

### Question 5:

a. Fit 1 is below:

```
> fit1=glm(SoreThroat$Y~SoreThroat$D+SoreThroat$T,data=SoreThroat, family = binomial(link = "logit"))
> summary(fit1)
```

Call:  
glm(formula = SoreThroat\$Y ~ SoreThroat\$D + SoreThroat\$T, family = binomial(link = "logit"),  
data = SoreThroat)

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-1.41734	1.09457	-1.295	0.19536
SoreThroat\$D	0.06868	0.02641	2.600	0.00931 **
SoreThroat\$T	-1.65895	0.92285	-1.798	0.07224 .

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 46.180 on 34 degrees of freedom  
Residual deviance: 30.138 on 32 degrees of freedom  
AIC: 36.138

Number of Fisher Scoring iterations: 5

b. Fit 2 is below:

```
> fit2=glm(SoreThroat$Y~SoreThroat$D+SoreThroat$T+SoreThroat$T:SoreThroat$D,data=SoreThroat, family = binomial(link = "logit"))
> summary(fit2)
```

Call:  
glm(formula = SoreThroat\$Y ~ SoreThroat\$D + SoreThroat\$T + SoreThroat\$T:SoreThroat\$D,  
family = binomial(link = "logit"), data = SoreThroat)

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	0.04979	1.46940	0.034	0.9730
SoreThroat\$D	0.02848	0.03429	0.831	0.4062
SoreThroat\$T	-4.47224	2.46707	-1.813	0.0699 .
SoreThroat\$D:SoreThroat\$T	0.07460	0.05777	1.291	0.1966

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 46.180 on 34 degrees of freedom  
Residual deviance: 28.321 on 31 degrees of freedom  
AIC: 36.321

Number of Fisher Scoring iterations: 6

c. The prediction equation for fit 2 is:

$$\text{Logit}(p(Y=1)) = 0.049 + 0.028 \times \text{Duration} - 4.472 \times \text{Type} + 0.074 \times (D \times T)$$

For Duration (D):

For every one minute increase in surgery, the estimated odds of sore throat is increased by  $\exp(0.028) = 1.02$  times while keeping other variables held constant.

For Type of device used (T):

For every other type of device used in surgery, the estimated odds of sore throat is increased by  $\exp(-4.472) = 0.011$  times while keeping other variables held constant.

For Duration in Duration interacting with Type of device:

For every one-minute increase in surgery for a fixed type of device, the estimated odds of sore throat is increased by  $\exp(0.077) = 1.077$  times while keeping other variables held constant.

For Type of device used in Duration interacting with Type of device:

For every change in device type used while keeping duration fixed, the estimated odds of sore throat is increased by  $\exp(-4.423) = 0.011$  times while keeping other variables held constant.

d. LRT test:

```
> anova(fit1, fit2, test="LRT")
```

Analysis of Deviance Table

Model 1: SoreThroat\$Y ~ SoreThroat\$D + SoreThroat\$T

Model 2: SoreThroat\$Y ~ SoreThroat\$D + SoreThroat\$T + SoreThroat\$T:SoreThroat\$D

	Resid. Df	Resid. Dev	Df	Deviance	Pr(>Chi)
1	32	30.138			
2	31	28.321	1	1.8169	0.1777

From above, the model with interaction term does not seem adequate in predicting the sore throat and hence fit 1 without interaction seems more adequate with p-value 0.177 > 0.10 at 10% level of significance with deviance of 1.8 for DF=1.

e. Fit 1 is selected without interaction term from comparing above.

```

> fit1=glm(SoreThroat$Y~SoreThroat$D+SoreThroat$T,data=SoreThroat, family = binomial(link = "logit"))
> summary(fit1)

Call:
glm(formula = SoreThroat$Y ~ SoreThroat$D + SoreThroat$T, family = binomial(link = "logit"),
    data = SoreThroat)

Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept)  -1.41734     1.09457  -1.295  0.19536
SoreThroat$D   0.06868     0.02641   2.600  0.00931 **
SoreThroat$T  -1.65895     0.92285  -1.798  0.07224 .
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 46.180  on 34  degrees of freedom
Residual deviance: 30.138  on 32  degrees of freedom
AIC: 36.138

Number of Fisher Scoring iterations: 5

```

For Duration (D):

For every one-minute increase in surgery, the estimated odds of sore throat is increased by  $\exp(0.068) = 1.07$  times while keeping other variables held constant.

For Type of device used (T):

For every other type of device used in surgery, the estimated odds of sore throat is increased by  $\exp(-1.658) = 0.190$  times while keeping other variables held constant.

f. Predictive power for model fit 1:

```

> predicted1 <- as.numeric(fitted(fit1) > phat)
> xtabs(~SoreThroat$Y + predicted1)

```

	predicted1	
SoreThroat\$Y	0	1
0	10	3
1	4	18

Fit 1 predicts 10 cases with no sore throats out of 13 and 18 cases with sore throats out of 22.

Predictive model for fit2:

```
> predicted2 <- as.numeric(fitted(fit2) > phat)
> xtabs(~SoreThroat$Y + predicted2)
      predicted2
SoreThroat$Y  0  1
      0    9  4
      1    2 20
```

Fit 2 predicts 9 cases with no sore throats out of 13 and 20 cases with sore throats out of 22.

Fit 1 seems adequate with better prediction compared with observed and fitted values.

- g. The predicted value is below:

```
[1] -0.39777
```

The probability for 39 minutes of duration of surgery with tracheal tube type of device the predicted value is -0.3977.

- h. The predicted value is below:

```
[1] 1.26118
```

The probability for a patient underwent surgery and experienced sore throat with duration of surgery as 39 minutes when a laryngeal mask airway used has a predicted value of 1.261.

- i. The predicted value for tracheal tube as device seems much lower than the predicted value for laryngeal mask as a device for sore throat as the patient with sore throat were much higher for mask airway than tube device type.