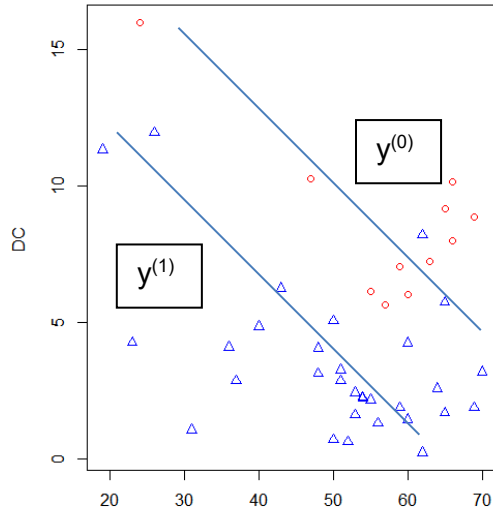


Test #3

Problem 1. For subjects undergoing stem cell transplants, dendritic cells (DCs) are critical to the generation of immunologic tumor responses. A study is conducted on 44 subjects who underwent a medical intervention and the outcome variable is the concentration of **DC cells**. One of the independent variables is the **age** of the subject (AGE), and the second independent variable is the **mobilization method** which is either Method 0 or Method 1. Data are stored in test31.csv



- Write the regression equation using the multiple linear regression with dummy variables (dummy variable x_2).
- Write the regression equations when the dummy variable takes the values $x_2 = 0$ and $x_2 = 1$.
- Sketch the regression lines
- Perform the hypothesis test to validate the regression model, that is, test the hypothesis $H_0: \beta_1 = 0$ vs $H_1: \beta_1 \neq 0$ using significance level $\alpha = 0.05$.
- Compute the 95% confidence interval of the regression coefficient β_1 .

SOLUTION

```
> Probl <- read.csv("C:/Users/test3_1.csv")
> x1 <- Probl$AGE
> x2 <- Probl$METHOD
> y <- Probl$DC
> relation <- lm(y~x1+x2, data = Probl)
> print(summary(relation))
```

Call:

```
lm(formula = y ~ x1 + x2, data = Probl)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-4.7075	-1.4134	-0.5344	1.4642	6.0338

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	15.23907	1.64913	9.241	1.41e-11	***
x1	-0.11585	0.02597	-4.461	6.23e-05	***
x2	-5.88035	0.84596	-6.951	1.93e-08	***

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

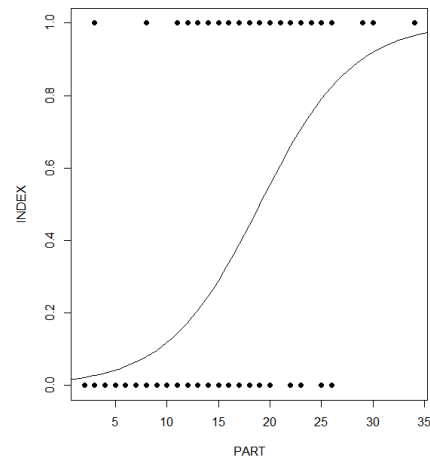
Residual standard error: 2.347 on 41 degrees of freedom
 Multiple R-squared: 0.5768, Adjusted R-squared: 0.5562
 F-statistic: 27.94 on 2 and 41 DF, p-value: 2.206e-08

