```
#Load in the packages
```

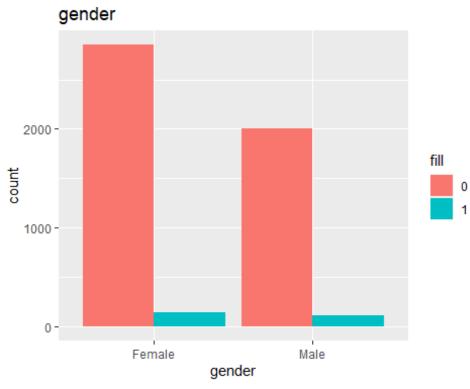
```
library(ggplot2)
library(tidyverse)
## -- Attaching packages ------
                                                  ----- tidyverse
1.3.1 --
## v tibble 3.1.6 v dplyr 1.0.7
## v tidyr 1.1.4 v stringr 1.4.0
## v readr 2.1.1 v forcats 0.5.1
## v purrr 0.3.4
## -- Conflicts -----
tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(dplyr)
#Load in the data
data <- read.csv("C:/Users/oriri/OneDrive/Desktop/DSC680/Project</pre>
2/stroke data.csv")
#cleaning the data
#removing column id
data <- data %>%
  select(-c(id)) %>%
  filter(gender != "Other") %>%
  mutate(stroke = as.factor(stroke))
#getting a count of all the stroke events
data %>%
  group_by(stroke) %>%
  count()
## # A tibble: 2 x 2
## # Groups:
                stroke [2]
## stroke
## <fct> <int>
## 1 0
              4860
```

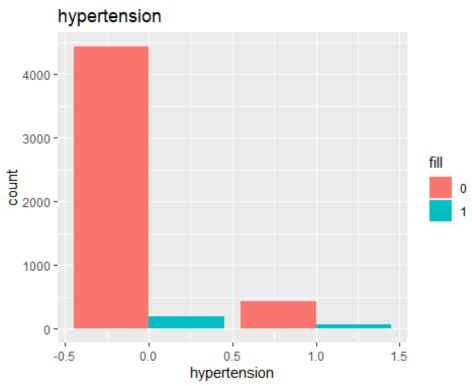
#Visualizing the categorical variables against stroke status

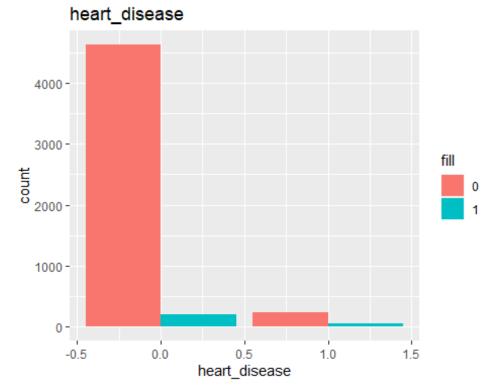
249

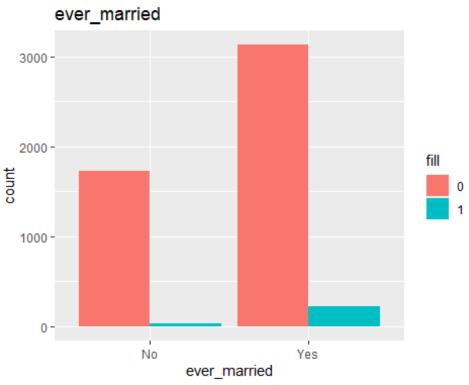
2 1

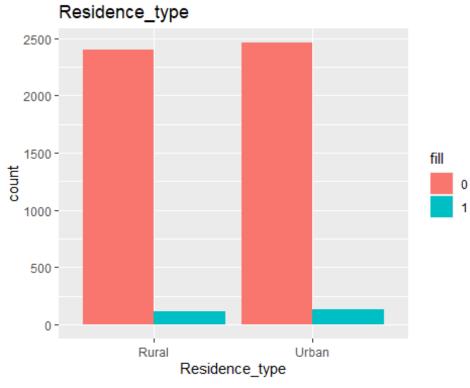
```
#putting all of the categorical variables in a list
cat_list <- c("gender", "hypertension", "heart_disease", "ever_married",
"Residence_type", "smoking_status")</pre>
```

