Final Project

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```
### Importing the data set
heartoriginal <- read.csv("~/Stat230/heart.csv")</pre>
```

###Uploading the dataset and removing incomplete cases.

```
###Dropping N/A values that would interfere with teh dataset
heartoriginal <- heartoriginal %>% drop_na()
heart <- filter(heartoriginal, Cholesterol > 0)
```

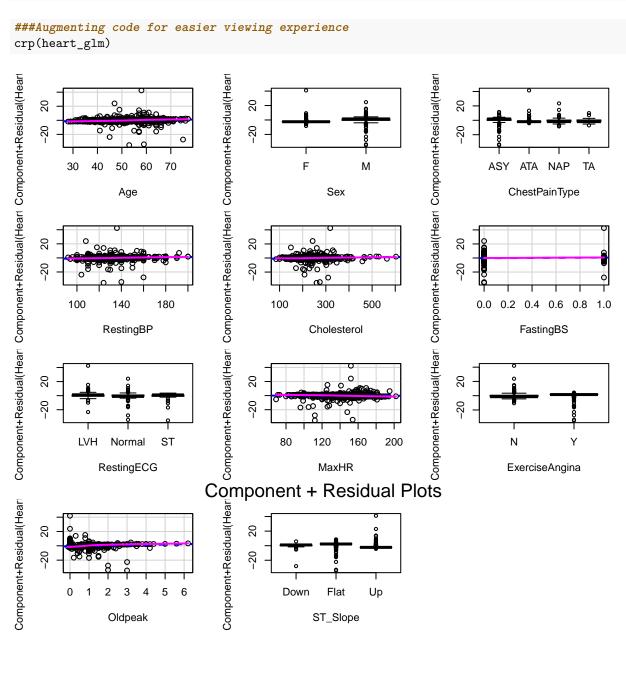
###Fitting the full model with all the predictors on the entire Dataset.

```
### Creating the full glm model for the dataset with all variables
heart_glm <- glm(HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol + FastingBS + RestingBS + Rest
summary(heart_glm)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + RestingBP +
          Cholesterol + FastingBS + RestingECG + MaxHR + ExerciseAngina +
         Oldpeak + ST_Slope, family = binomial, data = heart)
Deviance Residuals:
         Min
                                  1Q
                                              Median
                                                                              ЗQ
                                                                                                  Max
-2.6697 -0.3841 -0.1085
                                                                    0.4465
                                                                                           2.7371
Coefficients:
                                              Estimate Std. Error z value Pr(>|z|)
(Intercept)
                                         -5.4373046 1.7625169 -3.085 0.002036 **
Age
                                            0.0313784 0.0148105 2.119 0.034119 *
                                           1.8655490 0.3134065 5.952 2.64e-09 ***
SexM
ChestPainTypeATA -1.6731804  0.3544226  -4.721  2.35e-06 ***
ChestPainTypeNAP -1.5730121 0.3029404 -5.192 2.08e-07 ***
ChestPainTypeTA -1.6332529  0.4838117  -3.376  0.000736 ***
                                            0.0117792 0.0072988 1.614 0.106557
RestingBP
Cholesterol
                                           0.0024955 0.0019773 1.262 0.206928
FastingBS
                                            0.2923999 0.3311265 0.883 0.377212
RestingECGNormal -0.2297888 0.2842091 -0.809 0.418791
                                         RestingECGST
MaxHR
                                           0.0005807 0.0057810 0.100 0.919991
ExerciseAnginaY
                                           0.9073515  0.2671360  3.397  0.000682 ***
                                            0.4108355 0.1406671 2.921 0.003493 **
Oldpeak
                                      1.3038217 0.5197574 2.509 0.012124 *
ST_SlopeFlat
```

```
ST_SlopeUp
                 -1.2100372 0.5655279 -2.140 0.032382 *
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63
                            on 745
                                    degrees of freedom
                   483.58
                            on 730
Residual deviance:
                                    degrees of freedom
AIC: 515.58
Number of Fisher Scoring iterations: 6
```

###Augmenting code for easier viewing experience

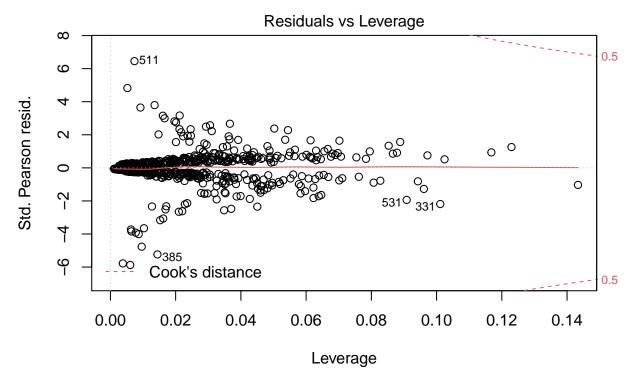
crp(heart_glm)



