## hw2

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## Question 4

## [1] 0.2530864

```
SAheart = read.table("http://www-stat.stanford.edu/~tibs/ElemStatLearn/datasets/SAheart.data",sep=",",h
train = SAheart[1:300 ,]
test = SAheart[301:462,]
log_model = glm(chd ~ ., family = binomial(link = "logit"), data = train)
summary(log_model)
##
## Call:
## glm(formula = chd ~ ., family = binomial(link = "logit"), data = train)
## Deviance Residuals:
                    Median
      Min
           10
                                30
                                        Max
## -1.8176 -0.8282 -0.4348 0.9484
                                     2.4491
## Coefficients:
##
                 Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                -5.013991 1.570094 -3.193 0.00141 **
                ## sbp
                 0.069165
## tobacco
                          0.032321
                                     2.140 0.03236 *
                 0.111462 0.074524 1.496 0.13474
## ldl
## adiposity
                 0.046678 0.037295
                                     1.252 0.21072
## famhistPresent 0.826525 0.280107 2.951 0.00317 **
                                     2.949 0.00319 **
## typea
                 0.046926
                            0.015911
                            0.053181 -1.190 0.23404
## obesity
                -0.063287
## alcohol
                 0.005111
                            0.005927 0.862 0.38849
                 0.039145
                            0.014631 2.675 0.00746 **
## age
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 394.29 on 299 degrees of freedom
## Residual deviance: 317.75 on 290 degrees of freedom
## AIC: 337.75
## Number of Fisher Scoring iterations: 4
fitted.results = predict(log_model, test, type = "response")
fitted.results <- ifelse(fitted.results > 0.5,1,0) #picking 0.5 as the boundry
log_error = mean((test$chd - fitted.results)^2)
log_error
```

1

```
##same as log_error = mean(fitted.results != test$chd)
std_log_error = sd((test$chd - fitted.results)^2)/ sqrt(nrow(test))
std_log_error
## [1] 0.03426547
LDA
lda_model = lda(chd~., data = train)
summary(lda_model)
##
          Length Class Mode
## prior
         2
                -none- numeric
## counts 2
                -none- numeric
## means 18
               -none- numeric
## scaling 9
                -none- numeric
## lev 2 -none- character
## svd
         1
               -none- numeric
## N
               -none- numeric
          1
          3 -none- call
## call
## terms 3 terms call
## xlevels 1
                -none- list
lda_pred = predict(lda_model, test)$class
lda_error = mean((test$chd - lda_pred)^2)
## Warning in Ops.factor(test$chd, lda_pred): '-' not meaningful for factors
lda_error
## [1] NA
lda_pred = as.numeric(lda_pred)
test$chd = as.numeric(test$chd)
std_lda_error = sd((test$chd - lda_pred)^2)/ sqrt(nrow(test))
std_lda_error
## [1] 0.08431974
QDA
qda_model = qda(chd~., data = train)
qda_model$means
##
                          ldl adiposity famhistPresent
         sbp tobacco
                                                         typea obesity
## 0 134.0105 2.629368 4.468421 23.29147
                                             0.2947368 53.04211 25.57868
## 1 138.9091 5.433818 5.451545 27.90682
                                             0.5545455 56.28182 26.76291
     alcohol
                  age
## 0 12.41268 38.02632
## 1 18.75418 49.47273
qda_pred = predict(qda_model, test)$class
```

qda\_pred = as.numeric(qda\_pred)

```
qda_error = mean((qda_pred - test$chd)^2)
std_qda_error = sd((qda_pred - test$chd)^2)/sqrt(nrow(test))

log = cbind("Logistic Regression",log_error,std_log_error)
lda = cbind("LDA",lda_error, std_lda_error)
qda = cbind("QDA",qda_error, std_qda_error)

summary_table = rbind(log, lda, qda)
colnames(summary_table) = c("Model", "Test Error", "Std Error")
```

Given that the test error and the standard errors are similar for all three models, I'd pick logistic since it's the easiest to interpret.

## Question 3