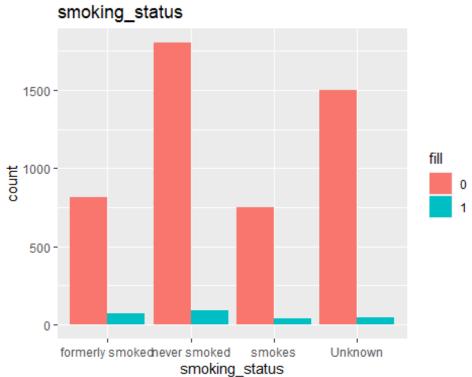












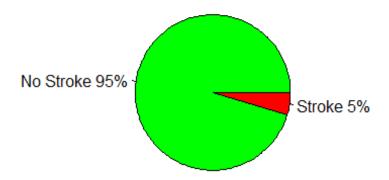
#getting

proportions of the categorical variables #gender

#Count of stroke by gender
data %>%

```
group_by(gender, stroke) %>%
  count()
## # A tibble: 4 x 3
## # Groups: gender, stroke [4]
## gender stroke
## <chr> <fct> <int>
## 1 Female 0
                  2853
## 2 Female 1
                    141
## 3 Male 0
                    2007
## 4 Male 1
                    108
#Female
slices <- c(2853, 141)
lbls <- c("No Stroke", "Stroke")</pre>
pct <- round(slices/sum(slices) * 100)</pre>
lbls <- paste(lbls, pct)</pre>
lbls <- paste(lbls,"%", sep = "")</pre>
pie(slices, labels = lbls, col=c("green", "red"),
main="Female Stroke Events")
```

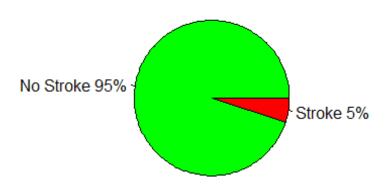
Female Stroke Events



```
#Male
slices <- c(2007, 108)
lbls <- c("No Stroke", "Stroke")
pct <- round(slices/sum(slices) * 100)
lbls <- paste(lbls, pct)
lbls <- paste(lbls,"%", sep = "")</pre>
```

```
pie(slices,labels = lbls, col=c("green", "red"),
    main="Male Stroke Events")
```

Male Stroke Events



```
#2 proportion test
prop.test(x = c(141, 108), n = c(2994, 2115))

##
## 2-sample test for equality of proportions with continuity correction
##
## data: c(141, 108) out of c(2994, 2115)
## X-squared = 0.34, df = 1, p-value = 0.5598
## alternative hypothesis: two.sided
## 95 percent confidence interval:
## -0.016439097 0.008499814
## sample estimates:
## prop 1 prop 2
## 0.04709419 0.05106383
```

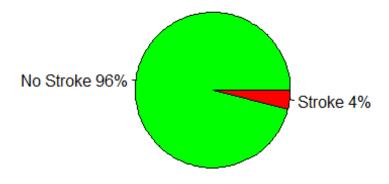
#hypertension

```
#Getting counts of stroke events by hyper tension
data %>%
   group_by(hypertension, stroke) %>%
   count()

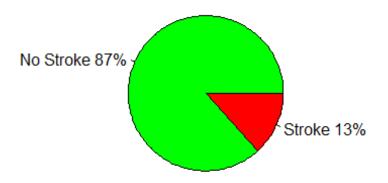
## # A tibble: 4 x 3
## # Groups: hypertension, stroke [4]
## hypertension stroke n
```

```
<int> <fct> <int>
##
                0 0
## 1
                           4428
## 2
                0 1
                            183
                1 0
                            432
## 3
## 4
                1 1
                           66
#No Hypertension
slices <- c(4428, 183)
lbls <- c("No Stroke", "Stroke")</pre>
pct <- round(slices/sum(slices) * 100)</pre>
lbls <- paste(lbls, pct)</pre>
lbls <- paste(lbls,"%", sep = "")</pre>
pie(slices,labels = lbls, col=c("green", "red"),
main="No Hypertension Stroke Events")
```

No Hypertension Stroke Events



Hypertension Stroke Events



```
#2 proportion test
prop.test(x = c(183, 66), n = c(4611, 498))

##

## 2-sample test for equality of proportions with continuity correction
##

## data: c(183, 66) out of c(4611, 498)

## X-squared = 81.573, df = 1, p-value < 2.2e-16

## alternative hypothesis: two.sided

## 95 percent confidence interval:
## -0.1242628 -0.0614220

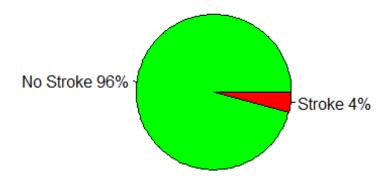
## sample estimates:
## prop 1 prop 2

## 0.0396877 0.1325301</pre>
```

