

```
> #Measuring the performance of Logistic Regression as a classifier
> #Using the Challenger data as an example
>
```

```
> library(faraway)
```

```
> library(caret)
```

```
Loading required package: lattice
```

```
Attaching package: 'lattice'
```

```
The following object is masked from 'package:faraway':
```

```
melanoma
```

```
Loading required package: ggplot2
```

```
Warning message:
```

```
package 'ggplot2' was built under R version 3.6.3
```

```
> library(pROC)
```

```
Type 'citation("pROC")' for a citation.
```

```
Attaching package: 'pROC'
```

```
The following objects are masked from 'package:stats':
```

```
cov, smooth, var
```

```
Warning message:
```

```
package 'pROC' was built under R version 3.6.3
```

```
> library(epiDisplay)
```

```
Loading required package: foreign
```

```
Loading required package: survival
```

```
Attaching package: 'survival'
```

```
The following object is masked from 'package:caret':
```

```
cluster
```

```
The following objects are masked from 'package:faraway':
```

```
rats, solder
```

```
Loading required package: MASS
```

```
Loading required package: nnet
```

```
Attaching package: 'epiDisplay'
```

```
The following object is masked from 'package:pROC':
```

```
ci
```

The following object is masked from 'package:ggplot2':

alpha

The following object is masked from 'package:lattice':

dotplot

Warning message:

package 'epiDisplay' was built under R version 3.6.3

>

>

> if (FALSE)

+ {"

+ O-Ring data analyzed using a logistic model in R

+ "}

>

> #read in the data which is in a csv file

> oring <- read.csv("C:/Users/jmard/Desktop/Computing and Graphics in Applied
Statistics2020/Lecture 08 14Feb2020/Challenger.csv",header = TRUE)

> oring

	Launch	Temp	TD
1	1	53	1
2	2	56	1
3	3	57	1
4	4	63	0
5	5	66	0
6	6	67	0
7	7	67	0
8	8	67	0
9	9	68	0
10	10	69	0
11	11	70	0
12	12	70	1
13	13	70	1
14	14	70	1
15	15	72	0
16	16	73	0
17	17	75	0
18	18	75	1
19	19	76	0
20	20	76	0
21	21	78	0
22	22	79	0
23	23	80	0
24	24	81	0

> nrow(oring)

[1] 24

> summary(oring)

Launch

Temp

TD

```

Min.      : 1.00      Min.      :53.00      Min.      :0.0000
1st Qu.: 6.75      1st Qu.:67.00      1st Qu.:0.0000
Median :12.50      Median :70.00      Median :0.0000
Mean    :12.50      Mean    :69.92      Mean    :0.2917
3rd Qu.:18.25      3rd Qu.:75.25      3rd Qu.:1.0000
Max.     :24.00      Max.     :81.00      Max.     :1.0000
>
> logistic <- glm(TD ~ Temp,data=oring,family=binomial(link='logit'))
> summary(logistic)

Call:
glm(formula = TD ~ Temp, family = binomial(link = "logit"), data = oring)

Deviance Residuals:
      Min       1Q   Median       3Q      Max
-1.2125  -0.8253  -0.4706   0.5907   2.0512

Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept) 10.87535     5.70291   1.907   0.0565 .
Temp        -0.17132     0.08344  -2.053   0.0400 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 28.975  on 23  degrees of freedom
Residual deviance: 23.030  on 22  degrees of freedom
AIC: 27.03

Number of Fisher Scoring iterations: 4

>
> windows(7,7)
> #generate ROC curve
> ROCresult <- roc(oring$TD ~ logistic$fitted)
Setting levels: control = 0, case = 1
Setting direction: controls < cases
> plot(ROCresult, legacy.axes = TRUE)
> names(ROCresult)
 [1] "percent"          "sensitivities"    "specificities"
 [4] "thresholds"       "direction"         "cases"
 [7] "controls"         "fun.sesp"         "auc"
[10] "call"            "original.predictor" "original.response"
[13] "predictor"       "response"         "levels"
> ROCresult$auc
Area under the curve: 0.7227
>
> logistic.display(logistic)

