

RMproject2

Objective 1: Display the ability to perform EDA and build a logistic regression model for interpretation purposes.

Perform your multiple logistic regression analysis and provide interpretation of the regression coefficients including hypothesis testing, and confidence intervals. For simplicity sake, you do not need to include interactions with this model. Comment on the practical vs statistical significance of the deemed important factors.

note the difference between a model that is interpretable versus a model that is complex. (A model with a lot of predictors can still be interpretable) Interactions models shouldn't be used here as you should display your ability to interpret the regression coefficients. Effects plots may be used in addition too but not at the exclusions of coefficient interpretation.

Assessing risk factors and predicting if a woman with osteoporosis will have a bone fracture within the first year after joining the study. ?glow_bonemed for data description of variables.

```
#install.packages("aplore3")
library(aplore3)
## Warning: package 'aplore3' was built under R version 4.3.3
?glow_bonemed
## starting httpd help server ... done
library(ggplot2)
library(corrplot)
## corrplot 0.92 loaded
library(tidyverse)
## — Attaching core tidyverse packages ————— tidyverse
2.0.0 —
## ✓ dplyr      1.1.4      ✓ readr      2.1.5
## ✓ forcats   1.0.0      ✓ stringr    1.5.1
## ✓ lubridate 1.9.3      ✓ tibble     3.2.1
## ✓ purrr     1.0.2      ✓ tidyr      1.3.1
## — Conflicts —————
tidyverse_conflicts() —
## ✗ dplyr::filter() masks stats::filter()
## ✗ dplyr::lag()     masks stats::lag()
## ⓘ Use the conflicted package (<http://conflicted.r-lib.org/>) to force all
conflicts to become errors
library(GGally)
## Registered S3 method overwritten by 'GGally':
```

```
## method from
## +.gg ggplot2
```

```
library(dplyr)
```

```
#EDA
```

```
data("glow_bonemed")
```

```
head(glow_bonemed) # see rows
```

```
## sub_id site_id phy_id priorfrac age weight height bmi premeno
momfrac
## 1 1 1 14 No 62 70.3 158 28.16055 No
No
## 2 2 4 284 No 65 87.1 160 34.02344 No
No
## 3 3 6 305 Yes 88 50.8 157 20.60936 No
Yes
## 4 4 6 309 No 82 62.1 160 24.25781 No
No
## 5 5 1 37 No 61 68.0 152 29.43213 No
No
## 6 6 5 299 Yes 67 68.0 161 26.23356 No
No

## armassist smoke raterisk fracscore fracture bonemed bonemed_fu bonetreat
## 1 No No Same 1 No No No No
## 2 No No Same 2 No No No No
## 3 Yes No Less 11 No No No No
## 4 No No Less 5 No No No No
## 5 No No Same 1 No No No No
## 6 No Yes Same 4 No No No No
```

```
tail(glow_bonemed)
```

```
## sub_id site_id phy_id priorfrac age weight height bmi premeno
momfrac
## 495 495 2 70 No 87 64.4 158 25.79715 No
Yes
## 496 496 5 287 Yes 79 63.5 157 25.76169 No
Yes
## 497 497 5 296 No 64 48.1 149 21.66569 No
No
```



```
## 1st Qu.: 59.90 1st Qu.:157.0 1st Qu.:23.27 Yes: 97 Yes: 65
Yes:188
## Median : 68.00 Median :161.5 Median :26.42
## Mean : 71.82 Mean :161.4 Mean :27.55
## 3rd Qu.: 81.30 3rd Qu.:165.0 3rd Qu.:30.79
## Max. :127.00 Max. :199.0 Max. :49.08
## smoke raterisk fracscore fracture bonemed bonemed_fu
## No :465 Less :167 Min. : 0.000 No :375 No :371 No :361
## Yes: 35 Same :186 1st Qu.: 2.000 Yes:125 Yes:129 Yes:139
## Greater:147 Median : 3.000
## Mean : 3.698
## 3rd Qu.: 5.000
## Max. :11.000
## bonetreat
## No :382
## Yes:118
##
##
##
##
```

```
sum(is.na(glow_bonemed))# missing values are 0 , if not could use to find
specifc na - sapply(glow_bonemed, function(x) sum(is.na(x)))
```

```
## [1] 0
```

```
str(glow_bonemed) # see structure, integer, factor, levels etc.
```

```
## 'data.frame': 500 obs. of 18 variables:
## $ sub_id : int 1 2 3 4 5 6 7 8 9 10 ...
## $ site_id : int 1 4 6 6 1 5 5 1 1 4 ...
## $ phy_id : int 14 284 305 309 37 299 302 36 8 282 ...
## $ priorfrac : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 2 1 2 2 1 ...
## $ age : int 62 65 88 82 61 67 84 82 86 58 ...
## $ weight : num 70.3 87.1 50.8 62.1 68 68 50.8 40.8 62.6 63.5 ...
## $ height : int 158 160 157 160 152 161 150 153 156 166 ...
## $ bmi : num 28.2 34 20.6 24.3 29.4 ...
## $ premeno : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ momfrac : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
```

```
## $ armassist : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
## $ smoke      : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 2 1 1 1 1 ...
## $ raterisk   : Factor w/ 3 levels "Less","Same",...: 2 2 1 1 2 2 1 2 2 1
...
## $ fracscore  : int  1 2 11 5 1 4 6 7 7 0 ...
## $ fracture   : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ bonemed    : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
## $ bonemed_fu: Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
## $ bonetreat  : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
```