```
heart_aug <- augment(heart_glm)</pre>
heart_aug %>%
slice_max(.hat, n = 1)
# A tibble: 1 x 18
  HeartDisease
                 Age Sex
                           ChestPainType RestingBP Cholesterol FastingBS
         <int> <int> <chr> <chr>
                                              <int>
             0
                  67 F
                           NAP
                                                115
                                                             564
                                                                         0
1
# ... with 11 more variables: RestingECG <chr>, MaxHR <int>,
   ExerciseAngina <chr>, Oldpeak <dbl>, ST_Slope <chr>, .fitted <dbl>,
    .resid < dbl>, .std.resid < dbl>, .hat < dbl>, .sigma < dbl>, .cooksd < dbl>
plot(heart_glm, which=4, id.n = 4)
```


Obs. number glm(HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol + Fa ...

plot(heart_glm, which=5, id.n = 4)

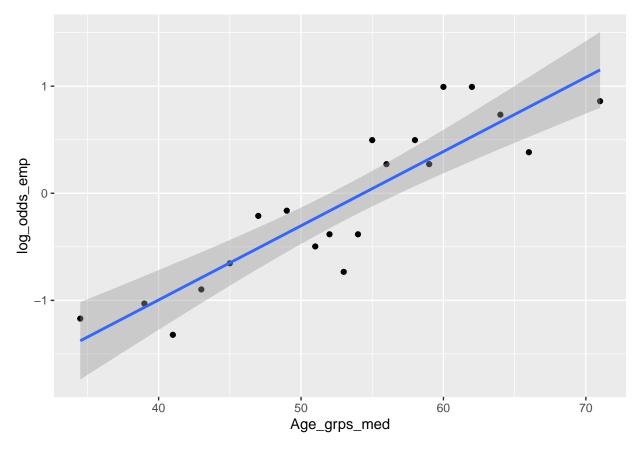


glm(HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol + Fa ..

```
heart_empLO
# A tibble: 20 x 4
   Age_grps Age_grps_med pi_emp log_odds_emp
      <int>
                    <dbl> <dbl>
                                          <dbl>
1
          1
                     34.5 0.237
                                         -1.17
 2
          2
                     39
                            0.263
                                         -1.03
 3
          3
                     41
                            0.211
                                         -1.32
 4
          4
                            0.289
                                         -0.898
                     43
 5
          5
                     45
                            0.342
                                         -0.654
 6
          6
                     47
                            0.447
                                         -0.211
 7
          7
                     49
                            0.459
                                         -0.163
 8
          8
                     51
                            0.378
                                         -0.496
                     52
                           0.405
                                         -0.383
```

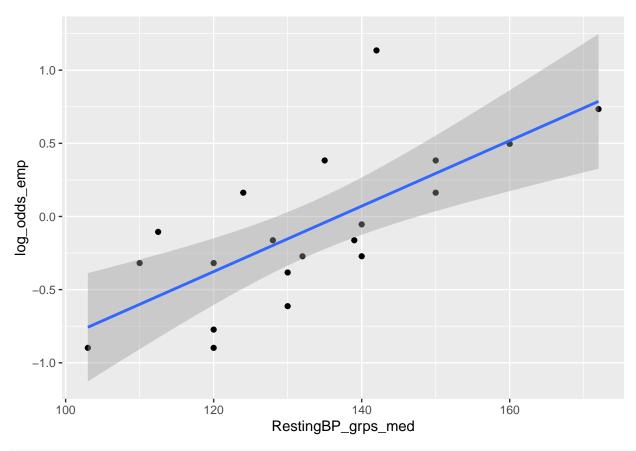
```
10
         10
                      53
                            0.324
                                          -0.734
11
          11
                      54
                            0.405
                                          -0.383
12
          12
                      55
                            0.622
                                          0.496
13
                            0.568
                                          0.272
          13
                      56
14
          14
                      58
                            0.622
                                          0.496
                     59
15
          15
                            0.568
                                          0.272
16
          16
                      60
                            0.730
                                          0.993
17
          17
                      62
                            0.730
                                          0.993
18
                            0.676
                                          0.734
         18
                      64
19
          19
                      66
                            0.595
                                          0.383
20
          20
                      71
                            0.703
                                          0.860
```

