

# CS-324 MACHINE LEARNING

COMPLEX ENGINEERING PROBLEM

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#### **Objective:**

The primary objective of this report is to investigate the effect of gene polymorphism on renal dysfunction in children following liver transplantation. Utilizing the provided hypothetical dataset of 60 pediatric liver transplant recipients using the logistic regression model, the report aims to:

#### Step1:

Filling missing values using group members' roll no. In mod 7. The members' roll numbers are 087,011,060 and for the fourth missing value taking the average of all three roll numbers i.e. 52.67 rounded off to 53.

- 1.  $87 \mod 7 = 3$
- 2.  $11 \mod 7 = 4$
- 3.  $60 \mod 7 = 4$
- 4.  $53 \mod 7 = 4$

(An updated Excel sheet is attached for reference).

#### Step2:

Performance of all questioned tasks.

30

a) Using some suitable software system, produce the binary logistic model from the given training data.

Software Used: Minitab

The given Excel file was uploaded and recoded as follows:

#### Summary

Male

# Original Recoded Number Value Value of Rows Female 0 30

Recoded data column Sex

#### Summary

Original	Recoded	Number
Value	Value	of Rows
Leu/Leu	0	21
Leu/Pro	1	24
Pro/Pro	2	15

Recoded data column Type

#### Summary

Original	Recoded	Number
Value	Value	of Rows
Absent	0	14
Absent	0	18
Present	1	16
Present	1	12

Recoded data column Disease

Then binary logistic model was produced. The following results were obtained.

#### Binary Logistic Regression: Disease versus TST, Sex, Type

#### Method

Link function Logit Categorical predictor coding (1, 0) Rows used 60

#### **Response Information**

Variable	Value	Count	
Disease	1	28	(Event)
	0	32	
	Total	60	

#### **Regression Equation**

$P(1) = \exp(Y')/(1 + \exp(Y'))$ <b>Sex Type</b>						
0	0	Y' = -4.921 + 0.3604 TST				
0	1	Y' = -3.031 + 0.3604  TST				
0	2	Y' = -3.142 + 0.3604  TST				
1	0	Y' = -4.827 + 0.3604 TST				
1	1	Y' = -2.937 + 0.3604 TST				
1	2	Y' = -3.048 + 0.3604  TST				

#### Coefficients

Term	Coef	SE Coef	<b>Z-Value</b>	P-Value	VIF
Constant	-4.92	1.40	-3.53	0.000	
TST	0.360	0.111	3.26	0.001	1.26
Sex					
1	0.094	0.638	0.15	0.883	1.05

## Type 1 1.890 0.815 2.32 0.020 1.69 2 1.779 0.903 1.97 0.049 1.66

#### **Odds Ratios for Continuous Predictors**

	Odds Ratio	95% CI
TST	1.4338	(1.1544, 1.7808)

#### **Odds Ratios for Categorical Predictors**

Level A	Level B	Odds Ratio	95% CI
Sex			
1	0	1.0986	(0.3145, 3.8373)
Type			
1	0	6.6179	(1.3392, 32.7031)
2	0	5.9215	(1.0096, 34.7314)
2	1	0.8948	(0.2006, 3.9917)

Odds ratio for level A relative to level B

#### **Model Summary**

Deviance	Deviance				Area Under
R-Sq	R-Sq(adj)	AIC	AICc	BIC	ROC Curve
25.26%	20.43%	71.97	73.08	82.44	0.8270

#### **Goodness-of-Fit Tests**

Test	DF	Chi-Square	P-Value
Deviance	55	61.97	0.241
Pearson	55	57.74	0.374
Hosmer-Lemeshow	8	8.78	0.361

#### **Analysis of Variance**

# Source DF Chi-Square P-Value Regression 4 12.68 0.013 TST 1 10.62 0.001 Sex 1 0.02 0.883 Type 2 5.87 0.053

#### Fits and Diagnostics for Unusual Observations

Obs	Observed Probability	Fit	Resid	Std Resid	
29	1.000	0.113	2.089	2.18	R
30	1.000	0.154	1.933	2.03	R

b) Compute the odds ratio for all the predictor variables (using the probability method) and interpret them appropriately.

