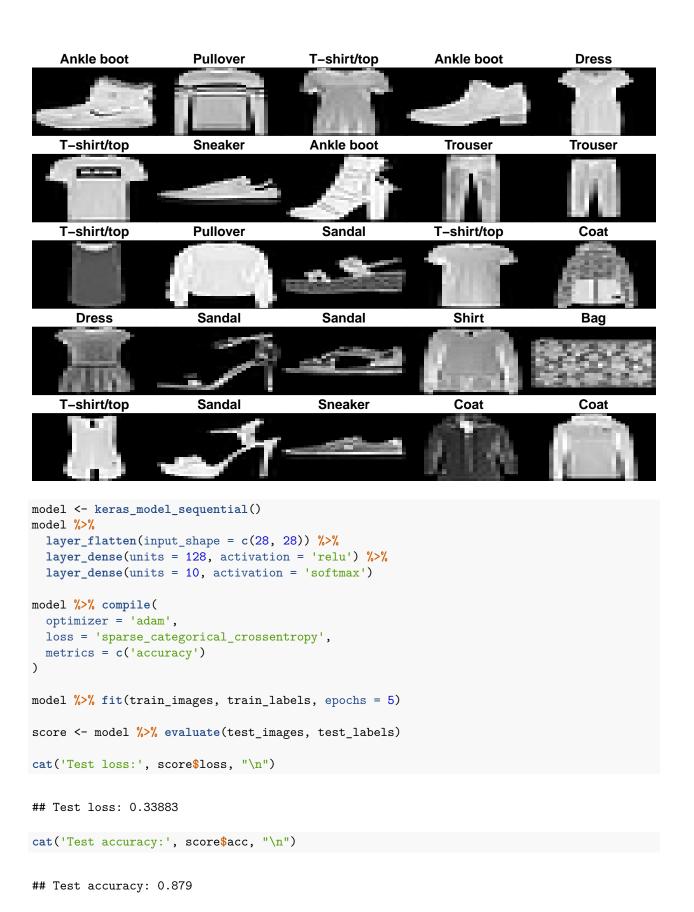
ggplot(image_1, aes(x = x, y = y, fill = value)) +
    geom_tile() +
    scale_fill_gradient(low = "white", high = "black", na.value = NA) +
    scale_y_reverse() +
    theme_minimal() +
    theme(panel.grid = element_blank()) +
    theme(aspect.ratio = 1) +
    xlab("") +
    ylab("")</pre>
```

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>
```



predictions <- model %>% predict(test\_images)