

## Question 1

a) Calculations done in Excel sheet attached

Source	SS	df	MS	Number of obs	=	2298
				F(3, 2294)	=	554
Model	190.873336	3.00	63.624445400	Prob > F	=	0.000
Residual	263.547174	2294.00	0.114885429	R-squared	=	0.420
				Adj R-squared	=	0.419
Total	454.420510	2297.00	0.197832177	Root MSE	=	0.339
lfare	Coef.	Std. Err.	t	P> t	[ 95% conf. interval ]	
ldist	0.4544197	0.128003	3.550	0.0004180	0.204	0.705
lpassen	-0.0678697	0.0084094	-8.071	0.000	-0.084	-0.051
bmktskr	0.3092456	0.0425167	7.274	0.000	0.226	0.393
_cons	2.241328	0.1218621	18.392	0.000	2.002	2.480

formulas

$$F(3, 2294) = \frac{\text{MS of Model}}{\text{MS of Residual}}$$

$$R^2 = 1 - \frac{\text{SSR}}{\text{SST}}$$

$$\text{Adj } R^2 = 1 - \frac{n-1}{n-k-1} \cdot \frac{\text{SSR}}{\text{SST}}$$

$$\text{Root MSE} = \sqrt{\text{MS of Residuals}}$$

p-values = Using 2tail tests (Excel)

b) Interpret the Coefficients on ldist and bmktskr in English sentence

Coefficient of ldist(0.4544197) is the elasticity of lfare with respect to ldist

Coefficient of bmktskr(0.3092456) is the elasticity of lfare with respect to bmktskr

ldist: When ldist increases by 1%, lfare increases by 0.4544197%, holding all other independent variables fixed.

bmktskr: When bmktskr increases by 1%, lfare increases by 0.3092456%, holding all other independent variables fixed.

- c) Test null hypothesis that the coefficient of bmktskr is equal to zero

1c) Test  $H_0: \beta_{\text{bmktskr}} = 0$

2-tailed t-test

$$t\text{ statistic} = \frac{0.3092456}{0.0425167} = 7.2735$$
$$P\text{-value} = 0.00 \text{ (calculated in 1a)}$$

P-value (0.00) is less than the p-value for 5% confidence level.  $0 < 0.05$

Hence,  $H_0: \beta_{\text{bmktskr}} = 0$  can be rejected.

However, the alternate hypothesis that coefficient of "bmktskr" is not zero cannot be rejected.

- d) Interpret the R-squared and the overall F-statistic
- R-squared

1d)  $R^2 = 0.420$

$R^2$  is the percentage of the dependent variable variation that a model can explain.

$$R^2 = \frac{\text{variance exp. by model}}{\text{Total var.}}$$

$R^2$  of 42% indicates that 42% variance of  $\log(\text{fare})$  of the dependent variable can be explained by the model with independent variables  $\log(\text{distance})$ ,  $\log(\text{market share})$ ,  $\log(\text{passengers})$ . where  $\log(\text{distance})$  is  $\log(\text{distance})$ ,  $\log(\text{market share})$  is market share and  $\log(\text{passengers})$  is no. of passengers.

#### F-statistic

f - statistic

$$f_{\text{stat}} = \frac{R^2/k}{(1-R^2)/(n-k-1)}$$

$$= \frac{0.420/3}{(1-0.42)/(2298-3-1)}$$

$$= \frac{0.42}{0.58} \times \frac{2294}{3}$$

$$\boxed{f_{\text{value}} \approx 554} = \frac{\text{MS of Regression Model}}{\text{MS of Residuals}}$$

P-value associated is 0.00, this when compared with 5% significance level ( $p < 0.05$ ). We can reject the null hypothesis that the independent variables are insignificant in determining airfare.

We can however not reject the alternate hypothesis that the variables  $\log(\text{distance})$ ,  $\log(\text{market share})$ ,  $\log(\text{passengers})$  are significant in determining  $\log(\text{airfare})$ .



Question 2

- a) First, test the hypothesis that  $H_0: \beta_0 = 0$  against the two-sided alternative. Then test  $H_0: \beta_1 = 1$  against the two-sided alternative. What do you conclude?

Q2

$$\text{price} = \beta_0 + \beta_1 \text{assess} + u$$

$$\hat{\text{price}} = -14.47 + 0.976 \text{ assess} + u$$

(16.27)      (0.049)

$$n = 88 \quad SSR = 165644.51 \quad R^2 = 0.820$$

②  $\Rightarrow H_0: \beta_0 = 0$  against 2 sided alternative

$$\hat{\beta}_0 = -14.47 \quad \beta_0 = 0 \quad S.E(\hat{\beta}_0) = 16.27$$

$$|t_{\text{stat}}| = \left| \frac{-14.47 - 0}{16.27} \right| = 0.8894$$

The critical value at 5% significance

level, two sided, d.o.f = 86 (88-1-1) is

1.98. As the absolute  $t_{\text{stat}}$  0.8894 < 1.98

the null hypothesis  $H_0$  cannot be rejected.

