

Homework 4

Andrew, Jasmine, Abell

2024-12-03

```
setwd("/cloud/project")

sleepdata <- read.csv("Sleep_health_and_lifestyle_dataset.csv", header = TRUE)

attach(sleepdata)
#this will allows to name variables the way they are

names(sleepdata)

## [1] "Person.ID"          "Gender"
## [3] "Age"                "Occupation"
## [5] "Sleep.Duration"     "Quality.of.Sleep"
## [7] "Physical.Activity.Level" "Stress.Level"
## [9] "BMI.Category"       "Blood.Pressure"
## [11] "Heart.Rate"         "Daily.Steps"
## [13] "Sleep.Disorder"

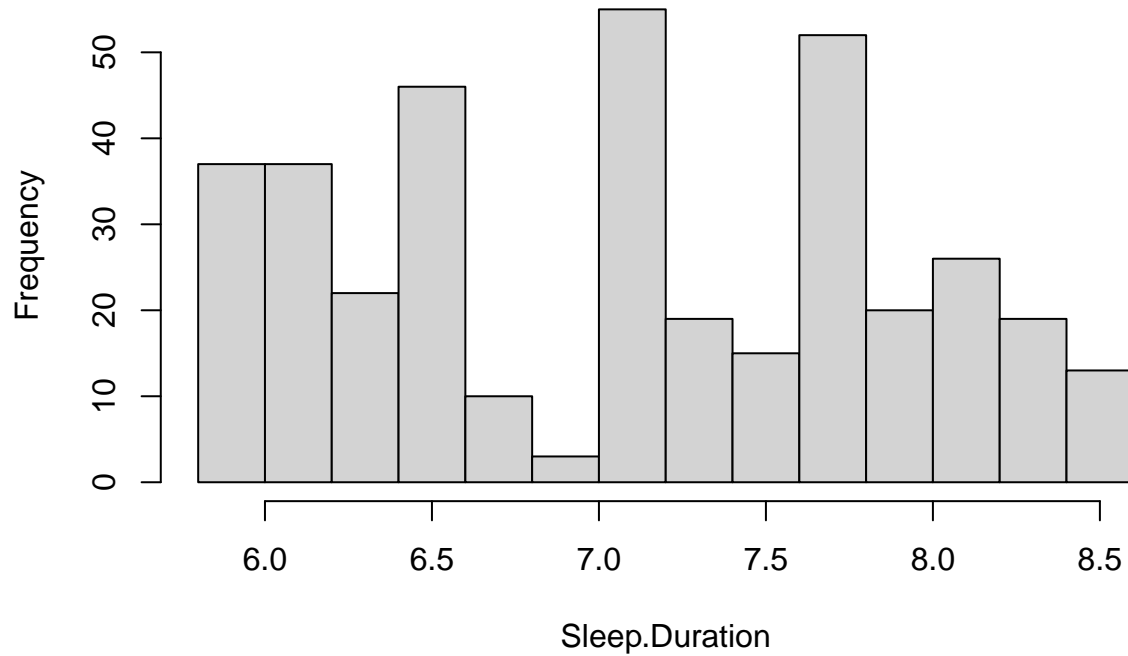
#Ho: data is not normal; ha: data is normal type I error set to 0.05 reject null
#hypothesis that the outcome is not normal and conclude data is normal
shapiro.test(Sleep.Duration)

##
##  Shapiro-Wilk normality test
##
## data:  Sleep.Duration
## W = 0.93577, p-value = 1.268e-11

#Histogram of the outcome shows to be a bimodal distribution.
#This could be an indication that sleep duration might differ
#by a particular group - maybe gender.
#Given our large sample size of n=374 observations, by the
#central limit theorem, we can conclude normality approximation.
#P-value is normal

hist(Sleep.Duration)
```