Using the equation computed by Minitab. Odd ratios are calculated.

#### $Odds_0$

Sex=0 Type=0 Y' = -4.921 + 0.3604 TST 
$$Y' = -4.921 + 0.3604(4)$$
 
$$= -3.4794$$
 
$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{3.4794}} = 0.0299$$
 
$$Odds_0 = \frac{P(1)}{1 - P(1)} = 0.0308$$

 $Odds_1$ 

Sex=1 Type=0 Y' = -4.827 + 0.3604 TST 
$$Y' = -4.827 + 0.3604(4)$$
$$= -3.3854$$
$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{3.3854}} = 0.0328$$
$$Odds_1 = \frac{P(1)}{1 - P(1)} = 0.339$$

**Odds Ratio** sex:

$$Odds \ Ratio = rac{Odds_1}{Odds_0}$$
 $Odds \ Ratio \ _{sex} = 1.0995$ 

#### Interpretation of *Odds Ratio* sex:

Males have approximately 9.95% higher odds of renal dysfunction than females, considering other features constant.

 $0dds_2$ 

$$= -3.4794$$

$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{3.4794}} = 0.0299$$

$$Odds_2 = \frac{P(1)}{1 - P(1)} = 0.0308$$

 $0dds_3$ 

Sex=0 Type=1 Y' = 
$$-3.031 + 0.3604$$
 TST TST=4

$$Y' = -3.031 + 0.3604(4)$$

$$= -1.5894$$

$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{1.5894}} = 0.1695$$

$$Odds_3 = \frac{P(1)}{1 - P(1)} = 0.2040$$

Odds Ratio type1vstype0:

$$Odds \ Ratio = rac{Odds_3}{Odds_2}$$
 $Odds \ Ratio \ _{type1} = 6.6249$ 

#### Interpretation of *Odds Ratio* type1vstpe0:

Individuals with the Leu/Pro genotype have approximately 6.62 times higher odds of renal dysfunction compared to those with the Leu/Leu genotype, holding all other features constant.

 $Odds_4$ 

$$Sex=0$$
 Type=0 Y' = -4.921 + 0.3604 TST

TST=4

$$Y' = -4.921 + 0.3604(4)$$

$$= -3.4794$$

$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{3.4794}} = 0.0299$$

$$Odds_2 = \frac{P(1)}{1 - P(1)} = 0.0308$$

 $0dds_5$ 

Sex=0 Type=2 Y' = -3.142 + 0.3604 TST

$$Y' = -3.142 + 0.3604(4)$$

$$= -1.7004$$

$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{1.5894}} = 0.1544$$

$$Odds_3 = \frac{P(1)}{1 - P(1)} = 0.1826$$

#### Odds Ratio type2vstpye0

$$Odds \ Ratio = rac{Odds_5}{Odds_4}$$
 $Odds \ Ratio \ _{type1} = 5.9286$ 

### Interpretation of *Odds Ratio* $_{type1vstpe0}$ :

Individuals with the Pro/Pro genotype have approximately 5.93 times higher odds of renal dysfunction compared to those with the Leu/Leu genotype, holding all other variables

$$0dds_6$$

Sex=0 Type=0 Y' = -4.921 + 0.3604 TST 
$$Y' = -4.921 + 0.3604(4)$$
 
$$= -3.4794$$
 
$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{3.4794}} = 0.0299$$
 
$$Odds_4 = \frac{P(1)}{1 - P(1)} = 0.0308$$

 $Odds_7$ 

Sex=0 Type=0 Y' = -4.921 + 0.3604 TST 
$$Y' = -4.921 + 0.3604(5)$$
 
$$= -3.1190$$
 
$$P(1) = \frac{1}{1 + e^{y'}} = \frac{1}{1 + e^{3.4794}} = 0.0423$$

$$Odds_0 = \frac{P(1)}{1 - P(1)} = 0.0442$$

Odds Ratio <sub>TST</sub>:

$$Odds \ Ratio = rac{Odds_7}{Odds_6}$$
 $Odds \ Ratio \ _{type1} = 1.4351$ 

#### Interpretation of *Odds Ratio* <sub>TST</sub>:

For each additional year since transplant (TST), the odds of renal dysfunction increase by approximately 43.51%. This means that each extra year post-transplant raises the odds of developing renal dysfunction by 1.4351 times compared to the previous year.

