# Heart Attack Risk Prediction. Final Project

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#Tidying (Yoonhye) #Removed unwanted columns

#Renamed columns:

```
untidied_dataset <- untidied_dataset %>%
  rename(
    heart_attack_risk = "Heart Attack Risk",
    blood_pressure = "Blood Pressure",
    heart_rate = "Heart Rate",
    family_history = "Family History",
    exercise_hrs_week = "Exercise Hours Per Week",
    stress_level = "Stress Level",
    sedentary_hrs_day = "Sedentary Hours Per Day",
    physical_days_week = "Physical Activity Days Per Week",
    sleep_hrs = "Sleep Hours Per Day",
    age = "Age",
    sex = "Sex",
    cholesterol = "Cholesterol",
    diet = "Diet",
    triglycerides = "Triglycerides"
)
```

#Split original dataset into "no family history" and "family history":

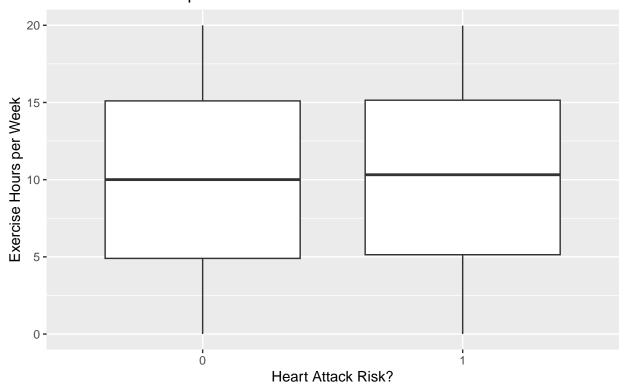
#Separated the blood\_pressure column into systolic and diastolic:

```
## Warning: 'combine()' was deprecated in dplyr 1.0.0.
## i Please use 'vctrs::vec_c()' instead.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

#EDA (Jiwon Jung) #boxplot for the exercise hours explanatory variable:

```
no_family_history_dataset %>%
ggplot() +
    geom_boxplot(
    mapping = aes(
        x =
            factor(heart_attack_risk),
        y = exercise_hrs_week))+
labs(
    title = "Boxplot of Heart Attack Risk and
    Exercise Hours per Week",
    x = "Heart Attack Risk?",
    y = "Exercise Hours per Week")
```

# Boxplot of Heart Attack Risk and Exercise Hours per Week



#summary statistics for exercise hours:

```
no_family_history_dataset %>%
  group_by(
    factor(heart_attack_risk)) %>%
    summarize(
    median = median(exercise_hrs_week),
    mean = mean(exercise_hrs_week),
    max = max(exercise_hrs_week),
    min = min(exercise_hrs_week)
## # A tibble: 2 x 5
##
     'factor(heart_attack_risk)' median mean
                                                        min
                                                max
                                  <dbl> <dbl> <dbl>
                                                      <dbl>
## 1 0
                                   10.0 9.99 20.0 0.00244
```

#Boxplot for sleep hours explanatory variable:

## 2 1

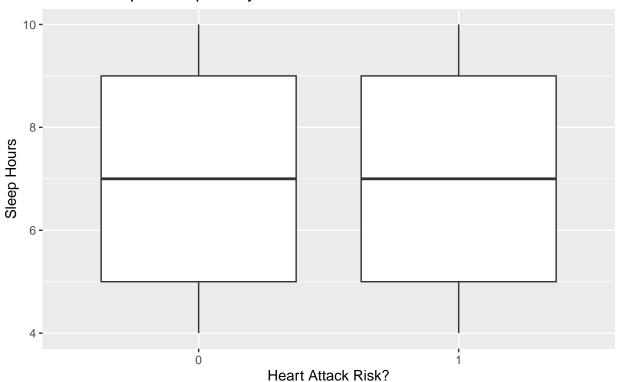
```
no_family_history_dataset %>%
    ggplot() +
    geom_boxplot(
        mapping = aes(
        x = factor(heart_attack_risk),
        y = sleep_hrs)) +
```

20.0 0.00511

10.3 10.2