$\Rightarrow H_0: \beta_1 = 1$  against 2 sided alternative.

$$\hat{\beta}_1 = 0.976 \quad \beta_1 = 1 \quad S.E(\hat{\beta}_1) = 0.049$$

$$|t_{\text{stat}}| = \left| \frac{0.976 - 1}{0.049} \right| = |-0.489| = 0.489$$

Critical value at 5% level, two sided, d.o.f = 86

is 1.98. As the absolute value of  $t_{\text{statistic}}$  is

very less compared to 1.98, we fail to reject

the null hypothesis  $H_0: \beta_1 = 1$

$\Rightarrow$  As we cannot reject both the  $H_0$ 's, the variable assess (assessment) is rational and significant in determining price.

b) Joint hypothesis,  $B_0 = 0$  and  $B_1 = 1$

2 ⑥ Joint hypothesis  $B_0 = 0$   $B_1 = 1$ , perform  $f$  test.

$$\text{Given } SSR_{ur} = 165644.51 \quad n = 88$$

$$SSR_r = 209448.99$$

$$f = \frac{(SSR_r - SSR_{ur}) / q}{SSR_{ur} / (n - k - 1)}$$

$$q = df_r - df_{ur} = 1$$

$$n - k - 1 = df_{ur} = 88 - 1 - 1 = 86.$$

$$f = \frac{(209448.99 - 165644.51) / 1}{(165644.51) / 86}$$

$$= \frac{43804.48}{1926.1} = 22.742$$

$$f_{\text{stat}} = 22.742$$

Critical value from  $f$  distribution for 5% level

$q = 1$  and  $n - k - 1 = 86$  is  $\sim 3.96$ . (3.9519)

As the value of  $f_{\text{stat}}$  calculated (22.742) is

very high than the critical value, we can

reject the null hypothesis  $H_0$ . As  $H_0$  is rejected

we can say that ~~both hypothesis~~ both the

$B_0$  and  $B_1$  (i.e) intercept and the coeff. assess (assessment)

is ~~not~~ significant statistically at the 5% level.

Also it is effective to use  $t$  statistic over

$f$  statistic as  $t$  is flexible for single variable

hypothesis and can be used against 1 sided alternatives



c) Test  $H_0: \beta_2 = 0, \beta_3 = 0, \beta_4 = 0$

$$2 \text{ c) } \text{price} = \beta_0 + \beta_1 \text{arears} + \beta_2 \text{lotsize} + \beta_3 \text{sqrft} + \beta_4 \text{bdrms} + u$$

$$\text{Test } H_0: \beta_2 = 0 \quad \beta_3 = 0 \quad \beta_4 = 0$$

$$\text{Given } R_{ur}^2 = 0.829$$

$$R_r^2 = 0.82 \quad (\text{from 2a})$$

$$f = \frac{(R_{ur}^2 - R_r^2) / q}{(1 - R_{ur}^2) / (n - k - 1)}$$

$$q = d_{ur} - d_{ur} = 3 \quad (88 - 2 - 88 + 5)$$

$$n - k - 1 = d_{ur} = 83 \quad (88 - 4 - 1)$$

$$f = \frac{(0.829 - 0.82) / 3}{(1 - 0.829) / 83} = 1.4561$$

Critical value from  $f$  distribution for 5% significance level,  $q=3$ ,  $n-k-1=83$  is  $\approx 2.7146$

As the  $f$  value calculated is less than critical value, we cannot reject  $H_0$  the null hypothesis. Thus, the variables lotsize, sqrft, bdrms are jointly insignificant at the 5% level.

- d) If the variance of price changes with assess, lotsize, sqft, or bdrms, what can you say about the F-test from part (c)

2 (d) In 2c we saw that the variables lotsize, sqft, bdrms are jointly insignificant at the 5% significance level.

Now if variance of price changes with all variables, we can say this is attributed to SSR. This is because as per MLR 5, it says that variance of  $y$  given  $x$ , does not depend on the values of the independent variables.

But this does not make sense, as according to homoskedasticity, it is required that variance of  $u$  does not depend on independent variables.

So when we are saying variance changes with  $x$  variables, then our assumption of homoskedasticity does not apply. That means the F-test from the part 2c does not validate the hypothesis. We need to then test each variable separately to see the significance of the variable in the model.



