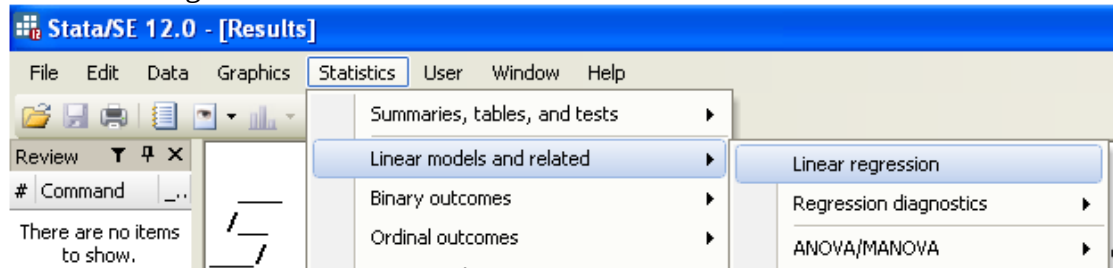


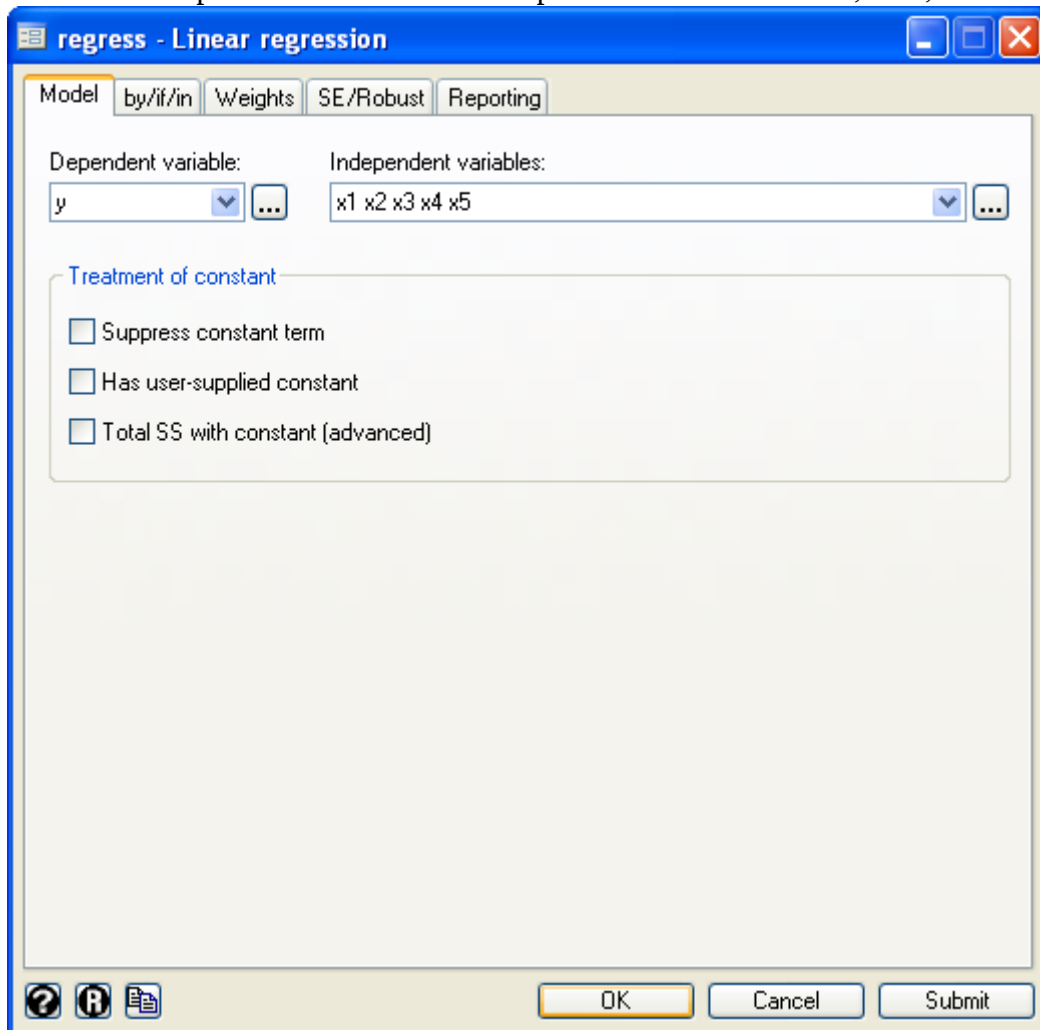
Multiple Regression Model

The model:
$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \beta_3 X_{3i} + \beta_4 X_{4i} + \beta_5 X_{5i} + u_i$$

To run a regression, from menu bar go to Statistics, choose Linear models and related, select Linear regression.



