

Quantitative Methods in Finance

Tutorial, Part 8:

Regression models with dummy explanatory variables.

Example 1: Based on quarterly data on profits and sales of industrial enterprises in the United States in the period 1965q1–1970q4 (data file `profit.dta`; both variables are expressed in billion USD at constant prices) estimate the following regression function:

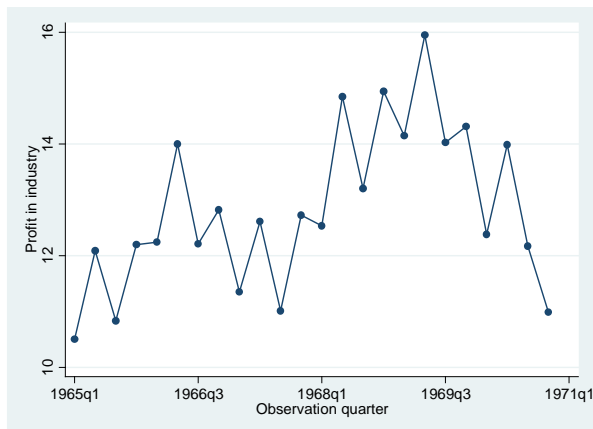
$$PROFIT_t = \beta_1 + \beta_2 SALES + u_t$$

and analyze the potential impact of the seasonal component on the link between the (numeric) variables by using dummy explanatory variables.

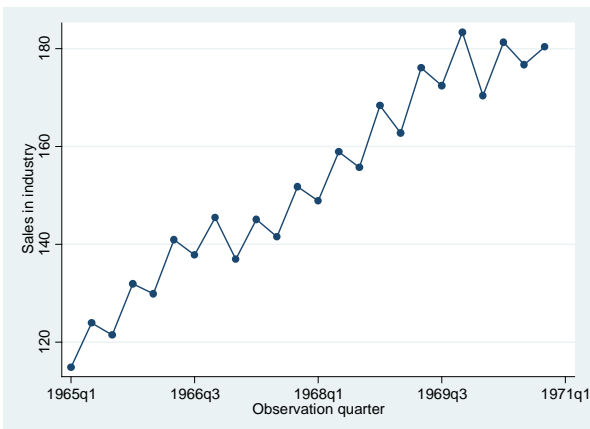
Computer printout of the results in Stata:

```
. tsset quarter
      time variable:  quarter, 1965q1 to 1970q4
              delta:  1 quarter
```

```
. twoway connected profit quarter
```



```
. twoway connected sales quarter
```



```
. gen q=quarter(dofq(quarter))
```

```
. tabulate q, gen(d)
```

q	Freq.	Percent	Cum.
1	6	25.00	25.00
2	6	25.00	50.00
3	6	25.00	75.00
4	6	25.00	100.00
Total	24	100.00	

. list quarter profit sales q d1-d4

	quarter	profit	sales	q	d1	d2	d3	d4
1.	1965q1	10.503	114.862	1	1	0	0	0
2.	1965q2	12.092	123.968	2	0	1	0	0
3.	1965q3	10.834	121.454	3	0	0	1	0
4.	1965q4	12.201	131.917	4	0	0	0	1
5.	1966q1	12.245	129.911	1	1	0	0	0
6.	1966q2	14.001	140.976	2	0	1	0	0
7.	1966q3	12.213	137.828	3	0	0	1	0
8.	1966q4	12.82	145.465	4	0	0	0	1
9.	1967q1	11.349	136.989	1	1	0	0	0
10.	1967q2	12.615	145.126	2	0	1	0	0
11.	1967q3	11.014	141.536	3	0	0	1	0
12.	1967q4	12.73	151.776	4	0	0	0	1
13.	1968q1	12.539	148.862	1	1	0	0	0
14.	1968q2	14.849	158.913	2	0	1	0	0
15.	1968q3	13.203	155.727	3	0	0	1	0
16.	1968q4	14.947	168.409	4	0	0	0	1
17.	1969q1	14.151	162.781	1	1	0	0	0
18.	1969q2	15.949	176.057	2	0	1	0	0
19.	1969q3	14.024	172.419	3	0	0	1	0
20.	1969q4	14.315	183.327	4	0	0	0	1
21.	1970q1	12.381	170.415	1	1	0	0	0
22.	1970q2	13.991	181.313	2	0	1	0	0
23.	1970q3	12.174	176.712	3	0	0	1	0
24.	1970q4	10.985	180.37	4	0	0	0	1