```
### Plotting the model
ggplot(heart_empLO, aes(x=Age_grps_med, y=log_odds_emp)) +
  geom_point() +
  geom_smooth(method = lm)
  `geom_smooth()` using formula 'y ~ x'
```



```
heart_empL02
# A tibble: 20 x 4
   RestingBP_grps RestingBP_grps_med pi_emp log_odds_emp
           <int>
                              <dbl> <dbl>
                                                 <dbl>
1
                              103 0.289
                                               -0.898
               1
 2
               2
                              110 0.421
                                               -0.318
 3
               3
                              112. 0.474
                                               -0.105
 4
               4
                              120 0.289
                                               -0.898
5
               5
                              120 0.316
                                               -0.773
6
               6
                              120 0.421
                                               -0.318
7
               7
                              124
                                  0.541
                                               0.163
8
               8
                              128
                                   0.459
                                               -0.163
9
               9
                              130 0.351
                                               -0.613
10
              10
                              130 0.405
                                               -0.383
11
              11
                              132 0.432
                                               -0.272
12
              12
                              135
                                   0.595
                                               0.383
13
              13
                              139 0.459
                                               -0.163
14
              14
                              140
                                  0.432
                                               -0.272
15
              15
                                   0.486
                                               -0.0541
                              140
16
              16
                              142
                                   0.757
                                                1.13
17
                              150
                                   0.541
                                                0.163
              17
18
              18
                              150
                                    0.595
                                                0.383
              19
19
                              160
                                    0.622
                                                0.496
20
              20
                              172 0.676
                                                0.734
```

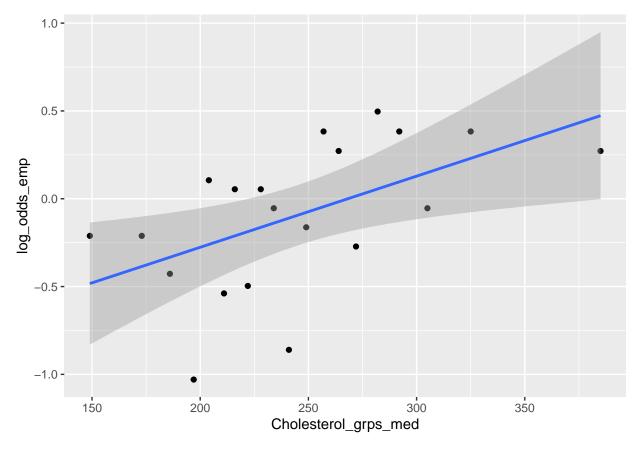
```
ggplot(heart_empLO2, aes(x=RestingBP_grps_med, y=log_odds_emp)) +
  geom_point() +
  geom_smooth(method = lm)
  `geom_smooth()` using formula 'y ~ x'
```



```
###Repeating the process above but for a different variable
heart_empLO3 <-heart %>%
  group_by(Cholesterol_grps) %>%
   summarize(Cholesterol_grps_med = median(Cholesterol),
             pi_emp = mean(HeartDisease),
                                                  # proportion died
             log_odds_emp = log(pi_emp/(1-pi_emp))) # log odds
heart_empLO3
# A tibble: 20 x 4
   Cholesterol_grps Cholesterol_grps_med pi_emp log_odds_emp
              <int>
                                   <dbl> <dbl>
                                                       <dbl>
 1
                  1
                                     149 0.447
                                                     -0.211
 2
                  2
                                     173 0.447
                                                     -0.211
 3
                  3
                                     186 0.395
                                                     -0.427
 4
                  4
                                                     -1.03
                                     197 0.263
 5
                  5
                                     204 0.526
                                                      0.105
 6
                  6
                                     211 0.368
                                                     -0.539
                                                      0.0541
 7
                  7
                                     216 0.514
 8
                  8
                                     222 0.378
                                                     -0.496
 9
                                     228 0.514
                                                   0.0541
```

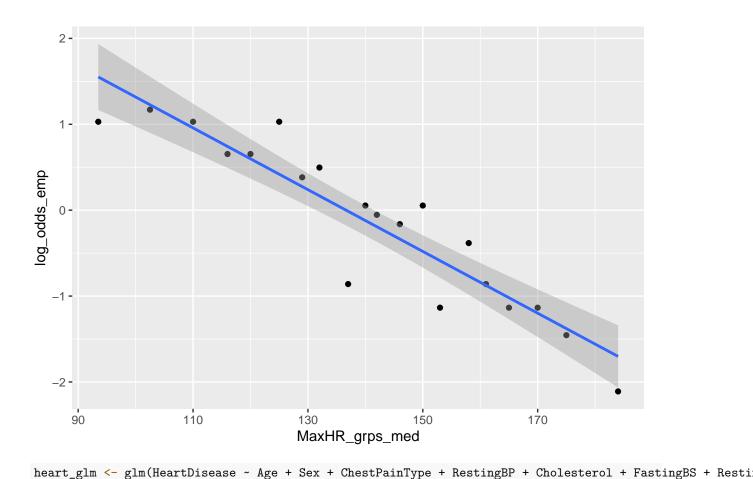
```
-0.0541
10
                 10
                                      234 0.486
11
                 11
                                      241 0.297
                                                      -0.860
                                                      -0.163
12
                 12
                                      249 0.459
13
                 13
                                                       0.383
                                      257 0.595
14
                 14
                                      264 0.568
                                                       0.272
15
                                                      -0.272
                 15
                                      272 0.432
16
                 16
                                      282 0.622
                                                       0.496
17
                 17
                                                       0.383
                                      292 0.595
                                      305 0.486
18
                 18
                                                      -0.0541
19
                 19
                                      325
                                                       0.383
                                          0.595
20
                 20
                                      385 0.568
                                                       0.272
```

