Myocardial Infarction Complications

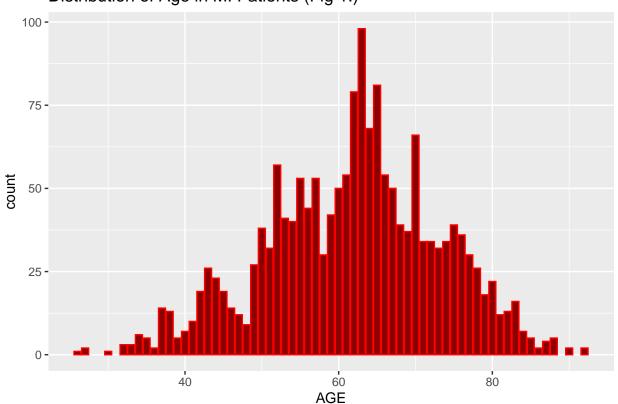
Lennox Garay

2023 - 11 - 04

```
data = read.csv("Myocardial.csv")
attach(data)
data = data %>% mutate_if(is.numeric, na.aggregate, FUN = median) %>%
  mutate_if(is.integer, na.aggregate, FUN = median) %>%
  mutate_if(is.double, na.aggregate, FUN = median)
cause_death = data.frame(deaths = data$LET_IS[which(data$LET_IS > 0)]) ## 16% of the patients with MI h
ggplot(data = cause_death, aes(x=factor(deaths), fill = deaths)) + geom_bar(stat = 'count') + xlab('Cau
  scale_x_discrete(name='Cause of Death', labels=c("0" = "Alive", '1'= 'Cardiogenic Shock', "2" = "Pulm
    90 -
                                                                                             deaths
                                                                                                  6
                                                                                                  5
                                                                                                  4
                                                                                                  3
                                                                                                  2
    30 -
     0 -
                                    Myocardial Rupture
             Cardiogenic Shock
                        Pulmonary Edema
                                                           Thromboembolism
                                               Heart Failure
                                                                      Asystole
                                                                                 V-Fib
                                       Cause of Death
response = data[,113:124]
```

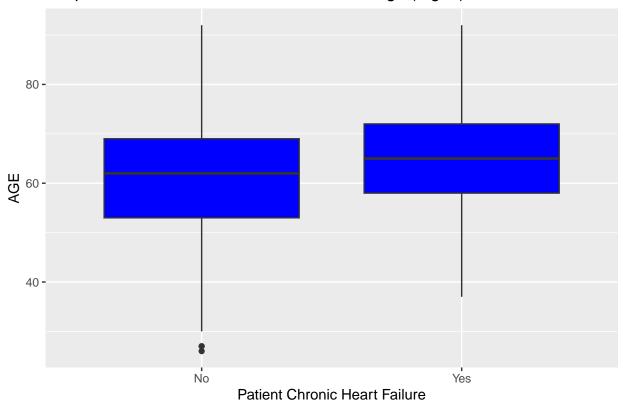
```
ggplot(data = data, aes(x = AGE)) + geom_histogram(stat = 'count', color = 'red', fill = 'darkred') + g
## Warning in geom_histogram(stat = "count", color = "red", fill = "darkred"):
## Ignoring unknown parameters: `binwidth`, `bins`, and `pad`
```

Distribution of Age in MI Patients (Fig 1.)



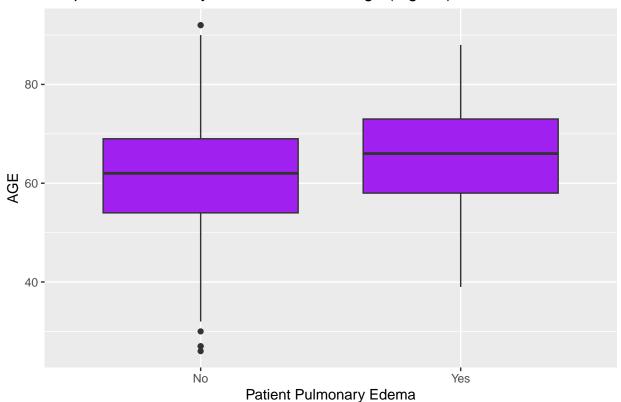
```
ggplot(data = data, aes(x = factor(ZSN), y = AGE)) +
  geom_boxplot(fill='blue') +
  ggtitle(label = 'Boxplot of Chronic Heart Failure vs Patient age (Fig 2.)') +
  scale_x_discrete(name='Patient Chronic Heart Failure', labels=c("0" = "No", '1'= 'Yes'))
```

Boxplot of Chronic Heart Failure vs Patient age (Fig 2.)



