## HW5

## 2024-11-30

```
library(data.table)
set.seed(123)
load_digits <- function(subset=NULL, normalize=TRUE) {</pre>
  df <- fread("digits.csv")</pre>
  df <- as.matrix(df)</pre>
  if (length(subset) > 0) {
    c <- dim(df)[2]
    1_col <- df[, c]</pre>
    index <- NULL</pre>
    for (i in 1:length(subset)){
      number <- subset[i]</pre>
      index <- c(index, which(l_col == number))</pre>
    df <- df[sort(index), ]</pre>
  digits \leftarrow df[, -1]
  labels <- df[, c]</pre>
  if (normalize) {
    digits <- digits - min(digits)</pre>
    digits <- digits / max(digits)</pre>
  for (i in 1:length(subset)) {
    labels[labels == subset[i]] <- i - 1</pre>
 return(list(digits, labels))
result <- load_digits(subset=c(1, 7), normalize=TRUE)</pre>
digits <- result[[1]]</pre>
labels <- result[[2]]</pre>
split_samples <- function(digits, labels) {</pre>
  num_samples <- dim(digits)[1]</pre>
  num_training <- round(num_samples * 0.7)</pre>
  indices <- sample(1:num_samples, size = num_samples)</pre>
  training_idx <- indices[1:num_training]</pre>
  testing_idx <- indices[-(1:num_training)]</pre>
  return(list(digits[training_idx, ], labels[training_idx], digits[testing_idx, ], labels[testing_idx])
}
result <- split_samples(digits, labels)</pre>
training_digits <- result[[1]]</pre>
training_labels <- result[[2]]</pre>
```

```
testing_digits <- result[[3]]</pre>
testing_labels <- result[[4]]</pre>
linear_svm <- function(training_digits, training_labels, testing_digits, testing_labels, C = 1) {</pre>
  n <- nrow(training_digits)</pre>
  d <- ncol(training_digits)</pre>
  w \leftarrow matrix(0, nrow = d, ncol = 1)
  for (epoch in 1:100) {
    for (i in 1:n) {
      # Compute the margin
      margin <- training_labels[i] * (as.vector(t(w)) %*% training_digits[i, ] + b)</pre>
      if (margin < 1) {</pre>
        w <- w + C * (training_labels[i] * matrix(training_digits[i, ], nrow = d, ncol = 1))
        b <- b + C * training_labels[i]</pre>
      }
    }
  }
  predict <- function(x) {</pre>
    if (as.vector(t(w)) %*% x + b >= 0) {
      return(1)
    } else {
      return(0)
    }
  }
  train_predictions <- apply(training_digits, 1, predict)</pre>
  test_predictions <- apply(testing_digits, 1, predict)</pre>
  training_accuracy <- mean(train_predictions == training_labels)</pre>
  testing_accuracy <- mean(test_predictions == testing_labels)</pre>
  return(list(training_accuracy = training_accuracy, testing_accuracy = testing_accuracy))
}
model_results <- linear_svm(training_digits, training_labels, testing_digits, testing_labels, C = 1)</pre>
cat("Training Accuracy: ", model_results$training_accuracy, "\n")
## Training Accuracy: 0.4782609
cat("Testing Accuracy: ", model_results$testing_accuracy, "\n")
## Testing Accuracy: 0.537037
C_{\text{values}} \leftarrow c(0.01, 0.1, 1, 10, 100)
results_c <- data.frame(C = C_values, Training_Accuracy = NA), Testing_Accuracy = NA)
for (i in 1:length(C_values)) {
  model_results <- linear_svm(training_digits, training_labels, testing_digits, testing_labels, C = C_v
  results_c$Training_Accuracy[i] <- model_results$training_accuracy</pre>
  results_c$Testing_Accuracy[i] <- model_results$testing_accuracy</pre>
}
print(results_c)
```

```
C Training_Accuracy Testing_Accuracy
                   0.4782609
                                       0.537037
## 1 1e-02
## 2 1e-01
                   0.4782609
                                       0.537037
## 3 1e+00
                   0.4782609
                                       0.537037
## 4 1e+01
                   0.4782609
                                       0.537037
## 5 1e+02
                   0.4782609
                                       0.537037
poly_kernel <- function(x1, x2, degree = 2) {</pre>
 return((x1 %*% t(x2) + 1) ^ degree)
kernel_svm <- function(training_digits, training_labels, testing_digits, testing_labels, C = 1) {</pre>
  n <- nrow(training_digits)</pre>
  alpha \leftarrow rep(0, n)
  threshold <- 0.001
  max iter <- 100
  degree <- 2
  for (iter in 1:max_iter) {
    for (i in 1:n) {
      output <- sum(alpha * training_labels * poly_kernel(training_digits, training_digits[i, , drop = ]
      error <- training_labels[i] - output</pre>
      if (!is.na(error) && abs(error) > threshold) {
        alpha[i] <- alpha[i] + C * error
    }
  }
  predict_kernel <- function(x) {</pre>
    output <- sum(alpha * training_labels * poly_kernel(training_digits, t(x), degree))</pre>
    return(ifelse(output >= 0, 1, 0))
  train_predictions <- apply(training_digits, 1, predict_kernel)</pre>
  test_predictions <- apply(testing_digits, 1, predict_kernel)</pre>
  training_accuracy <- mean(train_predictions == training_labels)</pre>
  testing_accuracy <- mean(test_predictions == testing_labels)</pre>
  return(list(training_accuracy = training_accuracy, testing_accuracy = testing_accuracy))
model_results_kernel <- kernel_svm(training_digits, training_labels, testing_digits, testing_labels, C
cat("Kernel SVM Training Accuracy: ", model_results_kernel$training_accuracy, "\n")
## Kernel SVM Training Accuracy: NA
cat("Kernel SVM Testing Accuracy: ", model_results_kernel$testing_accuracy, "\n")
```

## Kernel SVM Testing Accuracy: NA