```
ggplot(heart_empLO3, aes(x=Cholesterol_grps_med, y=log_odds_emp)) +
  geom_point() +
  geom_smooth(method = lm)
  `geom_smooth()` using formula 'y ~ x'
```



```
heart_empLO4 <-heart %>%
  group_by(MaxHR_grps) %>%
   summarize(MaxHR_grps_med = median(MaxHR),
                                                 # median MaxHR of groups
             pi_emp = mean(HeartDisease),
                                                   # proportion died
             log_odds_emp = log(pi_emp/(1-pi_emp)))
                                                        # log odds
heart_empL04
# A tibble: 20 x 4
   MaxHR_grps MaxHR_grps_med pi_emp log_odds_emp
        <int>
                      <dbl> <dbl>
                                            <dbl>
1
            1
                        93.5 0.737
                                           1.03
 2
            2
                       102.
                               0.763
                                           1.17
 3
            3
                       110
                               0.737
                                           1.03
 4
            4
                                           0.654
                       116
                               0.658
 5
            5
                       120
                               0.658
                                           0.654
 6
            6
                       125
                               0.737
                                           1.03
 7
            7
                       129
                               0.595
                                           0.383
 8
            8
                       132
                               0.622
                                           0.496
9
            9
                       137
                               0.297
                                          -0.860
10
           10
                       140
                               0.514
                                           0.0541
                       142
                               0.486
                                          -0.0541
11
           11
12
           12
                       146
                               0.459
                                          -0.163
13
           13
                       150
                                          0.0541
                               0.514
14
           14
                       153
                               0.243
                                          -1.13
15
           15
                       158
                               0.405
                                          -0.383
16
           16
                       161
                               0.297
                                          -0.860
17
           17
                       165
                               0.243
                                          -1.13
18
           18
                       170
                               0.243
                                          -1.13
19
           19
                       175
                               0.189
                                          -1.46
20
           20
                       184
                               0.108
                                          -2.11
```

```
ggplot(heart_empLO4, aes(x=MaxHR_grps_med, y=log_odds_emp)) +
  geom_point() +
  geom_smooth(method = lm)
  `geom_smooth()` using formula 'y ~ x'
```



summary(heart_glm) Call: glm(formula = HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol + FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak + ST_Slope, family = binomial, data = heart) Deviance Residuals: Min 1Q Median Max 0.4465 -2.6697 -0.3841 -0.1085 2.7371 Coefficients: Estimate Std. Error z value Pr(>|z|)-5.4373046 1.7625169 -3.085 0.002036 ** (Intercept) Age 0.0313784 0.0148105 2.119 0.034119 * SexM1.8655490 0.3134065 5.952 2.64e-09 *** ChestPainTypeATA -1.6731804 0.3544226 -4.721 2.35e-06 *** ChestPainTypeNAP -1.5730121 0.3029404 -5.192 2.08e-07 *** -3.376 0.000736 *** ChestPainTypeTA -1.6332529 0.4838117 RestingBP 0.0117792 0.0072988 1.614 0.106557 Cholesterol 1.262 0.206928 0.0024955 0.0019773 FastingBS 0.2923999 0.3311265 0.883 0.377212 RestingECGNormal -0.2297888 0.2842091 -0.809 0.418791 RestingECGST -0.443 0.657792 -0.1746017 0.3941671

0.0005807 0.0057810 0.100 0.919991

MaxHR