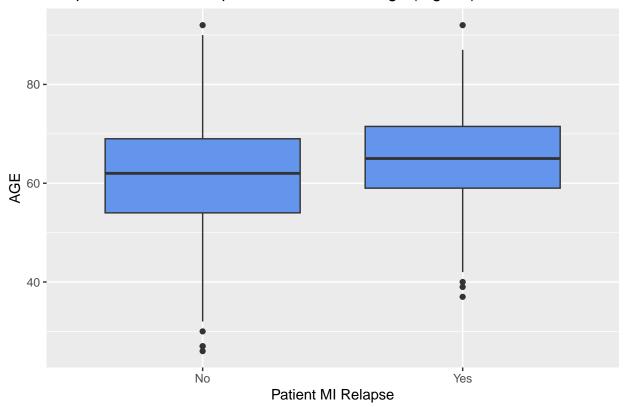
```
ggplot(data = data, aes(x = factor(OTEK_LANC), y = AGE)) +
geom_boxplot(fill='purple') +
ggtitle(label = 'Boxplot of Pulmonary Edema vs Patient age (Fig 2.1)') +
scale_x_discrete(name='Patient Pulmonary Edema', labels=c("0" = "No", '1'= 'Yes'))
```

Boxplot of Pulmonary Edema vs Patient age (Fig 2.1)



```
ggplot(data = data, aes(x = factor(REC_IM), y = AGE)) +
  geom_boxplot(fill='cornflowerblue') +
  ggtitle(label = 'Boxplot of Patient Relapse of MI vs Patient age (Fig 2.2)') +
  scale_x_discrete(name='Patient MI Relapse', labels=c("0" = "No", '1'= 'Yes'))
```

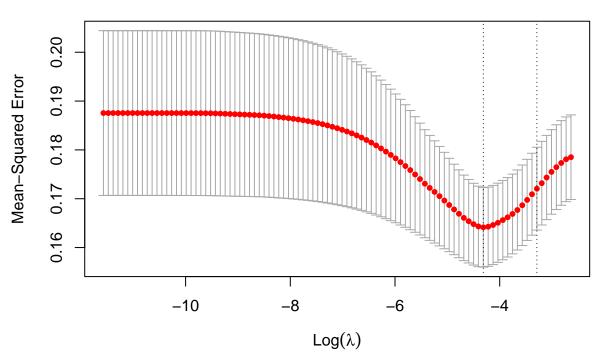
Boxplot of Patient Relapse of MI vs Patient age (Fig 2.2)



```
EDEMA_VS_MI_RELAPSE = matrix(c(OTEK_LANC, REC_IM), ncol = 2)
tetrachoric(EDEMA_VS_MI_RELAPSE)
## Call: tetrachoric(x = EDEMA_VS_MI_RELAPSE)
## tetrachoric correlation
     C1
## R1 1.00
## R2 0.39 1.00
##
## with tau of
## [1] 1.3 1.3
FAILURE_VS_RELAPSE = matrix(c(ZSN, REC_IM), ncol = 2)
tetrachoric(FAILURE_VS_RELAPSE)
## Call: tetrachoric(x = FAILURE_VS_RELAPSE)
## tetrachoric correlation
     C1 C2
##
## R1 1.00
## R2 0.19 1.00
##
## with tau of
## [1] 0.73 1.32
x = data[,2:112]
y = data[,'ZSN']
```

```
cv_model = cv.glmnet(as.matrix(x),y, alpha=1) ## 10-fold cross validation
best_lambda = cv_model$lambda.min ## 0.01339178
plot(cv_model)
```

