

STAT 224 Autumn 2022 Homework 2

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Question 1

```
fevdata = read.table("http://www.stat.uchicago.edu/~yibi/s224/data/fevdata.txt", h = TRUE)
fevdata$sex = factor(fevdata$sex, labels=c("Female","Male"))
fevdata$smoke = factor(fevdata$smoke, labels=c("Nonsmoker","Smoker"))
```

Q1a — 6 points

```
lmm.nosmoke = lm(fev ~ age, data=subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
summary(lmm.nosmoke)
```

```
##
## Call:
## lm(formula = fev ~ age, data = subset(fevdata, sex == "Male" &
##     smoke == "Nonsmoker"))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.4850 -0.3506 -0.0438  0.3511  1.8230
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.0576     0.1147   -0.5     0.62
## age           0.2882     0.0114   25.3    <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.556 on 308 degrees of freedom
## Multiple R-squared:  0.676, Adjusted R-squared:  0.674
## F-statistic: 641 on 1 and 308 DF, p-value: <2e-16
```

$$\hat{\beta}_0^{mn} = -0.0576, s.e.(\hat{\beta}_0^{mn}) = 0.1147$$

$$\hat{\beta}_1^{mn} = 0.2882, s.e.(\hat{\beta}_1^{mn}) = 0.0114$$

$$\hat{\sigma}^{mn} = 0.556, n = 310$$

Q1b — 8 points

$$t = \frac{\hat{\beta}_j - \beta_j^0}{s.e.(\hat{\beta}_j)}$$

i)

```
t = (-0.0576 - 0)/0.1147
n=nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
p=1
pval = pt(t,df=n-p-1,lower.tail = F)
print(paste0("t-stat: ",signif(t,digits = 5)))
```

```
## [1] "t-stat: -0.50218"
```

```
print(paste0("df: ",n-p-1))
```

```
## [1] "df: 308"
```

```
print(paste0("P-value: ",signif(pval,digits = 5)))
```

```
## [1] "P-value: 0.69205"
```

Fail to reject null hypothesis

ii)

```
t = (0.2882 - 0.1)/0.0114
n=nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
p=1
pval = 2*pt(abs(t),df=n-p-1,lower.tail = F)
print(paste0("t-stat: ",signif(t,digits = 5)))
```

```
## [1] "t-stat: 16.509"
```

```
print(paste0("df: ",n-p-1))
```

```
## [1] "df: 308"
```

```
print(paste0("P-value: ",signif(pval,digits = 5)))
```

```
## [1] "P-value: 2.6709e-44"
```

Reject null

iii)

```
t = (0.2882 - 0.1)/0.0114
n=nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
p=1
pval = pt(t,df=n-p-1)
print(paste0("t-stat: ",signif(t,digits = 5)))
```

```
## [1] "t-stat: 16.509"
```

```
print(paste0("df: ",n-p-1))
```

```
## [1] "df: 308"
```

```
print(paste0("P-value: ",signif(pval,digits = 5)))
```

```
## [1] "P-value: 1"
```

Fail to reject null

iv)

```
t = (0.2882 - 0.3)/0.0114
n=nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
p=1
pval = pt(t,df=n-p-1,lower.tail = F)
print(paste0("t-stat: ",signif(t,digits = 5)))
```

```
## [1] "t-stat: -1.0351"
```

```
print(paste0("df: ",n-p-1))
```

```
## [1] "df: 308"
```

```
print(paste0("P-value: ",signif(pval,digits = 5)))
```

```
## [1] "P-value: 0.84928"
```

Fail to reject null

Q1c — 4 points

$$\hat{\beta}_j \pm t_{(n-p-1, \frac{\alpha}{2})} * s.e.(\hat{\beta}_j)$$

```
beta_hat = 0.2882
se = 0.0114
n=nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
p=1
alpha = 0.1
t = qt(alpha/2,df=n-p-1,lower.tail = F)
print(paste0("confidence interval: ", '(',
             signif(beta_hat-t*se,digits=5), ',',
             signif(beta_hat+t*se,digits=5), ')'))
```

```
## [1] "confidence interval: (0.26939,0.30701)"
```

We are 90% confident that β_1^{mn} is between 0.26939 and 0.30701. In words, we are 90% confident that the lung capacity in liters of male nonsmokers increases by a value between 0.26939 and 0.30701 per year.