```
labs(
  title = "Boxplot of Heart Attack Risk
  and Sleep Hours per Day",
  x = "Heart Attack Risk?",
  y = "Sleep Hours")
```

# Boxplot of Heart Attack Risk and Sleep Hours per Day



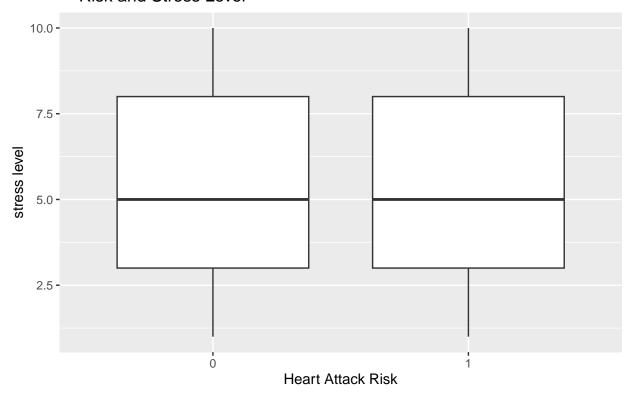
 $\# Summary \ statistics \ of sleep hours:$ 

```
no_family_history_dataset %>%
  group_by(
  factor(heart_attack_risk)) %>%
  summarize(
  median = median(sleep_hrs),
  mean = mean(sleep_hrs),
  max = max(sleep_hrs),
  min = min(sleep_hrs)
)
```

# Boxplot of stress level explanatory variable:

```
no_family_history_dataset %>%
    ggplot() +
geom_boxplot(
    mapping = aes(
        x = factor(heart_attack_risk),
        y = stress_level)) +
labs(
    title = "Boxplot of Heart Attack
    Risk and Stress Level",
    x = "Heart Attack Risk",
    y = "stress level")
```

#### Boxplot of Heart Attack Risk and Stress Level



#Summary statistics of stress level explanatory variable:

```
no_family_history_dataset %>%
group_by(
   factor(heart_attack_risk)) %>%
   summarize(
   median = median(stress_level),
   mean = mean(stress_level),
   max = max(stress_level),
   min = min(stress_level)
)
```

```
## # A tibble: 2 x 5
## 'factor(heart_attack_risk)' median mean max min
```

#Splitting the stress levels into "not stressed" and "stressed":

```
no_family_history_dataset <- no_family_history_dataset %>%
  mutate(
    count =
      ifelse(
        stress_level >= 1
        & !(stress_level >=6)
        & heart_attack_risk == 1,
        "no stress & risk",
      ifelse(
        stress_level >= 1
        & !(stress_level >=6)
        & heart_attack_risk == 0,
        "no stress & no risk",
      ifelse(
        stress_level >= 6
        & heart_attack_risk == 1,
        "stress & risk",
      ifelse(
        stress_level >= 6
        & heart_attack_risk == 0,
        "stress and no risk", NA))))
  )
```

#Counted how many people were stressed and at a risk of heart attacks:

```
no_family_history_dataset %>%
  group_by(factor(count)) %>%
  summarize(n = n())

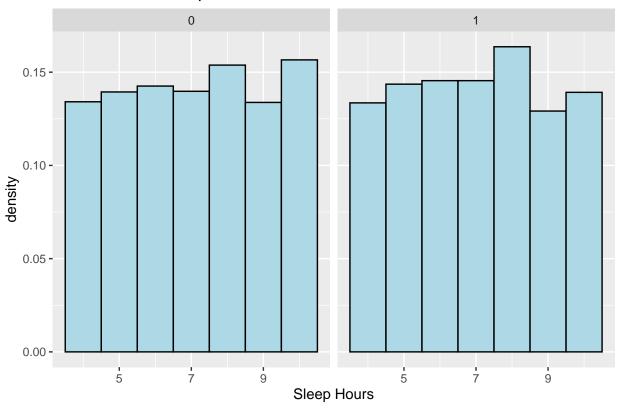
## # A tibble: 4 x 2
## 'factor(count)' n
## <fct> <int>
## 1 no stress & no risk 1465
## 2 no stress & risk 812
## 3 stress & risk 783
## 4 stress and no risk 1383
#Bar blot showing distribution of sleep Hours:
```

```
no_family_history_dataset$sleep_hrs <- as.numeric(no_family_history_dataset$sleep_hrs)</pre>
```

```
bins = 7,
fill = "light blue",
color = "black")+
labs(
  title = "Distribution of Sleep Hours",
  x = "Sleep Hours")+
facet_wrap(
  ~heart_attack_risk,
  scales = "free_x")
```

```
## Warning: The dot-dot notation ('..density..') was deprecated in ggplot2 3.4.0.
## i Please use 'after_stat(density)' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

