
Assignment On Lifetime Data Analysis AST 405

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Contents

Answer To Ques 01	2
Answer To Ques 02	3
Answer To Ques 03	6
Answer To Ques 04	7
Answer To Ques 05	7
Answer To Ques 06	8
Answer To Ques 07	9
Answer To Ques 08	10
Answer To Ques 09	11
R-code	12

Answer To Ques 01

For the continuous variables `age` and `bmi`, mean and standard deviation (SD) is obtained. And for categorical variable `arcus`, `behpat` and `chd69`, frequency and proportion (in percentage) is obtained as descriptive statistics, which are shown in the following table:

Table 1: Descriptive Statistics

	level	Overall
n		3154
age (mean (SD))		46.35 (5.56)
bmi (mean (SD))		24.48 (2.55)
arcus (%)	0	2219 (70.4)
	1	934 (29.6)
behpat (%)	A1	275 (8.7)
	A2	1290 (40.9)
	B3	1236 (39.2)
	B4	353 (11.2)
chd69 (%)	No	2900 (91.9)
	Yes	254 (8.1)

Answer To Ques 02

Effect of age and bmi : To check whether each of **age** and **bmi** has significant effect on **chol**, we need to do correlation test (pearson).

For each correlation test of **age** and **bmi** with **chol**, the hypotheses are,

$$H_o : \rho = 0$$

$$H_a : \rho \neq 0$$

where ρ is the population correlation coefficient.

The appropriate test statistic for testing the hypothesis is,

$$t_o = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$$

where r is the sample correlation coefficient and t_o follows t_{n-2} distribution under H_o .

Here we will reject H_o if the associated **p.value** is less than $\alpha = 0.05$, where α is the level of significance.

Now from Table: 2, we can see that both of the variables **age** and **bmi** is significantly associated with **chol** variable.

Table 2: Correlation Test (age, bmi) and T-test (arcus) with chol

term	estimate	statistic	p.value
age	0.081	4.550	<.001
bmi	0.060	3.385	<.001
arcus	-12.672	-7.604	<.001

Effect of arcus Now to check the effect of **arcus** (a categorical variable with level 0 and 1) on **chol**, we can do two sample t-test, where one sample is people with **arcus** 0 and other sample is people with **arcus** 1. Therefore, μ_o and μ_1 are the population mean of **arcus** 0 and **arcus** 1 group respectively.

Then our hypotheses are:

$$H_o : \mu_o = \mu_1$$

$$H_a : \mu_o \neq \mu_1$$

and appropriate test statistic (assuming unequal variance),

$$t_o = \frac{\bar{x}_o - \bar{x}_1}{\sqrt{\frac{s_o^2}{n_o} + \frac{s_1^2}{n_1}}}$$

where \bar{x} and s^2 denotes sample mean and sample variance with subscript 0 and 1 for **arcus** group 0 and 1 respectively and n_o and n_1 are the corresponding sample sizes.

Here we will reject H_o if the associated **p value** is less then $\alpha = 0.05$, where α is the level of significance.

Then from Table: 2, since **p.value** is less than 0.001 we can conclude that mean **chol** differs significantly for 0 and 1 group of **arcus**, that is, **arcus** has significant effect on **chol**.

Effect of behpat: To check the effect of **behpat** on **chol**, we can do oneway ANOVA. In this case the hypotheses are:

$$H_o : \mu_1 = \mu_2$$

$$H_a : \mu_i \neq \mu_j \quad \text{for at least on } i \neq j$$

and test statistic is $F_o = \frac{MS_{reg}}{MS_E}$. We will reject H_o if associated **p.value** is less than 0.05.

Table 3: One Way Analysis of Variance for chol on behpat

term	df	sumsq	meansq	statistic	p.value
behpat	3	30741.67	10247.223	5.475	<.001
Residuals	3135	5867632.50	1871.653	NA	NA

Then from Table: 3, we conclude that mean **chol** differs significantly over the levels of **behpat**.

Answer To Ques 03

To examine the association between **behat** and **chd69**, the hypotheses are:

H_o : There's no association between **behat** and **chd69**.

H_a : There's association between **behat** and **chd69**.

Here, the test statistics is:

$$\chi^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(O_{ij} - E_{ij})^2}{E_{ij}}$$

Where, O_{ij} is observed frequency and E_{ij} is the expected frequency and r, c is the row and column number of contingency table 4

Here, we would reject H_o if associated **p.value** corresponding to test statistics is less the 0.05.

	CHD	
	No	Yes
behat		
A1	246	29
A2	1158	132
B3	1164	72
B4	332	21

Table 4: Contingency table

statistic	p.value
20.978	<.001

Table 5: chi-square test

Now from the Table 5, since **p.value** is less than 0.001, we can conclude that there's an significant association between **behat** and **chd69**.

Answer To Ques 04

Table 6: Estimate of Model Parameters

term	estimate	std.error	statistic	p.value
(Intercept)	196.971	6.483	30.381	<.001
age	0.632	0.139	4.550	<.001

The fitted regression line is:

$$\widehat{\text{chol}} = 196.97 + 0.63(\text{age}) \quad (1)$$

Interpretation of regression coefficients:

The cholesterol level is expected to increase by 0.63 units for 1-year increase of age.