```
> confint(relation, level=0.95)
                2.5 %      97.5 %
(Intercept) 11.9085761 18.5695634
x1           -0.1682921 -0.0634029
x2           -7.5887930 -4.1719009
```

SOLUTION Prob 1

- a) $y = 15.239 - 0.116 x_1 - 5.880 x_2$**
- b) @ $x_2=0$: $y^{(0)} = 15.239 - 0.116 x_1$
@ $x_2=1$: $y^{(1)} = 9.359 - 0.116 x_1$**
- c) sketch is on the page above**
- d) Test hypothesis $H_0: \beta_1 = 0$ vs $H_1: \beta_1 \neq 0$
We reject H_0 since p-value = $6.23e-05 < 0.05$**
- e) C.I. = $-0.116 \pm t(0.95;41)*0.026$; $t(0.95;41)=2.020$
C.I = **(-0.168,-0.063)****

Problem 2. A study explores the relationship between the Hospital Anxiety and Depression **index** (INDEX) and **participation** in a post-surgery rehabilitation program (PART=1, if participated, and PART=0, if not). We wish to predict the likelihood of participation to the program if we know the patient Hospital Anxiety and Depression index. Data are stored in test32.csv



- a) Compute and report the logistic regression equation.
- b) Test the null hypothesis $H_0: \beta_1 = 0$ vs $H_1: \beta_1 \neq 0$ at significance level 0.05
- c) Compute the odds ratio.
- d) Compute the 95% confidence interval for odds ratio
- e) Can you conclude the odds that a woman with a high index score will participate are higher than the odds that a woman with a low index score will participate in a rehabilitation program? Explain.

SOLUTION

```
> Prob2 <- read.csv("C:/Users/test3_2.csv")
> INDEX <- Prob2$INDEX #Y
> PART <- Prob2$PART #X
> logit_mod <- glm(PART~INDEX, family="binomial", data = Prob2)
> summary(logit_mod)
```

Call:

```
glm(formula = PART ~ INDEX, family = "binomial", data = prob2)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.8676	-0.7524	-0.4772	0.7558	2.6802

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-4.23095	0.65608	-6.449	1.13e-10	***
INDEX	0.22241	0.03806	5.844	5.11e-09	***

```
> (coefficients(logit_mod))
```

```
(Intercept)      INDEX  
0.01453861  1.24908442
```

```
> exp(confint(logit_mod, level=0.95))
```

	2.5 %	97.5 %
(Intercept)	0.003646643	0.0483399
INDEX	1.164784203	1.3532355

SOLUTION Prob2

- a) $\ln(p/(1-p)) = -4.231 + 0.222 x1$.**
- b) Since the p-value is 5.11e-09, we reject H0 and accept the alternative hypothesis that beta1 is different from 0**
- c) odds ratio = 1.249**
- d) CI = (1.165, 1.353)**
- e) Since the p-value=5.11e-09, we rejected the null hypothesis. This can also be seen by the CI of the odds ratio above not containing the value 1. Hence, we conclude that the odds that a woman with a high index score will participate are higher than the odds that a woman with a low index score will participate in a rehabilitation program.**

Problem 3. A pharmaceutical company administered three different vaccines (type A, B, C) to 6 individuals each and measured the antibody presence in their blood after a week. Results are stored in the file test33.csv

- (i) Apply an appropriate non-parametric method to test the alternative hypothesis that the antibody responses are different at level of significance 0.05.
- (ii) If the test is significant, run a post-hoc analysis and analyze the outcome.

SOLUTION

Part (i): We apply the Kruskal–Wallis test

```
> Data$Type = factor(Data$Type, levels=unique(Data$Type))  
> kruskal.test(Antibodies ~ Type, data = Data)
```

```
Kruskal-Wallis rank sum test
```

```
data: Antibodies by Type
```

```
Kruskal-Wallis chi-squared = 8.2222, df = 2, p-value = 0.01639
```

Conclusion: Since the p-value is less than 0.05, we accept the alternative hypothesis that the Antibody values are different in the 3 Types

Part (ii): Since the Kruskal–Wallis test is significant, we run the Dunn test as post-hoc test

```
> dunnTest(Antibodies ~ Type, data=Data, method="bh")  
Dunn (1964) Kruskal-Wallis multiple comparison  
p-values adjusted with the Benjamini-Hochberg method.
```

	Comparison	Z	P.unadj	P.adj
1	A - B	2.8118380	0.004925931	0.01477779
2	A - C	0.9192547	0.357962356	0.35796236
3	B - C	-1.8925832	0.058413313	0.08761997

SOLUTION Prob3

- (i) Since the p-value = 0.01639 is less than 0.05, we accept the alternative hypothesis**
- (ii) Since only the p-value of the A-B comparison is less than 0.05, we conclude that only the difference between the antibodies associated with Types A and B are significantly different.**