MPH Practicum

Table 1: Demographics

category	cancer	noncancer		
Gender				
Male				
Female				
Median Age				
Smoking History				
Never				
Former				
Current				

Table 2: Exclusion criteria

	n
Initial	184
Excluded	
UTUC	5
No cysto	1
No follow up	4
Insufficient RNA for biomarker analysis	7
Duplicates	9
Total	158

Table 3: Smoking count

smoking	smoker	n
never	0	50
former	1	78
current	1	30

```
kable(smoking_count,
    booktabs = T,
    align = "1",
    caption = "Smoking count"
)
```

```
kable(cysto_result_risk,
    booktabs = T,
    align = "l",
    caption = "Cysto results vs risk"
    )
```

```
kable(smoking_hx,
    booktabs = T,
    align = "1",
    caption = "Smoking history"
    )
```

Table 4: Cysto results vs risk

visual_suspicion	risk	bca	n
no	NoBC	0	81
yes	NoBC	0	11
yes	low	1	7
yes	intermediate	1	12
yes	high	1	47

Table 5: Smoking history

smoking	packyrs	yearquit	n
never	0	0	50
former	0	1	3
former	1	1	75
current	1	0	30

#logistic regression cancer = biomarkers

Coefficients:

```
log_biomarkers <- glm(bca ~ ROBO1 + CRH + IGF2, family = "binomial", data = detection)
summary(log_biomarkers)
##
## Call:
## glm(formula = bca ~ ROBO1 + CRH + IGF2, family = "binomial",
##
      data = detection)
##
## Deviance Residuals:
      Min
            1Q
                    Median
                                  3Q
                                          Max
## -2.4177 -0.4177 -0.2003
                             0.2754
                                       2.3159
## Coefficients:
              Estimate Std. Error z value Pr(>|z|)
##
                          3.30546
                                   6.535 6.36e-11 ***
## (Intercept) 21.60163
              -0.10807
                          0.07557 -1.430 0.152722
## ROB01
## CRH
              -0.28474
                          0.07745 -3.676 0.000236 ***
              -0.27209
                          0.09409 -2.892 0.003831 **
## IGF2
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 214.737 on 157 degrees of freedom
## Residual deviance: 90.608 on 154 degrees of freedom
## AIC: 98.608
##
## Number of Fisher Scoring iterations: 6
#logistic regression cancer = model3
log_model3 <- glm(bca ~ MODEL3, family = "binomial", data = detection)</pre>
summary(log_model3)
##
## glm(formula = bca ~ MODEL3, family = "binomial", data = detection)
## Deviance Residuals:
      Min
                1Q
                    Median
                                  3Q
                                          Max
## -2.1730 -0.4505 -0.2786 0.4570
                                       2.4043
```

```
Estimate Std. Error z value Pr(>|z|)
## (Intercept) -3.4050
                           0.5000 -6.809 9.80e-12 ***
## MODEL3
                5.7850
                           0.7659 7.553 4.26e-14 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 214.74 on 157 degrees of freedom
## Residual deviance: 108.60 on 156 degrees of freedom
## AIC: 112.6
## Number of Fisher Scoring iterations: 5
#logistic regression cancer = biomarkers + smoking(binary)
log_smoker <- glm(bca ~ ROB01 + CRH + IGF2 + smoker, family = "binomial", data = detection)
summary(log smoker)
##
## glm(formula = bca ~ ROBO1 + CRH + IGF2 + smoker, family = "binomial",
      data = detection)
##
## Deviance Residuals:
       Min
                  1Q
                        Median
                                      3Q
                                               Max
## -2.73058 -0.34312 -0.08536 0.18277
                                           2.25117
##
## Coefficients:
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) 21.93236
                          3.82240
                                   5.738 9.59e-09 ***
## ROB01
              -0.07658
                          0.07960 -0.962 0.336025
## CRH
                          0.08798 -3.878 0.000105 ***
              -0.34115
## IGF2
              -0.31032
                          0.10409 -2.981 0.002871 **
## smoker
              2.50849
                          0.83424
