## HW8

# Rosemary Kinuthia

### 4/17/2018

K-nearest neighbor

```
Let's try a variation on the NHANES data set again.
 library(tidyverse)
 ## - Attaching packages -
                                                       - tidyverse 1.2.1 --
 ## ✔ ggplot2 2.2.1
                         ✓ purrr
                                  0.2.4
                                  0.7.4
 ## 🗸 tibble 1.4.2

✓ dplyr

 ## / tidyr 0.8.0

✓ stringr 1.3.0

 ## ✔ readr
             1.1.1

✓ forcats 0.3.0

 ## - Conflicts -
                                                — tidyverse_conflicts() —
 ## * dplyr::filter() masks stats::filter()
 ## * dplyr::lag() masks stats::lag()
 library(class)
 library(rpart)
 library(NHANES)
 library(RColorBrewer)
 library(plot3D)
 library(parallel)
 library(randomForestSRC)
 ##
 ##
     randomForestSRC 2.5.1
 ##
 ##
     Type rfsrc.news() to see new features, changes, and bug fixes.
 ##
 ##
 ## Attaching package: 'randomForestSRC'
 ## The following object is masked from 'package:purrr':
 ##
 ##
        partial
 library(ggRandomForests)
 ## Attaching package: 'ggRandomForests'
```

```
## The following object is masked from 'package:randomForestSRC':
##
##
       partial.rfsrc
library(mosaic)
## Loading required package: lattice
## Loading required package: ggformula
##
## New to ggformula? Try the tutorials:
   learnr::run_tutorial("introduction", package = "ggformula")
   learnr::run_tutorial("refining", package = "ggformula")
## Loading required package: mosaicData
## Loading required package: Matrix
## Attaching package: 'Matrix'
## The following object is masked from 'package:tidyr':
##
##
       expand
##
## The 'mosaic' package masks several functions from core packages in order to add
## additional features. The original behavior of these functions should not be affected by this.
##
## Note: If you use the Matrix package, be sure to load it BEFORE loading mosaic.
##
## Attaching package: 'mosaic'
## The following object is masked from 'package:Matrix':
##
##
       mean
## The following objects are masked from 'package:dplyr':
##
##
       count, do, tally
## The following object is masked from 'package:purrr':
##
##
       cross
```

```
## The following objects are masked from 'package:stats':
##
## binom.test, cor, cor.test, cov, fivenum, IQR, median,
## prop.test, quantile, sd, t.test, var
```

```
## The following objects are masked from 'package:base':
##
## max, mean, min, prod, range, sample, sum
```

Create the NHANES dataset again, just like we did in class, only using sleep trouble (variable name = SleepTrouble) as the dependent variable, instead of Diabetes.

```
# Create the NHANES dataset again
people <- NHANES %>% dplyr::select(Age, Gender, SleepTrouble, BMI, HHIncome, PhysActive)
#%>% na.omit()
glimpse(people)
```

#### Problem 1

What is the marginal distribution of sleep trouble?

```
tally(~ SleepTrouble, data = people, format = "percent")
```

```
## SleepTrouble
## No Yes <NA>
## 57.99 19.73 22.28
```

Recall from our prior work, the packages work better if the dataset is a dataframe, and the variables are numeric.

```
class(people)
```

```
## [1] "tbl_df" "tbl" "data.frame"
```

```
# Convert back to dataframe
people <- as.data.frame(people)
glimpse(people)</pre>
```

```
# Convert factors to numeric - the packages just seem to work better that way
people$Gender <- as.numeric(people$Gender)
people$SleepTrouble <- as.numeric(people$SleepTrouble)
people$HHIncome <- as.numeric(people$HHIncome)
people$PhysActive <- as.numeric(people$PhysActive)

people <- na.omit(people)</pre>
glimpse(people)
```

Apply the k-nearest neighbor procedure to predict SleepTrouble from the other covariates, as we did for Diabetes. Use k = 1, 3, 5, and 20.

#### Problem 2

```
# Apply knn procedure to predict SleepTrouble
# Let's try different values of k to see how that affects performance
knn.1 <- knn(train = people, test = people, cl = as.numeric(people$SleepTrouble), k = 1)
knn.3 <- knn(train = people, test = people, cl = people$SleepTrouble, k = 3)
knn.5 <- knn(train = people, test = people, cl = people$SleepTrouble, k = 5)
knn.20 <- knn(train = people, test = people, cl = people$SleepTrouble, k = 20)
#knn.1</pre>
```

Now let's see how well these classifiers work overall

#### Problem 3

```
# Calculate the percent predicted correctly
# How well do these classifiers (k = 1, 3, 5, 20) work?
100*sum(people$SleepTrouble == knn.1)/length(knn.1)
```

```
## [1] 100
```

```
100*sum(people$SleepTrouble == knn.3)/length(knn.3)
```

```
## [1] 92.25522
 100*sum(people$SleepTrouble == knn.5)/length(knn.5)
 ## [1] 88.58889
 100*sum(people$SleepTrouble == knn.20)/length(knn.20)
 ## [1] 78.8262
Problem 4
What about success overall?
 table(knn.1, people$SleepTrouble)
 ##
 ## knn.1
             1
 ##
        1 5239
 ##
            0 1798
 table(knn.3, people$SleepTrouble)
 ##
 ## knn.3
           1
 ##
        1 5070 376
        2 169 1422
 table(knn.5, people$SleepTrouble)
 ##
 ## knn.5
             1
 ##
        1 5023 587
        2 216 1211
 table(knn.20, people$SleepTrouble)
 ##
 ## knn.20
            1
```

The github repository for this assignment can be accessed via this link (https://github.com/RosemaryKinuthia/2018week11.git

##

##

1 5104 1355

2 135 443

(https://github.com/RosemaryKinuthia/2018week11.git)) [https://github.com/RosemaryKinuthia/2018week11.git (https://github.com/RosemaryKinuthia/2018week11.git)]