3.1 There were 64 countries in 1992 that competed in the Olympics and won at least one medal. Let MEDALS be the total number of medals won, and let GDPB be GDP (billions of 1995 dollars). A linear regression model explaining the number of medals won is $MEDALS = \beta_1 + \beta_2 GDPB + \epsilon$. The estimated relationship is

 $\widehat{\mathit{MEDALS}} = b_1 + b_2 \mathit{GDPB} = 7.61733 + 0.01309 \mathit{GDPB}$

Ho: B2 =0 H,: B2 >0

b. If the null hypothesis Ho: Bz=c is true, it has a t-distribution with N-2 degrees of freedom and $t = \frac{b_k - c}{se(b_k)} \sim t_{(N-2)}$ By (a), we know that k=1, c=0And N=64, se(b2)=0.00215

Therefore, the test statistic in part (a):

t = \\ \frac{b_2 - 0}{0.00215} = \frac{0.01309}{0.00215} \quad \text{and its distribution to t(62)}

C. Is the alternative hypothesis is true, we reject the null hypothesis, then the t-statistic $t = \frac{b_2 - C}{se(b_0)}$ does not have a t-distribution with N-2 degrees of

> And E[b2]>0, the test-statistic t>0 Hence, the distribution shifted to the right.

d=0.01, the critical value for the right tail rejection region is the 99th percentile of the t-distribution with 69-2=62 degrees of Sreedom.

t (299,62)= 2-388

If the test statistic t< 2.388, we reject the alternative hypothesis, otherwise, we reject the null hypothesis.

e. since t=6.088 > t(0.99,62) ⇒ we accept Hi

> the medals and GDPB has positive relationship. 0=0.01, P(making Type | error)=0.01,

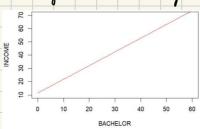
only 1% to reject Ho is it was true.

percentage of the population with a bachelor's degree or more for the 50 U.S. States plus the Distri of Columbia, a total of N = 51 observations. The results from a simple linear regression of INCOM on BACHELOR are

 $\widehat{INCOME} = (a) + 1.029BACHELOR$ se (2.672) (c)
t (4.31) (10.75)

Since, $t = \frac{b_1}{se(b_1)} = 4.31$ and $3b_1 = 2.672$ Thus, 61=431x2612=11.51632

- (1) since the slope b2 = 1.029 >0, it is increasing.
- 2 According to the above answer, when the bachelor increase, the income also increase => positive relationship
- 3) It is increasing at an constant rate Since it is an linear model, which has neither increasing rate nor decreasing rate



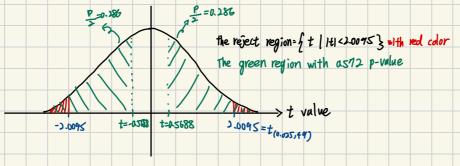
C. Since $t = \frac{b_2}{5eb_2}$ and $b_2 = 1.029$, t = 10.15Thus, se(b2)= 1029 = 0.09512

Ho: B1=10,

$$\frac{1}{7} t = \frac{b_1 - 10}{50(b_1)} = \frac{11.5163240}{2.672} = \frac{1.52}{2.672} = 0.5688$$

t(0.025, 49) = -2.0095

if the fest-statistic t with 2008st = 2009s: We do not reject Ho otherwise, t <- 20095 or t > 20095: we reject Ho.



Since as12 >> 0.05, we fail to reject H.

5. x=299, then t(298,49)=3.619 the interval=[b2-t(aggs.f9) seb2,

b2+ t(0,995,99) sebs]

=[1.029-2.679 -0.09572, 1.029+2679x0.0952] =[01125, 12854]

we have 99% consident that the real slope is contained in this interval.

Ho: Bz = 1, t&=10.75, Hi: B=1 Since $\alpha = 0.05$,

t(0.995,49)= 2.009575 And the test statistic t= 1029-1 =0303

Since 0.303 < 2.009575, which means the test statistic is not in the critical region.

Thus, we fail to reject Ho. No evidence to say that the

slope β_2 is not one.

ssion model $WAGE = \beta_1 + \beta_2 EDUC + e$. Where WAGE is hourly wage rate in US 2013 dollars. *EDUC* is years of schooling. The model is esti an urban area, and again for individuals in a rural area.