Answer To Ques 05

Table 7: Estimate of Model Parameters

term	estimate	std.error	statistic	p.value
(Intercept)	228.352	1.097	208.102	<.001
dibpatType B	-4.147	1.546	-2.682	0.007

The fitted regression line is:

$$\widehat{\text{chol}} = 228.35 - 4.15(\text{dibpat}_{\text{Type B}}) \quad (2)$$

Interpretation of regression coefficients:

- Mean cholesterol level of Type A dibpat subjects is 228.35.
- Also, mean cholesterol level of Type A dibpat subjects is 4.15 unit higher than Type B dibpat subjects.

Answer To Ques 06

Table 8: Estimate of Model Parameters

term	estimate	std.error	statistic	p.value
(Intercept)	235.073	2.618	89.778	<.001
behpatA2	-8.150	2.883	-2.827	0.005
behpatB3	-10.226	2.894	-3.533	<.001
behpatB4	-13.122	3.491	-3.759	<.001

The fitted regression line is:

$$\widehat{\text{chol}} = 235.07 - 8.15(\text{behpat}_{\text{A2}}) - 10.23(\text{behpat}_{\text{B3}}) - 13.12(\text{behpat}_{\text{B4}}) \quad (3)$$

Interpretation of regression coefficients:

Since **Type A1** Behavior pattern subjects are reference group,

- Mean cholesterol level of **Type A1** subjects is 235.07 unit.
- Mean cholesterol level of **Type A2** subjects is -8.15 unit lower than that of **Type A1** subjects.
- Mean cholesterol level of **Type B3** subjects is -10.23 unit lower than that of **Type A1** subjects.
- Mean cholesterol level of **Type B4** subjects is -13.12 unit lower than that of **Type A1** subjects.

Answer To Ques 07

Table 9: Estimate of Model Parameters

term	estimate	std.error	statistic	p.value
(Intercept)	200.191	6.637	30.163	<.001
age	0.600	0.140	4.302	<.001
dibpatType B	-3.469	1.550	-2.238	0.025

The fitted regression line is:

$$\widehat{\text{chol}} = 200.19 + 0.6(\text{age}) - 3.47(\text{dibpat}_{\text{Type B}}) \quad (4)$$

Interpretation of regression coefficients:

Since **Type A1** subjects are reference group,

- Mean cholesterol level of subjects with age 0 and **Type A** pattern is 200.19 unit.
- Mean cholesterol level increases about 0.6 for 1-year increase of age, holding dibpat fixed.
- Mean cholesterol level of **Type B2** subjects is -3.47 unit lower than that of **Type A** dibpat subjects, holding the subject's age fixed.

Here, Both the regression coefficients corresponding to age and dibpat (Eq 4) have changed from the case of simple linear regression in Eq 1 and Eq 2. And also the value of R^2_{adj} increased.

Answer To Ques 08

Table 10: Estimate of Model Parameters

term	estimate	std.error	statistic	p.value
(Intercept)	222.252	1.171	189.755	<.001
age40	0.632	0.139	4.550	<.001

The new fitted model after subtracting 40 from the variable `age` is:

$$\widehat{\text{chol}} = 222.25 + 0.63(\text{age40}) \quad (5)$$

Interpretation of regression coefficients:

- Since the explanatory variable is age minus 40, we can say, the mean cholesterol level of a 40 year old subject is 222.25.
- Mean cholesterol level is expected to increase by 0.63 for 1 year increase of age.

The main difference of this model (Eq 1) compared to model 5 is that, we can interpret the intercept term for this model logically.

Answer To Ques 09

Table 11: Estimate of Model Parameters

term	estimate	std.error	statistic	p.value
(Intercept)	223.702	1.597	140.043	<.001
dibpatType B	-3.644	2.175	-1.675	0.094
smokeYes	8.728	2.189	3.988	<.001
dibpatType B:smokeYes	0.265	3.085	0.086	0.932

In this case, the fitted model is:

$$\widehat{\text{chol}} = 223.7 - 3.64(\text{dibpat}_{\text{Type B}}) + 8.73(\text{smoke}_{\text{Yes}}) + 0.27(\text{dibpat}_{\text{Type B}} \times \text{smoke}_{\text{Yes}}) \quad (6)$$

Interpretation of regression coefficients:

Here the reference groups are **Non-smoker** and **dibpat Type A** subjects. So,

- The mean cholesterol level of Type A non smoker subject is 223.7 unit.
- Among the non-smokers, mean cholesterol level of dibpat Type B subjects is 3.64 unit lower compared to Type A subject.
- Among the Type A subjects, mean cholesterol level of smokers is 8.73 unit higher than that of non-smokers.
- Difference of mean cholesterol level between smokers and non-smokers is 0.27 unit higher in Type B dibpat subjects compared to that of Type A dibpat subjects.

