ml

$March\ 7,\ 2023$

Classifying with different k values

• Use different values of ${\tt k}$, check and interpret the predictions: ${\tt k}=1,$ 5, 11, 15, 21, 27.

k value	False negatives	False positives	Percent classified incorrectly
1	1	3	4 percent
5	2	0	2 percent
11	3	0	3 percent
15	3	0	3 percent
21	2	0	2 percent
27	4	0	4 percent

Figure 1: Sample results - kNN on wbcd for different k

• Check for different k by looping over a function check_k:

```
1. restart the dataset wbcd.
2. normalize it to obtain wbcd_n.
  normalize <- function(x) {</pre>
    return ((x-min(x))/(max(x)-min(x)))
  wbcd_n <- as.data.frame(lapply(wbcd[2:31],FUN=normalize))</pre>
3. split wbcd_n in training and test data:
  wbcd_train <- wbcd_n[1:469,] # all normalized columns</pre>
  wbcd_test <- wbcd_n[470:569,] # all normalized columns</pre>
4. create training and testing labels
  wbcd_train_labels <- wbcd[1:469,1] # from the original dataset
  wbcd_test_labels <- wbcd[470:569,1] # from the original dataset
5. define function with k as argument
  library(class)
  wbcd_check_k <- function(x=1) {
    wbcd_test_pred <- knn(train = wbcd_train, # training data</pre>
                            test = wbcd test, # test data
                            cl = wbcd_train_labels, # class factor
                            k = x) # nearest neighbors
    print(table(x = wbcd_test_labels,
                 y = wbcd_test_pred))
    mean(wbcd_test_pred==wbcd_test_labels)
6. loop over the k values 1,5,11,15,21,27:
  ## get the Wisconsin breast cancer data as data frame:
  wbcd <- read.csv(file="http://bit.ly/3khqmkp")</pre>
  ## drop the first (ID) column:
  wbcd <- wbcd[-1]
  ## recode target class as labeled 2-level factor
  wbcd$diagnosis |> factor(c("B","M"),c("Benign","Malignant")) -> wbcd$diagnosi
  # wbcd$diagnosis |> str()
  normalize <- function(x) {</pre>
    return ((x-min(x))/(max(x)-min(x)))
  wbcd_n <- as.data.frame(lapply(wbcd[2:31],FUN=normalize))</pre>
```

```
wbcd_train <- wbcd_n[1:469,] # all normalized columns</pre>
wbcd_test <- wbcd_n[470:569,] # all normalized columns</pre>
wbcd_train_labels <- wbcd[1:469,1] # from the original dataset</pre>
wbcd_test_labels <- wbcd[470:569,1] # from the original dataset</pre>
library(class)
wbcd_check_k <- function(x=1) {</pre>
  wbcd_test_pred <- knn(train = wbcd_train, # training data</pre>
                         test = wbcd_test, # test data
                         cl = wbcd_train_labels, # class factor
                         k = x) # nearest neighbors
  print(table(x = wbcd_test_labels,
               y = wbcd_test_pred))
  mean(wbcd_test_pred==wbcd_test_labels)
k_{vals} < c(1,5,11,15,21,27)
result <- rep(NA,length(k_vals))</pre>
names(result) <- k_vals</pre>
j <- 0
for (i in k_vals) {
  j < -j + 1
  result[j] <- wbcd_check_k(i)</pre>
  print(paste("Accuracy for k =",i,":",
               format(result[j],digits=6)))
## Find best result
## cat("Best result:", format(max(result),digits=3),
       "for k =", which(result==max(result)),"\n")
##
   У
х
            Benign Malignant
  Benign
                 58
                            3
  Malignant
                  1
                           38
[1] "Accuracy for k = 1 : 0.96"
             Benign Malignant
х
                 61
                            0
  Benign
                  2
                           37
  Malignant
[1] "Accuracy for k = 5 : 0.98"
             Benign Malignant
х
```

```
Benign
                61
                            0
                 3
  Malignant
                           36
[1] "Accuracy for k = 11 : 0.97"
            Benign Malignant
х
                61
  Benign
  Malignant
                 3
                           36
[1] "Accuracy for k = 15 : 0.97"
            Benign Malignant
х
                61
  Benign
                            0
  Malignant
                 2
                           37
[1] "Accuracy for k = 21 : 0.98"
   У
            Benign Malignant
х
                61
  Benign
                            0
                 4
  Malignant
[1] "Accuracy for k = 27 : 0.96"
```

