Homework 8

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Link to repository: (https://github.com/jynakay/Assignments (https://github.com/jynakay/Assignments)) [https://github.com/jynakay/Assignments (https://github.com/jynakay/Assignments)].

```
library(tidyverse)
library(class)
library(npart)
library(NHANES)
library(RColorBrewer)
library(plot3D)
library(parallel)
library(randomForestSRC)
library(ggRandomForests)
library(mosaic)
library(dplyr)
```

Problem 1

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

What is the marginal distribution of sleep trouble?

```
#create dataset
people <- NHANES %>% dplyr::select(Age, Gender, SleepTrouble, BMI, HHIncome, PhysActive)
glimpse(people)
```

```
#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.

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

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

```
glimpse(people)
```

```
#convert variables to numeric
people$Age <- as.numeric(people$Age)
people$Gender <- as.numeric(people$Gender)
people$SleepTrouble <- as.numeric(people$SleepTrouble)
people$BMI <- as.numeric(people$BMI)
people$HHIncome <- as.numeric(people$HHIncome)
people$PhysActive <- as.numeric(people$PhysActive)</pre>
people <- na.omit(people)
glimpse(people)
```

Problem 2

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

```
#Apply k-nearest neighbor approach to predict SleepTrouble for k = 1, 3, 5, 20
knn.1 <- knn(train = people, test = people, cl = 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)</pre>
```

Problem 3

Now let's see how well these classifiers work overall.

```
# Calculate the percent predicted correctly
100*sum(people$SleepTrouble == knn.1)/length(knn.1)
```

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

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

```
## [1] 92.01364
```

```
100*sum(people$SleepTrouble == knn.5)/length(knn.5)
```

```
## [1] 88.68836
```

```
100*sum(people$SleepTrouble == knn.20)/length(knn.20)
```

```
## [1] 78.62726
```

Problem 4

What about success overall?

```
# Another way to look at success rate against increasing k
table(knn.1, people$SleepTrouble)
```

```
##
## knn.1 1 2
## 1 5239 0
## 2 0 1798
```

```
table(knn.3, people$SleepTrouble)
```

```
## ## knn.3 1 2 ## 1 5062 385 ## 2 177 1413
```

```
table(knn.5, people$SleepTrouble)
```

```
table(knn.20, people$SleepTrouble)
```