Q1d — 5 points

```
aggregate(age ~ sex + smoke, data=fevdata, mean)
```

```
##      sex      smoke      age
## 1 Female Nonsmoker  9.366
## 2  Male Nonsmoker  9.687
## 3 Female   Smoker 13.256
## 4  Male   Smoker 13.923
```

```
aggregate(age ~ sex + smoke, data=fevdata, sd)
```

```
##      sex      smoke      age
## 1 Female Nonsmoker  2.693
## 2  Male Nonsmoker  2.778
## 3 Female   Smoker  2.245
## 4  Male   Smoker  2.465
```

$$\hat{\beta}_0 + \hat{\beta}_1 x_0 \pm t_{(n-2, \frac{\alpha}{2})} \hat{\sigma} \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{\sum (x_i - \bar{x})^2}}$$

```
est = lmm.nosmoke$coefficients[1] + lmm.nosmoke$coefficients[2]*18
est
```

```
## (Intercept)
##          5.13
```

Estimate: $\hat{\beta}_0 + \hat{\beta}_1 x_0 = 5.13$

```
n = nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
t = qt(0.05/2, n-2, lower.tail = F)
sig_hat = 0.556
x_0 = 18
x_bar = 9.687
sd = 2.778
num = (x_0 - x_bar)^2
denom = (n-1)*sd^2
int = t*sig_hat*sqrt((1/n) + (num/denom))
int
```

```
## [1] 0.1963
```

```
print(paste0('CI: (', signif(est-int, digits=5), ', ', signif(est+int, digits=5), ')'))
```

```
## [1] "CI: (4.9339, 5.3266)"
```

Q1e — 4 points

$$\hat{\beta}_0 + \hat{\beta}_1 x_0 \pm t_{(n-2, \frac{\alpha}{2})} \hat{\sigma} \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{\sum (x_i - \bar{x})^2}}$$

```
est = lmm.nosmoke$coefficients[1] + lmm.nosmoke$coefficients[2]*14
est
```

```
## (Intercept)
##          3.977
```

Estimate: $\hat{\beta}_0 + \hat{\beta}_1 x_0 = 3.977$

```

n = nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
t = qt(0.05/2,n-2,lower.tail = F)
sig_hat = 0.556
x_0 = 14
x_bar = 9.687
sd = 2.778
num = (x_0-x_bar)^2
denom = (n-1)*sd^2
int = t*sig_hat*sqrt((1/n)+(num/denom))
int

```

```
## [1] 0.1149
```

```
print(paste0('CI: (',signif(est-int,digits=5),',',signif(est+int,digits=5),')'))
```

```
## [1] "CI: (3.8625,4.0923)"
```

The interval for Q1d is wider.

Q1f — 5 points

$$\hat{\beta}_0 + \hat{\beta}_1 x_0 \pm t_{(n-2, \frac{\alpha}{2})} \hat{\sigma} \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{\sum (x_i - \bar{x})^2}}$$

```

est = lmm.nosmoke$coefficients[1] + lmm.nosmoke$coefficients[2]*14
est

```

```
## (Intercept)
##          3.977
```

Estimate: $\hat{\beta}_0 + \hat{\beta}_1 x_0 = 3.977$

```

n = nrow(subset(fevdata, sex == "Male" & smoke == "Nonsmoker"))
t = qt(0.05/2,n-2,lower.tail = F)
sig_hat = 0.556
x_0 = 14
x_bar = 9.687
sd = 2.778
num = (x_0-x_bar)^2
denom = (n-1)*sd^2
int = t*sig_hat*sqrt(1+(1/n)+(num/denom))
int

```

```
## [1] 1.1
```

```
print(paste0('CI: (', signif(est-int, digits=5), ', ', signif(est+int, digits=5), ')'))
```

```
## [1] "CI: (2.8774,5.0775)"
```

The interval for this question is much larger since we are making a point prediction for a specific individual rather than for the average.