# Distribution of Sleep Hours

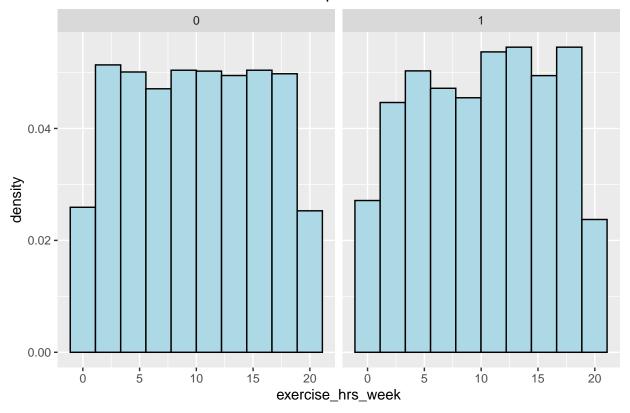


#PMF plot showing distribution of Exercise Hours per week:

```
no_family_history_dataset %>%
    ggplot() +
    geom_histogram(
        mapping = aes(
            x = exercise_hrs_week,
            y = ..density..),
        bins = 10,
```

```
fill = "light blue",
  color = "black")+
labs(
  title =
    "PMF distribution of Exercise hours per week")+
facet_wrap(
  ~heart_attack_risk,
  scales = "free_x")
```

# PMF distribution of Exercise hours per week



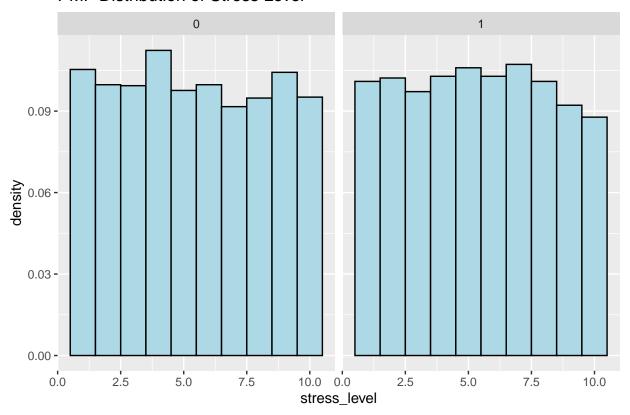
#Histogram showing distribution of stress level:

no\_family\_history\_dataset\$stress\_level <- as.numeric(no\_family\_history\_dataset\$stress\_level)</pre>

```
no_family_history_dataset %>%
    ggplot() +
    geom_histogram(
        mapping = aes(
            x = stress_level,
            y = ..density..),
        bins = 10,
        fill = "light blue",
        color = "black")+
    labs(
        title =
            "PMF Distribution of Stress Level")+
    facet_wrap(
```

```
"heart_attack_risk,
scales = "free_x")
```

#### PMF Distribution of Stress Level



#count of people with heart attack risk among those with no family history.

```
no_family_history_dataset %>%
group_by(factor(heart_attack_risk)) %>%
summarize(n = n())
```

```
## # A tibble: 2 x 2
## 'factor(heart_attack_risk)' n
## <fct> <int>
## 1 0 2848
## 2 1 1595
```

#count of people with heart attack risk among those with family history.

```
family_history_dataset %>%
  group_by(
  factor(heart_attack_risk)) %>%
  summarize(n = n())
```

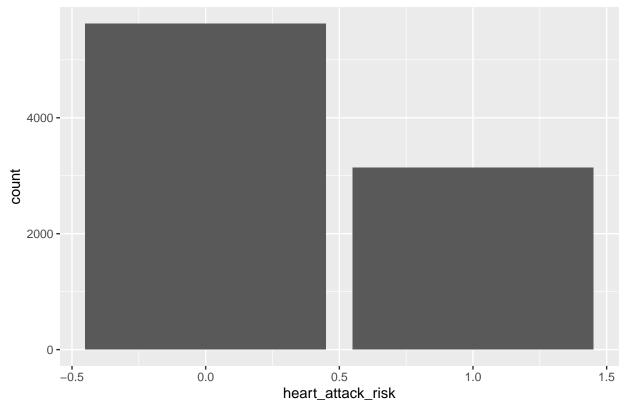
```
## # A tibble: 2 x 2
## 'factor(heart_attack_risk)' n
```

```
## <fct> <int>
## 1 0 2776
## 2 1 1544
```

#Barplot showing the distribution of people with different combinations of heart attack risk and family history #Why is the count messed up!