```
ExerciseAnginaY 0.9073515 0.2671360 3.397 0.000682 ***
Oldpeak
                 ST SlopeFlat
                1.3038217 0.5197574
                                      2.509 0.012124 *
ST_SlopeUp
                -1.2100372 0.5655279 -2.140 0.032382 *
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 483.58 on 730 degrees of freedom
AIC: 515.58
Number of Fisher Scoring iterations: 6
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm2 <- glm(HeartDisease ~ Sex + ChestPainType + RestingBP + Cholesterol + FastingBS + RestingECG
anova(heart_glm2, heart_glm, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Sex + ChestPainType + RestingBP + Cholesterol +
   FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
   ST_Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol +
   FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
       731
              488.14
1
2
       730
               483.58 1
                           4.559 0.03275 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
1-pchisq(483.58, 730)
[1] 1
488.14 - 483.58 ##LRT
[1] 4.56
0.016550/0.2087463
[1] 0.07928284
2*pnorm(0.07928284)
[1] 1.063192
```

The estimated coefficient for Age is 0.0313784, and the standard error is 0.0148105. The p-value that we get for the goodness of fit test is 1. The p-value that we get for the Wald's test is 0.034119. The p-value that we get for the drop-in-deviance test is 0.03275. This means that Age is significant and that we should keep this in our model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart3_glm <- glm(HeartDisease ~ Age + ChestPainType + RestingBP + Cholesterol + FastingBS + RestingECG
anova(heart3_glm, heart_glm, test = "Chisq")</pre>
```

```
Analysis of Deviance Table

Model 1: HeartDisease ~ Age + ChestPainType + RestingBP + Cholesterol +
FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
ST_Slope

Model 2: HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol +
FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
ST_Slope
Resid. Df Resid. Dev Df Deviance Pr(>Chi)

1 731 523.59
2 730 483.58 1 40.005 2.533e-10 ***

---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

In this model, the estimated coefficient for Sex is 1.8655490 with a standard error of 0.3134065. The p-value that we get for the Wald's test is 2.64e-09. In this drop-in-deviance test, the p-value is significant here, showing that we do not take Sex out of the model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart4_glm <- glm(HeartDisease ~ Age + Sex + RestingBP + Cholesterol + FastingBS + RestingECG + MaxHR +
anova(heart4_glm, heart_glm, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + RestingBP + Cholesterol + FastingBS +
   RestingECG + MaxHR + ExerciseAngina + Oldpeak + ST_Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol +
   FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
   ST Slope
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
       733
1
              525.90
2
       730
               483.58 3 42.319 3.433e-09 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The p-value that we get for the Wald's test for each ChestPainType is 2.64e-09. In this drop-in-deviance test, the p-value is significant here, showing that we do not take ChestPainType out of the model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm5 <- glm(HeartDisease ~ Age + Sex + ChestPainType + Cholesterol + FastingBS + RestingECG + Max
summary(heart_glm5)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + Cholesterol +
    FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
   ST_Slope, family = binomial, data = heart)
Deviance Residuals:
                 Median
   Min
             1Q
                               3Q
                                       Max
-2.7222 -0.3890 -0.1173 0.4445
                                     2.7024
Coefficients:
                  Estimate Std. Error z value Pr(>|z|)
```

```
(Intercept)
                -4.2217418 1.5882242 -2.658 0.007857 **
                 0.0352513 0.0146593
                                       2.405 0.016186 *
Age
SexM
                 1.8280371 0.3086990 5.922 3.19e-09 ***
ChestPainTypeATA -1.6288593  0.3514200  -4.635  3.57e-06 ***
ChestPainTypeNAP -1.5713404  0.3031520  -5.183  2.18e-07 ***
ChestPainTypeTA -1.5549800 0.4795066 -3.243 0.001183 **
Cholesterol
                 0.0028097 0.0019746 1.423 0.154747
FastingBS
                 0.3602864 0.3265471
                                      1.103 0.269888
RestingECGNormal -0.2195001 0.2831660 -0.775 0.438243
RestingECGST
                -0.1345451 0.3926224 -0.343 0.731837
MaxHR
                 0.0006743 0.0057665 0.117 0.906910
ExerciseAnginaY 0.9451880 0.2657907
                                      3.556 0.000376 ***
                 0.4295708 0.1388026
Oldpeak
                                        3.095 0.001969 **
ST_SlopeFlat
                1.3003924 0.5166849 2.517 0.011843 *
ST_SlopeUp
                -1.1782705 0.5612939 -2.099 0.035799 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.6 on 745 degrees of freedom
Residual deviance: 486.2 on 731 degrees of freedom
AIC: 516.2
Number of Fisher Scoring iterations: 6
anova(heart_glm5, heart_glm, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + ChestPainType + Cholesterol + FastingBS +
   RestingECG + MaxHR + ExerciseAngina + Oldpeak + ST_Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol +
   FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
       731
               486.20
       730
               483.58 1 2.6213 0.1054
```