#heart disease

```
## 2
                  0 1
                              202
## 3
                  1 0
                              229
## 4
                  1 1
                               47
#No Hypertension
slices <- c(4631, 202)
lbls <- c("No Stroke", "Stroke")</pre>
pct <- round(slices/sum(slices) * 100)</pre>
lbls <- paste(lbls, pct)</pre>
lbls <- paste(lbls,"%", sep = "")</pre>
pie(slices, labels = lbls, col=c("green", "red"),
main="No Heart Disease Stroke Events")
```

No Heart Disease Stroke Events



Heart Disease Stroke Events



```
#2 proportion test
prop.test(x = c(202, 47), n = c(4833, 276))

##

## 2-sample test for equality of proportions with continuity correction
##

## data: c(202, 47) out of c(4833, 276)

## X-squared = 90.229, df = 1, p-value < 2.2e-16

## alternative hypothesis: two.sided
## 95 percent confidence interval:
## -0.17511205 -0.08187568

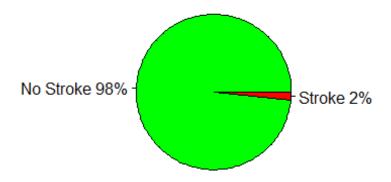
## sample estimates:
## prop 1 prop 2

## 0.04179599 0.17028986</pre>
```

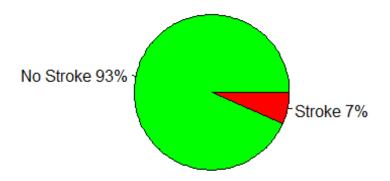
#ever married

```
## 2 No
                              29
## 3 Yes
                   0
                            3133
## 4 Yes
                   1
                             220
#Never Married
slices <- c(1727, 29)
lbls <- c("No Stroke", "Stroke")</pre>
pct <- round(slices/sum(slices) * 100)</pre>
lbls <- paste(lbls, pct)</pre>
lbls <- paste(lbls,"%", sep = "")</pre>
pie(slices,labels = lbls, col=c("green", "red"),
main="Never Married Stroke Events")
```

Never Married Stroke Events



Married Stroke Events

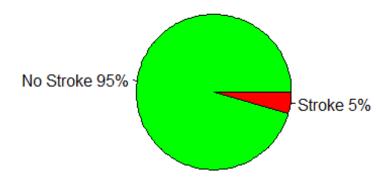


```
#2 proportion test
prop.test(x = c(29, 220), n = c(1156, 3353))

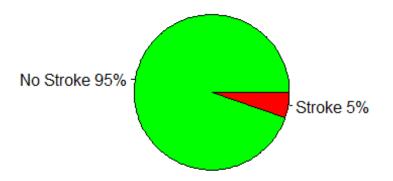
##
## 2-sample test for equality of proportions with continuity correction
##
## data: c(29, 220) out of c(1156, 3353)
## X-squared = 26.289, df = 1, p-value = 2.939e-07
## alternative hypothesis: two.sided
## 95 percent confidence interval:
## -0.05341703 -0.02763572
## sample estimates:
## prop 1 prop 2
## 0.02508651 0.06561288
```

#Residence_type

Rural Residence Stroke Events



Urban Stroke Events



```
#2 proportion test
prop.test(x = c(114, 220), n = c(2513, 2596))

##

## 2-sample test for equality of proportions with continuity correction
##

## data: c(114, 220) out of c(2513, 2596)

## X-squared = 31.77, df = 1, p-value = 1.735e-08

## alternative hypothesis: two.sided

## 95 percent confidence interval:
## -0.05322594 -0.02553738

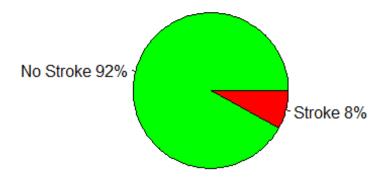
## sample estimates:
## prop 1 prop 2

## 0.04536411 0.08474576
```