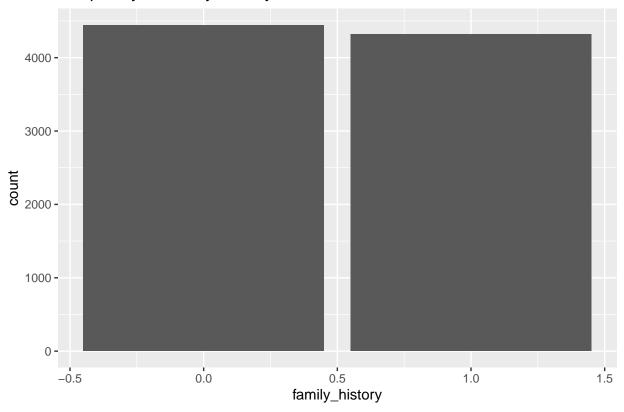
```
untidied_dataset %>%
ggplot()+
geom_bar(
    mapping = aes(
        x = heart_attack_risk)
)+
labs(title = "Frequency of Heart Attack Risk")
```

# Frequency of Heart Attack Risk



```
untidied_dataset %>%
ggplot()+
geom_bar(
    mapping = aes(
        x = family_history)
)+
labs(title = "Frequency of Family History")
```

# Frequency of Family History



#Modeling (Giselle Rahimi) #Modeling (exercise\_hrs\_week)

```
split <- initial_split(no_family_history_dataset, prop = 0.7)
trainData <- training(split)
testData <-testing(split)</pre>
```

```
trainData %>%
summarize(
  total = n(),
  missing = sum(is.na(heart_attack_risk)),
  fraction_missing = sum(is.na(heart_attack_risk)/total)
)
```

```
## # A tibble: 1 x 3
## total missing fraction_missing
## <int> <int> <dbl>
## 1 3110 0 0
```

```
model_1 <-
  glm(
  heart_attack_risk ~ exercise_hrs_week,
  data = trainData,
  family = binomial()
)</pre>
```

```
model_1_preds <-
  trainData %>%
  add_predictions(
    model_1,
   type = "response"
  ) %>%
  mutate(
    outcome =
     if_else(
        pred >= 0.36,
                       true = 1,
                       false = 0))
model_1_preds %>%
  mutate(
    correct = if_else(
     outcome == heart_attack_risk,
     true = 1,
     false = 0
  ) %>%
  summarize(
   total_correct = sum(correct),
    accuracy = total_correct/n()
## # A tibble: 1 x 2
## total_correct accuracy
##
            <dbl>
                      <dbl>
             1887
                      0.607
## 1
logistic_cv1 <-</pre>
  cv.glm(
   trainData,
  model_1, K = 5
logistic_cv1$delta
## [1] 0.2304086 0.2303714
#Modeling (sleep_hrs)
model_2 <-
  glm(
 heart_attack_risk ~ sleep_hrs,
 data = trainData,
  family = binomial()
model_2_preds <-
trainData %>%
```

```
add_predictions(
    model_2,
    type = "response"
  ) %>%
  mutate(
    outcome =
     if_else(
        pred >0.36,
                       true = 1,
                       false = 0))
model_2_preds %>%
  mutate(
    correct =
      if_else(
      outcome == heart_attack_risk,
      true = 1,
      false = 0
    )
  ) %>%
  summarize(
   total_correct = sum(correct),
    accuracy = total_correct/n()
## # A tibble: 1 x 2
## total_correct accuracy
##
            <dbl>
                      <dbl>
## 1
             1636
                      0.526
logistic_cv2 <-</pre>
  cv.glm(
   trainData,
   model_2,
 K = 5)
logistic_cv2$delta
## [1] 0.2301777 0.2301645
#Third model:
model_3 <-
  glm(
 heart_attack_risk ~ stress_level,
 data = trainData,
 family = binomial()
model_3_preds <-</pre>
trainData %>%
```

```
add_predictions(
    model_3,
    type = "response"
  ) %>%
  mutate(
    outcome =
     if_else(
        pred >0.36,
                       true = 1,
                       false = 0))
model_3_preds %>%
  mutate(
    correct =
      if_else(
      outcome == heart_attack_risk,
      true = 1,
      false = 0
    )
  ) %>%
  summarize(
   total_correct = sum(correct),
    accuracy = total_correct/n()
## # A tibble: 1 x 2
## total_correct accuracy
##
             <dbl>
                      <dbl>
## 1
             1673
                      0.538
logistic_cv3 <-</pre>
  cv.glm(
   trainData,
   model_3,
  K = 5)
logistic_cv3$delta
## [1] 0.2303995 0.2303624
#Fourth model:
model_4 <-
 heart_attack_risk ~ exercise_hrs_week + sleep_hrs + stress_level,
 data = trainData,
  family = binomial()
model_4_preds <-
trainData %>%
```

```
add_predictions(
   model_4,
   type = "response"
  ) %>%
 mutate(
   outcome =
     if_else(pred >= 0.36,
                      true = 1,
                      false = 0))
model_4_preds %>%
summarize(
 min = min(outcome),
 max = max(outcome)
## # A tibble: 1 x 2
      min max
## <dbl> <dbl>
## 1 0 1
model_4_preds %>%
summarize(
 min = min(pred)
## # A tibble: 1 x 1
##
      min
##
   <dbl>
## 1 0.347
model_4_preds %>%
arrange(desc(pred))