Histogram of Sleep.Duration



```
#creates a new variable systolic extracting the first 3 digits of
#the bloodpressure
```

```
is.character(Blood.Pressure)
```

```
## [1] TRUE
```

```
sleepdata$systolic = substr(Blood.Pressure, 1, 3)
```

```
sleepdata$systolic = as.numeric(sleepdata$systolic)
```

```
sleepdata$diastolic = substr(Blood.Pressure, 5, 6)
```

```
sleepdata$diastolic = as.numeric(sleepdata$diastolic)
```

```
install.packages("leaps")
```

```
library(leaps)
```

```
#Now we run the regsubsets to find the best model
```

```
output <- regsubsets(Sleep.Duration ~ Gender + Age + Occupation + Quality.of.Sleep + Physical.Activity.1 +
                    Stress.Level + BMI.Category + Heart.Rate + Daily.Steps +
                    Sleep.Disorder + systolic + diastolic, data=sleepdata, nvmax=12)
```

```
summOut1 <- summary(output)
```

```
summOut1
```

```
## Subset selection object
```

```
## Call: regsubsets.formula(Sleep.Duration ~ Gender + Age + Occupation +
```

```

##      Quality.of.Sleep + Physical.Activity.Level + Stress.Level +
##      BMI.Category + Heart.Rate + Daily.Steps + Sleep.Disorder +
##      systolic + diastolic, data = sleepdata, nvmax = 12)
## 24 Variables (and intercept)
##
##                                     Forced in Forced out
## GenderMale                        FALSE      FALSE
## Age                              FALSE      FALSE
## OccupationDoctor                   FALSE      FALSE
## OccupationEngineer                 FALSE      FALSE
## OccupationLawyer                   FALSE      FALSE
## OccupationManager                  FALSE      FALSE
## OccupationNurse                    FALSE      FALSE
## OccupationSales Representative     FALSE      FALSE
## OccupationSalesperson              FALSE      FALSE
## OccupationScientist                FALSE      FALSE
## OccupationSoftware Engineer        FALSE      FALSE
## OccupationTeacher                  FALSE      FALSE
## Quality.of.Sleep                   FALSE      FALSE
## Physical.Activity.Level             FALSE      FALSE
## Stress.Level                       FALSE      FALSE
## BMI.CategoryNormal Weight           FALSE      FALSE
## BMI.CategoryObese                  FALSE      FALSE
## BMI.CategoryOverweight              FALSE      FALSE
## Heart.Rate                         FALSE      FALSE
## Daily.Steps                       FALSE      FALSE
## Sleep.DisorderNone                 FALSE      FALSE
## Sleep.DisorderSleep Apnea          FALSE      FALSE
## systolic                          FALSE      FALSE
## diastolic                         FALSE      FALSE
## 1 subsets of each size up to 12
## Selection Algorithm: exhaustive
##      GenderMale Age OccupationDoctor OccupationEngineer OccupationLawyer
## 1  ( 1 ) " "      " " " "      " "      " "
## 2  ( 1 ) " "      " " "*"      " "      " "
## 3  ( 1 ) " "      " " "*"      "*"      " "
## 4  ( 1 ) " "      " " "*"      "*"      " "
## 5  ( 1 ) " "      " " "*"      "*"      " "
## 6  ( 1 ) " "      " " "*"      "*"      " "
## 7  ( 1 ) " "      " " "*"      "*"      " "
## 8  ( 1 ) " "      " " "*"      "*"      "*"
## 9  ( 1 ) " "      " " "*"      "*"      "*"
## 10 ( 1 ) " "      " " "*"      "*"      "*"
## 11 ( 1 ) " "      " " "*"      "*"      "*"
## 12 ( 1 ) " "      "*" "*"      "*"      "*"
##
##      OccupationManager OccupationNurse OccupationSales Representative
## 1  ( 1 ) " "      " "      " "
## 2  ( 1 ) " "      " "      " "
## 3  ( 1 ) " "      " "      " "
## 4  ( 1 ) " "      " "      " "
## 5  ( 1 ) " "      " "      " "
## 6  ( 1 ) " "      " "      " "
## 7  ( 1 ) " "      " "      " "
## 8  ( 1 ) " "      " "      " "
## 9  ( 1 ) " "      " "      " "