c) Re-compute the odds ratios in sec (b) using the exponential formulas.

$$Odd\ Ratios = e^{bi}$$
 $Odd\ Ratio_{sex} = e^{0.094} = 1.0984$ 
 $Odd\ Ratio_{type1vstype0} = e^{1.890} = 6.6194$ 
 $Odd\ Ratio_{type2vstype0} = e^{1.779} = 5.9239$ 
 $Odd\ Ratio_{TST} = e^{0.360} = 1.4333$ 

d) Compare the odds in favor of the patients having three years since transplant with the odds in favor of the patients having seven years since transplant. Also, interpret it properly.

Order Ratio for more than 1 unit change is calculated as,

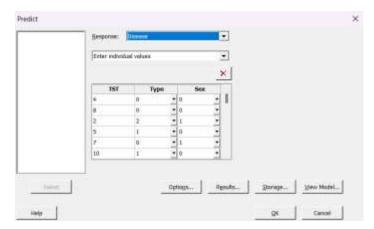
$$Odd\ Ratios = e^{cbi}$$
 $c = change \qquad bi = coefficient$ 
 $TST_0 = 3 \qquad TST_1 = 7$ 
 $c = TST_1 - TST_0$ 
 $c = 4$ 
 $Odd\ Ratio = e^{4(0.360)}$ 
 $Odd\ Ratio = 4.2207$ 

#### **Interpretation:**

This means that the odds of renal dysfunction are approximately 4.22 times higher for patients who have been transplanted for 7 years compared to those who have been transplanted for 3 years.

e) Evaluate the performance of the model in (a) from the given test data (see Excel sheet).

The given test data was fed to Minitab for predictions.



The following predictions were made by Minitab

#### **Prediction for Disease**

#### **Regression Equation**

```
\begin{array}{lll} P(1) &=& \exp(Y')/(1 + \exp(Y')) \\ Y' &=& -4.92 + 0.360 \, \text{TST} + 0.000000 \, \text{Sex\_0} + 0.094 \, \text{Sex\_1} + 0.000000 \, \text{Type\_0} + 1.890 \, \text{Type\_1} \\ &+& 1.779 \, \text{Type\_2} \end{array}
```

#### **Settings**

Variable	Setting
TST	4
Type	0
Sex	0

#### **Prediction**

Fitted			
Probability	SE Fit	95% CI	
0.0298979	0.0295554	(0.0041651,	
		0.185067)	

#### **Settings**

Variable	Setting
TST	8
Type	0
Sex	0

### Prediction

rittea			
Probability	SE Fit	95% CI	
0.115250	0.0735573	(0.0307066,	
		0.348799)	

#### **Settings**

Variable	Setting
TST	2
Type	2
Sex	1

#### **Prediction**

# Fitted SE Fit 95% CI 0.0888552 0.0841557 (0.0125547, 0.427915)

#### **Settings**

Variable	Setting
TST	5
Type	1
Sex	0

#### **Prediction**

#### **Fitted**

Probability	SE Fit	95% CI	
0.226273	0.118737	(0.0718405,	
		0.524928)	

#### **Settings**

Variable	Setting
TST	7
Type	0
Sex	1

#### **Prediction**

#### **Fitted**

<b>Probability</b>	SE Fit	95% CI	
0.0907493	0.0748122	(0.0166011,	
		0.371101)	

#### **Settings**

Variable	Setting
TST	10
Type	1
Sex	0

#### **Prediction**

#### **Fitted**

Probability	SE Fit	95% CI
0.639289	0.139132	(0.351999,
		0.852560)