In this ANOVA test, when analyzing the term for RestingBP, the drop-in-deviance p-value is 0.1054, which shows that RestingBP is not significant to the model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm6 <- glm(HeartDisease ~ Age + Sex + ChestPainType + FastingBS + RestingECG + MaxHR + ExerciseA
summary(heart_glm6)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + FastingBS +
   RestingECG + MaxHR + ExerciseAngina + Oldpeak + ST_Slope,
    family = binomial, data = heart)
Deviance Residuals:
   Min
                  Median
                                3Q
              1Q
                                        Max
-2.7022 -0.3823 -0.1218
                            0.4420
                                     2.7468
```

```
Coefficients:
                 Estimate Std. Error z value Pr(>|z|)
(Intercept)
               -3.4139099 1.4742536 -2.316 0.020575 *
                0.0340911 0.0146149 2.333 0.019668 *
Age
SexM
                ChestPainTypeATA -1.6422280 0.3505312 -4.685 2.80e-06 ***
ChestPainTypeNAP -1.5993727  0.3016820  -5.302  1.15e-07 ***
ChestPainTypeTA -1.5801583 0.4771498 -3.312 0.000927 ***
FastingBS
                RestingECGNormal -0.2503346 0.2821015 -0.887 0.374868
RestingECGST
              -0.1721090 0.3900980 -0.441 0.659072
MaxHR
                ExerciseAnginaY
                0.9538006  0.2654084  3.594  0.000326 ***
Oldpeak
                0.4243075 0.1388137
                                   3.057 0.002238 **
ST_SlopeFlat
               1.3266653 0.5125717 2.588 0.009646 **
ST_SlopeUp
               -1.1793038 0.5574743 -2.115 0.034392 *
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 488.28 on 732 degrees of freedom
AIC: 516.28
Number of Fisher Scoring iterations: 5
anova(heart_glm6, heart_glm5, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + ChestPainType + FastingBS + RestingECG +
   MaxHR + ExerciseAngina + Oldpeak + ST_Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + Cholesterol + FastingBS +
   RestingECG + MaxHR + ExerciseAngina + Oldpeak + ST_Slope
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
       732
              488.28
1
       731
              486.20 1 2.0758 0.1497
```

The p-value for this drop-in-deviance test for the term Cholesterol is 0.1497, which shows that the smaller model is better, as this term is not significant.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm7 <- glm(HeartDisease ~ Age + Sex + ChestPainType + RestingECG + MaxHR + ExerciseAngina + Oldp
summary(heart_glm7)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + RestingECG +
   MaxHR + ExerciseAngina + Oldpeak + ST_Slope, family = binomial,
    data = heart)
Deviance Residuals:
   Min
              1Q
                 Median
                                3Q
                                        Max
-2.7372 -0.3905 -0.1207
                            0.4492
                                     2.8723
```

```
Coefficients:
                  Estimate Std. Error z value Pr(>|z|)
(Intercept)
                -3.4534928 1.4666683 -2.355 0.018540 *
Age
                 0.0367148  0.0144116  2.548  0.010847 *
SexM
                 1.8054616  0.3020732  5.977  2.27e-09 ***
ChestPainTypeATA -1.6400974  0.3496293  -4.691  2.72e-06 ***
ChestPainTypeNAP -1.5838085 0.3008390 -5.265 1.40e-07 ***
ChestPainTypeTA -1.5368769 0.4732512 -3.247 0.001164 **
RestingECGNormal -0.2648430 0.2808551 -0.943 0.345687
                -0.1301001 0.3874668 -0.336 0.737044
RestingECGST
MaxHR
                 0.0007577 0.0056988 0.133 0.894224
ExerciseAnginaY
                 0.9543314  0.2648203  3.604  0.000314 ***
                 0.4147557 0.1388159 2.988 0.002810 **
Oldpeak
ST_SlopeFlat
                 1.2711619 0.5106020 2.490 0.012791 *
ST_SlopeUp
                -1.2599518 0.5547673 -2.271 0.023139 *
___
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 489.76 on 733 degrees of freedom
AIC: 515.76
Number of Fisher Scoring iterations: 5
anova(heart_glm7, heart_glm6, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + ChestPainType + RestingECG + MaxHR +
   ExerciseAngina + Oldpeak + ST_Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + FastingBS + RestingECG +
   MaxHR + ExerciseAngina + Oldpeak + ST_Slope
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
       733
               489.76
       732
               488.28 1 1.4843 0.2231
```

For testing the term FastingBS, we get a drop-in-deviance p-value of 0.2231. This shows that this term is not significant to the model.

