HW9_XU

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```
#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
y_train<-mnist$train$y</pre>
x test<-mnist$test$x
y_test<-mnist$test$y</pre>
x_train<-array_reshape(x_train,c(nrow(x_train),784))
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()
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
## ______
## dropout (Dropout)
                             (None, 256)
## dense_1 (Dense)
                             (None, 128)
                                                      32896
## ______
## dropout_1 (Dropout) (None, 128) 0
## dense 2 (Dense) (None, 10)
## 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
                                                                                                              30
                                                             15
                                                                                             25
                                                           epoch
model %>% evaluate(x_test, y_test)
## $loss
## [1] 0.1111397
##
## $accuracy
## [1] 0.981
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} 
##
##
         \begin{smallmatrix} 137 \end{smallmatrix} ] \ 0 \ 1 \ 6 \ 4 \ 6 \ 7 \ 3 \ 1 \ 7 \ 1 \ 8 \ 2 \ 0 \ 2 \ 9 \ 9 \ 5 \ 5 \ 1 \ 5 \ 6 \ 0 \ 3 \ 4 \ 4 \ 6 \ 5 \ 4 \ 6 \ 5 \ 4 \ 5 \ 1 \ 4
```

```
[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 5 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 2 7 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 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] 1 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
 \hbox{\tt \#\#} \quad [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 0 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
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("")
```