                                   3.007 0.002639 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
      Null deviance: 214.737 on 157 degrees of freedom
## Residual deviance: 78.957 on 153 degrees of freedom
## AIC: 88.957
## Number of Fisher Scoring iterations: 7
#logistic regression cancer = biomarkers + smoking(factor)
log_smoking <- glm(bca ~ ROBO1 + CRH + IGF2 + smoking, family = "binomial", data = detection)</pre>
summary(log_smoking)
##
## Call:
## glm(formula = bca ~ ROB01 + CRH + IGF2 + smoking, family = "binomial",
##
     data = detection)
```

```
##
## Deviance Residuals:
      Min
               1Q
                     Median
## -2.95643 -0.30678 -0.07892 0.18497
                                     2.34767
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
                         4.08118 5.594 2.22e-08 ***
               22.83002
## (Intercept)
                        0.08321 -1.205 0.22810
## ROB01
               -0.10029
## CRH
               ## IGF2
               2.737 0.00619 **
                         0.85759
## smokingformer 2.34762
## smokingcurrent 3.13124
                         1.05069
                                 2.980 0.00288 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 214.737 on 157 degrees of freedom
\mbox{\tt \#\#} Residual deviance: 77.869 on 152 degrees of freedom
## AIC: 89.869
## Number of Fisher Scoring iterations: 7
```

```
\#R-squared cancer = biomarkers
pR2(log_biomarkers)['McFadden']
## fitting null model for pseudo-r2
## McFadden
## 0.5780512
#R-squared cancer = model3
pR2(log_model3)['McFadden']
## fitting null model for pseudo-r2
## McFadden
## 0.4942864
#R-squared cancer = biomarkers + smoking(binary)
pR2(log_smoker)['McFadden']
## fitting null model for pseudo-r2
## McFadden
## 0.6323096
#R-squared cancer = biomarkers + smoking(factor)
pR2(log_smoking)['McFadden']
## fitting null model for pseudo-r2
## McFadden
## 0.6373741
```

```
#c statistics cancer = biomarkers
predictions_biomarkers <- predict(log_biomarkers, type = "response")</pre>
roc(detection$bca, predictions biomarkers)
## Setting levels: control = 0, case = 1
## Setting direction: controls < cases
##
## Call:
## roc.default(response = detection$bca, predictor = predictions_biomarkers)
## Data: predictions_biomarkers in 92 controls (detection$bca 0) < 66 cases (detection$bca 1).
## Area under the curve: 0.9463
#c statistics cancer = model3
predictions_model3 <- predict(log_model3, type = "response")</pre>
roc(detection$bca, predictions_model3)
## Setting levels: control = 0, case = 1
## Setting direction: controls < cases
##
## Call:
## roc.default(response = detection$bca, predictor = predictions_model3)
## Data: predictions_model3 in 92 controls (detection$bca 0) < 66 cases (detection$bca 1).
## Area under the curve: 0.9269
#c statistics cancer = biomarkers + smoking(binary)
predictions_smoker <- predict(log_smoker, type = "response")</pre>
roc(detection$bca, predictions_smoker)
## Setting levels: control = 0, case = 1
## Setting direction: controls < cases
##
## Call:
## roc.default(response = detection$bca, predictor = predictions_smoker)
## Data: predictions smoker in 92 controls (detection$bca 0) < 66 cases (detection$bca 1).
## Area under the curve: 0.9601
#c statistics cancer = biomarkers + smoking(factor)
predictions_smoking <- predict(log_smoking, type = "response")</pre>
roc(detection$bca, predictions_smoking)
## Setting levels: control = 0, case = 1
## Setting direction: controls < cases
```

```
##
## Call:
## roc.default(response = detection$bca, predictor = predictions_smoking)
##
## Data: predictions_smoking in 92 controls (detection$bca 0) < 66 cases (detection$bca 1).
## Area under the curve: 0.9615</pre>
```

```
#BIC cancer = biomarkers
BIC(log_biomarkers)

## [1] 110.8582

#BIC cancer = model3
BIC(log_model3)

## [1] 118.7204

#BIC cancer = biomarkers + smoking(binary)
BIC(log_smoker)

## [1] 104.2695

#BIC cancer = biomarkers + smoking(factor)
BIC(log_smoking)
```