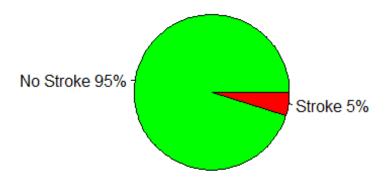
#smoking_status

```
## 3 never smoked
                                1802
## 4 never smoked 1
                                  90
## 5 smokes
                       0
                                 747
                       1
## 6 smokes
                                   42
## 7 Unknown
                       0
                                1497
## 8 Unknown
                                  47
#Formerly Smoked
slices <- c(814, 70)
lbls <- c("No Stroke", "Stroke")</pre>
pct <- round(slices/sum(slices) * 100)</pre>
lbls <- paste(lbls, pct)</pre>
lbls <- paste(lbls,"%", sep = "")
pie(slices,labels = lbls, col=c("green", "red"),</pre>
main="Formerly Smoked Stroke Events")
```

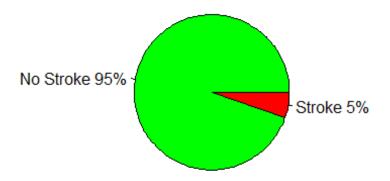
Formerly Smoked Stroke Events



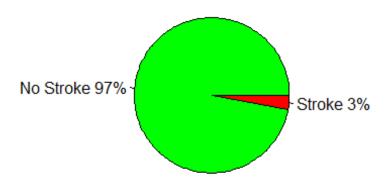
Never Smoked Stroke Events



Smokes Stroke Events



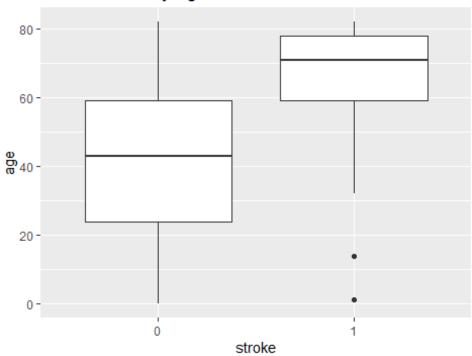
Unknown Stroke Events



#Investigating Continuous Variables

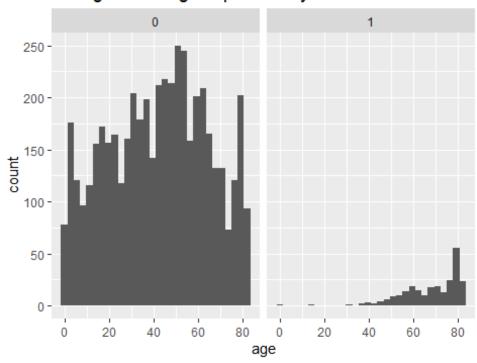
```
numeric_vars %>%
  ggplot(aes(x = stroke, y = age)) +
  geom_boxplot() +
  ggtitle("Stroke Event By Age")
```

Stroke Event By Age



```
numeric_vars %>%
  ggplot(aes(x = age)) +
  geom_histogram() +
  facet_wrap(~stroke) +
  ggtitle("Histograms of Age seperated by Stroke Event")
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

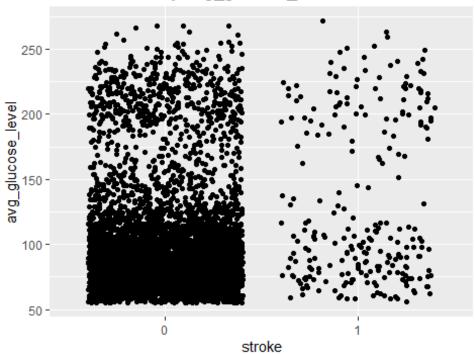
Histograms of Age seperated by Stroke Event



#avg_glucose_level

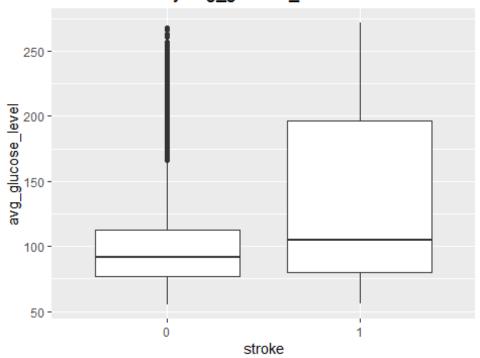
```
numeric_vars %>%
  ggplot(aes(x = stroke, y = avg_glucose_level)) +
  geom_jitter() +
  ggtitle("Stroke Event By avg_glucose_level")
```

Stroke Event By avg_glucose_level



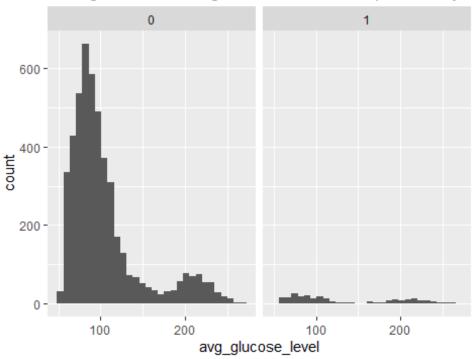
```
numeric_vars %>%
  ggplot(aes(x = stroke, y = avg_glucose_level)) +
  geom_boxplot() +
  ggtitle("Stroke Event By avg_glucose_level")
```

Stroke Event By avg_glucose_level



