Final Report

Sabir Meah, Michael Miller, Liyang Yuan

2022-12-16

```
set.seed(7423)
```

Data Read-In

```
no dementia \leftarrow array(NA, dim = c(3200, 128, 128))
for (i in 1:3200){
  no_dementia[i,,] <- readJPEG(paste0(getwd(), "/Data/Non_Demented/non_", i, ".jpg"))</pre>
verymild_dementia \leftarrow array(NA, dim = c(2240, 128, 128))
for (i in 1:2240){
  verymild_dementia[i,,] <- readJPEG(pasteO(getwd(), "/Data/Very_Mild_Demented/verymild_", i, ".jpg"))</pre>
mild_dementia \leftarrow array(NA, dim = c(896, 128, 128))
for (i in 1:896){
  mild_dementia[i,,] <- readJPEG(paste0(getwd(), "/Data/Mild_Demented/mild_", i, ".jpg"))</pre>
}
moderate_dementia <- array(NA, dim = c(64, 128, 128))
for (i in 1:64){
  moderate_dementia[i,,] <- readJPEG(paste0(getwd(), "/Data/Moderate_Demented/moderate_", i, ".jpg"))</pre>
}
lenNoDem <- 3200
lenVeryMildDem <- 2240</pre>
lenMildDem <- 896
lenModDem <- 64
totalLen <- lenNoDem + lenVeryMildDem + lenMildDem + lenModDem
class <- c(rep("No Dementia", lenNoDem),</pre>
           rep("Very Mild Dementia", lenVeryMildDem),
           rep("Mild Dementia", lenMildDem),
           rep("Moderate Dementia", lenModDem))
#dementia_data <- list(no_dementia, verymild_dementia, mild_dementia, moderate_dementia)
dementia_data \leftarrow array(NA, dim = c(6400, 128, 128))
dementia_data[1:lenNoDem,,] <- no_dementia</pre>
dementia_data[3201:(lenNoDem + lenVeryMildDem),,] <- verymild_dementia</pre>
dementia_data[5441:(lenNoDem + lenVeryMildDem + lenMildDem),,] <- mild_dementia
dementia_data[6337:totalLen,,] <- moderate_dementia</pre>
```

Model: 4 classes with one hot encoding

```
lenNoDem <- 3200
lenVeryMildDem <- 2240</pre>
lenMildDem <- 896
lenModDem <- 64
img rows <- 128
img_cols <- 128
totalLen <- lenNoDem+lenVeryMildDem+lenMildDem+lenModDem
noDemClass <- cbind(rep(1,lenNoDem),rep(0,lenNoDem),rep(0,lenNoDem),rep(0,lenNoDem))
veryMildDemClass <- cbind(rep(0,lenVeryMildDem),rep(1,lenVeryMildDem),rep(0,lenVeryMildDem),rep(0,lenVeryMildDem)
mildDemClass <- cbind(rep(0,lenMildDem),rep(0,lenMildDem),rep(1,lenMildDem),rep(0,lenMildDem))
modDemClass <- cbind(rep(0,lenModDem),rep(0,lenModDem),rep(0,lenModDem)),rep(1,lenModDem))</pre>
classMat <- rbind(noDemClass,veryMildDemClass,mildDemClass,modDemClass)</pre>
noDemTestIdx <- sample(lenNoDem, round(lenNoDem/4.0), replace = FALSE)
noDemTrainIdx <- setdiff(1:lenNoDem, noDemTestIdx)</pre>
veryMildDemTestIdx <- sample((lenNoDem+1):(lenNoDem+lenVeryMildDem),</pre>
                                round(lenVeryMildDem/4.0), replace = FALSE)
veryMildDemTrainIdx <- setdiff((lenNoDem+1):(lenNoDem+lenVeryMildDem),</pre>
                                veryMildDemTestIdx)
mildDemTestIdx <- sample((lenNoDem+lenVeryMildDem+1):</pre>
                            (lenNoDem+lenVeryMildDem+lenMildDem),
                            round(lenMildDem/4.0), replace = FALSE)
mildDemTrainIdx <- setdiff((lenNoDem+lenVeryMildDem+1):</pre>
                            (lenNoDem+lenVeryMildDem+lenMildDem),
                            mildDemTestIdx)
modDemTestIdx <- sample((lenNoDem+lenVeryMildDem+lenMildDem+1):</pre>
                          (lenNoDem+lenVeryMildDem+lenMildDem+lenModDem),
                          round(lenModDem/4.0), replace = FALSE)
modDemTrainIdx <- setdiff((lenNoDem+lenVeryMildDem+lenMildDem+1):</pre>
                            (lenNoDem+lenVeryMildDem+lenMildDem+lenModDem),
                           modDemTestIdx)
x_train <- dementia_data[c(noDemTrainIdx,veryMildDemTrainIdx,</pre>
                             mildDemTrainIdx, modDemTrainIdx),,]
y_train <- classMat[c(noDemTrainIdx,veryMildDemTrainIdx,</pre>
                       mildDemTrainIdx,modDemTrainIdx),]
x_test <- dementia_data[c(noDemTestIdx,veryMildDemTestIdx,</pre>
                            mildDemTestIdx,modDemTestIdx),,]
y_test <- classMat[c(noDemTestIdx,veryMildDemTestIdx,</pre>
                      mildDemTestIdx,modDemTestIdx),]
shuffleIdx <- sample(1:round(3.0*totalLen/4.0))</pre>
x_train <- x_train[shuffleIdx,,]</pre>
y_train <- y_train[shuffleIdx,]</pre>
x_train <- array_reshape(x_train, c(nrow(x_train), img_rows, img_cols, 1))</pre>
```

```
x_test <- array_reshape(x_test, c(nrow(x_test), img_rows, img_cols, 1))</pre>
input_shape <- c(img_rows, img_cols, 1)</pre>
batch_size <- 32
num_classes <- 4</pre>
epochs <- 100
cnn_model <- keras_model_sequential() %>%
  layer_conv_2d(filters = 64, kernel_size = c(3,3), activation = 'relu', input_shape = input_shape) %>%
  layer_max_pooling_2d(pool_size = c(2, 2)) %>%
  layer_dropout(rate = 0.25) %>%
  layer_conv_2d(filters = 64, kernel_size = c(3,3), activation = 'relu', input_shape = input_shape) %>%
  layer_max_pooling_2d(pool_size = c(2, 2)) %>%
  layer_dropout(rate = 0.25) %>%
  layer_flatten() %>%
  layer_dense(units = 128, activation = 'relu', kernel_regularizer=regularizer_11_12(11=1e-4,12=1e-5),)
  layer dropout(rate = 0.25) %>%
  layer_dense(units = num_classes, activation = 'softmax')
## Loaded Tensorflow version 2.9.3
cnn_model %>% compile(
 loss = loss_categorical_crossentropy,
 optimizer = optimizer_adam(),
 metrics = c('accuracy')
cnn_history <- cnn_model %>% fit(
 x_train, y_train,
 batch_size = batch_size,
 epochs = epochs,
  validation_split = 0.2
```

Process Model Results

```
predictProbs = predict(cnn_model, x_test)
predictClass <- matrix(NA, nrow = nrow(x_test), ncol = num_classes)
for (i in 1:nrow(x_test)) {
    classVec <- rep(0, num_classes)
        classVec[which(predictProbs[i,] == max(predictProbs[i,]))] <- 1
        predictClass[i,] <- classVec
}

testAccuracyNoDem <- 0.0
testAccuracyVeryMildDem <- 0.0
testAccuracyMildDem <- 0.0
testAccuracyModDem <- 0.0
diffMat <- y_test - predictClass

for (i in 1:length(noDemTestIdx)) {
    testAccuracyNoDem <- testAccuracyNoDem+ifelse(min(diffMat[i,]) == 0,1,0)</pre>
```

| No.Dementia | Very.Mild.Dementia | Mild.Dementia | Moderate.Dementia |
|-------------|--------------------|---------------|-------------------|
| 0.97875 | 0.9464286 | 0.9598214 | 1 |