(Intercept) -3.867397 1.390712 -2.781 0.005421 **

```
Age
              SexM
              ChestPainTypeTA -1.530705 0.471104 -3.249 0.001157 **
MaxHR
              0.001452 \quad 0.005544 \quad 0.262 \ 0.793353
ExerciseAnginaY 0.946540 0.263994 3.585 0.000336 ***
Oldpeak
              ST_SlopeFlat
             1.240071 0.509260 2.435 0.014890 *
ST_SlopeUp
             -1.289846 0.553381 -2.331 0.019762 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 490.68 on 735 degrees of freedom
AIC: 512.68
Number of Fisher Scoring iterations: 5
anova(heart_glm8, heart_glm7, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + ChestPainType + MaxHR + ExerciseAngina +
   Oldpeak + ST Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + RestingECG + MaxHR +
   ExerciseAngina + Oldpeak + ST_Slope
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
      735
            490.68
1
      733
            489.76 2 0.91379 0.6332
```

For testing the term RestingECG, we get a drop-in-deviance p-value of 0.6332. This shows that this term is not significant to the model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm9 <- glm(HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina + Oldpeak + ST_Slope, famil
summary(heart_glm9)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina +
   Oldpeak + ST_Slope, family = binomial, data = heart)
Deviance Residuals:
   Min
           1Q
               Median
                         3Q
                                Max
-2.7265 -0.3928 -0.1227 0.4556
                             2.9284
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
             -3.60085 0.94673 -3.803 0.000143 ***
(Intercept)
              Age
SexM
              ChestPainTypeNAP -1.57932 0.29584 -5.338 9.38e-08 ***
```

```
ChestPainTypeTA -1.51087 0.46339 -3.260 0.001112 **
ExerciseAnginaY 0.93541 0.26059 3.590 0.000331 ***
Oldpeak
               ST_SlopeFlat
               ST_SlopeUp
              -1.27465 0.55154 -2.311 0.020828 *
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 490.74 on 736 degrees of freedom
AIC: 510.74
Number of Fisher Scoring iterations: 5
anova(heart_glm9, heart_glm8, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina + Oldpeak +
   ST Slope
Model 2: HeartDisease ~ Age + Sex + ChestPainType + MaxHR + ExerciseAngina +
   Oldpeak + ST_Slope
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
      736
              490.74
      735
           490.68 1 0.068734 0.7932
```

For testing the term MaxHR, we get a drop-in-deviance p-value of 0.7932. This shows that this term is not significant to the model.

For testing the term ExerciseAngina, we get a drop-in-deviance p-value of 0.000364. This shows that this term is significant to the model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm11 <- glm(HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina + ST_Slope, family = binom
anova(heart_glm11, heart_glm9, test = "Chisq")
Analysis of Deviance Table</pre>
```

For testing the term OldPeak, we get a drop-in-deviance p-value of 0.001653. This shows that this term is significant to the model.

For testing the term ST_Slope, we get a drop-in-deviance p-value of 2.2e-163. This shows that this term is significant to the model.

```
### Doing drop-in-deviance tests to test the significance of the variable
heart_glm <- glm(HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol + FastingBS + RestingBS + RestingBP + Cholesterol + FastingBS + RestingBP + Cholesterol + FastingBS + RestingBP + Cholesterol + FastingBP + RestingBP + Rest
summary(heart_glm)
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + RestingBP +
            Cholesterol + FastingBS + RestingECG + MaxHR + ExerciseAngina +
            Oldpeak + ST_Slope, family = binomial, data = heart)
Deviance Residuals:
            Min
                                           1Q
                                                       Median
                                                                                                    3Q
                                                                                                                             Max
-2.6697 -0.3841 -0.1085 0.4465
                                                                                                                    2.7371
Coefficients:
                                                          Estimate Std. Error z value Pr(>|z|)
                                                    -5.4373046 1.7625169 -3.085 0.002036 **
(Intercept)
Age
                                                        0.0313784 0.0148105 2.119 0.034119 *
SexM
                                                        ChestPainTypeATA -1.6731804  0.3544226  -4.721  2.35e-06 ***
ChestPainTypeNAP -1.5730121 0.3029404 -5.192 2.08e-07 ***
ChestPainTypeTA -1.6332529  0.4838117  -3.376  0.000736 ***
RestingBP
                                                        0.0117792 0.0072988 1.614 0.106557
```

```
Cholesterol
                 0.0024955 0.0019773 1.262 0.206928
FastingBS
                 0.2923999  0.3311265  0.883  0.377212
RestingECGNormal -0.2297888 0.2842091 -0.809 0.418791
RestingECGST
             -0.1746017 0.3941671 -0.443 0.657792
MaxHR
                 0.0005807 0.0057810 0.100 0.919991
ExerciseAnginaY 0.9073515 0.2671360 3.397 0.000682 ***
Oldpeak
                 0.4108355 0.1406671 2.921 0.003493 **
                1.3038217 0.5197574 2.509 0.012124 *
ST_SlopeFlat
ST_SlopeUp
                -1.2100372 0.5655279 -2.140 0.032382 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 483.58 on 730
                                   degrees of freedom
AIC: 515.58
Number of Fisher Scoring iterations: 6
heart_glm13 <- glm(HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina + Oldpeak + ST_Slope, fami
anova(heart_glm13, heart_glm, test = "Chisq")
Analysis of Deviance Table
Model 1: HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina + Oldpeak +
Model 2: HeartDisease ~ Age + Sex + ChestPainType + RestingBP + Cholesterol +
   FastingBS + RestingECG + MaxHR + ExerciseAngina + Oldpeak +
   ST_Slope
 Resid. Df Resid. Dev Df Deviance Pr(>Chi)
       736
               490.74
       730
               483.58 6 7.1638 0.306
```