plots

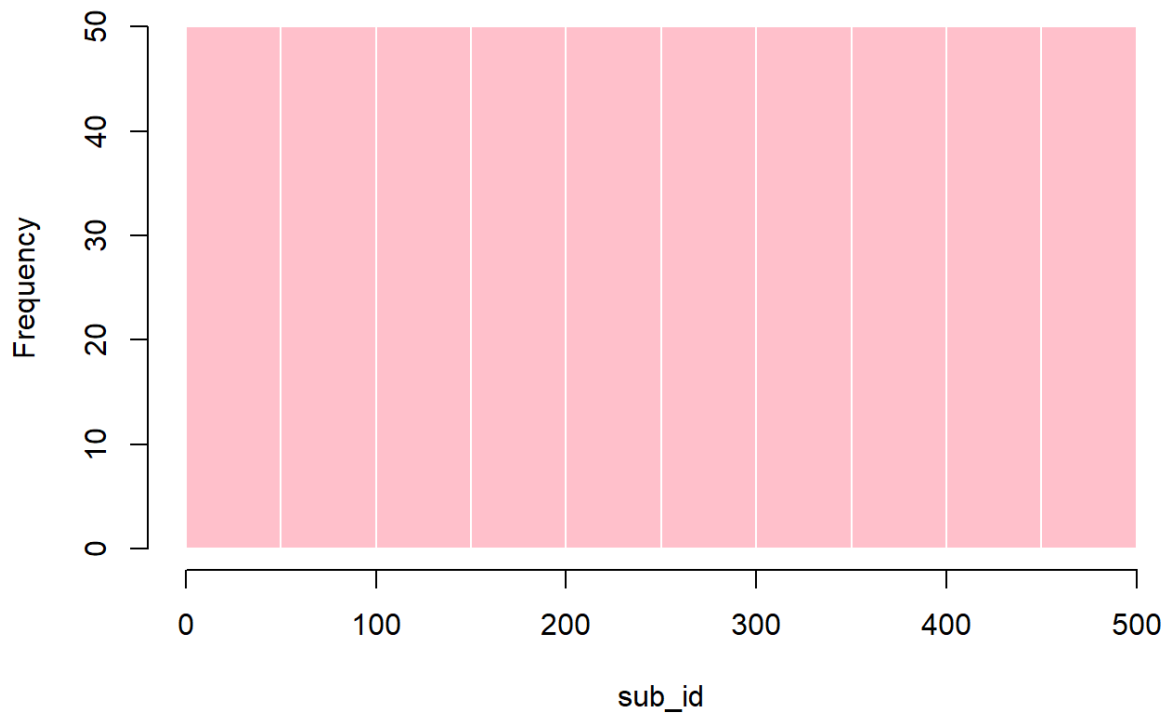
```
#simple histograms

#hist(glow_bonemed$age, main = "Age Distribution", xlab = "Age", col =
"pink", border = "white") # added color, choose a few simple plots for slide
show

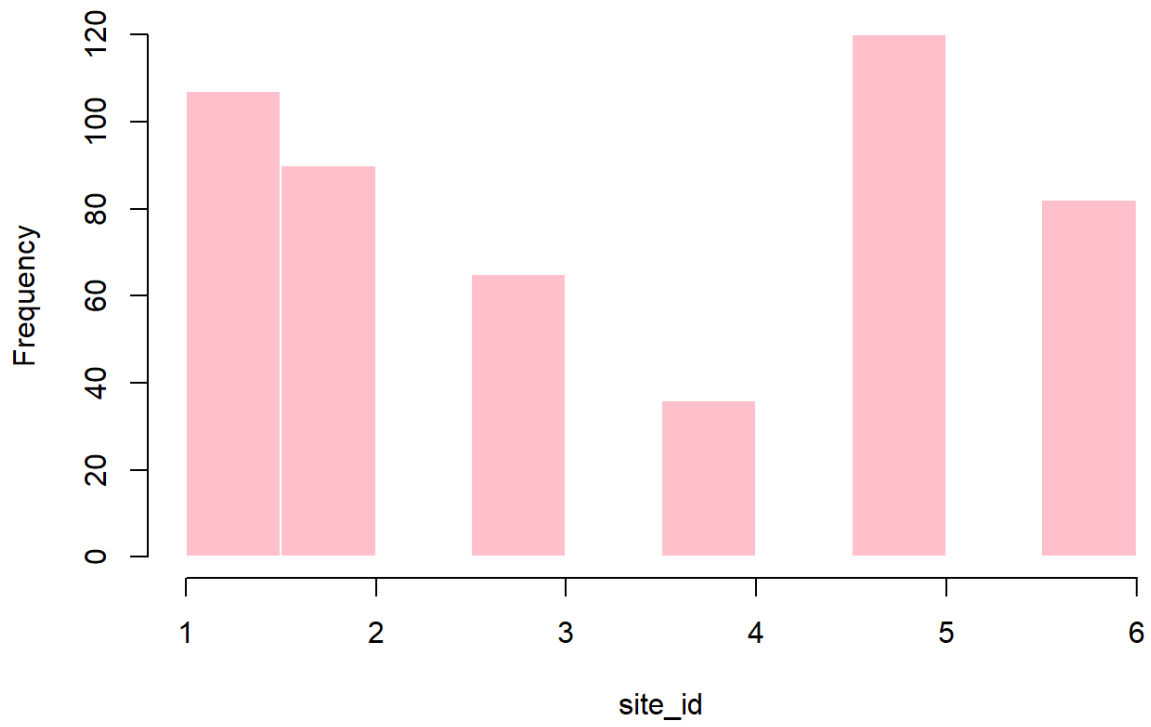
#to loop through all numeric Identify numeric variables
numeric_variables <- sapply(glow_bonemed, is.numeric)

# Loop histograms
lapply(names(glow_bonemed)[numeric_variables], function(var_name) {
  hist(glow_bonemed[[var_name]],
    main = paste(var_name, "Distribution"),
    xlab = var_name,
    col = "pink",
    border = "white")
})
```