```

```

## 10 ( 1 ) " " " " "*"
## 11 ( 1 ) " " " " "*"
## 12 ( 1 ) " " " " "*"
## OccupationSalesperson OccupationScientist OccupationSoftware Engineer
## 1 ( 1 ) " " " " " "
## 2 ( 1 ) " " " " " "
## 3 ( 1 ) " " " " " "
## 4 ( 1 ) " " " " " "
## 5 ( 1 ) " " " " " "
## 6 ( 1 ) " " " " " "
## 7 ( 1 ) "*" " " " " " "
## 8 ( 1 ) "*" " " " " " "
## 9 ( 1 ) "*" " " " " " "
## 10 ( 1 ) "*" " " " " " "
## 11 ( 1 ) "*" " " " " " "
## 12 ( 1 ) "*" " " " " " "
## OccupationTeacher Quality.of.Sleep Physical.Activity.Level
## 1 ( 1 ) " " "*" " "
## 2 ( 1 ) " " "*" " "
## 3 ( 1 ) " " "*" " "
## 4 ( 1 ) " " "*" " "
## 5 ( 1 ) " " "*" "*"
## 6 ( 1 ) "*" "*" "*"
## 7 ( 1 ) " " "*" "*"
## 8 ( 1 ) " " "*" "*"
## 9 ( 1 ) " " "*" " "
## 10 ( 1 ) " " "*" "*"
## 11 ( 1 ) " " "*" " "
## 12 ( 1 ) " " "*" "*"
## Stress.Level BMI.CategoryNormal Weight BMI.CategoryObese
## 1 ( 1 ) " " " " " "
## 2 ( 1 ) " " " " " "
## 3 ( 1 ) " " " " " "
## 4 ( 1 ) " " " " " "
## 5 ( 1 ) " " " " " "
## 6 ( 1 ) "*" " " " " " "
## 7 ( 1 ) "*" " " " " " "
## 8 ( 1 ) "*" " " " " " "
## 9 ( 1 ) "*" " " " " " "
## 10 ( 1 ) "*" " " "*"
## 11 ( 1 ) "*" " " "*"
## 12 ( 1 ) "*" " " " "
## BMI.CategoryOverweight Heart.Rate Daily.Steps Sleep.DisorderNone
## 1 ( 1 ) " " " " " "
## 2 ( 1 ) " " " " " "
## 3 ( 1 ) " " " " " "
## 4 ( 1 ) " " "*" " " " "
## 5 ( 1 ) " " " "*" " "
## 6 ( 1 ) " " " " " "
## 7 ( 1 ) " " "*" " " " "
## 8 ( 1 ) " " "*" " " " "
## 9 ( 1 ) " " "*" " " " "
## 10 ( 1 ) " " "*" " " " "
## 11 ( 1 ) " " "*" " " " "

```

```
## 12 ( 1 ) " " " " "*" " "
##      Sleep.DisorderSleep Apnea systolic diastolic
## 1 ( 1 ) " " " " " "
## 2 ( 1 ) " " " " " "
## 3 ( 1 ) " " " " " "
## 4 ( 1 ) " " " " " "
## 5 ( 1 ) " " " " " "
## 6 ( 1 ) " " " " " "
## 7 ( 1 ) " " " " " "
## 8 ( 1 ) " " " " " "
## 9 ( 1 ) " " "*" "*"
## 10 ( 1 ) " " " " " "
## 11 ( 1 ) " " "*" "*"
## 12 ( 1 ) " " "*" "*"

