

HW EE325

1.) This file CEOSAL2.DTA contains data on 177 chief executive officers and can be used to examine the effects of firm performance on CEO salary

(i) `. regress lsalary lsales lmktval`

Source	SS	df	MS			
Model	19.3365617	2	9.66828083	Number of obs =	177	
Residual	45.3096514	174	.260400295	F(2, 174) =	37.13	
Total	64.6462131	176	.367308029	Prob > F =	0.0000	
				R-squared =	0.2991	
				Adj R-squared =	0.2911	
				Root MSE =	.51029	

lsalary	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lsales	.1621283	.0396703	4.09	0.000	.0838315	.2404252
lmktval	.106708	.050124	2.13	0.035	.0077787	.2056372
_cons	4.620917	.2544083	18.16	0.000	4.118794	5.123041

Ans : The model of the constant elasticity: $\log(\text{salary}) = \beta_0 + \beta_1 \log(\text{sales}) + \beta_2 \log(\text{mktval}) + u$

The estimated model would be: $\log(\text{salary}) = 4.6209 + 0.1621 \log(\text{sales}) + 0.1067 \log(\text{mktval})$

(ii) . regress lsalary lsales lmktval profits

Source	SS	df	MS			
Model	19.3509799	3	6.45032663	Number of obs =	177	
Residual	45.2952332	173	.261822157	F(3, 173) =	24.64	
Total	64.6462131	176	.367308029	Prob > F =	0.0000	
				R-squared =	0.2993	
				Adj R-squared =	0.2872	
				Root MSE =	.51169	

lsalary	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lsales	.1613683	.0399101	4.04	0.000	.0825949	.2401416
lmktval	.0975286	.0636886	1.53	0.128	-.0281782	.2232354
profits	.0000357	.000152	0.23	0.815	-.0002643	.0003356
_cons	4.686924	.3797294	12.34	0.000	3.937425	5.436423

Ans (i) Profits cannot be included in log form because the company can make loss and thus the profit will be negative. But the number after log cannot be negative.

(ii) No, because these firm performance variables can explain the variation in CEO salaries only 29.93%. There is still left 70.07% that cannot be explained.

(iii) . regress lsalary lsales lmktval profits ceoten

Source	SS	df	MS			
Model	20.5768102	4	5.14420254	Number of obs =	177	
Residual	44.0694029	172	.256217459	F(4, 172) =	20.08	
Total	64.6462131	176	.367308029	Prob > F =	0.0000	
				R-squared =	0.3183	
				Adj R-squared =	0.3024	
				Root MSE =	.50618	

lsalary	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lsales	.1622339	.0394826	4.11	0.000	.0843012	.2401667
lmktval	.1017598	.063033	1.61	0.108	-.022658	.2261775
profits	.0000291	.0001504	0.19	0.847	-.0002677	.0003258
ceoten	.0116847	.005342	2.19	0.030	.0011403	.022229
_cons	4.55778	.3802548	11.99	0.000	3.807213	5.308347

Ans The estimated model is:

$$\log(\text{salary}) = 4.56 + 0.16\log(\text{sales}) + 0.10\log(\text{mktval}) + 0.00\text{profits} + 0.01\text{ceoten}$$

This means that if CEO tenure increases by 1 year, salary would increase by 1%.

(iv) `. correlate profits lmktval`
`(obs=177)`

	profits	lmktval
profits	1.0000	
lmktval	0.7769	1.0000

Ans (i) Yes, these variables are highly correlated because the correlation coefficient between `lmktval` and `profits` is 0.7769.

(ii) For the OLS estimators, since the `log(mktval)` and `profits` are highly correlated meaning that if `log(mktval)` changes, `profits` will also change. Moreover you cannot hold any of these factors constant by given a change in one variable. Thus you cannot observe the pure effect of each variable, `log(mktval)` and `profits`, on CEO salaries

2.) Use the data in `WAGE2.DTA` for this exercise

Source	SS	df	MS			
Model	25.6953242	3	8.56510806	Number of obs =	935	
Residual	139.960959	931	.150334005	F(3, 931) =	56.97	
Total	165.656283	934	.177362188	Prob > F =	0.0000	
				R-squared =	0.1551	
				Adj R-squared =	0.1524	
				Root MSE =	.38773	

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.0748638	.0065124	11.50	0.000	.062083	.0876446
exper	.0153285	.0033696	4.55	0.000	.0087156	.0219413
tenure	.0133748	.0025872	5.17	0.000	.0082974	.0184522
_cons	5.496696	.1105282	49.73	0.000	5.279782	5.713609

(i) Ans $H_0: \beta_2 = \beta_3$

$H_a: \beta_2 \neq \beta_3$

```
(ii) . test exper = tenure
      ( 1)  exper - tenure = 0

            F( 1, 931) = 0.17
            Prob > F = 0.6805
```

Ans At the 5% significance level, we accept $H_0: \beta_2 = \beta_3$ because it does not fall into the rejection region as the significance level of the test is 68.05%.