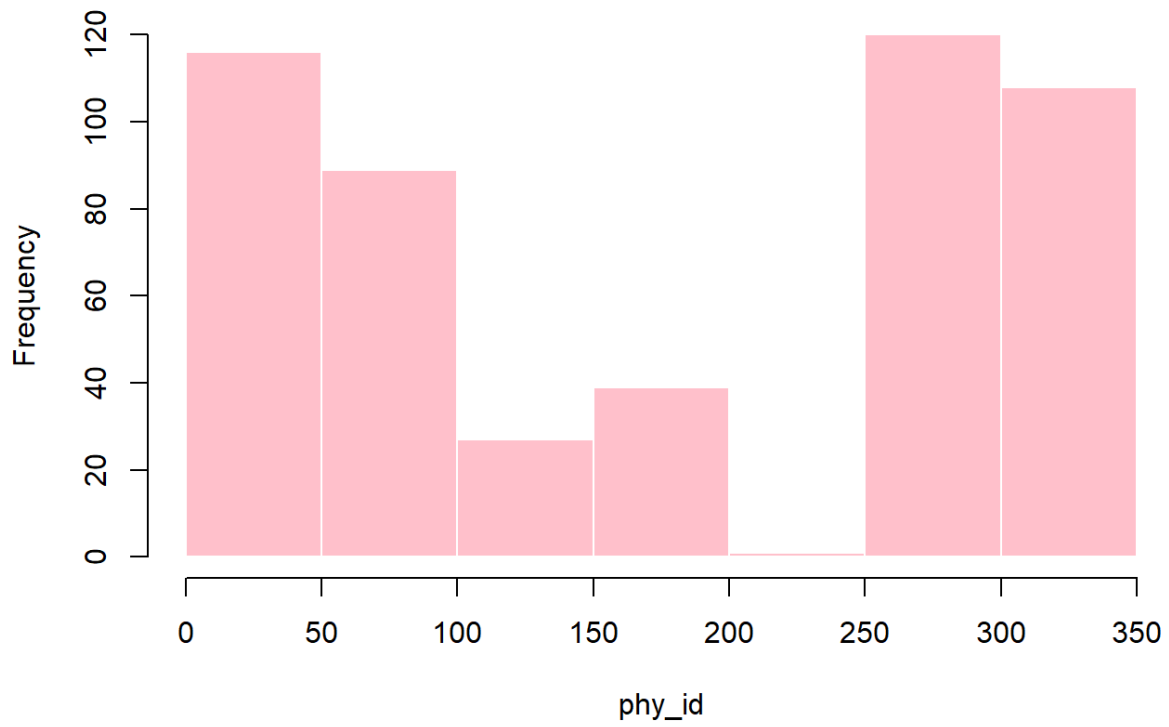
sub_id Distribution



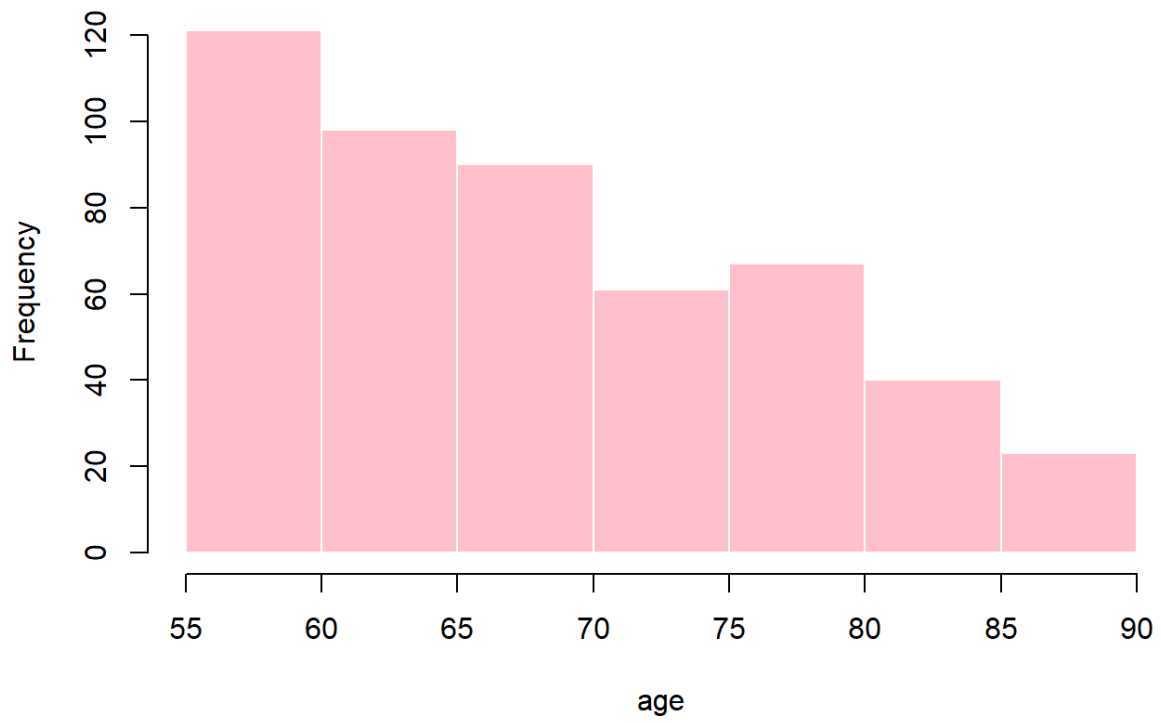
site_id Distribution



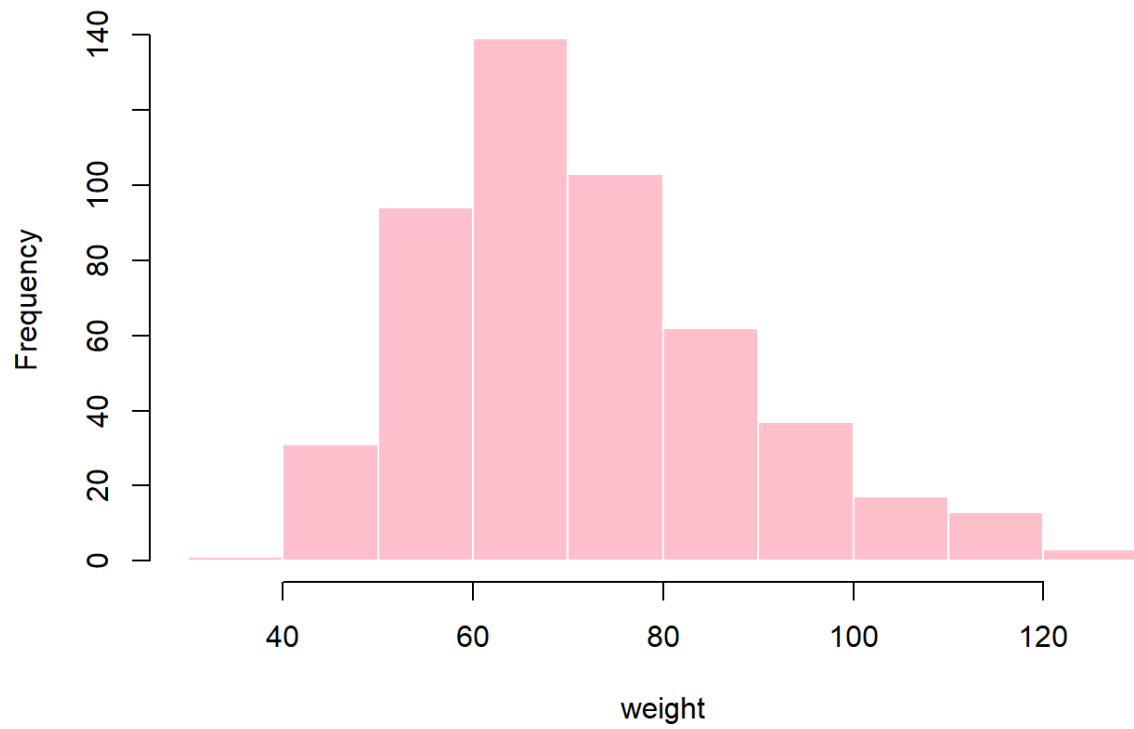
phy_id Distribution

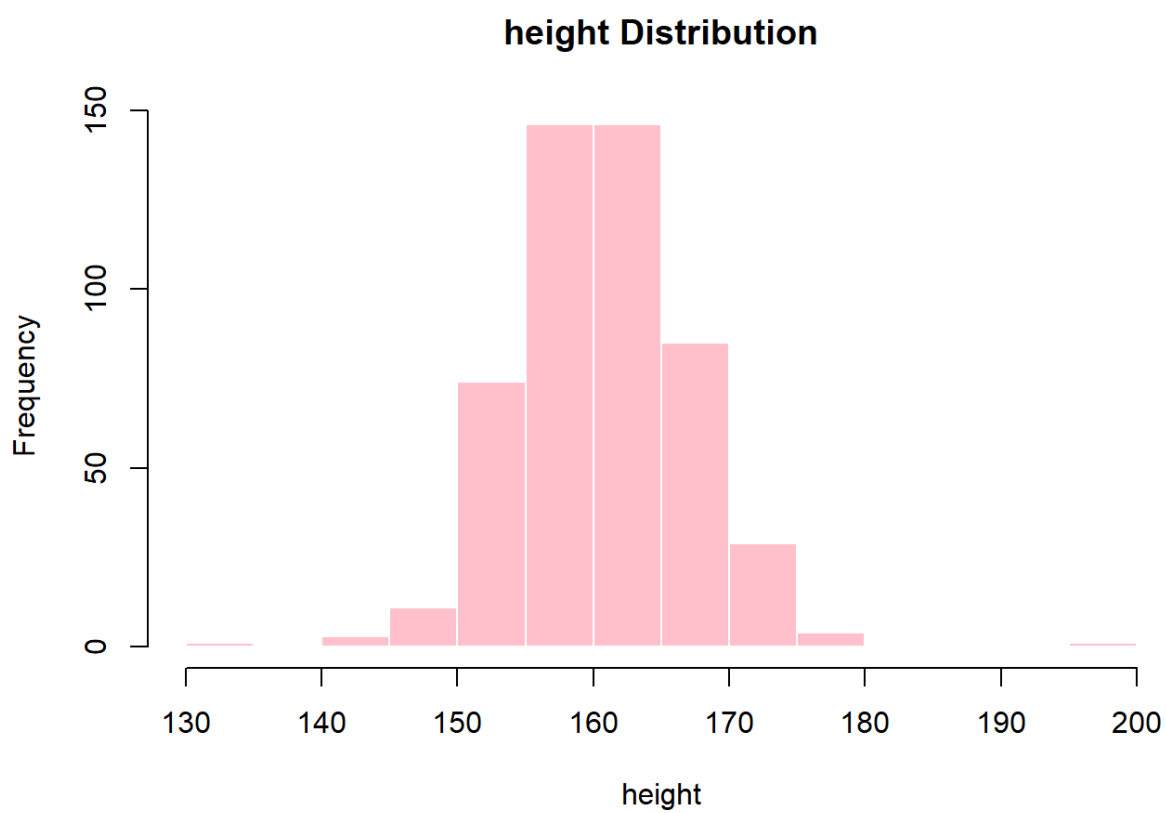


age Distribution

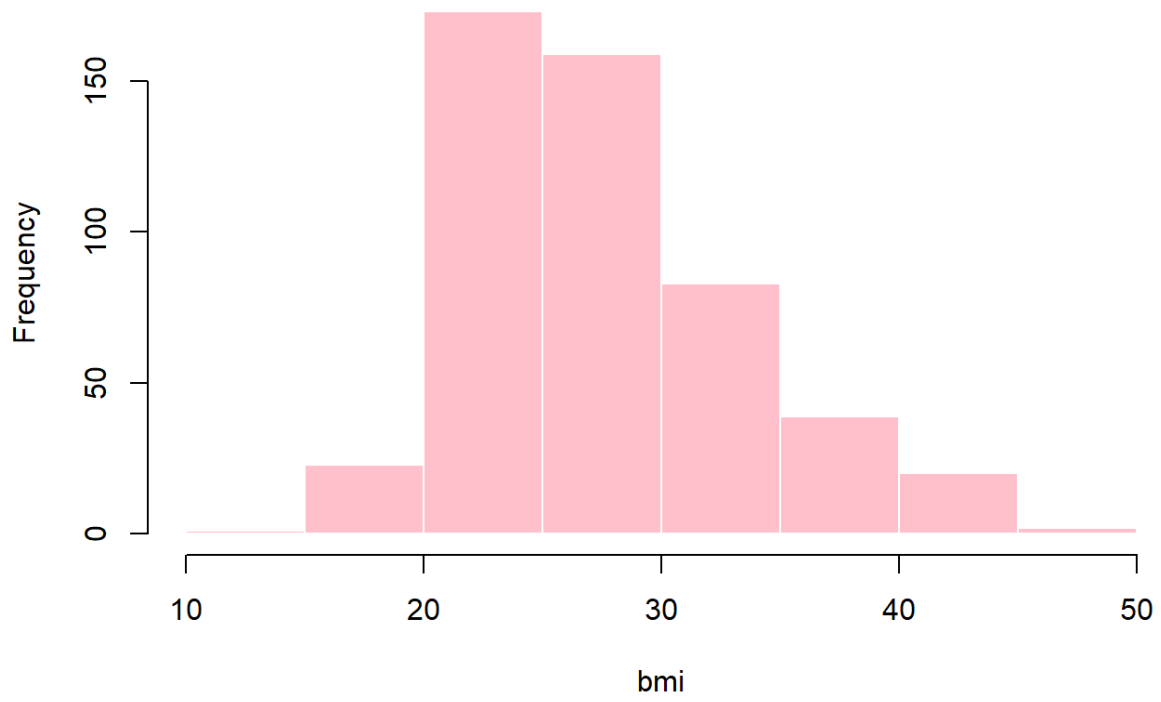


weight Distribution

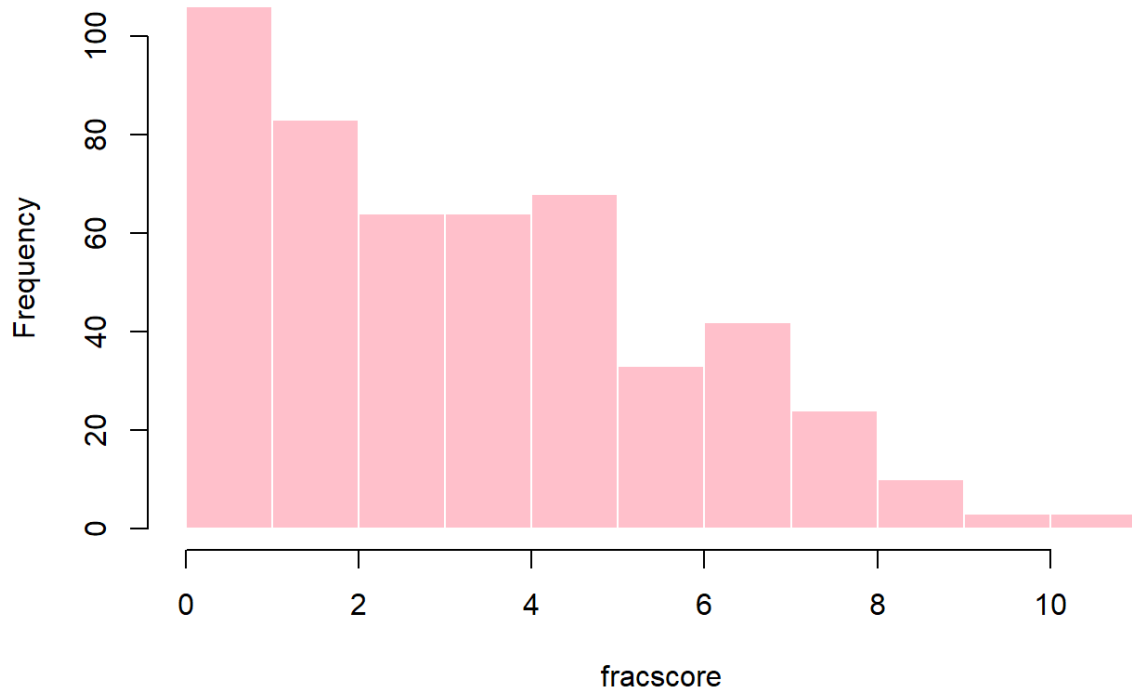




bmi Distribution



fracscore Distribution



```
## [[1]]
## $breaks
## [1] 0 50 100 150 200 250 300 350 400 450 500
##
## $counts
## [1] 50 50 50 50 50 50 50 50 50 50
##
## $density
## [1] 0.002 0.002 0.002 0.002 0.002 0.002 0.002 0.002 0.002 0.002
##
## $mids
## [1] 25 75 125 175 225 275 325 375 425 475
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
```

```
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## [[2]]
## $breaks
## [1] 1.0 1.5 2.0 2.5 3.0 3.5 4.0 4.5 5.0 5.5 6.0
##
## $counts
## [1] 107 90 0 65 0 36 0 120 0 82
##
## $density
## [1] 0.428 0.360 0.000 0.260 0.000 0.144 0.000 0.480 0.000 0.328
##
## $mids
## [1] 1.25 1.75 2.25 2.75 3.25 3.75 4.25 4.75 5.25 5.75
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## [[3]]
## $breaks
## [1] 0 50 100 150 200 250 300 350
##
## $counts
## [1] 116 89 27 39 1 120 108
```

```
##
## $density
## [1] 0.00464 0.00356 0.00108 0.00156 0.00004 0.00480 0.00432
##
## $mids
## [1] 25 75 125 175 225 275 325
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE
##
## attr("class")
## [1] "histogram"
##
## [[4]]
## $breaks
## [1] 55 60 65 70 75 80 85 90
##
## $counts
## [1] 121 98 90 61 67 40 23
##
## $density
## [1] 0.0484 0.0392 0.0360 0.0244 0.0268 0.0160 0.0092
##
## $mids
## [1] 57.5 62.5 67.5 72.5 77.5 82.5 87.5
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE
```

```

##
## attr(,"class")
## [1] "histogram"
##
## [[5]]
## $breaks
## [1] 30 40 50 60 70 80 90 100 110 120 130
##
## $counts
## [1] 1 31 94 139 103 62 37 17 13 3
##
## $density
## [1] 0.0002 0.0062 0.0188 0.0278 0.0206 0.0124 0.0074 0.0034 0.0026 0.0006
##
## $mids
## [1] 35 45 55 65 75 85 95 105 115 125
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## [[6]]
## $breaks
## [1] 130 135 140 145 150 155 160 165 170 175 180 185 190 195 200
##
## $counts
## [1] 1 0 3 11 74 146 146 85 29 4 0 0 0 1
##
## $density

```



```

## [1] 0.0004 0.0000 0.0012 0.0044 0.0296 0.0584 0.0584 0.0340 0.0116 0.0016
## [11] 0.0000 0.0000 0.0000 0.0004
##
## $mids
## [1] 132.5 137.5 142.5 147.5 152.5 157.5 162.5 167.5 172.5 177.5 182.5
187.5
## [13] 192.5 197.5
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## [[7]]
## $breaks
## [1] 10 15 20 25 30 35 40 45 50
##
## $counts
## [1] 1 23 173 159 83 39 20 2
##
## $density
## [1] 0.0004 0.0092 0.0692 0.0636 0.0332 0.0156 0.0080 0.0008
##
## $mids
## [1] 12.5 17.5 22.5 27.5 32.5 37.5 42.5 47.5
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE

```

```

##
## attr("class")
## [1] "histogram"
##
## [[8]]
## $breaks
## [1] 0 1 2 3 4 5 6 7 8 9 10 11
##
## $counts
## [1] 106 83 64 64 68 33 42 24 10 3 3
##
## $density
## [1] 0.212 0.166 0.128 0.128 0.136 0.066 0.084 0.048 0.020 0.006 0.006
##
## $mids
## [1] 0.5 1.5 2.5 3.5 4.5 5.5 6.5 7.5 8.5 9.5 10.5
##
## $xname
## [1] "glow_bonemed[[var_name]]"
##
## $equidist
## [1] TRUE
##
## attr("class")
## [1] "histogram"

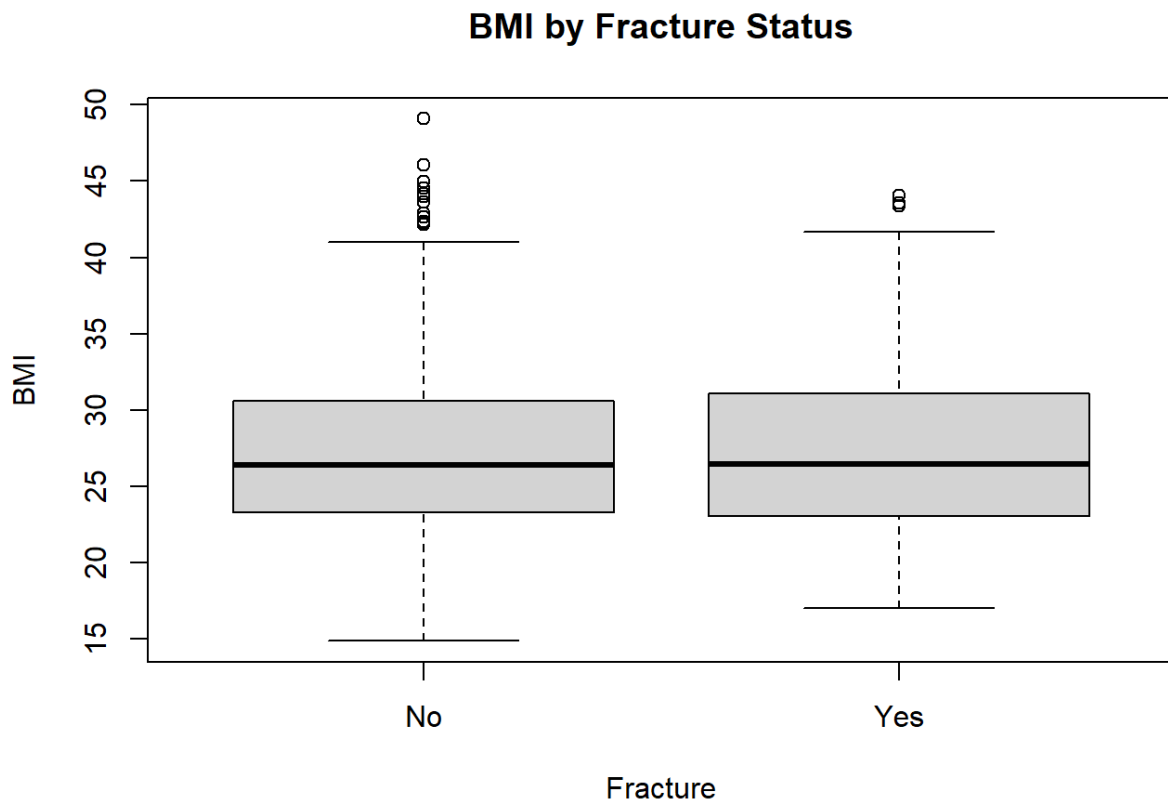
```

#simple boxplots

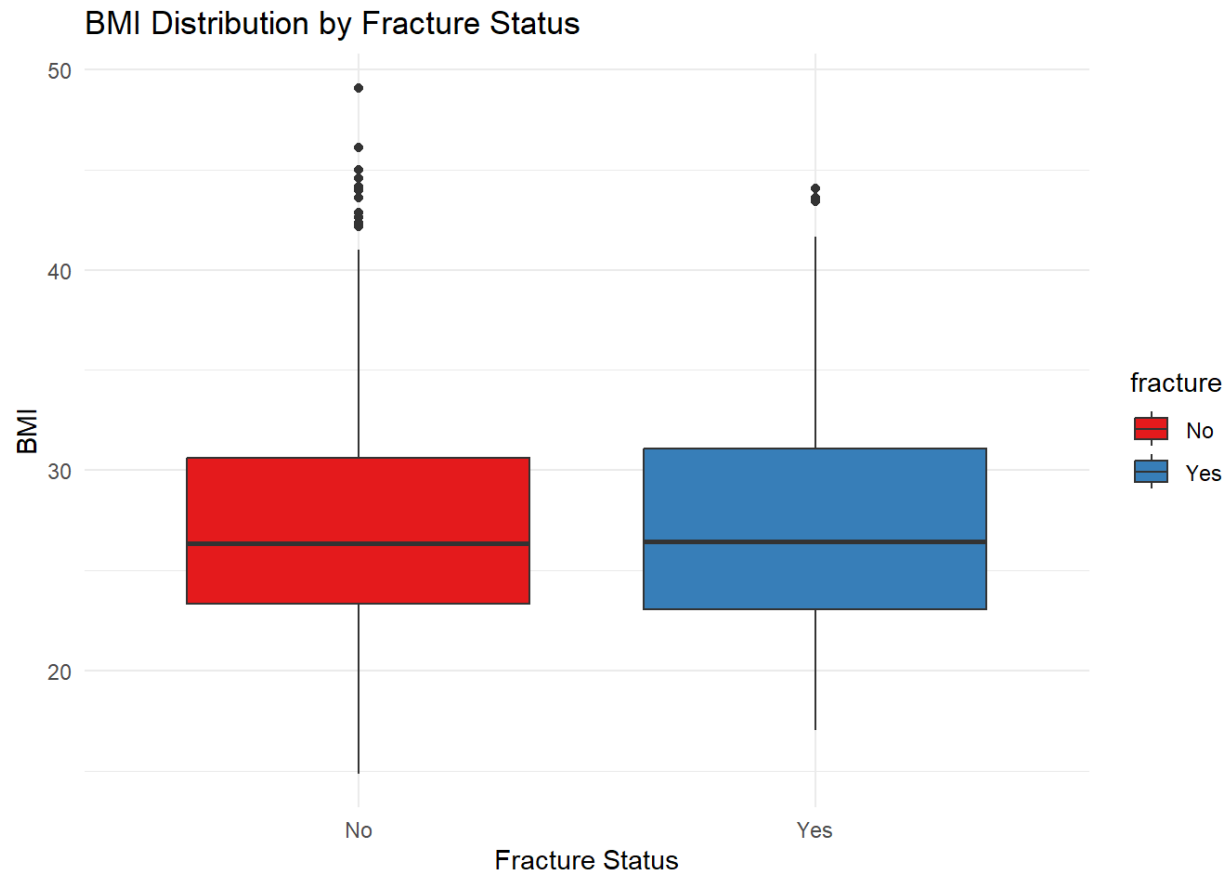
```

boxplot(bmi ~ fracture, data=glow_bonemed, main="BMI by Fracture Status",
xlab="Fracture", ylab="BMI")

```



```
# Creating boxplots for BMI by fracture status
ggplot(glow_bonemed, aes(x = fracture, y = bmi, fill = fracture)) +
  geom_boxplot() +
  scale_fill_brewer(palette = "Set1") +
  labs(y = "BMI", x = "Fracture Status", title = "BMI Distribution by
Fracture Status") +
  theme_minimal()
```



```
# position="dodge" places the bars side by side instead of stacking them

#corr matrix
numeric_vars <- c("age", "weight", "height", "bmi", "fracscore")

cor_matrix <- cor(glow_bonemed[,numeric_vars])
corrplot(cor_matrix, method = "circle")
```

```
#scatterplot matrix
numeric_variables <- c("age", "weight", "height", "bmi", "fracscore")
ggpairs(glow_bonemed[, c(numeric_variables, "fracture")], aes(colour = fracture))

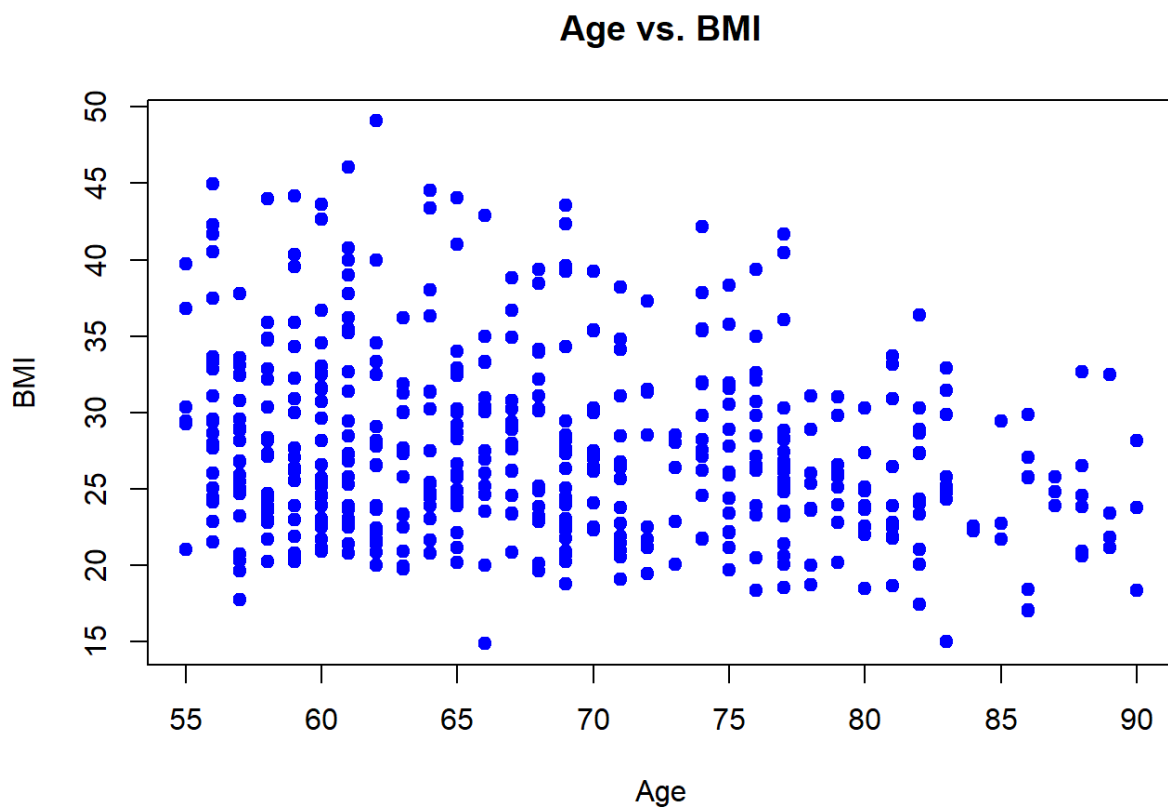
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
# as age increases fracture score increases
```

```
# simple scatterplot
```

```
plot(glow_bonemed$age, glow_bonemed$bmi, main = "Age vs. BMI", xlab = "Age",
     ylab = "BMI", pch = 19, col = "blue")
```



```
head(glow_bonemed)
```

```
##   sub_id site_id phy_id priorfrac age weight height      bmi premeno
momfrac
## 1      1      1     14        No  62   70.3   158 28.16055      No
## 2      2      4    284        No  65   87.1   160 34.02344      No
```

```

## 3      3      6    305      Yes  88   50.8    157 20.60936      No
Yes
## 4      4      6    309      No  82   62.1    160 24.25781      No
No
## 5      5      1     37      No  61   68.0    152 29.43213      No
No
## 6      6      5    299      Yes  67   68.0    161 26.23356      No
No

##   armassist smoke raterisk fracscore fracture bonemed bonemed_fu bonetreat
## 1      No    No    Same           1      No      No      No      No
## 2      No    No    Same           2      No      No      No      No
## 3     Yes    No   Less           11     No      No      No      No
## 4      No    No   Less            5     No      No      No      No
## 5      No    No    Same            1     No      No      No      No
## 6      No   Yes    Same            4     No      No      No      No

str(glow_bonemed)
## 'data.frame':   500 obs. of  18 variables:
##  $ sub_id      : int  1 2 3 4 5 6 7 8 9 10 ...
##  $ site_id     : int  1 4 6 6 1 5 5 1 1 4 ...
##  $ phy_id      : int  14 284 305 309 37 299 302 36 8 282 ...
##  $ priorfrac   : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 2 1 2 2 1 ...
##  $ age         : int  62 65 88 82 61 67 84 82 86 58 ...
##  $ weight      : num  70.3 87.1 50.8 62.1 68 68 50.8 40.8 62.6 63.5 ...
##  $ height      : int  158 160 157 160 152 161 150 153 156 166 ...
##  $ bmi         : num  28.2 34 20.6 24.3 29.4 ...
##  $ premeno     : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
##  $ momfrac     : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
##  $ armassist   : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
##  $ smoke       : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 2 1 1 1 1 ...
##  $ raterisk    : Factor w/ 3 levels "Less","Same",...: 2 2 1 1 2 2 1 2 2 1
...
##  $ fracscore  : int  1 2 11 5 1 4 6 7 7 0 ...
##  $ fracture    : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
##  $ bonemed     : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
##  $ bonemed_fu : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
##  $ bonetreat  : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...

```

```

#bonemed Bone medications at enrollment (1: No, 2: Yes)
#bonemed_fu Bone medications at follow-up (1: No, 2: Yes)
#bonetreat Bone medications both at enrollment and follow-up (1: No, 2: Yes)
#priorfrac History of Prior Fracture (1: No, 2: Yes)
# age Age at Enrollment (Years)
#fracscore Fracture Risk Score (Composite Risk Score)
#fractureAny fracture in first year (1: No, 2: Yes)

#if want to change, use mutate or pivot
#tidy data (Reshape wide to long), use pivot
#glow_bonemed_long <- glow_bonemed %>%
  # pivot_longer(cols = c(age, weight, height, bmi, fracscore), names_to =
"variable", values_to = "value")

```

logistic regression model #glm (Generalized Linear Model)/binomial family/ fitting a logistic regression model/ predict the log odds of the probability of the outcome occurring

```

str(glow_bonemed) # no codes 1, yes coded 2
## 'data.frame':   500 obs. of  18 variables:
##  $ sub_id      : int   1 2 3 4 5 6 7 8 9 10 ...
##  $ site_id     : int   1 4 6 6 1 5 5 1 1 4 ...
##  $ phy_id      : int  14 284 305 309 37 299 302 36 8 282 ...
##  $ priorfrac   : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 2 1 2 2 1 ...
##  $ age         : int   62 65 88 82 61 67 84 82 86 58 ...
##  $ weight      : num   70.3 87.1 50.8 62.1 68 68 50.8 40.8 62.6 63.5 ...
##  $ height      : int  158 160 157 160 152 161 150 153 156 166 ...
##  $ bmi         : num   28.2 34 20.6 24.3 29.4 ...
##  $ premeno     : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
##  $ momfrac     : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
##  $ armassist   : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
##  $ smoke       : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 2 1 1 1 1 ...
##  $ raterisk    : Factor w/ 3 levels "Less","Same",...: 2 2 1 1 2 2 1 2 2 1 ...
##  $ fracscore   : int    1 2 11 5 1 4 6 7 7 0 ...
##  $ fracture    : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
##  $ bonemed     : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...