#### **Settings**

Variable	Setting
TST	5
Type	0
Sex	1

#### **Prediction**

#### Fitted

By comparing the given test data and predictions done by the software we have TP=0, TN=3, FN=3, FP=1. Considering the threshold to be 0.5, predictions lesser than the threshold are considered "Absent" and values greater than the threshold are "Present"

#### Accuracy:

$$Accuracy = \frac{TP + TN}{TP + TN + FP + FN}$$

$$Accuracy = \frac{0+3}{0+3+3+1}$$

$$Accuracy = 0.4286$$

#### Precision:

$$Precision = \frac{TP}{TP + FP}$$

$$Precision = \frac{0}{0+1}$$

$$Precision = 0$$

Recall:

$$Recall = \frac{TP}{TP + FN}$$
 
$$Recall = \frac{0}{0+3}$$
 
$$Recall = 0$$

Error Rate:

$$Error = 1 - Accuracy$$
  
 $Error = 1 - 0.4286$   
 $Error = 0.5714$ 

Specificity:

$$Specificity = \frac{TN}{TN + FP}$$
$$Specificity = \frac{3}{3 + 1}$$
$$Specifity = 0.75$$

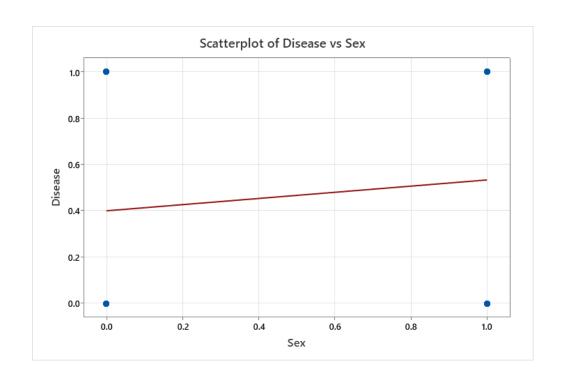
F1 - Score:

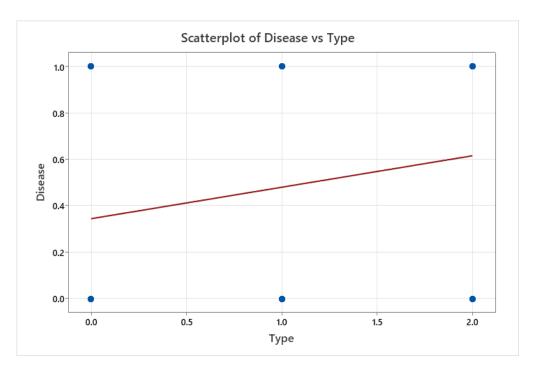
$$F1-Score = 2 \times \frac{Precision \times Recall}{Precision + Recall}$$
  
Cannot be computed

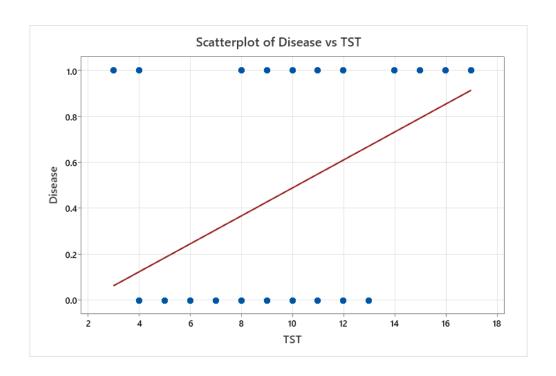
According to the model's performance measures, it can accurately identify diseases that are absent (high specificity), but it has trouble predicting diseases that are present (low sensitivity). This implies that to increase the model's ability to forecast the presence of disease, it might be necessary to change the threshold or take into account new variables.

Scatterplots to visualize regression trends for the given data model:

Scatterplot of Disease vs Sex, Type, TST







### The diagnostic plot of the model:

