Homework 4

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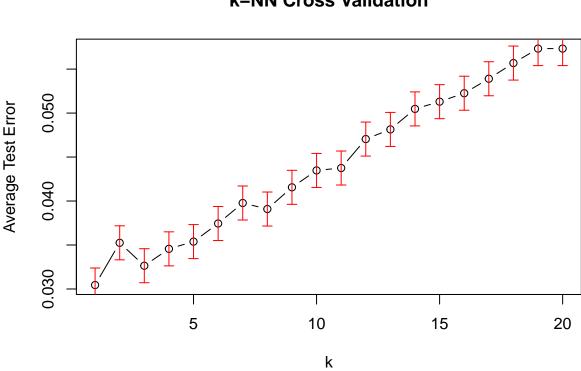
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```
train_data <- read.table("/Users/yuningli/Desktop/zip.train.gz")</pre>
test data<- read.table("/Users/yuningli/Desktop/zip.test.gz")</pre>
library(class)
library(caret)
## Loading required package: ggplot2
## Loading required package: lattice
# Extract labels and features
train_labels <- train_data[, 1]</pre>
train_features <- train_data[, -1]</pre>
test_labels <- test_data[, 1]</pre>
test_features <- test_data[, -1]</pre>
# Function to normalize features (if needed)
normalize <- function(data) {</pre>
 (data - min(data)) / (max(data) - min(data))
}
# Normalize features to [0, 1] (assuming grayscale values)
train_features_normalized <- apply(train_features, 2, normalize)</pre>
test_features_normalized <- apply(test_features, 2, normalize)</pre>
# Train k-NN model
k \leftarrow 5 # You can adjust the value of k
knn_model <- knn(train_features_normalized, test_features_normalized, train_labels, k = k)
# Evaluate the model
error_rate <- mean(knn_model != test_labels)</pre>
cat('Error rate:', error_rate, '\n')
## Error rate: 0.05480817
# Extract labels and features
labels <- train data[, 1]</pre>
features <- train_data[, -1]</pre>
# Convert labels to factor
labels <- as.factor(labels)</pre>
# Set up control parameters for 5-fold cross-validation
```

```
folds <- createFolds(labels, k = 5, list = TRUE)</pre>
k_values <- 1:20
\# Perform k-fold cross-validation
cv_errors <- numeric(length(k_values))</pre>
for (i in seq_along(k_values)) {
 k <- k_values[i]</pre>
  current_errors <- numeric(length(folds))</pre>
  for (j in seq_along(folds)) {
    train_indices <- unlist(folds[-j])</pre>
    test_indices <- folds[[j]]</pre>
    knn_model <- knn(features[train_indices, ], features[test_indices, ], labels[train_indices], k = k)</pre>
    current_errors[j] <- mean(knn_model != labels[test_indices])</pre>
  cv_errors[i] <- mean(current_errors)</pre>
# Print the estimated average test error for each 'k'
print(data.frame(k = k_values, avg_error = cv_errors))
##
      k avg_error
## 1
     1 0.03044854
## 2 2 0.03524917
## 3
      3 0.03264304
      4 0.03456320
## 4
## 5
      5 0.03538653
      6 0.03744367
## 6
## 7
      7 0.03977563
## 8 8 0.03908986
## 9
       9 0.04155880
## 10 10 0.04347868
## 11 11 0.04375303
## 12 12 0.04704474
## 13 13 0.04814232
## 14 14 0.05047447
## 15 15 0.05129742
## 16 16 0.05225764
## 17 17 0.05390354
## 18 18 0.05568662
## 19 19 0.05733233
## 20 20 0.05733233
# Calculate standard errors
se <- sd(cv_errors) / sqrt(length(cv_errors))</pre>
# Plot the estimated average test error as a function of 'k' with error bars
plot(k_values, cv_errors, type = 'b', xlab = 'k', ylab = 'Average Test Error', main = 'k-NN Cross Valid
```

```
# Add error bars
arrows(k_values, cv_errors - se, k_values, cv_errors + se, angle = 90, code = 3, length = 0.05, col = ":
```

k-NN Cross Validation



```
# Find the index of the minimum cross-validated error
min_error_index <- which.min(cv_errors)</pre>
# Identify the minimum error and its associated 'k' value
min_error <- cv_errors[min_error_index]</pre>
best_k <- k_values[min_error_index]</pre>
# Calculate the standard error
se <- sd(cv_errors) / sqrt(length(cv_errors))</pre>
# Identify the largest 'k' within one standard error of the minimum
selected_k <- k_values[cv_errors <= min_error + se][1]</pre>
cat('Minimum Cross-validated Error:', min_error, '\n')
## Minimum Cross-validated Error: 0.03044854
cat('Best k Value:', best_k, '\n')
## Best k Value: 1
cat('Selected k using One-Standard Error Rule:', selected_k, '\n')
## Selected k using One-Standard Error Rule: 1
```

```
# Extract labels and features
train_labels <- train_data[, 1]</pre>
train features <- train data[, -1]
# Convert labels to factor
train_labels <- as.factor(train_labels)</pre>
# Fit the final k-NN model using the full training dataset and 'k=1'
final_k <- 1 # Use the selected 'k' from the one-standard error rule
final_knn_model <- knn(train_features, train_features, train_labels, k = final_k)
# Print the final k-NN model details
cat('Final k-NN Model (k =', final_k, '):\n')
## Final k-NN Model (k = 1):
print(final knn model)
##
      [1] 6 5 4 7 3 6 3 1 0 1 7 0 1 1 7 7 4 8 0 1 4 8 7 4 8 7 3 7 4 1 3 6 7 4 1 3 7
##
     [38] 7 4 5 4 2 7 4 1 3 7 7 4 0 6 3 2 0 8 6 6 2 0 8 7 8 2 0 9 0 2 2 0 8 1 2 0 8
##
     [75] 3 3 2 8 2 2 0 8 1 4 4 8 9 8 9 6 7 6 1 9 7 0 8 0 4 6 8 0 0 3 0 8 0 9 0 3 8
     \begin{smallmatrix} [112] \end{smallmatrix} 0 1 2 2 9 0 6 6 5 9 2 0 9 1 4 1 2 7 1 0 9 0 8 0 7 9 1 3 0 4 4 3 5 1 6 8 5 
   [149] \ 4 \ 4 \ 6 \ 8 \ 4 \ 4 \ 8 \ 6 \ 4 \ 0 \ 2 \ 3 \ 9 \ 8 \ 6 \ 8 \ 9 \ 3 \ 5 \ 6 \ 8 \ 0 \ 2 \ 2 \ 6 \ 8 \ 4 \ 1 \ 0 \ 2 \ 7 \ 1 \ 0 \ 2 \ 2 \ 7 \ 1
##
   [186] \ 0 \ 9 \ 2 \ 7 \ 0 \ 4 \ 8 \ 0 \ 8 \ 7 \ 2 \ 7 \ 1 \ 3 \ 2 \ 7 \ 3 \ 2 \ 2 \ 2 \ 7 \ 1 \ 0 \ 2 \ 2 \ 8 \ 5 \ 4 \ 2 \ 2 \ 7 \ 8 \ 7 \ 0 \ 2 \ 7 \ 1
##
##
    [223] 0 2 6 0 2 7 6 0 8 2 7 1 1 0 1 7 7 6 4 4 6 2 9 1 1 9 0 3 1 1 8 1 0 3 1 6 1
## [260] 1 7 5 5 3 6 5 0 6 8 8 0 0 0 1 4 9 5 7 0 0 0 1 6 0 1 9 8 6 0 1 9 8 6 8 0 0
## [297] 9 6 8 5 4 4 9 4 0 8 6 4 8 1 2 1 2 1 2 6 8 0 2 1 3 0 2 3 2 3 5 8 0 2 7 9 0
## [334] 0 0 0 6 8 9 7 7 9 4 9 1 2 9 4 5 3 8 5 7 1 8 0 1 4 6 5 5 3 2 0 1 0 6 6 6 6
##
    [371] 0 9 0 2 6 0 5 5 9 6 1 7 7 7 6 1 8 2 0 6 0 5 2 1 6 1 8 6 6 6 0 9 0 2 6 0 2
## [408] 5 6 0 9 0 2 6 0 5 2 1 6 0 5 4 6 6 4 7 3 6 1 8 2 1 6 0 9 0 2 6 0 4 3 5 6 0
## [445] 5 6 1 8 2 0 6 0 9 0 2 6 1 8 0 1 6 0 5 2 2 6 1 7 6 1 6 0 4 7 3 1 8 2 6 0 9
## [482] 0 2 6 1 7 0 1 6 0 5 2 2 0 9 2 7 6 0 9 0 2 6 1 8 2 1 6 1 9 1 3 6 0 9 0 2 6
   [519] 1 9 0 0 4 6 2 6 1 8 0 1 4 6 9 9 2 9 7 1 0 2 8 0 7 0 2 9 7 1 0 2 8 2 3 4 2
## [556] 8 2 1 2 2 8 6 5 8 2 8 7 3 9 2 8 2 1 7 2 8 2 7 2 0 0 9 2 2 8 3 6 2 8 3 0 3
## [593] 2 8 6 0 1 2 8 1 0 2 8 2 6 5 2 8 2 6 2 8 4 3 2 8 2 3 2 8 1 1 2 3 2 2 6 2 1
   [630] \ 2 \ 2 \ 2 \ 1 \ 1 \ 6 \ 2 \ 7 \ 0 \ 2 \ 2 \ 4 \ 0 \ 1 \ 2 \ 3 \ 4 \ 5 \ 2 \ 3 \ 0 \ 6 \ 0 \ 9 \ 3 \ 0 \ 0 \ 3 \ 0 \ 0 \ 8 \ 8 \ 3 \ 0 \ 0 \ 8 \ 3
##
    [667] 3 0 2 4 0 3 0 8 1 5 3 0 2 7 4 3 0 8 0 9 3 0 0 3 0 2 3 0 2 4 7 8 1 1 1 1 2
## [704] 1 3 7 2 1 0 3 8 0 3 7 1 3 0 8 5 1 6 8 5 7 0 8 9 1 1 9 8 0 0 0 5 2 0 0 1 1
## [741] 2 0 0 3 7 5 9 4 0 2 9 7 4 1 0 2 1 3 0 0 2 2 1 5 0 2 1 0 9 0 2 1 4 6 0 1 7
##
   [778] 2 0 0 1 9 0 6 0 2 1 3 8 0 1 9 7 0 0 1 8 4 1 0 1 3 6 4 0 2 6 0 1 0 1 8 3 0
## [815] 9 1 0 4 9 8 0 7 2 0 0 6 5 0 9 0 0 6 3 0 0 0 9 0 7 0 0 9 3 4 4 0 6 9 0 0 7
## [852] 0 8 0 6 6 0 0 9 1 8 6 5 4 0 0 6 0 4 0 0 7 5 8 0 0 2 3 0 0 6 6 0 0 9 0 7 0
## [889] 0 9 2 3 0 0 9 1 2 0 0 6 9 7 0 7 5 4 2 3 5 5 7 1 8 8 6 0 0 1 8 0 2 0 2 0 4
##
    [926] 8 2 8 0 6 0 2 1 3 2 9 7 1 0 4 0 1 8 5 0 2 3 8 5 3 4 5 8 9 1 0 7 3 8 1 5 2
## [963] 4 6 2 0 6 0 1 0 7 4 6 9 1 2 2 3 2 0 4 8 8 3 8 6 6 6 1 4 9 7 0 0 6 9 7 4 0
## [1000] 5 9 8 0 0 2 9 0 9 8 4 0 8 4 9 5 5 0 4 8 2 3 2 2 6 9 8 4 8 0 8 4 0 7 2 0 8
## [1037] 5 0 0 6 5 3 7 0 6 9 2 5 7 6 5 7 1 9 6 5 4 6 8 5 4 4 6 8 0 0 9 6 8 8 0 3 6
## [1074] 8 1 1 4 8 5 0 2 7 8 5 7 1 8 8 5 2 2 7 1 1 8 9 3 1 9 0 1 1 9 9 7 7 1 9 7 1
## [1111] 4 1 9 1 1 9 7 1 4 1 9 7 2 6 0 0 0 6 6 2 2 2 4 6 3 1 6 4 6 3 1 4 6 6 5 4 7
## [1148] 3 9 0 0 1 9 0 0 4 5 7 0 0 6 2 7 0 8 0 9 7 0 0 9 4 0 2 9 6 4 0 5 0 5 4 1 0
## [1185] 1 7 2 4 0 7 9 4 1 2 2 8 4 2 5 4 1 4 1 7 0 1 4 1 5 3 1 4 0 9 1 4 4 3 0 1 4
## [1222] 2 1 0 1 4 3 1 1 4 1 4 7 2 4 1 9 4 2 5 0 1 4 2 5 0 1 4 2 5 0 1 4 1 1 3 9 1
## [1259] 4 7 0 3 2 1 2 3 2 1 2 0 2 2 1 2 6 8 2 1 2 0 2 1 2 6 2 1 4 0 4 2 1 7 0 1 1
## [1296] 2 1 2 2 8 3 0 3 6 6 3 0 3 2 6 3 0 3 5 6 0 3 4 9 3 0 3 0 2 6 3 0 3 4 4 3 0
```

```
## Levels: 0 1 2 3 4 5 6 7 8 9
# Extract labels and features from the test dataset
test_labels <- test_data[, 1]</pre>
test_features <- test_data[, -1]</pre>
# Convert labels to factor
test_labels <- as.factor(test_labels)</pre>
# Use the final k-NN model to make predictions on the test dataset
test_predictions <- knn(train_features, test_features, train_labels, k = final_k)
# Create a confusion matrix
conf_matrix <- table(Actual = test_labels, Predicted = test_predictions)</pre>
# Print the confusion matrix
cat('Confusion Matrix:\n')
## Confusion Matrix:
print(conf_matrix)
##
        Predicted
                                            9
## Actual
          0
