

# Advanced Data Analysis -HW2

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1

```
library(openxlsx)

data.1 <- read.xlsx("~/Desktop/2019 Spring/Advanced Data Analysis/hw/HW2/HW2(Diets).xlsx")
```

a)

```
model.1.1 <- lm(Response~as.factor(Diet),data=data.1)
anova(model.1.1)
```

```
## Analysis of Variance Table
##
## Response: Response
##      Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(Diet)  3 0.52652 0.175506  4.4891 0.01185 *
## Residuals      25 0.97740 0.039096
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

$H_0 : \mu_1 = \mu_2 = \mu_3 = \mu_4$  vs  $H_1$ :Not  $H_0$

From the anova table above, we have p-value = 0.012 < 0.05, so we reject  $H_0$  under significance level  $\gamma = 0.05$

In conclusion, there are at least two means that are not equal.

b)

```
t.test(Response~Diet, data=data.1[1:15,],conf.level=0.95)

##
## Welch Two Sample t-test
##
## data: Response by Diet
## t = 1.1625, df = 9.1239, p-value = 0.2745
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1325412 0.4139698
## sample estimates:
## mean in group 1 mean in group 2
##      3.745714      3.605000
```

Our test :  $H_0 : \mu_1 - \mu_2 = 0$

According to the test, we have 95% confidence interval for  $\mu_1 - \mu_2$  is (-0.13,0.41)

c) & d)

```
mu_1 <- mean(data.1$Response[data.1$Diet==1])
mu_2 <- mean(data.1$Response[data.1$Diet==2])
mu_3 <- mean(data.1$Response[data.1$Diet==3])
mu_4 <- mean(data.1$Response[data.1$Diet==4])

n1 <- sum(data.1$Diet==1)
n2 <- sum(data.1$Diet==2)
n3 <- sum(data.1$Diet==3)
n4 <- sum(data.1$Diet==4)

L <- (mu_1+mu_2-mu_3-mu_4)/2
L

## [1] -0.08505952

# From the anova table in part a), we know that MSE is 0.0391
SE_L <- sqrt(0.0391*sum(0.25/n1+0.25/n2+0.25/n3+0.25/n4))

t <- qt(1-0.05/2,df=29-4)

# hypothesis test
abs(L) > t*SE_L

## [1] FALSE

# confidence interval
L-SE_L*t

## [1] -0.2373727
L+SE_L*t

## [1] 0.06725369
```

part c):

The confidence interval for L is : (-0.237,0.067)

part d):

Our test:  $H_0 : L = \frac{\mu_1 + \mu_2 - \mu_3 - \mu_4}{2} = 0$  vs.  $H_1 : L \neq 0$

$$\hat{L} = \frac{\hat{\mu}_1 + \hat{\mu}_2 - \hat{\mu}_3 - \hat{\mu}_4}{2} = -0.08$$

Since  $|\hat{L}| < t_{\gamma/2}(df = n - k = 29 - 4)$ , we fail to reject  $H_0$

## 2

a)

```
# Define dummy variables
D <- function(x,i) {
  temp <- rep(NA,length(x))

  for(j in 1:length(x)) {
    if (x[j] == i) {temp[j] <- 1}
    else if(x[j] == 4) {temp[j] <- -1}
  }
}
```

```

    else {temp[j] <- 0}
  }

  return(temp)
}

D1 <- D(data.1$Diet,1)
D2 <- D(data.1$Diet,2)
D3 <- D(data.1$Diet,3)

model.2.1 <- lm(data.1$Response~D1+D2+D3)
summary(model.2.1)

##
## Call:
## lm(formula = data.1$Response ~ D1 + D2 + D3)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.3857 -0.0950 -0.0525  0.1250  0.4443
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.71789    0.03698  100.549  <2e-16 ***
## D1             0.02783    0.06450   0.431   0.6698
## D2            -0.11289    0.06173  -1.829   0.0794 .
## D3            -0.11955    0.06801  -1.758   0.0910 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1977 on 25 degrees of freedom
## Multiple R-squared:  0.3501, Adjusted R-squared:  0.2721
## F-statistic: 4.489 on 3 and 25 DF,  p-value: 0.01185

```

From the definition and the table above, we have:

$$\hat{\mu} = 3.72$$

$$\hat{\alpha}_1 = 0.028$$

$$\hat{\alpha}_2 = -0.113$$

$$\hat{\alpha}_3 = -0.120$$

$$\hat{\alpha}_4 = -(\hat{\alpha}_1 + \hat{\alpha}_2 + \hat{\alpha}_3) = -(0.028 - 0.113 - 0.120) = 0.205$$

b)

```

model.2.2 <- lm(data.1$Response~1)
anova(model.2.2,model.2.1)

## Analysis of Variance Table
##
## Model 1: data.1$Response ~ 1
## Model 2: data.1$Response ~ D1 + D2 + D3
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)

```

```
## 1      28 1.5039
## 2      25 0.9774 3    0.52652 4.4891 0.01185 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Our test:  $H_0 : \alpha_1 = \alpha_2 = \alpha_3 = 0$  vs.  $H_\alpha$ : Not  $H_0$

From the anova table, we have  $\text{p-value} = 0.012 < 0.05$ , so reject  $H_0$

The  $\text{p-value}$  is exactly the same as the value in Problem 1 and we reached the same conclusion.

### 3

a)

Source	df	SS	MS	F
feed	2	23.43	11.715	17.564
error	7	4.67	0.667	
total	9	28.10		

b)

```
p.value <- 1-pf(17.69,df1=2,df2=7)
p.value
```

```
## [1] 0.001831386
```

Our test:  $H_0 : \mu_1 = \mu_2 = \mu_3$  vs.  $H_\alpha$ : Not  $H_0$

From the calculation above, we know that  $\text{p-value} = 0.0018 < 0.05$ , so reject  $H_0$

Therefore, we know that different chicken feeds have different effect.