[1] 108.2446

```
#ROC cancer = biomarkers
ggroc(roc(detection$bca, predictions_biomarkers)) +
  theme_minimal() +
  ggtitle("ROC log_biomarkers") +
  geom_segment(aes(x = 1, xend = 0, y = 0, yend = 1), color="grey", linetype="dashed")
```

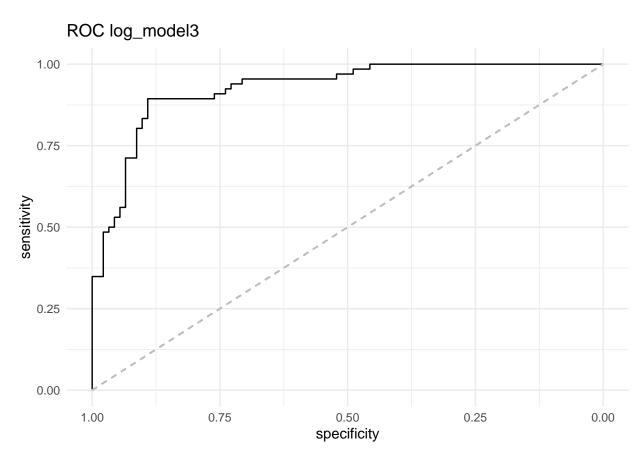
Setting levels: control = 0, case = 1

Setting direction: controls < cases

0.75 0.50 1.00 0.75 0.50 0.00 1.00 0.75 0.50 0.25 0.00 specificity

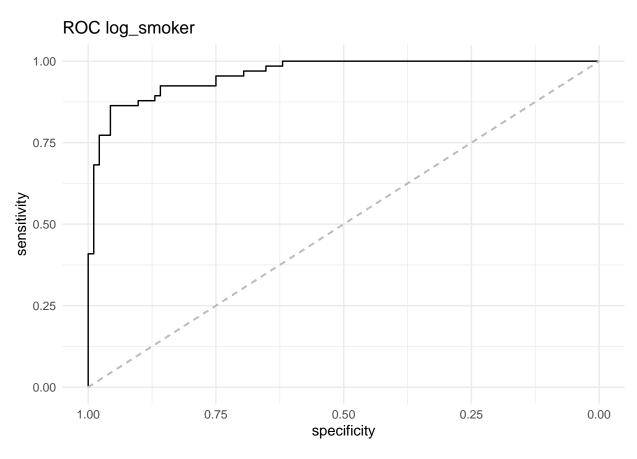
```
#ROC cancer = model3
ggroc(roc(detection$bca, predictions_model3)) +
   theme_minimal() +
   ggtitle("ROC log_model3") +
   geom_segment(aes(x = 1, xend = 0, y = 0, yend = 1), color="grey", linetype="dashed")

## Setting levels: control = 0, case = 1
## Setting direction: controls < cases</pre>
```



```
#ROC cancer = biomarkers + smoking(binary)
ggroc(roc(detection$bca, predictions_smoker)) +
  theme_minimal() +
  ggtitle("ROC log_smoker") +
  geom_segment(aes(x = 1, xend = 0, y = 0, yend = 1), color="grey", linetype="dashed")
```

Setting levels: control = 0, case = 1
Setting direction: controls < cases</pre>



```
#ROC cancer = biomarkers + smoking(factor)
ggroc(roc(detection$bca, predictions_smoking)) +
   theme_minimal() +
   ggtitle("ROC log_smoking") +
   geom_segment(aes(x = 1, xend = 0, y = 0, yend = 1), color="grey", linetype="dashed")
```

Setting levels: control = 0, case = 1
Setting direction: controls < cases</pre>