 $\widehat{WAGE} = -10.76 + 2.46EDUC, N = 986$ (se) (2.27) (0.16)

 $\widehat{WAGE} = -4.88 + 1.80EDUC$, N = 214 (se) (3.29) (0.24) native that it is greater than 1.80. Use the $\alpha = 0.05$ level of significance. Show all step

is the test-statistic t > t(0.95,984), the test-statistic $t = \frac{2.46 - 1.8}{0.16} = 4.125$ then we reject to H.: B271.8 otherwise, if $t \leq t_{(0.95,984)}$, then Since N=005 and N=986 => t(a95,984)=1.6464

since t>t(0.95,984) > we reject Ho, and has

evidence that \$2 is greater than 18

71=4125 t(0.75,984)-1.6464

we fail to reject Ho

Since $t = \frac{\text{wage} - (\beta_1 + \beta_2 \lambda)}{\text{se}(\text{wage})} vt_{(212)}$

and 0.95= P(tao25,22) < + < t(a995,20)

> 0.95=P(wage - t(0.025,212) x se wage < 13,782x

< wage + t (0975,712 * 50 (wage))

wage = -4.88+ 1.8x16= 23.92 Since se(wage)=0.833 and too.975.212)=1.9712 Therefore, the 95% interval is (wage-se(wage) *taus, 212),

wage + 50 (wage) * t (205,212) = [2392-0.833x1.9912,23,92+0.833x1.9912]

= [2,28,25,57]

wage = -1076+2.46 × 16 = 28.6 $se(wage)=(2.17)^{2}+256x(016)^{2}+32x(-0.345)$

Thus, se(wage)=0.81639 and t (0.98,984)=1.962378 Therefore, the 95% interval is (wage - se (wage) x t (0.05,789).

wage + Se(wage) * t(0.05, 984)) = [28.6-0.81639x1.962378, 28.6+0.81639x1.962378]

=[26,99193,30,20207]

the interval is narrower (since the standard error of whan regression is smaller)

Ho: β1≥4 and == 0.01 H,: B,<4

the test statistic $t = \frac{-4.88.4}{3.19} = -2.699$

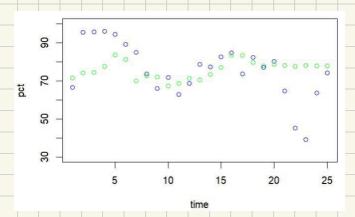
t (a01,212) =-2344

Since ==-2.699 1-2.344, we accept H1,

and have evidence to say that B, < 4.

- 3.19 The owners of a motel discovered that a defective product was used during construction. It took 7 months to correct the defects during which approximately 14 rooms in the 100-unit motel were taken out of service for 1 month at a time. The data are in the file *motel*.
 - a. Plot $MOTEL_PCT$ and $COMP_PCT$ versus TIME on the same graph. What can you say about the occupancy rates over time? Do they tend to move together? Which seems to have the higher occupancy rates? Estimate the regression model $MOTEL_PCT = \beta_1 + \beta_2 COMP_PCT + e$. Construct a 95% interval estimate for the parameter β_2 . Have we estimated the association between $MOTEL_PCT$ and $COMP_PCT$ relatively precisely, or not? Explain your reasoning.
 - **b.** Construct a 90% interval estimate of the expected occupancy rate of the motel in question, $MOTEL_PCT$, given that $COMP_PCT = 70$.
 - c. In the linear regression model $MOTEL_PCT = \beta_1 + \beta_2 COMP_PCT + e$, test the null hypothesis $H_0: \beta_2 \le 0$ against the alternative hypothesis $H_0: \beta_2 > 0$ at the $\alpha = 0.01$ level of significance. Discuss your conclusion. Clearly define the test statistic used and the rejection region.
 - **d.** In the linear regression model $MOTEL_PCT = \beta_1 + \beta_2 COMP_PCT + e$, test the null hypothesis $H_0: \beta_2 = 1$ against the alternative hypothesis $H_0: \beta_2 \neq 1$ at the $\alpha = 0.01$ level of significance. If the null hypothesis were true, what would that imply about the motel's occupancy rate versus their competitor's occupancy rate? Discuss your conclusion. Clearly define the test statistic used and the rejection region.
 - e. Calculate the least squares residuals from the regression of MOTEL_PCT on COMP_PCT and plot them against TIME. Are there any unusual features to the plot? What is the predominant sign of the residuals during time periods 17–23 (July, 2004 to January, 2005)?