111 111 109 106 105 100 95 89 71 42 17 10 2

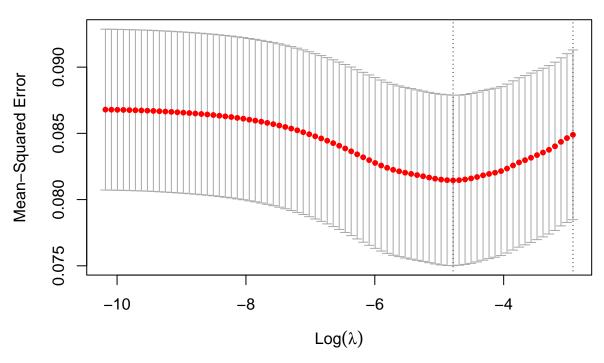


```
## LASSO
lasso_model = glmnet(x,y, alpha=1, lambda = best_lambda) ##alpha=1 is LASSO
coef(lasso_model)
## 112 x 1 sparse Matrix of class "dgCMatrix"
## (Intercept)
                 -0.0601954774
## AGE
                  0.0032515506
## SEX
                 -0.0006022468
## INF_ANAM
## STENOK AN
## FK_STENOK
                 -0.0064042039
## IBS_POST
## IBS_NASL
                 -0.0343223303
## GB
## SIM_GIPERT
## DLIT_AG
                  0.0757012149
## ZSN_A
## nr_11
```

```
## nr_01
## nr_02
## nr_03
## nr_04
               0.2647561330
## nr_07
## nr_08
## np_01
## np_04
## np_05
## np_07
              -0.0042956653
## np_08
## np_09
               0.4184150360
## np_10
## endocr_01 0.0879128520
## endocr_02 .
## endocr_02
## endocr_03
## zab_leg_01
                 0.1241515734
## zab_leg_02
## zab_leg_03
## zab_leg_04
## zab_leg_06
## S_AD_KBRIG
## D_AD_KBRIG
## S_AD_ORIT
## D_AD_ORIT
## O_L_POST
## K_SH_POST
                -0.1513260593
## MP_TP_POST
## SVT_POST
## GT_POST
## FIB_G_POST
## ant_im
## lat_im
               0.0253219456
## inf_im
## post_im
## IM_PG_P
## ritm_ecg_p_01 .
## ritm_ecg_p_02  0.0672579473
## ritm_ecg_p_04 .
## ritm_ecg_p_06  0.1305754934
## ritm_ecg_p_07 .
## ritm_ecg_p_08
## n_r_ecg_p_01
## n_r_ecg_p_02 -0.0192638936
## n_r_ecg_p_03
## n_r_ecg_p_04
## n_r_ecg_p_05
## n_r_ecg_p_06
                0.0696467377
## n_r_ecg_p_08
## n_r_ecg_p_09
## n_r_ecg_p_10
## n_p_ecg_p_01
## n_p_ecg_p_03
## n_p_ecg_p_04
```

```
## n_p_ecg_p_05
## n_p_ecg_p_06
## n_p_ecg_p_07
## n_p_ecg_p_08
## n_p_ecg_p_09
## n_p_ecg_p_10
## n_p_ecg_p_11
## n_p_ecg_p_12
## fibr_ter_01
## fibr_ter_02
## fibr_ter_03
## fibr_ter_05
## fibr_ter_06
## fibr_ter_07
## fibr_ter_08
## GIPO_K
## K_BLOOD
## GIPER_NA
## NA_BLOOD
## ALT_BLOOD
                  0.0253636194
## AST_BLOOD
## KFK_BLOOD
## L_BLOOD
## ROE
## TIME_B_S
## R_AB_1_n
## R_AB_2_n
## R_AB_3_n
                  0.0061743908
## NA_KB
## NOT_NA_KB
## TID KR
                  0.0209955362
## LID_KB
                  0.0085734823
## NITR_S
## NA_R_1_n
## NA_R_2_n 0.0624263680
## NA_R_3_n 0.0985947304
## NOT_NA_1_n 0.0157578267
## NOT_NA_2_n
## NOT_NA_3_n
## LID_S_n
## B_BLOK_S_n
## ANT_CA_S_n
## GEPAR_S_n
                 -0.0099349438
## ASP_S_n
## TIKL_S_n
## TRENT_S_n
x = data[,2:112]
y = data[,'REC_IM']
cv_model = cv.glmnet(as.matrix(x),y, alpha=1) ## 10-fold cross validation
best_lambda = cv_model$lambda.min ## 0.01339178
plot(cv_model)
```