```

```
## $ bonemed_fu: Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
## $ bonetreat : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...

levels(glow_bonemed$fracture)

## [1] "No" "Yes"

# Convert factor to its underlying integer codes 1,2
as.numeric(glow_bonemed$fracture)

## [1] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [38] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [75] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [112] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [149] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [186] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [223] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [260] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [297] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [334] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [371] 1 1 1 1 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [408] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [445] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [482] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2

str(glow_bonemed)

## 'data.frame': 500 obs. of 18 variables:
## $ sub_id : int 1 2 3 4 5 6 7 8 9 10 ...
## $ site_id : int 1 4 6 6 1 5 5 1 1 4 ...
## $ phy_id : int 14 284 305 309 37 299 302 36 8 282 ...
## $ priorfrac : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 2 1 2 2 1 ...
```



```
## $ age      : int  62 65 88 82 61 67 84 82 86 58 ...
## $ weight   : num  70.3 87.1 50.8 62.1 68 68 50.8 40.8 62.6 63.5 ...
## $ height   : int  158 160 157 160 152 161 150 153 156 166 ...
## $ bmi      : num  28.2 34 20.6 24.3 29.4 ...
## $ premeno  : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ momfrac  : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
## $ armassist : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 1 1 1 1 1 ...
## $ smoke    : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 2 1 1 1 1 ...
## $ raterisk  : Factor w/ 3 levels "Less","Same",...: 2 2 1 1 2 2 1 2 2 1
...
## $ fracscore : int  1 2 11 5 1 4 6 7 7 0 ...
## $ fracture  : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ bonemed   : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
## $ bonemed_fu: Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...
## $ bonetreat : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 2 1 1 ...

#dependent variable is fracture : Any fracture in first year (1: No, 2: Yes)
#is it predicted by: "priorfrac"      "age"      "weight"
# "height"      "bmi"      "premeno"      "momfrac"
"armassist"      "smoke"
# "raterisk"      "fracscore"      "bonemed"      "bonemed_fu"
"bonetreat"

modell1 <- glm(formula = fracture ~ priorfrac + age + weight + height + bmi +
premeno +
      momfrac + armassist + smoke + raterisk + fracscore + bonemed +
      bonemed_fu + bonetreat,
      family = binomial, data = glow_bonemed)

summary(modell1)

##
## Call:
## glm(formula = fracture ~ priorfrac + age + weight + height +
##      bmi + premeno + momfrac + armassist + smoke + raterisk +
##      fracscore + bonemed + bonemed_fu + bonetreat, family = binomial,
##      data = glow_bonemed)
```

```
##
## Coefficients:
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -17.07822   12.80864  -1.333  0.18242
## priorfracYes    0.52585    0.39600   1.328  0.18420
## age           0.01618    0.05629   0.288  0.77371
## weight        -0.13254    0.08907  -1.488  0.13673
## height         0.08019    0.08120   0.988  0.32340
## bmi           0.36724    0.23143   1.587  0.11255
## premenoYes     0.19494    0.29255   0.666  0.50520
## momfracYes     0.61382    0.42858   1.432  0.15208
## armassistYes   0.16779    0.61066   0.275  0.78349
## smokeYes      -0.27693    0.54660  -0.507  0.61241
## rateriskSame   0.36390    0.28982   1.256  0.20926
## rateriskGreater 0.53726    0.31964   1.681  0.09280 .
## fracscore     0.09721    0.28046   0.347  0.72889
## bonemedYes     1.50963    0.65911   2.290  0.02200 *
## bonemed_fuYes  1.61953    0.50153   3.229  0.00124 **
## bonetreatYes  -2.66789    0.83779  -3.184  0.00145 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 562.34  on 499  degrees of freedom
## Residual deviance: 488.15  on 484  degrees of freedom
## AIC: 520.15
##
## Number of Fisher Scoring iterations: 4
confint(modell1)
## Waiting for profiling to be done...
##           2.5 %      97.5 %
## (Intercept)  -42.22434329  8.02050531
## priorfracYes  -0.24736648  1.30847429
```

```
## age            -0.09386000  0.12727158
## weight         -0.31025521  0.03961953
## height         -0.07963981  0.23885108
## bmi            -0.08069659  0.82854835
## premenoYes     -0.39043906  0.76026365
## momfracYes     -0.22673386  1.45795795
## armassistYes   -1.02811562  1.37123892
## smokeYes       -1.39807971  0.76400830
## rateriskSame   -0.20000696  0.93956144
## rateriskGreater -0.08728069  1.16926797
## fracscore      -0.45491366  0.64695642
## bonemedYes      0.20460939  2.84614275
## bonemed_fuYes   0.64862664  2.63786064
## bonetreatYes   -4.34824962 -1.03068346
```

```
#AIC (Akaike Information Criterion) is 520.15
```

```
# interpretations
```

```
#"bonemed" - 1.50963, p val 0.02200 CI[0.20460939 , 2.84614275]
```

```
# patient on bonemed at enrollment, log odds of having fracture expected to
increase by approx 1.51 compared to those not on meds, other factor beeing
consistent
```

```
#"bonemed_fu" 1.61953 , p val 0.0012 CI[0.64862664 , 2.63786064]
```

```
# patient on bonemed at follow up, log odds of having fracture expected to
increase by approx 1.62 compared to those not on meds, other factor being
consistent high p value, very stats sig.
```

```
#"bonetreat" -2.66789 , p vale 0.00145 CI [-4.34824962 , -1.03068346]
```

```
# patient on bonemed at enrollment & follow up, log odds of having fracture
expected to decrease by approx 5.67 compared to those not on meds, other
factor being consistent high p value, very stats sig. CI is very wide so
uncertianty about this.
```

```
#model fit
```

```
# Predict probabilities of fracture
```

```
#predicted_probs <- predict(modell, type = "response")
```

```

# Adding predicted prob back to the dataset
#glow_bonemed$predicted_prob = predicted_probs
#head(glow_bonemed)

#too
model2 <- glm(formula = fracture ~age + bonemed +
               bonemed_fu + bonetreat,
               family = binomial, data = glow_bonemed)

summary(model2)

##
## Call:
## glm(formula = fracture ~ age + bonemed + bonemed_fu + bonetreat,
##      family = binomial, data = glow_bonemed)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -4.51622    0.85025  -5.312 1.09e-07 ***
## age           0.04472    0.01209   3.699 0.000216 ***
## bonemedYes    1.40723    0.64196   2.192 0.028374 *
## bonemed_fuYes 1.82394    0.47822   3.814 0.000137 ***
## bonetreatYes -2.59557    0.81171  -3.198 0.001386 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 562.34  on 499  degrees of freedom
## Residual deviance: 519.29  on 495  degrees of freedom
## AIC: 529.29
##
## Number of Fisher Scoring iterations: 4
confint(model2)

```

```
## Waiting for profiling to be done...
##               2.5 %       97.5 %
## (Intercept)  -6.20871813 -2.86949335
## age          0.02112541  0.06860985
## bonemedYes   0.13661628  2.71394369
## bonemed_fuYes 0.90229270  2.80188014
## bonetreatYes -4.22572102 -1.00798370
# AIC: 529.29 but now age is stat sig.
```