## # A tibble: 3,110 x 18
       age sex cholesterol systolic diastolic heart_rate family_history
##
     <dbl> <chr>
                     <dbl> <chr>
                                      <chr>
                                               <dbl> <dbl>
      75 Female
                         334 157
                                      103
                                                       92
## 1
                                                                      0
## 2 48 Male
                        212 180
                                      78
                                                      108
                                                                      0
## 3 90 Female
                       244 100
                                      60
                                                       85
                                                                      0
## 4 37 Male
                        272 132
                                      88
                                                       44
                                                                      0
                        327 100
      27 Male
## 5
                                      105
                                                       89
                                                                      0
## 6 19 Female
                        296 98
                                      93
                                                       97
                                                                      0
## 7
      67 Male
                         251 147
                                      107
                                                       69
                                                                      0
## 8
       20 Female
                         165 119
                                      79
                                                       96
                                                                      0
## 9
       60 Female
                         335 111
                                      74
                                                       62
                                                                      0
## 10
        90 Female
                         160 170
                                                       56
## # i 3,100 more rows
## # i 11 more variables: exercise_hrs_week <dbl>, stress_level <dbl>,
## #
      sleep_hrs <dbl>, heart_attack_risk <dbl>, triglycerides <dbl>,
      physical_days_week <dbl>, sedentary_hrs_day <dbl>, diet <chr>, count <chr>,
## #
      pred <dbl>, outcome <dbl>
```

```
model_4_preds %>%
  mutate(
    correct =
     if else(
      outcome == heart_attack_risk,
      true = 1,
     false = 0
    )
  ) %>%
  summarize(
   total_correct = sum(correct),
    accuracy = total_correct/n()
## # A tibble: 1 x 2
## total_correct accuracy
##
         <dbl>
                     <dbl>
## 1
             1648
                      0.530
logistic_cv4 <-</pre>
  cv.glm(
    trainData,
         model_4,
  K = 5)
logistic_cv4$delta
## [1] 0.2309549 0.2308533
\#\# Model 5:
model_5 <-
  glm(
 heart_attack_risk ~ stress_level + exercise_hrs_week,
 data = trainData,
  family = binomial()
model_5_preds <-
  trainData %>%
  add_predictions(
    model_5,
    type = "response"
  ) %>%
  mutate(
    outcome =
      if_else(
        condition = pred >0.36,
                      true = 1,
                       false = 0))
```

```
model_5_preds %>%
  mutate(
    correct =
     if else(
      condition = heart_attack_risk == outcome,
      true = 1,
      false = 0
    )
  ) %>%
  summarize(
   total_correct = sum(correct),
    accuracy = total_correct/n()
## # A tibble: 1 x 2
## total_correct accuracy
##
             <dbl>
                      <dbl>
             1676
                      0.539
## 1
logistic_cv5 <-</pre>
  cv.glm(
    trainData,
         model_5,
  K = 5)
logistic_cv5$delta
```