3.) Use the data in DISCRIM.DTA to answer this question.

```
(i) . regress lpsoda prpblck lincome prppov
```

Source	SS	df	MS			
Model	.250340622	3	.083446874	Number of obs =	401	
Residual	2.62840943	397	.006620679	F(3, 397) =	12.60	
Total	2.87875005	400	.007196875	Prob > F =	0.0000	
				R-squared =	0.0870	
				Adj R-squared =	0.0801	
				Root MSE =	.08137	

lpsoda	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
prpblck	.0728072	.0306756	2.37	0.018	.0125003	.1331141
lincome	.1369553	.0267554	5.12	0.000	.0843552	.1895553
prppov	.38036	.1327903	2.86	0.004	.1192999	.6414201
_cons	-1.463333	.2937111	-4.98	0.000	-2.040756	-.8859092

Ans The usual form is :

$$\log(\text{psoda}) = -1.463333 + 0.728072\text{prpblck} + 0.1369553\text{lincome} + 0.38036\text{prppov}$$

So given 0.05 level of significant, t value of the prpblck is 2.37, it falls into the rejection region. This is because the p-value of prpblck is equal to 0.018 which is less than the rejection area of 0.05. Therefore we reject H_0 at 0.05 level of significant which means that the estimated β_1 is statistically different from zero.

With 0.01 level of significant

```
. regress lpsoda prpbck lincome prppov, level (99)
```

Source	SS	df	MS			
Model	.250340622	3	.083446874	Number of obs = 401		
Residual	2.62840943	397	.006620679	F(3, 397) = 12.60		
				Prob > F = 0.0000		
				R-squared = 0.0870		
				Adj R-squared = 0.0801		
Total	2.87875005	400	.007196875	Root MSE = .08137		

lpsoda	Coef.	Std. Err.	t	P> t	[99% Conf. Interval]	
prpbck	.0728072	.0306756	2.37	0.018	-.0065895	.152204
lincome	.1369553	.0267554	5.12	0.000	.067705	.2062056
prppov	.38036	.1327903	2.86	0.004	.0366628	.7240571
_cons	-1.463333	.2937111	-4.98	0.000	-2.223536	-.7031291

It can be concluded that given 0.01 level of significant, t value of the prpbck is 2.37, it does not fall into the rejection region. As the p-value of prpbck is equal to 0.018 which is greater than the rejection area of 0.01, thus we accept H_0 at 0.01 level of significant which means that the estimated β_1 is equal to zero.

(ii)

```
. regress lpsoda prpbck lincome prppov lhseval
```

Source	SS	df	MS			
Model	.529488085	4	.132372021	Number of obs = 401		
Residual	2.34926197	396	.00593248	F(4, 396) = 22.31		
				Prob > F = 0.0000		
				R-squared = 0.1839		
				Adj R-squared = 0.1757		
Total	2.87875005	400	.007196875	Root MSE = .07702		

lpsoda	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
prpbck	.0975502	.0292607	3.33	0.001	.0400244	.155076
lincome	-.0529904	.0375261	-1.41	0.159	-.1267657	.0207848
prppov	.0521229	.1344992	0.39	0.699	-.2122989	.3165447
lhseval	.1213056	.0176841	6.86	0.000	.0865392	.1560721
_cons	-.8415149	.2924318	-2.88	0.004	-1.416428	-.2666019

Ans The coefficient of lhseval is 0.1213056. This means that if hseval increase by 1%, holding other factors constant, psoda will increase by 0.1213056%.

$$H_0: \beta_4 = 0$$

$$H_a: \beta_4 \neq 0$$

At 0.05 significance level, the t value is 6.86, which falls into the rejection region because the p-value is 0.000 less than 0.05. So we reject H_0 at 0.05 level of significance.

(iii) $H_0: \beta_1 = \beta_2 = 0$

$H_a: \beta_1 \neq \beta_2 \neq 0$

```
. test (lincome=0)(prppov=0)
```

```
( 1)  lincome = 0
```

```
( 2)  prppov = 0
```

```
      F( 2, 396) = 3.52  
      Prob > F = 0.0304
```

Given 0.05 level of significance, we can conclude that the two variables are jointly significant because the p-value is approximately equal to 0.0304 which is less than 0.05, thus it falls into the rejection region. Therefore we reject H_0 which means that $\beta_1 \neq \beta_2 \neq 0$, or $\ln\text{income}$ and prppov are jointly significant.

(iv) The second regression which added $\log(\text{hseval})$ in to the model is more reliable because the adjusted R-squared increases when adding this variable. The original adjusted R-squared before we added $\log(\text{hseval})$ is equal to 0.0801, which means that only 8.01% of variation in $\log(\text{psoda})$ can be explained by prpblck , $\log(\text{income})$ and prppov . Later on the adjusted R-squared is improved to 0.1757 when we include the $\log(\text{hseval})$ which means that 17.57% of the variation in $\log(\text{psoda})$ is explained by prpblck , $\log(\text{income})$, prppov and $\log(\text{hseval})$.