## AE-MNIST.R

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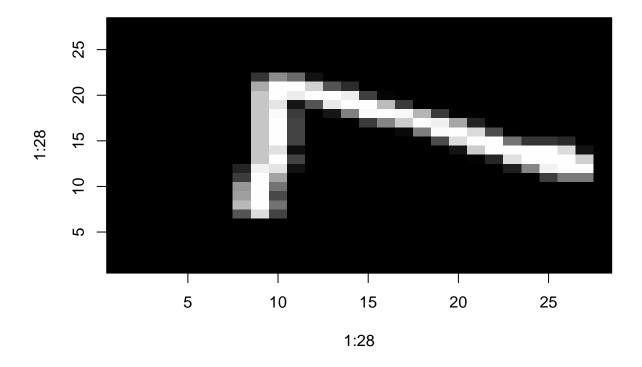
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```
# Autoenconder (AE) digits from MNIST dataset
library(keras)
#### Data
mnist <- dataset_mnist()</pre>
## Loaded Tensorflow version 2.7.0
x_train <- mnist$train$x</pre>
y_train <- mnist$train$y</pre>
x_test <- mnist$test$x</pre>
y_test <- mnist$test$y</pre>
# reshape
dim(x_train) <- c(nrow(x_train), 784)</pre>
dim(x_test) \leftarrow c(nrow(x_test), 784)
################ Selection several digits
cifra <- c(0:9)
x_train_cifra<-x_train[which(y_train %in% cifra),]</pre>
y_train_cifra<-y_train[which(y_train %in% cifra)]</pre>
x_test_cifra<-x_test[which(y_test %in% cifra),]</pre>
y_test_cifra<-y_test[which(y_test %in% cifra)]</pre>
dim(x_train_cifra)
## [1] 60000
length(y_train_cifra)
## [1] 60000
dim(x_test_cifra)
## [1] 10000
                784
length(y_test_cifra)
## [1] 10000
```

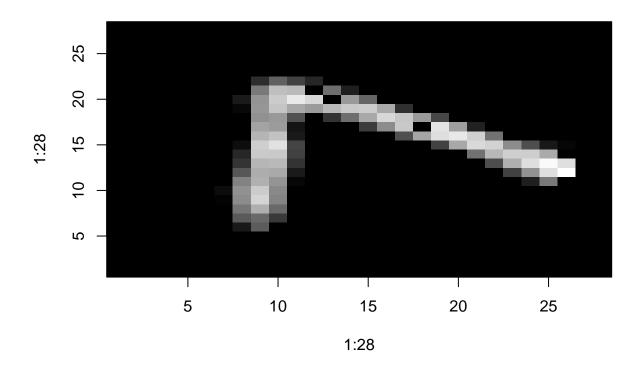
```
sort(unique(y_train_cifra))
## [1] 0 1 2 3 4 5 6 7 8 9
sort(unique(y_test_cifra))
## [1] 0 1 2 3 4 5 6 7 8 9
########## rescale
x_train_cifra <- x_train_cifra / 255</pre>
x_test_cifra <- x_test_cifra / 255</pre>
#y_train_cifra <- to_categorical(y_train_cifra, 10)</pre>
#y_test_cifra <- to_categorical(y_test_cifra, 10)</pre>
######################### Autoencoder
#### Encoder
model_enc <- keras_model_sequential()</pre>
model_enc %>%
 layer_dense(units = 128, activation = "relu", input_shape = ncol(x_train)) %>%
 layer_dense(units = 64, activation = "relu") %>%
 layer_dense(units = 32, activation = "relu")
summary(model_enc)
## Model: "sequential"
## Layer (type)
                           Output Shape
                                                         Param #
## -----
## dense_2 (Dense)
                                (None, 128)
                                                           100480
##
## dense_1 (Dense)
                                (None, 64)
                                                           8256
                                                           2080
## dense (Dense)
                                (None, 32)
##
## Total params: 110,816
## Trainable params: 110,816
## Non-trainable params: 0
#### Decoder
model_dec <- keras_model_sequential()</pre>
model dec %>%
 layer_dense(units = 64, activation = "relu", input_shape = c(32)) %>%
 layer dense(units = 128, activation = "relu") %>%
 layer_dense(units = ncol(x_train), activation = "relu")
summary(model_dec)
## Model: "sequential_1"
## Layer (type)
                        Output Shape
                                                       Param #
```

```
dense 5 (Dense)
                      (None, 64)
                                        2112
##
 dense 4 (Dense)
                      (None, 128)
                                        8320
##
##
## dense_3 (Dense)
                      (None, 784)
                                        101136
##
## Total params: 111,568
## Trainable params: 111,568
## Non-trainable params: 0
## ______
#### Autoencoder
model<-keras_model_sequential()</pre>
model %>%model_enc%>%model_dec
## Model: "sequential_2"
## Layer (type)
                    Output Shape
## sequential (Sequential)
                     (None, 32)
                                        110816
##
## sequential_1 (Sequential)
                      (None, 784)
                                        111568
##
## Total params: 222,384
## Trainable params: 222,384
## Non-trainable params: 0
summary(model)
## Model: "sequential_2"
## Layer (type)
                 Output Shape
## -----
## sequential (Sequential)
                    (None, 32)
                                        110816
##
## sequential_1 (Sequential)
                     (None, 784)
                                        111568
## Total params: 222,384
## Trainable params: 222,384
## Non-trainable params: 0
## ______
model %>% compile(
loss = "mean_squared_error",
 #optimizer = optimizer_rmsprop(),
optimizer = "adam",
```

```
metrics = c("mean_squared_error")
)
history <- model %>% fit(
  x= x_train_cifra, y = x_train_cifra, # Autoencoder
  epochs = 15, batch_size = 128,
 validation_split = 0.2
###### Prediction
# Autoencoder
output_cifra<-predict(model,x_test_cifra)</pre>
dim(output_cifra)
## [1] 10000
# From input to encoder
enc_output_cifra<-predict(model_enc,x_test_cifra)</pre>
dim(enc_output_cifra)
## [1] 10000
                32
# From encoder to decoder
dec_output_cifra<-predict(model_dec,enc_output_cifra)</pre>
dim(dec_output_cifra)
## [1] 10000
# Check
idx<-1
#x_test_cifra[idx,]
im<-matrix(x_test_cifra[idx,], nrow=28, ncol=28)</pre>
image(1:28, 1:28, im, col=gray((0:255)/255))
```



## #output\_cifra[idx,] im<-matrix(output\_cifra[idx,], nrow=28, ncol=28) image(1:28, 1:28, im, col=gray((0:255)/255))</pre>



```
#dec_output_cifra[idx,]
im<-matrix(dec_output_cifra[idx,], nrow=28, ncol=28)
image(1:28, 1:28, im, col=gray((0:255)/255))

#which.max(enc_output_cifra[idx,])
#which.min(enc_output_cifra[idx,])
#which(enc_output_cifra[idx,]==cifra)

# Encoder results
im<-matrix(enc_output_cifra[idx,], nrow=8, ncol=4)
image(1:8, 1:4, im, col=gray((0:255)/255))</pre>
```