a.



blue: motel-pct

Both of them tend to move together

Motel _ pct seems to have higher occupancy rate

> lowerbound [1] 0.4452978 > upperbound [1] 1.283981

According to the picture, the 95% considert interval is [-0.199,184] It is relatively precisely, since B2 is approximately 1.

b

> lowerbound2 [1] 76.97651 > upperbound2 [1] 86.87297

According to the picture, the 95% considert interval = [76.98, 86.88]

> test_t [1] 4.26536

[1] 2.499867

The test-statistic $T = \frac{b_2}{50(b_3)} \sim t_{A0}(23)$ $\Rightarrow t = 4.26$ and $t_{A01}(33) = 3.499$ Since $t = 4.26 > 2.499 = t_{A01}(23)$ \Rightarrow We reject Ho, there is evidence that $\beta_2 > 0$

d

> test_t2 [1] -0.6677491 > tc4 [1] 2.807336

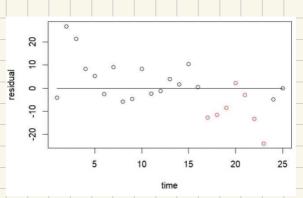
The test-statistic $T = \frac{b_2 - 1}{50(b_2)} \sim t_{0.01}(23)$ $\Rightarrow t = -0.667$ and $t_{0.01}(23) = 2.807$

If Ho is true, there is no evidence to says that \$2\$|, that means motel-pct and comp-pct has similar tendency.

Since -2.809 < -0.667 < 2.807, we Sail to reject H_0 .

Thus, motel_pct and comp_pct has similar tendency.

e.



During the time period 17-23, the residuals are negative, which means the actual motel occupancy is lower than expected value.

```
MPCT<-c(motel_motel_pct)
                                                              34
                                                                  #C
   CPCT<-c(motel$comp_pct)</pre>
                                                              35
                                                                   alpha3<-0.01
    TIME<-c(motelStime)
                                                                  tc3<-qt(1-alpha3, length(MPCT)-2)#reject H0 if t>tc3
 5
                                                                  test_t<-(b2/seb2)
                                                              37
 6
                                                                   print("reject HO if t>tc3")
                                                              38
    plot(TIME, CPCT, xlab="time", ylab="pct", ylim=c(30, 100))
                                                                   test t
    points (TIME, MPCT, col="blue")
                                                              39
   points(TIME, CPCT, col="green")
                                                              40
                                                                  tc3
   MC<-data.frame(CPCT, MPCT)
                                                              41
11 modMC<-lm(MPCT~CPCT, data=MC)
                                                              42
                                                                   #d
12 smodMC<-summary(modMC)
                                                              43
                                                                   alpha4<-0.01
13 b1<-coef(modMC)[[1]]
                                                                  tc4<-qt(1-alpha4/2, length(MPCT)-2)
14 b2<-coef(modMC)[[2]]
                                                              45
                                                                  test_t2<-((b2-1)/seb2)
15 alpha=0.975
                                                                  print("reject H0 if tc4>t>-tc4")
16 tc<-gt(alpha, length(TIME)-2)
17 seb2<-coef(smodMC)[2, 2]
                                                                   test_t2
                                                              47
18 lowerbound<-b2-tc*seb2</p>
                                                              48
                                                                  tc4
   upperbound<-b2+tc*seb2
19
                                                              49
   lowerbound
20
                                                              50
                                                                   #e
   upperbound
                                                                  e<-resid(modMC)
                                                              51
22
                                                                   plot(TIME, e, xlab="time", ylab="residual")
23
    #b
                                                                   points(c(17:23), e[17:23], col="red")
24 alpha2<-0.95
                                                              53
25 tc2<-qt(alpha2, length(MPCT)-2)
                                                                   curve(0*x, col="black", add=TRUE)
                                                              54
26 theCPCT<-70
                                                              CC
27 theMPCT<-b1+b2*theCPCT
   seMPCT<-((sd(MPCT))/(sqrt(length(MPCT))))
29 lowerbound2<-theMPCT-tc2*seMPCT</p>
   upperbound2<-theMPCT+tc2*seMPCT
30
31 lowerbound2
```

upperbound2