. regress profit sales

Source	SS	df	MS	Number of obs = 24			
Model	16.4051235	1	16.4051235	F(1, 22)	=	11.70	
Residual	30.8438618	22	1.40199372	Prob > F	=	0.0024	
Total	47.2489853	23	2.05430371	R-squared	=	0.3472	
				Adj R-squared	=	0.3175	
				Root MSE	=	1.1841	

profit	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sales	.0409541	.0119724	3.42	0.002	.0161249	.0657832
_cons	6.597974	1.840287	3.59	0.002	2.781452	10.4145

. regress profit sales d1-d4

note: d4 omitted because of collinearity

Source	SS	df	MS	Number of obs = 24			
Model	24.8290561	4	6.20726403	F(4, 19)	=	5.26	
Residual	22.4199291	19	1.17999627	Prob > F	=	0.0050	
Total	47.2489853	23	2.05430371	R-squared	=	0.5255	
				Adj R-squared	=	0.4256	
				Root MSE	=	1.0863	

profit	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sales	.0382462	.0114809	3.33	0.004	.0142163	.062276
d1	-.1838562	.6542925	-0.28	0.782	-1.553306	1.185594
d2	1.139036	.6307096	1.81	0.087	-.1810546	2.459126
d3	-.4016617	.6361179	-0.63	0.535	-1.733072	.9297484
d4	(omitted)					
_cons	6.872219	1.892072	3.63	0.002	2.912067	10.83237

. regress profit sales d2-d4

Source	SS	df	MS	Number of obs = 24		
Model	24.8290561	4	6.20726403	F(4, 19)	=	5.26
Residual	22.4199291	19	1.17999627	Prob > F	=	0.0050
Total	47.2489853	23	2.05430371	R-squared	=	0.5255
				Adj R-squared	=	0.4256
				Root MSE	=	1.0863

profit	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sales	.0382462	.0114809	3.33	0.004	.0142163	.062276
d2	1.322892	.6384745	2.07	0.052	-.0134505	2.659234
d3	-.2178055	.6322552	-0.34	0.734	-1.541131	1.10552
d4	.1838562	.6542925	0.28	0.782	-1.185594	1.553306
_cons	6.688363	1.711366	3.91	0.001	3.106432	10.27029

. test d3=d4=0

(1) d3 - d4 = 0
(2) d3 = 0

F(2, 19) = 0.20
Prob > F = 0.8197

. regress profit sales d2

Source	SS	df	MS	Number of obs = 24		
Model	24.35495	2	12.177475	F(2, 21)	=	11.17
Residual	22.8940352	21	1.09019215	Prob > F	=	0.0005
Total	47.2489853	23	2.05430371	R-squared	=	0.5155
				Adj R-squared	=	0.4693
				Root MSE	=	1.0441

profit	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
sales	.0393105	.010575	3.72	0.001	.0173186	.0613024
d2	1.331353	.4930214	2.70	0.013	.3060584	2.356647
_cons	6.515581	1.623084	4.01	0.001	3.140194	9.890968

■

Example 2: Based on the data for 150 industrial enterprises we were estimating the following production function: $\ln Q_i = \beta_1 + \beta_2 \ln L_i + \beta_3 \ln K_i + u_i$.

We divided the enterprises according to their ownership and got the following results:

- ♦ state ownership: $\widehat{\ln Q} = 3.63 + 0.53 \ln L + 0.48 \ln K$;
- ♦ mixed ownership: $\widehat{\ln Q} = 3.24 + 0.18 \ln L + 0.68 \ln K$;
- ♦ private ownership: $\widehat{\ln Q} = 4.73 + 0.23 \ln L + 0.55 \ln K$.

Based on the estimated production functions, write out a single regression model with explicit values of regression coefficients that will allow for a simultaneous estimation of the above three production functions.

Example 3: For 100 four-person households, we gathered data on their expenditures for food (EXP in 1,000 monetary units) and disposable income ($INCOME$ in 1,000 monetary units). We estimated the following regression functions:

$$\widehat{EXP} = 43.523 + 0.175 INCOME \quad R^2 = 0.452 ;$$

$$\begin{aligned} \widehat{EXP} &= 31.251 + 4.317 D1 + 13.251 D2 + 0.163 INCOME \\ se(b_j): & (5.44) \quad (2.15) \quad (4.25) \quad (0.05) \quad R^2 = 0.642. \end{aligned}$$

We divided the households into three groups: agricultural, mixed, and non-agricultural. We also defined two dummy variables: $D1$ equals 1, if the household is mixed, and $D2$ equals 1, if the household is non-agricultural.

- a) Interpret the estimated regression coefficients in both regression functions.
- b) Check the assumption that the type of household does not affect the expenditures.
- c) Can we claim that at a given level of income, the average expenditures for food in mixed households are higher than that of agricultural households?
- d) Explain why we used only two dummy variables, although we divided households into three groups.