```

Logistic regression predicting TD

	OR(95%CI)	P(Wald's test)	P(LR-test)
Temp (cont. var.)	0.84 (0.72,0.99)	0.04	0.015

Log-likelihood = -11.5152

No. of observations = 24

AIC value = 27.0305

```
>
> #now save the graph in a pdf file
> pdf(file="C:/users/jmard/Desktop/Computing and Graphics in Applied
Statistics2020/Output/Logistic_as_a_ClassifierR_Figure.pdf")
>
> plot(TD~Temp,data=oring,xlab="Temperature",ylab="Thermal Distress")
> lines(oring$Temp,logistic$fitted, type="l", col="red")
> title(main="O-Ring Data with Fitted Logistic Regression Line")
>
> #Assessing the predictive ability of the model
> #would like to see how the model is doing as a classifier
> #Our decision boundary will be 0.5. If predicted probability of P(TD|Temp) >
0.5 then predicted.TD = 1 otherwise predicted.TD=0
> #Note that for some applications, thresholds different than 0.5 could be a
better option
>
> #This analysis is provided for instructional purposes only
> #we should be using test data and should be performing Cross Validation
> #we are using the training data set so overfitting is a concern
>
> predicted.TD <- predict(logistic,data=oring,type='response') #using the
type='response' option generates P(TD|at each level of Temp)
> predicted.TD <- ifelse(predicted.TD > 0.5,1,0) #predicted.TD is 1 if predicted
P(TD|Temp) > 0.50, 0 otherwise
>
> table(predicted.TD, oring$TD)

predicted.TD  0  1
              0 16  4
              1  1  3

>
> misClasificError <- mean(predicted.TD != oring$TD) # != is the symbol for not
equal
> misClasificError
[1] 0.2083333
>
> #check misClassificationError rate
> 5/24
[1] 0.2083333
>
> print(paste('Accuracy',1-misClasificError))
```

```

[1] "Accuracy 0.791666666666667"
>
> oring <- cbind(oring,predicted.TD)
> data.frame(oring)
  Launch Temp TD predicted.TD
1      1   53  1           1
2      2   56  1           1
3      3   57  1           1
4      4   63  0           1
5      5   66  0           0
6      6   67  0           0
7      7   67  0           0
8      8   67  0           0
9      9   68  0           0
10     10   69  0           0
11     11   70  0           0
12     12   70  1           0
13     13   70  1           0
14     14   70  1           0
15     15   72  0           0
16     16   73  0           0
17     17   75  0           0
18     18   75  1           0
19     19   76  0           0
20     20   76  0           0
21     21   78  0           0
22     22   79  0           0
23     23   80  0           0
24     24   81  0           0
>
> predicted.TD <- as.factor(predicted.TD) #need to be sure this variable is a
factor
> oring$TD <- as.factor(oring$TD) #need to be sure this variable is a factor
>
> confusionMatrix(oring$TD,predicted.TD) #found in the caret library
Confusion Matrix and Statistics

```

	Reference	
Prediction	0	1
0	16	1
1	4	3

Modeling 0 as an event

```

Accuracy : 0.7917 =19/24
95% CI : (0.5785, 0.9287)
No Information Rate : 0.8333
P-Value [Acc > NIR] : 0.8005

```

Kappa : 0.4231 measure of agreement

Mcnemar's Test P-Value : 0.3711 test of agreement - paired observations

```
Sensitivity : 0.8000 =16/20
Specificity : 0.7500 = 3/4
Pos Pred Value : 0.9412 =16/17
Neg Pred Value : 0.4286 =3/7
Prevalence : 0.8333
Detection Rate : 0.6667
Detection Prevalence : 0.7083
Balanced Accuracy : 0.7750
```

```
'Positive' Class : 0
```

```
>
> dev.off()
null device
  1
>
```