```
numeric_vars %>%
  ggplot(aes(x = avg_glucose_level)) +
  geom_histogram() +
  facet_wrap(~stroke) +
  ggtitle("Histograms of Average Glucose Level seperated by Stroke Event")
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

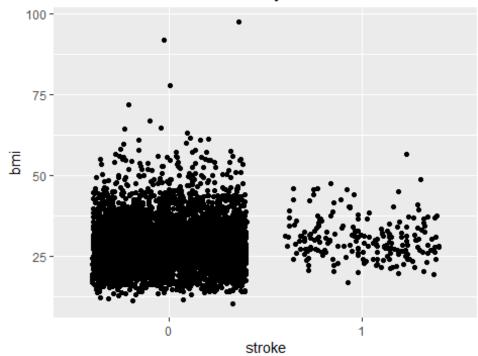
Histograms of Average Glucose Level seperated by S



#bmi

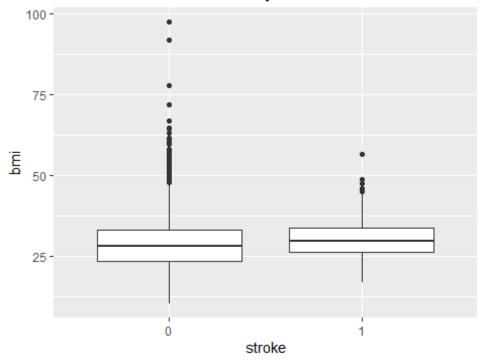
```
numeric_vars %>%
  filter(bmi != "N/A") %>%
  mutate(bmi = as.numeric(bmi)) %>%
  ggplot(aes(x = stroke, y = bmi))+
  geom_jitter() +
  ggtitle("Jitter Plot of Stroke Event by BMI")
```

Jitter Plot of Stroke Event by BMI



```
numeric_vars %>%
  filter(bmi != "N/A") %>%
  mutate(bmi = as.numeric(bmi)) %>%
  ggplot(aes(x = stroke, y = bmi))+
  geom_boxplot() +
  ggtitle("Box Plot of Stroke Event by BMI")
```

Box Plot of Stroke Event by BMI



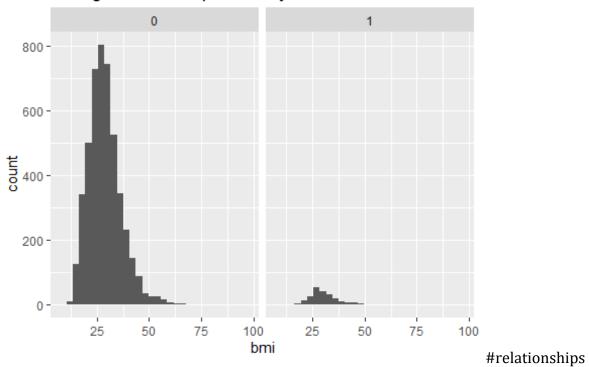
```
numeric_vars %>%
  mutate(bmi = as.numeric(bmi)) %>%
  ggplot(aes(x = bmi)) +
  geom_histogram() +
  facet_wrap(~stroke) +
  ggtitle("Histograms BMI seperated by Stroke Event")

## Warning in mask$eval_all_mutate(quo): NAs introduced by coercion

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

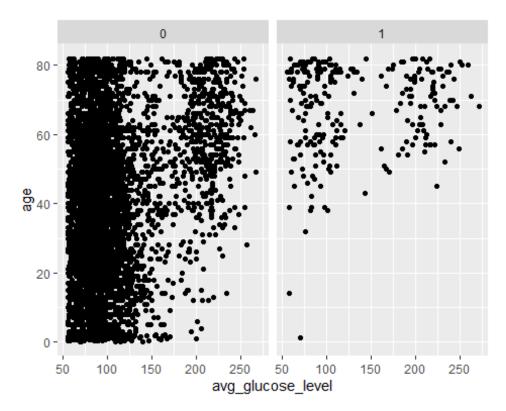
## Warning: Removed 201 rows containing non-finite values (stat_bin).
```