Taking out all of the insignificant terms gives us a drop-in-deviance p-value of 0.306.

```
#Summarizing the final model for the dataset
summary(heart_glm13)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina +
   Oldpeak + ST_Slope, family = binomial, data = heart)
Deviance Residuals:
  Min
          1Q
              Median
                        3Q
                              Max
-2.7265 -0.3928 -0.1227 0.4556
                            2.9284
Coefficients:
            Estimate Std. Error z value Pr(>|z|)
            -3.60085 0.94673 -3.803 0.000143 ***
(Intercept)
             Age
SexM
             ChestPainTypeNAP -1.57932 0.29584 -5.338 9.38e-08 ***
```

```
ChestPainTypeTA -1.51087 0.46339 -3.260 0.001112 **
ExerciseAnginaY 0.93541
                          0.26059 3.590 0.000331 ***
                0.42148
Oldpeak
                          0.13782 3.058 0.002227 **
ST SlopeFlat
               1.23923
                          0.51047 2.428 0.015198 *
                          0.55154 -2.311 0.020828 *
ST_SlopeUp
               -1.27465
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 490.74 on 736 degrees of freedom
AIC: 510.74
Number of Fisher Scoring iterations: 5
### 95% Confidence term for all coefficient
confint(heart glm13)
Waiting for profiling to be done...
                    2.5 %
                              97.5 %
(Intercept)
               -5.48437757 -1.76505093
                0.01334485 0.06503033
Age
SexM
                1.21625973 2.39270126
ChestPainTypeATA -2.34842497 -0.99380289
ChestPainTypeNAP -2.16919520 -1.00697143
ChestPainTypeTA -2.43603526 -0.61385476
ExerciseAnginaY 0.42347136 1.44700022
Oldpeak
               0.15650326 0.69790711
ST_SlopeFlat
                0.22458553 2.23291298
               -2.37035853 -0.20214809
ST_SlopeUp
### Wald-Z Tests for all parameters
summary(heart glm13)
Call:
glm(formula = HeartDisease ~ Age + Sex + ChestPainType + ExerciseAngina +
   Oldpeak + ST_Slope, family = binomial, data = heart)
Deviance Residuals:
   Min
           1Q
               Median
                            ЗQ
                                    Max
-2.7265 -0.3928 -0.1227 0.4556
                                 2.9284
Coefficients:
               Estimate Std. Error z value Pr(>|z|)
               -3.60085 0.94673 -3.803 0.000143 ***
(Intercept)
Age
                0.03891
                          0.01316 2.958 0.003098 **
SexM
                ChestPainTypeATA -1.65550 0.34450 -4.806 1.54e-06 ***
ChestPainTypeTA -1.51087
                          0.46339 -3.260 0.001112 **
ExerciseAnginaY 0.93541
                          0.26059 3.590 0.000331 ***
Oldpeak
                0.42148
                          0.13782 3.058 0.002227 **
ST_SlopeFlat
               1.23923
                          0.51047 2.428 0.015198 *
```

```
ST_SlopeUp -1.27465 0.55154 -2.311 0.020828 *
---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1032.63 on 745 degrees of freedom
Residual deviance: 490.74 on 736 degrees of freedom
AIC: 510.74

Number of Fisher Scoring iterations: 5
```