```

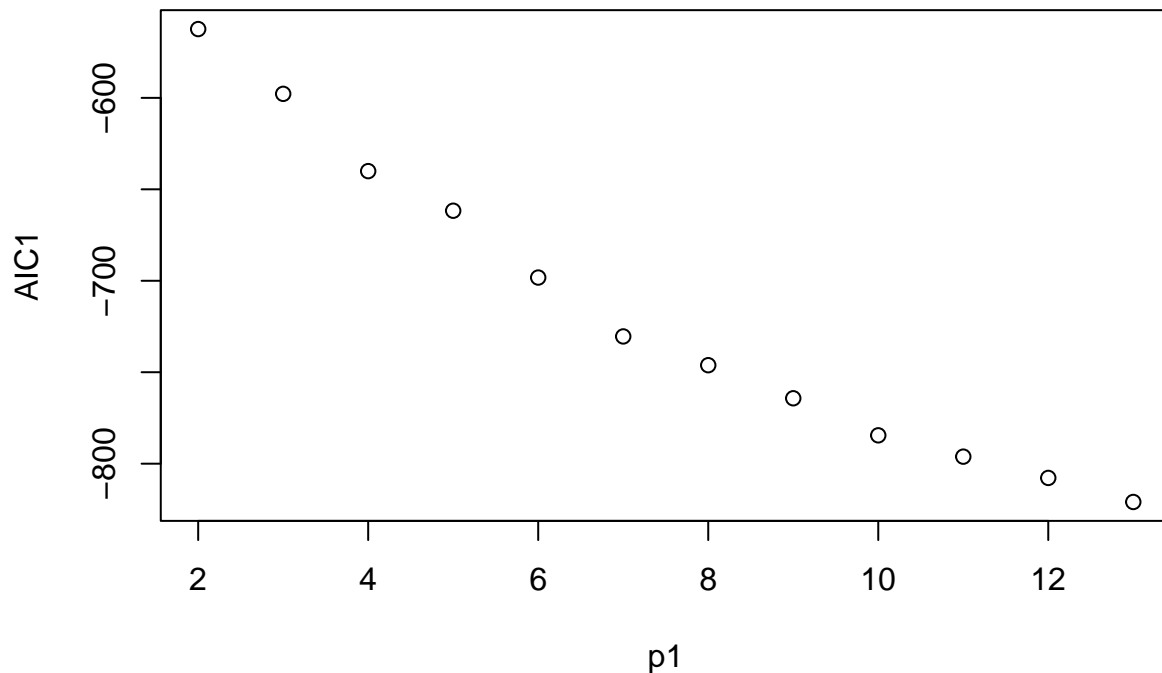
```
n1 <- length(Sleep.Duration)
n1
```

```
## [1] 374
```

```
p1 <- apply(summOut1$which, 1, sum)
```

```
aic1 <- summOut1$bic - log(n1) * p1 + 2 * p1
```

```
plot(p1, aic1, ylab = "AIC1")
```



```
#best model is the one with all the predictors as it has the lowest AIC
modell1 <- lm(Sleep.Duration ~ Gender + Age + Occupation + Quality.of.Sleep +
  Physical.Activity.Level +
  Stress.Level + BMI.Category + Heart.Rate + Daily.Steps +
  Sleep.Disorder + systolic + diastolic, data=sleepdata)

summary(modell1)
```

```
##
## Call:
## lm(formula = Sleep.Duration ~ Gender + Age + Occupation + Quality.of.Sleep +
##     Physical.Activity.Level + Stress.Level + BMI.Category + Heart.Rate +
##     Daily.Steps + Sleep.Disorder + systolic + diastolic, data = sleepdata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.71735 -0.14289 -0.03386  0.13013  0.97101
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    6.5255127   1.1566767   5.642 3.49e-08 ***
## GenderMale    -0.0163363   0.0852287  -0.192 0.848107
## Age           0.0274491   0.0065377   4.199 3.41e-05 ***
## OccupationDoctor  0.8323400   0.0861232   9.665 < 2e-16 ***
## OccupationEngineer 0.7772756   0.0867901   8.956 < 2e-16 ***
## OccupationLawyer  0.7294035   0.0989878   7.369 1.26e-12 ***
## OccupationManager 0.1027142   0.2601035   0.395 0.693160
## OccupationNurse   0.2427417   0.1128672   2.151 0.032187 *
## OccupationSales Representative 1.4483922   0.2358399   6.141 2.23e-09 ***
## OccupationSalesperson 0.6357357   0.1080311   5.885 9.35e-09 ***
## OccupationScientist 0.4568531   0.1701922   2.684 0.007614 **
## OccupationSoftware Engineer 0.6326634   0.1507845   4.196 3.45e-05 ***
## OccupationTeacher 0.2883608   0.0883850   3.263 0.001213 **
## Quality.of.Sleep 0.2860928   0.0561463   5.095 5.71e-07 ***
## Physical.Activity.Level 0.0092998   0.0015524   5.991 5.20e-09 ***
## Stress.Level    -0.1628751   0.0341770  -4.766 2.77e-06 ***
## BMI.CategoryNormal Weight -0.0338319   0.0682966  -0.495 0.620653
## BMI.CategoryObese -0.6002143   0.1938760  -3.096 0.002121 **
## BMI.CategoryOverweight -0.3467385   0.1028555  -3.371 0.000832 ***
## Heart.Rate      0.0332898   0.0101959   3.265 0.001203 **
## Daily.Steps     -0.0001284   0.0000219  -5.863 1.05e-08 ***
## Sleep.DisorderNone -0.1020383   0.0602559  -1.693 0.091268 .
## Sleep.DisorderSleep Apnea -0.0549659   0.0674340  -0.815 0.415567
## systolic        -0.1212507   0.0164668  -7.363 1.30e-12 ***
## diastolic        0.1359929   0.0221031   6.153 2.09e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2442 on 349 degrees of freedom
## Multiple R-squared:  0.9119, Adjusted R-squared:  0.9058
## F-statistic: 150.4 on 24 and 349 DF, p-value: < 2.2e-16
```

```
table(Occupation)
```

```
## Occupation
##      Accountant      Doctor      Engineer
##          37          71          63
##      Lawyer      Manager      Nurse
##          47           1          73
## Sales Representative Salesperson Scientist
##           2           32           4
##      Software Engineer      Teacher
##           4           40
```

```
table(BMI.Category)
```

```
## BMI.Category
##      Normal Normal Weight      Obese      Overweight
##      195          21          10          148
```

```
table(Sleep.Disorder)
```

```
## Sleep.Disorder
##      Insomnia      None Sleep Apnea
##      77          219          78
```