## [1] 0.2303419 0.2303112

## Model 1 has the highest accuracy!

Accuracy and cross validation error on test dataset:

```
model_1_preds_test %>%
mutate(
    correct = if_else(
    outcome == heart_attack_risk,
```

```
true = 1,
      false = 0
    )
  ) %>%
  summarize(
    total_correct = sum(correct),
    accuracy = total_correct/n()
## # A tibble: 1 x 2
   total_correct accuracy
##
            <dbl>
                       <dbl>
## 1
               823
                       0.617
logistic_cv1_test <-</pre>
  cv.glm(
    testData,
    model_1, K = 5
## Warning in y - yhat: longer object length is not a multiple of shorter object
## length
## Warning in y - yhat: longer object length is not a multiple of shorter object
## length
## Warning in y - yhat: longer object length is not a multiple of shorter object
## length
## Warning in y - yhat: longer object length is not a multiple of shorter object
## length
## Warning in y - yhat: longer object length is not a multiple of shorter object
## length
logistic_cv1_test$delta
## [1] 0.2285452 0.2288119
#Hypothesis testing: (Sukyoung Yoon)
#Change the heart_attack_risk variable to a character object.
  no_family_history_dataset$heart_attack_risk <-</pre>
  as.character(
    no_family_history_dataset$heart_attack_risk)
#Separated the stress_level variable into two categories (True indicates stress and False indicates no stress)
no_family_history_dataset <-</pre>
  no_family_history_dataset %>%
  mutate(
```

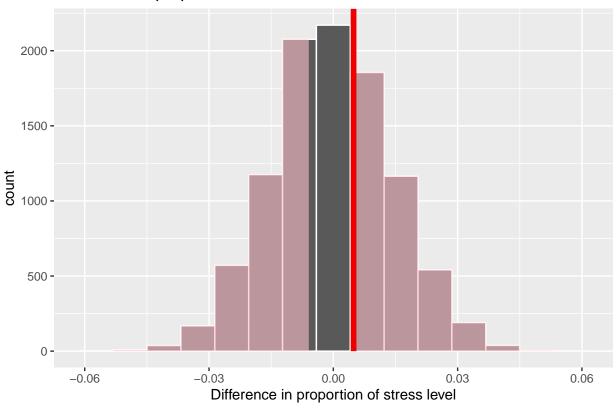
```
stress =
  ifelse(
    stress_level >= 6,
    "True",
  ifelse(
    stress_level >= 1,
    "False", NA
)))
```

#### Changing the stress variable to a character

```
no_family_history_dataset$stress <-</pre>
 as.character(
   no_family_history_dataset$stress)
str(no_family_history_dataset)
## tibble [4,443 x 17] (S3: tbl_df/tbl/data.frame)
## $ age
                       : num [1:4443] 67 21 90 84 20 38 50 60 66 45 ...
## $ sex
                       : chr [1:4443] "Male" "Female" "Male" "Male" ...
## $ cholesterol : num [1:4443] 208 324 358 220 145 166 303 145 340 294 ...
## $ systolic
                     : chr [1:4443] "158" "174" "102" "131" ...
                     : chr [1:4443] "88" "99" "73" "68" ...
## $ diastolic
## $ heart_rate : num [1:4443] 72 72 84 107 68 56 104 71 69 66 ...
## $ family_history : num [1:4443] 0 0 0 0 0 0 0 0 0 0 ...
## $ exercise_hrs_week : num [1:4443] 4.17 2.08 4.1 3.43 16.87 ...
                     : num [1:4443] 9 9 7 4 5 9 1 8 1 9 ...
## $ stress_level
## $ sleep hrs
                       : num [1:4443] 6 4 10 7 4 6 5 7 10 6 ...
## $ heart_attack_risk : chr [1:4443] "0" "0" "1" "1" ...
                    : num [1:4443] 286 587 284 370 790 402 517 247 747 360 ...
## $ triglycerides
## $ physical_days_week: num [1:4443] 0 4 4 6 7 0 1 7 1 4 ...
## $ sedentary hrs day : num [1:4443] 6.615 9.463 0.627 10.544 11.349 ...
                       : chr [1:4443] "Average" "Healthy" "Healthy" "Average" ...
## $ diet
## $ count
                       : chr [1:4443] "stress and no risk" "stress and no risk" "stress & risk" "no st
                       : chr [1:4443] "True" "True" "True" "False" ...
## $ stress
#Test for a difference in proportions (True - False) with stress
heart_null <- no_family_history_dataset %>%
 specify(
   heart_attack_risk ~ stress,
   success = "1") %>%
 hypothesize(
   null ="independence") %>%
 generate(
   reps = 10000,
   type = "permute") %>%
 calculate(
   stat = "diff in props",
   order = c("True", "False"))
```

```
heart_obs_stat <-
 no_family_history_dataset %>%
 specify(
   heart_attack_risk ~ stress,
   success = "1") %>%
 calculate(
   stat = "diff in props",
   order = c("True", "False"))
heart_null %>%
 get_p_value(
   obs_stat = heart_obs_stat,
  direction = "both")
## # A tibble: 1 x 1
## p_value
##
     <dbl>
## 1 0.759
heart_null %>%
 visualize() +
 shade_p_value(
   obs_stat = heart_obs_stat,
   direction = "both") +
   title = "Difference in proportion of stress level",
   x = "Difference in proportion of stress level",
  y = "count")
```