Q1g — 6 points

```
lmm.nosmoke.female = lm(fev ~ age, data=subset(fevdata, sex == "Female" & smoke == "Nonsmoker"))
summary(lmm.nosmoke.female)
```

```
##
## Call:
## lm(formula = fev ~ age, data = subset(fevdata, sex == "Female" &
##     smoke == "Nonsmoker"))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.0984 -0.2826 -0.0135  0.2374  1.0972
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.67387     0.08913   7.56  5.9e-13 ***
## age          0.18209     0.00915  19.91 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.411 on 277 degrees of freedom
## Multiple R-squared:  0.589, Adjusted R-squared:  0.587
## F-statistic: 396 on 1 and 277 DF, p-value: <2e-16
```

$$\hat{\beta}_0^{fn} = 0.67387, \hat{\beta}_1^{fn} = 0.18209$$

$$\hat{\beta}_j \pm t_{(n-p-1, \frac{\alpha}{2})} * s.e.(\hat{\beta}_j)$$

```
beta_hat = 0.18209
se = 0.00915
n=nrow(subset(fevdata, sex == "Female" & smoke == "Nonsmoker"))
p=1
alpha = 0.1
t = qt(alpha/2, df=n-p-1, lower.tail = F)
print(paste0("confidence interval: ", '(',
             signif(beta_hat-t*se, digits=5), ', ',
             signif(beta_hat+t*se, digits=5), ')'))
```

```
## [1] "confidence interval: (0.16699,0.19719)"
```

The confidence interval for boys is (0.26939, 0.30701) while it is (0.16699,0.19719) for girls. Since there is no overlap, the lung capacity for boys most likely grows faster than for girls.

Question 2

```
NLSY = read.table("http://www.stat.uchicago.edu/~yibi/s224/data/NLSY.txt", header=T)
NLSYm = subset(NLSY, Gender == "male")
```

Q2a — 6 points

```
lm1 = lm(log(Income2005) ~ AFQT, data=NLSYm)
lm2 = lm(log(Income2005) ~ AFQT + Edu2006, data=NLSYm)
```

```
print(lm1$coefficients[1])
```

```
## (Intercept)
##           3.241
```

```
lm2$coefficients[1]
```

```
## (Intercept)
##           2.476
```

For the first model the coefficient of **AFQT** is 3.241 while for the second model it is 2.476.

The coefficients are different since in the second model we also have education as a covariate. This means that the interpretation of the regression coefficient for **AFQT** is different, since in the second model it is interpreted as the change in log income from an increase of 1 percentile on the AFQT for a given level of education. This is different than for the first model which is just the change in log income from an increase of 1 percentile on the AFQT.

Q2b — 5 points

```
yres = lm(log(Income2005) ~ Edu2006, data=NLSYm)$res
tres = lm(AFQT ~ Edu2006, data=NLSYm)$res
lm(yres ~ tres)$coef
```

```
## (Intercept)      tres
## -3.023e-17    6.738e-03
```


Q2c — 2 points

```
sst = sum((log(NLSYm$Income2005) - mean(log(NLSYm$Income2005)))^2)
ssr = sum((lm1$fitted.values - mean(log(NLSYm$Income2005)))^2)
ssr/sst
```

```
## [1] 0.1221
```

```
sst = sum((log(NLSYm$Income2005) - mean(log(NLSYm$Income2005)))^2)
ssr = sum((lm2$fitted.values - mean(log(NLSYm$Income2005)))^2)
ssr/sst
```

```
## [1] 0.1491
```

The multiple R^2 values obtained mean that model 1 can explain 12.21% of the variation in Y and that model 2 can explain 14.91% of the variation in Y.