                  2
                      3
                         4
                             5
                                     7
                                        8
       0 355
##
              0
                  2
                      0
                         0
                             0
                                 0
                                            1
                                     1
##
       1
           0 255
                  0
                      0
                          6
                             0
                                     1
             1 183
##
       2
          6
                      2
                         1
                             0
                                 0
                                     2
                                            0
##
          3 0
                  2 154
                         0
                                 0
                                            2
       3
                             5
                                    0
##
       4
          0
                  1
                      0 182
                                 2
                                     2
                                            8
             3
                             1
                                        1
       5
          2 1
                         0 145
                                 2
##
                  2
                      4
                                     0
                                        3
                                            1
##
       6
          0 0 1
                        2
                             3 164
                                    0
                      0
##
       7
          0 1
                  1
                      1 4
                             0
                                 0 139
                                        0
                                            1
##
         5 0
                                 0
                                     1 148
                                            3
       8
                  1
                      6
                         1
                             1
##
                  1
                      0
                         2
                             0
                                 0
                                     4
                                         1 169
# Calculate the conditional test error using zero-one loss
conditional_test_error <- 1 - sum(diag(conf_matrix)) / sum(conf_matrix)</pre>
cat('\nConditional Test Error (Zero-One Loss):', conditional_test_error, '\n')
```

##
Conditional Test Error (Zero-One Loss): 0.05630294