# Difference in proportion of stress level



```
no_family_history_dataset <-
no_family_history_dataset %>%
mutate(
    yes_no =
    ifelse(
        heart_attack_risk == 0,
        "yes",
    ifelse(
        heart_attack_risk == 1,
        "no", NA
))))
```

```
unique(
  no_family_history_dataset$sleep_hrs)
```

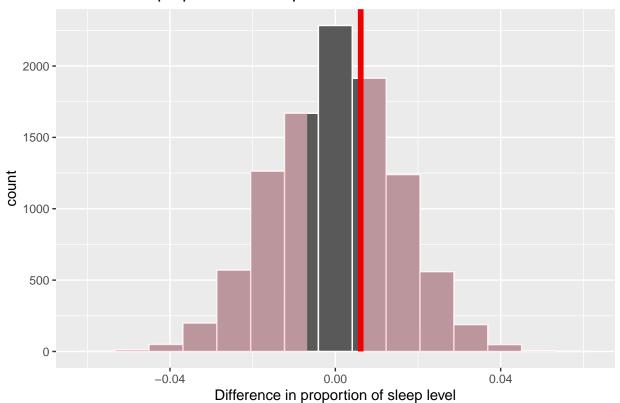
```
## [1] 6 4 10 7 5 8 9
```

#separating sleep into high and low

```
no_family_history_dataset <-</pre>
  no_family_history_dataset %>%
  mutate(
    sleep =
      ifelse(
        sleep_hrs >= 7,
        "High",
      ifelse(
        sleep_hrs < 7,</pre>
        "Low", NA
  )))
heart_null2 <- no_family_history_dataset %>%
  specify(
    yes_no ~ sleep,
    success = "yes") %>%
  hypothesize(
    null = "independence") %>%
  generate(
    reps = 10000,
    type = "permute") %>%
  calculate(
    stat = "diff in props",
    order=c("High", "Low"))
heart_obs_2 <-
  no_family_history_dataset %>%
  specify(
    yes_no ~ sleep,
    success = "yes") %>%
  calculate(
    stat = "diff in props",
    order=c("High", "Low"))
heart_null2 %>%
  get_p_value(
    obs_stat = heart_obs_2,
    direction = "both")
## # A tibble: 1 x 1
##
   p_value
       <dbl>
##
## 1 0.704
heart_null2 %>%
  visualize() +
  shade_p_value(
    obs_stat =
     heart_obs_2,
    direction = "both") +
    title = "Difference in proportion of sleep level",
```

```
x = "Difference in proportion of sleep level",
y = "count")
```

### Difference in proportion of sleep level



#Hypothesis test difference in proprotions for exercise\_hrs:

#Separating exercise hours into high and low categories.

```
heart_null3 <-
  no_family_history_dataset %>%
  specify(
    yes_no ~ exercise,
    success = "yes") %>%
  hypothesize(
    null = "independence") %>%
  generate(
    reps = 10000,
```

```
type = "permute") %>%
  calculate(
   stat = "diff in props",
   order=c("High", "Low"))
heart_obs_3 <-
 no_family_history_dataset %>%
 specify(
   yes_no ~ exercise,
   success = "yes") %>%
 calculate(
   stat = "diff in props",
   order=c("High", "Low"))
heart_null3 %>%
 get_p_value(
   obs_stat = heart_obs_3,
 direction = "both")
## # A tibble: 1 x 1
## p_value
## <dbl>
## 1 0.183
heart_null3 %>%
 visualize() +
 shade_p_value(
   obs_stat = heart_obs_3,
   direction = "both") +
 labs(
   title = "Difference in proportion of
   exercise hour category",
   x = "Difference in proportion
   of exercise hour category",
   y = "count")
```

# Difference in proportion of exercise hour category

