

## Multiple Regression in practice:

1. [15 Pts.] Consider the least squares fit of  $y$  to the continuous explanatory variables  $x$ ,  $z$ , and  $w$ . Provided below is the matrix  $(X^T X)^{-1}$ :

$$\begin{pmatrix} 2.56 & -0.26 & -0.28 & -0.07 \\ -0.26 & 0.03 & 0.03 & 0.00 \\ -0.28 & 0.03 & 0.04 & 0.00 \\ -0.07 & 0.00 & 0.00 & 0.01 \end{pmatrix}$$

$$\Psi(\beta) = \sigma^2 (X^T X)^{-1}$$

Also provided is the summary of the least squares fit.

`lm(formula = y ~ x + z + w, data = myD)`

Coefficients:	$\hat{\beta}_i$	Estimate	$\hat{SE}(\hat{\beta}_i)$	$t$	$t$ value
(Intercept)	-3.64797	1.61404		-2.260	
x	0.09884	XXXXXX		5.518	
z	0.38789	0.19174		2.023	
w	-0.21235	0.09320	XXXXXX		

$$t = \frac{\hat{\beta}_i}{\hat{SE}(\hat{\beta}_i)}$$

$$\hat{\sigma} = \sqrt{\frac{e^T e}{n-(p+1)}} = \sqrt{\frac{ESS}{n-(p+1)}}$$

i.e.

$$\begin{aligned} R^2 &= \frac{RegSS}{TSS} \\ &= \frac{TSS - ESS}{TSS} \end{aligned}$$

$$ESS = \left(\frac{1}{R^2} - 1\right) RegSS$$

$$F = \frac{RegSS/p}{ESS/(n-(p+1))} = \frac{R^2}{1-R^2} \cdot \frac{n-(p+1)}{p}$$

(a) What are the five missing values indicated by XXXXX in the above summary?

$$P = 3 \text{ in this case}$$

(b) Explain how to test whether the coefficient for  $z$  is zero.

① null hypothesis ( $H_0$ )

② t stat

③ If value lies in the extreme tail of the null distribution, reject  $H_0$ .

(c) The standard deviations (normalized by  $n - 1$ ) for the variables in this study are as follows:

Variable	SD
$y$	1.23
$x$	10.14
$z$	0.96
$w$	0.91

Compute, compare, and comment on the standardized regression coefficients for  $x$  and  $w$ .

Std coefficient of  $x$

$$\hat{\beta}_x' = \hat{\beta}_x \cdot \frac{SD(x)}{SD(y)}$$

## Dummy Variable Regression in practice:

1. Let's imagine that 80 students took a particular course at Berkeley of whom 20 were freshmen, 20 were sophomores, 20 were juniors and 20 were seniors. In R, I have saved the final scores (out of 100) for the 20 freshmen in the vector  $g_1$ , for the 20 sophomores in  $g_2$ , juniors in  $g_3$  and seniors in  $g_4$ . Consider the following output:

```
> mean(g1)          mean & SD
[1] 57.96
> sd(g1)           cal as (n-1)
[1] 3.92
> mean(g2)
[1] 64.13
> sd(g2)
[1] 3.91
> mean(g3)
[1] 67.60
> sd(g3)
[1] 6.92
> mean(g4)
[1] 71.22
> sd(g4)
[1] 5.77
```

Question	Total points
Q1	12
Q2	6
Q3	9
Q4	6
Q5	12
Q6	14
Q7	12
	60

Also, for  $i = 1, \dots, 80$ , let

- $y_i$ : Final score of the  $i^{th}$  student in the class.
- $x_{i1}$ : Takes the value 1 if the  $i^{th}$  student is a freshman and 0 otherwise.
- $x_{i2}$ : Takes the value 1 if the  $i^{th}$  student is a sophomore and 0 otherwise.
- $x_{i3}$ : Takes the value 1 if the  $i^{th}$  student is a junior and 0 otherwise.
- $x_{i4}$ : Takes the value 1 if the  $i^{th}$  student is a senior and 0 otherwise.

I fit the linear model:

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \beta_3 x_{i3} + \beta_4 x_{i4} + e_i, i = 1, \dots, n$$

to this data via R to obtain the following output:

① The mean of  
Base Case  
Residuals:  
Call:  
lm(formula = y ~ x1 + x2 + x3 + x4)

All dummy!

R actually run  $y \sim x_1 + x_2 + x_3 + \text{intercept}$   
(senior has excluded, seen as Base Case)

②  $t = \frac{\beta_i}{\text{SE}(\beta_i)}$   
Coefficients: (1 not defined because of singularities)  
Estimate Std. Error t value Pr(>|t|)  
(Intercept) XXXXX 1.18 60.24 < 2e-16 \*\*\*  
x1 -13.26 1.67 -7.93 1.5e-11 \*\*\*  
x2 -7.09 1.67 -4.24 6.2e-05 \*\*\*  
x3 -3.62 1.67 -2.16 0.034 \*x4 NA NA NA NA

$$\sqrt{VIF} = \sqrt{\sum V_{ii}}$$

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1  
Residual standard error: XXXXX on 76 degrees of freedom  
Multiple R-squared: 0.4734, Adjusted R-squared: XXXXX  
F-statistic: 22.77 on 3 and 76 DF, p-value: 1.279e-10

$$\begin{aligned} \hat{\sigma}^2 &= \frac{\text{Residual SS}}{n-(p+1)} \\ ESS &= \sum_i (\text{Size of group}_i - 1) \times \text{Sample Variance}_i \\ &= (20-1) \times 3.82 + (20-1) \times 9.1 + \dots \\ &= 1 - \left( \frac{n-1}{n-(p+1)} \cdot \frac{\text{Residual SS}}{\text{TSS}} \right) \\ &= 1 - \left( \frac{n-1}{n-(p+1)} \cdot (1-R^2) \right) \end{aligned}$$

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(b) Fill in the 3 missing values in the R output with proper reasoning. (6 points).

(a) Why does the R output above say “1 not defined because of singularities”? Give reasons for your answer and suggest a way to fix the problem. (2 points)

(explain perfect collinearity)



(c) Explain why the standard error estimates for the coefficients of  $x_1$ ,  $x_2$ , and  $x_3$  are all the same. (3 points).

$$SE(\beta) = \sqrt{\text{diagonal of } \hat{\sigma}^2(X^T X)^{-1}}$$

$$X = \begin{pmatrix} \text{intercept} & x_1 & x_2 & x_3 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$X^T X = \begin{pmatrix} 80 & 20 & 20 & 20 \\ 20 & 20 & 0 & 0 \\ 20 & 0 & 20 & 0 \\ 20 & 0 & 0 & 20 \end{pmatrix}$$