R-code

```
knitr::opts_chunk$set(  
  echo = FALSE,  
  message = FALSE,  
  warning = FALSE  
)  
  
## ----- package setup -----  
  
library(dplyr)  
library(purrr)  
library(knitr)  
library(broom)  
library(tableone)  
library(kableExtra)  
library(equationomatic)  
  
## ----- data setup -----  
  
load(here::here("data", "wcgs.Rdata"))  
sid <- 011  
set.seed(sid)  
mydat <- sample_n(wcgs, size = n(), replace = TRUE)  
  
## ----- utility functions -----  
  
kab_tab <- function(tab, ...) {  
  knitr::kable(tab,  
    format = "latex",  
    booktabs = TRUE,  
    digits = 3,  
    ...)  
}
```

```

p_format <- function(pval) {
  ifelse(pval < .001, "<.001", as.character(round(pval, 3)))
}

mod_tab <- function(mod, ...) {
  mod %>%
    tidy() %>%
    mutate(p.value = p_format(p.value)) %>%
    kable_tab(aligned = "lrrrr",
              caption = "Estimate of Model Parameters", ...) %>%
    kable_styling(latex_options = "HOLD_position")
}

reg_eq <- function(mod, ref, ...) {
  extract_eq(mod,
             use_coefs = TRUE,
             intercept = "beta",
             wrap = TRUE,
             label = paste0("eq", ref),
             ...)
}

params <- function(mod, param, dec = 2) {
  round(mod$coefficients[[param]], dec)
}

```

```
## ----- Code for Question-01 -----
```

```
tab <- CreateTableOne(  
  data = mydat,  
  vars = c("age", "bmi", "arcus", "behpat", "chd69"),  
  factorVars = "arcus",  
  addOverall = TRUE  
)  
  
tab_p <- print(tab, showAllLevels = TRUE, printToggle = FALSE)  
  
kab_tab(tab_p,  
  caption = "Descriptive Statistics") %>%  
  kable_styling(latex_options = "hold_position")
```

```
## ----- Code for Question-02 -----
```

```
arcus <- t.test(chol ~ arcus, data = mydat) %>%  
  tidy() %>%  
  mutate(term = "arcus") %>%  
  select(term, estimate, statistic, p.value)  
  
mydat %>%  
  select(age, bmi) %>%  
  map(~ cor.test(x = .x, y = mydat$chol)) %>%  
  map_dfr(broom::tidy, .id = "term") %>%  
  select(term:p.value) %>%  
  bind_rows(arcus) %>%  
  mutate(p.value = p_format(p.value)) %>%  
  kab_tab(align = "lrrr",  
    caption = "Correlation Test (age, bmi) and T-test (arcus) with chol") %>%  
  kable_styling(latex_options = "hold_position")
```

```

anova(lm(chol ~ behpat, data = mydat)) %>%
  tidy() %>%
  mutate(p.value = p_format(p.value)) %>%
  kab_tab(align = "lrrrrr",
          caption = "One Way Analysis of Variance for chol on behpat") %>%
  kable_styling(latex_options = "HOLD_position")

```

----- Code for Question-03 -----

```
df_cont <- mydat %>% janitor::tabyl(behpat, chd69)
```

```

tab_cont <- df_cont %>%
  kab_tab() %>%
  add_header_above(header = c(" " = 1, "CHD" = 2))

```

```

tab_chi <- df_cont %>%
  janitor::chisq.test() %>%
  tidy() %>%
  select(statistic, p.value) %>%
  mutate(p.value = p_format(p.value)) %>%
  kab_tab()

```

```

tab_side <- c(
  "\\begin{table}[H]
    \\begin{minipage}{.5\\linewidth}
      \\centering",
  tab_cont,
  "\\caption{Contingency table}
  \\label{Table-04}
\\end{minipage}%
  \\begin{minipage}{.5\\linewidth}
    \\centering",
  tab_chi,
  "\\caption{chi-square test}
  \\label{Table-05}

```



```

\\end{minipage}
\\end{table}"
)

## ----- Code for Question-04 -----

m1 <- lm(chol ~ age, data = mydat)
m1 %>% mod_tab()

reg_eq(m1, 1)

## ----- Code for Question-05 -----

m2 <- lm(chol ~ dibpat, data = mydat)
m2 %>% mod_tab()

reg_eq(m2, 2)

## ----- Code for Question-05 -----

m3 <- lm(chol ~ behpat, data = mydat)
m3 %>% mod_tab()

reg_eq(m3, 3)

## ----- Code for Question-08 -----

m4 <- lm(chol ~ age + dibpat, data = mydat)
m4 %>% mod_tab()

reg_eq(m4, 4)

## ----- Code for Question-08 -----

mydat %>%

```

```
mutate(age40 = age - 40) %>%
  lm(chol ~ age40, data = .) -> m5

m5 %>% mod_tab()

reg_eq(m5, 5)
## ----- Code for Question-08 -----

m6 <- lm(chol ~ dibpat * smoke, data = mydat)
m6 %>% mod_tab()
reg_eq(m6, 6)
```