Histograms BMI seperated by Stroke Event

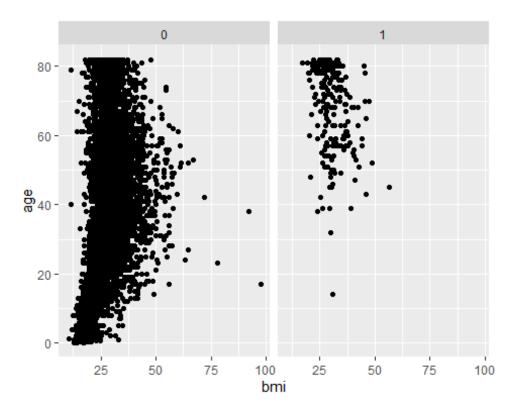


between numeric variables

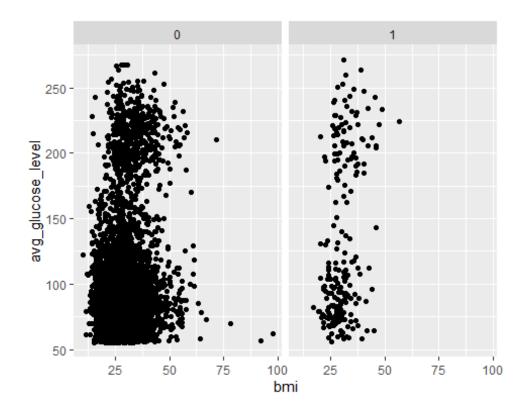
```
ggplot(data, aes(x = avg_glucose_level, y = age)) +
  geom_point() +
  facet_wrap(~stroke)
```



```
numeric_vars %>%
  filter(bmi != "N/A") %>%
  mutate(bmi = as.numeric(bmi)) %>%
  ggplot(aes(x = bmi, y = age)) +
  geom_point() +
  facet_wrap(~stroke)
```



```
numeric_vars %>%
  filter(bmi != "N/A") %>%
  mutate(bmi = as.numeric(bmi)) %>%
  ggplot(aes(x = bmi, y = avg_glucose_level)) +
  geom_point() +
  facet_wrap(~stroke)
```



#New dataset for the updated models

```
model_df <- data %>%
  filter(gender != "Other") %>%
  mutate(stroke = as.factor(stroke)) %>%
  filter(bmi != "N/A") %>%
  mutate(bmi = as.numeric(bmi)) %>%
  filter(age > 45) %>%
  filter(smoking_status != "Unknown") %>%
  select(-c(bmi, work_type))
```