### Question 3

#### a) Question 3a

Question 3

(a) The  $Y_i$ 's here are  $(x_i - \bar{x})^2$  and mean of each is equal to  $\text{Var}(x_i)$ , so by LLN the denominator converges to  $\text{Var}(x_i)$

$$\begin{aligned} \text{(b)} \quad E[(x_i - \bar{x})u_i] &= E[x_i u_i] - \bar{x} E[u_i] \\ &= E[x_i E[u_i | x]] - \bar{x} E[E[u_i | x]] \\ &= E[x_i \cdot E[u_i]] - \bar{x} E[E[u_i]] \\ &= E[x_i \cdot 0] - \bar{x} E[0] \\ &= 0 \end{aligned}$$

from above,  $Y_i$  i.e.  $(x_i - \bar{x})u_i$  have mean zero.



b) 3b

$$\begin{aligned}
 (b) \quad E[(x_i - \bar{x})u_i] &= E[x_i u_i] - \bar{x} E[u_i] \\
 &= E[x_i E[u_i | x]] - \bar{x} E[E[u_i | x]] \\
 &= E[x_i \cdot E[u_i]] - \bar{x} E[E[u_i]] \\
 &= E[x_i \cdot 0] - \bar{x} E[0] \\
 &= 0
 \end{aligned}$$

from above,  $y_i$  i.e.  $(x_i - \bar{x})u_i$  have mean zero.

c) 3c

$$\begin{aligned}
 (c) \quad \text{Var}[(x_i - \bar{x})u_i] &= \text{Var}(E[(x_i - \bar{x})u_i | x]) + E[\text{Var}((x_i - \bar{x})u_i | x)] \\
 &= \text{Var}[E[x_i u_i - \bar{x} u_i]] + E[\text{Var}((x_i - \bar{x})u_i | x)] \\
 &= \text{Var}[E[x_i u_i] - \bar{x} E[u_i]] + \text{Var}[(x_i - \bar{x})u_i] \\
 &= 0 + (x_i - \bar{x})^2 \text{Var}(u_i) + (x_i - \bar{x})^2 \text{Var}(u_i) + \dots \\
 &= 0 + E[(x_i - \bar{x})^2] \cdot \text{Var}(u_i) \\
 &= \sigma^2 \cdot \sum_{i=1}^n (x_i - \bar{x})^2 \\
 &= \sigma^2 \text{Var}(x_i)
 \end{aligned}$$

We used homoskedasticity  
 $\text{Var}(u_i) = \text{Var}(u) = \sigma^2$

d) 3d

(d)

We know

$$\sqrt{n} (\hat{\beta}_1 - \beta_1) = \sqrt{n} \frac{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}) u_i}{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$$

from 2b. we found  $E(\text{numerator}) = 0$

from 2c we found variance of numerators  $u_i$  turns is equal to  $\sigma^2 \text{Var}(x_i)$ .

$\Rightarrow$  Now using the Central limit theorem, we can say that, the distribution of  $\sqrt{n} (\hat{\beta}_1 - \beta_1)$  converges to  $N(0, \sigma^2 \text{Var}(x))$

$\Rightarrow$  Denominator term  $a_n = \left( \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 \right)^{-1}$

converges to  $\approx \text{Var}(x_i)$

$$\text{Now (we have)} \frac{\text{numerator}}{\text{denominator}} = \frac{N(0, \sigma^2 \text{Var}(x))}{\text{Var}(x_i)}$$

$$= N\left(0, \frac{\sigma^2}{\text{Var}(x_i)}\right)$$

Continued...



Given fact that, If a random variable  $X$  is such that  $\sqrt{n}X$  converges in distribution to  $N(0, \sigma^2)$ , then for an  $a_n$  such that  $\text{plim } a_n = a$ ,  $\sqrt{n}a_n X$  converges in distribution to  $N(0, a^2 \sigma^2)$ .

In our case  $\sqrt{n}X = \frac{\text{numerator}}{\text{denominator}}$  where  $\text{plim } a_n = a$ .

The  $\sqrt{n}a_n X$  (i.e.)  $\sqrt{n}(\hat{\beta}_1 - \beta_1)$  converges in distribution to  $N\left(0, \frac{\sigma^2}{\text{Var}(x_i)}\right)$ .

Thus we can say

$\sqrt{n}(\hat{\beta}_1 - \beta_1)$  is asymptotically distributed  $N\left(0, \frac{\sigma^2}{\text{Var}(x_i)}\right)$

$$\left(\frac{\sigma^2}{\text{Var}(x_i)}, 0\right)$$