108 107 108 107 102 98 94 77 62 32 12 4 2 1



```
## LASSO
lasso_model = glmnet(x,y, alpha=1, lambda = best_lambda) ##alpha=1 is LASSO
coef(lasso_model)
## 112 x 1 sparse Matrix of class "dgCMatrix"
##
                 -0.0376688120
## (Intercept)
## AGE
                  0.0010370397
## SEX
                 -0.0044599629
## INF_ANAM
## STENOK_AN
                  0.0075441960
## FK_STENOK
## IBS_POST
                  0.0059883330
## IBS_NASL
## GB
## SIM_GIPERT
## DLIT_AG
## ZSN_A
## nr_11
## nr_01
## nr_02
## nr_03
                  0.0402408641
## nr_04
## nr_07
```

```
## nr_08
               0.6323311551
## np_01
## np_04
## np_05
## np_07
## np_08
## np_09
## np_10 0.0594247505
## endocr_01 0.0059498843
## endocr_02
## endocr_03
## zab_leg_01
                -0.0207001131
              0.0334371408
## zab_leg_02
## zab_leg_03
## zab_leg_04
## zab_leg_06
## S_AD_KBRIG
## D AD KBRIG
## S_AD_ORIT
## D_AD_ORIT
## O_L_POST
## K_SH_POST
                -0.0547746265
## MP_TP_POST
## SVT_POST
## GT_POST
               0.1650972264
## FIB_G_POST 0.0036998215
               0.0015787898
## ant_im
## lat_im
                 0.0003735782
## inf_im
## post_im
## IM_PG_P
## ritm_ecg_p_01 .
## ritm_ecg_p_02 -0.0107063723
## ritm_ecg_p_04
## ritm_ecg_p_06
## ritm_ecg_p_07  0.0067364141
## ritm_ecg_p_08 .
## n_r_ecg_p_01
## n_r_ecg_p_02
## n_r_ecg_p_03
## n_r_ecg_p_04
## n_r_ecg_p_05
## n_r_ecg_p_06 -0.0376731704
## n_r_ecg_p_08
                 0.0917148846
## n_r_ecg_p_09
## n_r_ecg_p_10
## n_p_ecg_p_01
## n_p_ecg_p_03
## n_p_ecg_p_04
## n_p_ecg_p_05
## n_p_ecg_p_06
## n_p_ecg_p_07
## n_p_ecg_p_08
## n_p_ecg_p_09
```

```
## n_p_ecg_p_10
                 0.0374898418
## n_p_ecg_p_11
## n_p_ecg_p_12
## fibr_ter_01
## fibr_ter_02
## fibr_ter_03
## fibr_ter_05
## fibr_ter_06
## fibr_ter_07
## fibr_ter_08
## GIPO_K
## K_BLOOD
## GIPER_NA
                 0.0288109190
## NA_BLOOD
## ALT_BLOOD
## AST_BLOOD
## KFK_BLOOD
## L_BLOOD
                 0.0027686056
## ROE
## TIME_B_S
## R_AB_1_n
## R_AB_2_n
## R_AB_3_n
                 0.1251436955
## NA_KB
## NOT_NA_KB
## LID_KB
                -0.0030615932
## NITR_S
                 0.0019354032
## NA_R_1_n
                 0.0035326449
## NA_R_2_n
## NA_R_3_n
                 0.0336397371
                 0.0067432103
## NOT_NA_1_n
## NOT_NA_2_n
                 0.0021473474
## NOT_NA_3_n
## LID_S_n
## B_BLOK_S_n
## ANT_CA_S_n
              -0.0092336644
## GEPAR_S_n
                 0.0130147926
## ASP_S_n
## TIKL_S_n
## TRENT_S_n
                -0.0082192527
#x, response
newd = cor(x[sapply(x,is.numeric)])
melted = melt(newd)
ggplot(melted, aes(x = Var1, y = Var2, fill = value)) + geom_tile() + scale_fill_gradient(high = "red",
  labs(title = "Correlation Heatmap of Covariates") +
 theme(text = element_text(size=4), axis.text.x = element_text(angle=90, hjust = .5, vjust = .5))
```