#Random Forest modeling

```
library(randomForest)
## Warning: package 'randomForest' was built under R version 4.1.3
## randomForest 4.7-1
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:dplyr':
##
## combine
```

```
## The following object is masked from 'package:ggplot2':
##
##
       margin
library(datasets)
library(caret)
## Warning: package 'caret' was built under R version 4.1.3
## Loading required package: lattice
## Attaching package: 'caret'
## The following object is masked from 'package:purrr':
##
##
       lift
set.seed(11)
#creating a train and test set
stroke <- model_df %>%
  filter(stroke == '1')
dt <- sort(sample(nrow(stroke), nrow(stroke)*.8))</pre>
train <- stroke[dt,]</pre>
test <- stroke[-dt,]</pre>
no stroke <- model df %>%
  filter(stroke == '0') %>%
  sample_n(500)
dt2 <- sort(sample(nrow(no_stroke), nrow(no_stroke)*.8))</pre>
train2 <- no_stroke[dt2,]</pre>
test2 <- no_stroke[-dt2,]</pre>
training <- rbind(train, train2)</pre>
testing <- rbind(test, test2)</pre>
#training the model
rf <- randomForest(stroke~., data = training)</pre>
#Prediction and Confusion Matrix
p1 <- predict(rf, training)</pre>
confusionMatrix(p1,training$stroke)
## Confusion Matrix and Statistics
##
##
              Reference
## Prediction
                 0
                     1
##
            0 400
                    63
##
                 0
                   74
```

```
##
##
                  Accuracy : 0.8827
##
                    95% CI: (0.8524, 0.9087)
##
       No Information Rate: 0.7449
##
       P-Value [Acc > NIR] : 1.778e-15
##
##
                     Kappa: 0.6363
##
##
   Mcnemar's Test P-Value : 5.662e-15
##
##
               Sensitivity: 1.0000
##
               Specificity: 0.5401
            Pos Pred Value: 0.8639
##
##
            Neg Pred Value: 1.0000
##
                Prevalence: 0.7449
            Detection Rate: 0.7449
##
##
      Detection Prevalence: 0.8622
##
         Balanced Accuracy: 0.7701
##
          'Positive' Class: 0
##
##
p2 <- predict(rf,testing)</pre>
confusionMatrix(p2, testing$stroke)
## Confusion Matrix and Statistics
##
             Reference
##
## Prediction 0 1
##
            0 97 33
            1 3 2
##
##
##
                  Accuracy : 0.7333
##
                    95% CI: (0.6504, 0.8057)
##
       No Information Rate: 0.7407
##
       P-Value [Acc > NIR] : 0.6213
##
##
                     Kappa: 0.0376
##
    Mcnemar's Test P-Value: 1.343e-06
##
##
##
               Sensitivity: 0.97000
               Specificity: 0.05714
##
##
            Pos Pred Value: 0.74615
            Neg Pred Value: 0.40000
##
##
                Prevalence: 0.74074
##
            Detection Rate: 0.71852
##
      Detection Prevalence: 0.96296
##
         Balanced Accuracy: 0.51357
##
```

```
##
           'Positive' Class : 0
##
#logistic Regression
library(fastDummies)
## Warning: package 'fastDummies' was built under R version 4.1.3
set.seed(11)
#creating a train and test set
stroke <- model_df %>%
  filter(stroke == '1')
stroke <- dummy cols(stroke, select columns = c('gender', 'ever married',</pre>
'Residence_type'))
dt <- sort(sample(nrow(stroke), nrow(stroke)*.8))</pre>
train <- stroke[dt,]
test <- stroke[-dt,]</pre>
no_stroke <- model_df %>%
  filter(stroke == '0') %>%
  sample n(500)
no stroke <- dummy cols(no stroke, select columns = c('gender',
'ever_married', 'Residence_type'))
dt2 <- sort(sample(nrow(no_stroke), nrow(no_stroke)*.8))</pre>
train2 <- no stroke[dt2,]
test2 <- no_stroke[-dt2,]</pre>
training <- rbind(train, train2)</pre>
testing <- rbind(test, test2)</pre>
log <- glm(stroke~hypertension + heart_disease, data = training, family =</pre>
"binomial")
summary(log)
##
## Call:
## glm(formula = stroke ~ hypertension + heart disease, family = "binomial",
##
       data = training)
##
## Deviance Residuals:
       Min
                       Median
                                     3Q
                                             Max
                 10
## -1.1961 -0.6761 -0.6761
                                 1.1588
                                          1.7822
##
## Coefficients:
##
                  Estimate Std. Error z value Pr(>|z|)
                               0.1271 -10.697 < 2e-16 ***
## (Intercept)
                   -1.3596
                               0.2205
                                         2.942 0.00326 **
                    0.6488
## hypertension
```

```
## heart disease
                ## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 609.92 on 536 degrees of freedom
## Residual deviance: 592.80 on 534 degrees of freedom
## AIC: 598.8
##
## Number of Fisher Scoring iterations: 4
glm.probs <- predict(log, type = 'response')</pre>
glm.pred <- ifelse(glm.probs > .5, "1", "0")
table(glm.pred, training$stroke)
##
## glm.pred 0
                1
         0 387 127
##
##
         1 13 10
mean(glm.pred == training$stroke)
## [1] 0.7392924
glm.probs2 = predict(log, newdata = testing, type = 'response')
glm.pred2 = ifelse(glm.probs2 > .5, "1", "0")
table(glm.pred2, testing$stroke)
##
## glm.pred2 0 1
          0 97 34
##
##
          1 3 1
mean(glm.pred2 == testing$stroke)
## [1] 0.7259259
```