#interpretation of the significant variables from this model:

#Sleep duration increases significantly by 0.027 units for every unit

#increase in age, adjusting for everything else

#Sleep Duration increases significantly by 0.027 units for every unit

#increase in age, adjusting for everything else

#Sleep Duration increases significantly by 0.83 units for Doctors

vs accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.78 units for Engineers

vs. Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.73 units for Lawyers

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.24 units for Nurses

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 1.45 units for Sales Reps

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.63 units for SalesPerson

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.46 units for Scientists

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.63 units for SoftwareEngineers

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.29 units for Teachers

#vs Accountants, adjusting for everything else

#Sleep Duration increases significantly by 0.29 units for every unit

#increase in quality of sleep score, adjusting for everything else

#Sleep Duration increases significantly by 0.009 units for every unit

#increase in physical activity level score, adjusting for everything else

#Sleep Duration decreases significantly by 0.16 units for every unit

#increase in stress level score, adjusting for everything else

```

#Increase in daily steps, adjusting for everything else

#Sleep Duration decreases significantly by 0.121 units for every unit
#Increase in systolic reading, adjusting for everything else

#Sleep Duration increases significantly by 0.13 units for every unit
#increase in diastolic reading, adjusting for everything else

#we check for multicollinearity using vif and tolerance

install.packages("car")
library(car)

vif(model1)

```

```

##              GVIF Df GVIF^(1/(2*Df))
## Gender          11.386299  1      3.374359
## Age             20.107382  1      4.484126
## Occupation      2933.052016 10      1.490617
## Quality.of.Sleep  28.246011  1      5.314698
## Physical.Activity.Level  6.539635  1      2.557271
## Stress.Level     23.003395  1      4.796185
## BMI.Category     110.831848  3      2.191681
## Heart.Rate       11.120014  1      3.334668
## Daily.Steps       7.851678  1      2.802085
## Sleep.Disorder    11.655508  2      1.847706
## systolic         101.805454  1     10.089869
## diastolic        115.998380  1     10.770254

```

```

#if the vif shows greater than 10, it means that there is such a strong
#relationship between the variables, like that these may be collinear
#If collinear, this will bias the results of the model from our results
#we see that systolic and diastolic might be collinear

```

```

#tolerance the inverse of vif; we run this as an extra check
1/vif(model1)

```

```

##              GVIF      Df GVIF^(1/(2*Df))
## Gender          0.0878248485 1.0000000    0.29635257
## Age             0.0497329793 1.0000000    0.22300892
## Occupation      0.0003409418 0.1000000    0.67086327
## Quality.of.Sleep  0.0354032289 1.0000000    0.18815746
## Physical.Activity.Level  0.1529137380 1.0000000    0.39104186
## Stress.Level     0.0434718439 1.0000000    0.20849903
## BMI.Category     0.0090226773 0.3333333    0.45627069
## Heart.Rate       0.0899279456 1.0000000    0.29987989
## Daily.Steps      0.1273613137 1.0000000    0.35687717
## Sleep.Disorder    0.0857963433 0.5000000    0.54121172
## systolic         0.0098226564 1.0000000    0.09910932
## diastolic        0.0086208100 1.0000000    0.09284832

```

```

#With these results we look for the last column to be >0.10, if its less, than
#this means its collinear, the two variables are systolic and diastolic

```