• The Resulting table

k value	False negatives	False positives	% correct
1	1	3	96
5	2	0	98
11	3	0	97
15	3	0	97
21	2	0	98
27	4	0	96

- It is unwise to tailor the approach too closely to the test data: the difference of 100 patient records is not very large compared to the likely test data sets in production.
- To be certain that a learner will generalize to future data, you can create several sets of 100 patients at random and repeatedly retest.

Fixing the algorithm

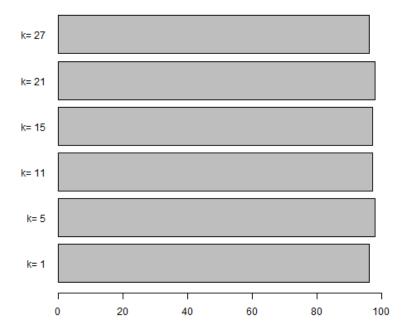
- Problem in ??: updating the result vector **r** inside so that the returned vector does not have **k[length(k)]** = 7 values but only three:
 - loopindex i assumes values 1,5,7

- calls f(i) and returns a single value
- I want to put the value into r but not into r[i] (which is NA after the first loop, because length(r) is 3.
- Create dummy index j and increase it

```
f <- function(x=1) {
  return (mean(rnorm(x)))
                                        \#k < -c(1,5,7)
cat("k:",k,"\n")
r <- rep(NA,length(k))
                                        #cat("r:",r,"\n")
j <- 0
for (i in k) { # loop runs three times
  j <- j + 1
 r[j] <- f(i) # assign value of k to r
  print(i) # print k
  print(r) # print r
}
r
Error in cat("k:", k, "\n") : object 'k' not found
Error: object 'k' not found
Error in k : object 'k' not found
Error: object 'r' not found
```

Plot results vector (barplot or histogram)

• Plot the results vector. Not a very interesting plot.

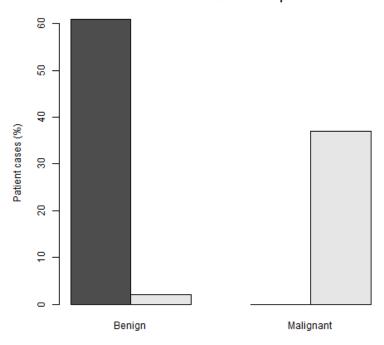


Plot cross table results

- Use capture.output to stop gmodels::CrossTable from printing to the screen, but save the table in cross_table, a list.
- Use table instead for cleaner, shorter output:

```
}
  cf(21)
             У
              Benign Malignant
    Benign
                  61
                   2
   Malignant
                             37
  [1] 0.98
• Without function, to test the plotting:
 wbcd_test_pred <- knn(train = wbcd_train, # training data</pre>
                         test = wbcd_test, # test data
                         cl = wbcd_train_labels, # class factor
                         k = 21) # nearest neighbors
  cross_table <- table(x = wbcd_test_labels,</pre>
                        y = wbcd_test_pred)
  ##
          write(cross_table, file='./data/crosstable.txt')
  mean(wbcd_test_pred==wbcd_test_labels)
  cross_table
  [1] 0.98
              Benign Malignant
                  61
    Benign
    Malignant
                    2
                             37
  barplot(cross_table,
          beside=TRUE,
          ylab="Patient cases (%)",
          main="Wisconsin Breast Cancer Data prediction")
```

Wisconsin Breast Cancer Data prediction



• ChatGPT

```
library(gmodels)

my_function <- function(x, y) {
    ## Create a table
    mytable <- table(x, y)
    ## Suppress output of CrossTable, save the result to an R object
    result <- capture.output(gmodels::CrossTable(x = x, y = y, prop.chisq = FALSE))
    ## Return the result
    return(result)
}

## Call the function
x <- c(1, 1, 2, 3, 3, 3)
y <- c("A", "A", "B", "B", "C", "C")
result <- my_function(x, y)
## View the result</pre>
```