From regress – Linear regression window, specify dependent and independent variables in Dependent variable: and Independent variables: boxes, then, click OK.



The results will be as follows:

```
. reg y x1 x2 x3 x4 x5
```

Source	SS	df	MS	Number of obs = 23		
Model	1129.30649	5	225.861299	F(5, 17) = 57.63		
Residual	66.622243	17	3.91895547	Prob > F = 0.0000		
Total	1195.92874	22	54.3603971	R-squared = 0.9443		
				Adj R-squared = 0.9279		
				Root MSE = 1.9796		

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	.0048894	.0049619	0.99	0.338	-.0055794	.0153581
x2	-.6518875	.1743999	-3.74	0.002	-1.019839	-.2839359
x3	.2432418	.0895442	2.72	0.015	.05432	.4321636
x4	.1043175	.0706436	1.48	0.158	-.0447275	.2533625
x5	-.0711103	.0983811	-0.72	0.480	-.2786762	.1364556
_cons	38.59691	4.214487	9.16	0.000	29.70512	47.4887

Hypotheses Testing

To test whether we should eliminate x_4 and x_5 out of the model, or $H_0: \beta_4 = \beta_5 = 0$.

```
. test (x4=0) (x5=0)
```

```
( 1) x4 = 0
( 2) x5 = 0
```

```
F( 2, 17) = 1.17
Prob > F = 0.3354
```

To test whether $\beta_4 = \beta_5$.

```
. test x4 = x5
```

```
( 1) x4 - x5 = 0
```

```
F( 1, 17) = 1.27
Prob > F = 0.2746
```

```
. regress y x1 x2 x3
```

Source	SS	df	MS	Number of obs = 23		
Model	1120.17021	3	373.390069	F(3, 19) = 93.65		
Residual	75.7585287	19	3.98729099	Prob > F = 0.0000		
Total	1195.92874	22	54.3603971	R-squared = 0.9367		
				Adj R-squared = 0.9267		
				Root MSE = 1.9968		

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	.0108762	.0023812	4.57	0.000	.0058922	.0158602
x2	-.5410846	.1579697	-3.43	0.003	-.8717189	-.2104503
x3	.1740546	.0625307	2.78	0.012	.0431764	.3049328
_cons	38.6472	3.6496	10.59	0.000	31.0085	46.2859

To test whether $\beta_2 + \beta_3 = 1$

```
. test x2+x3=1
```

```
( 1) x2 + x3 = 1
```

```
F( 1, 19) = 132.06
Prob > F = 0.0000
```

Chow Test

```
. g dummy=(obs>1981)
```

```
. g d_inc=inc*dummy
```

```
. reg sav inc
```

Source	SS	df	MS	Number of obs =
Model	76621.7867	1	76621.7867	26
Residual	23248.3	24	968.679166	F(1, 24) = 79.10
Total	99870.0867	25	3994.80347	Prob > F = 0.0000
				R-squared = 0.7672
				Adj R-squared = 0.7575
				Root MSE = 31.124

sav	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
inc	.0376791	.0042366	8.89	0.000	.0289353 .046423
_cons	62.42267	12.76075	4.89	0.000	36.08578 88.75957

```
. reg sav inc if obs<=1981
```

Source	SS	df	MS	Number of obs =
Model	16456.2587	1	16456.2587	12
Residual	1785.03254	10	178.503254	F(1, 10) = 92.19
Total	18241.2912	11	1658.2992	Prob > F = 0.0000
				R-squared = 0.9021
				Adj R-squared = 0.8924
				Root MSE = 13.361

sav	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
inc	.0803319	.0083665	9.60	0.000	.0616901 .0989737
_cons	1.016115	11.63771	0.09	0.932	-24.91432 26.94655

```
. reg sav inc if obs>1981
```

Source	SS	df	MS	Number of obs =
Model	2614.39647	1	2614.39647	14
Residual	10005.2214	12	833.768451	F(1, 12) = 3.14
Total	12619.6179	13	970.739837	Prob > F = 0.1020
				R-squared = 0.2072
				Adj R-squared = 0.1411
				Root MSE = 28.875

sav	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
inc	.0148624	.0083932	1.77	0.102	-.0034248 .0331496
_cons	153.4947	32.71227	4.69	0.001	82.22075 224.7686

```
. reg sav inc dum d_inc
```

Source	SS	df	MS	Number of obs =
Model	88079.8327	3	29359.9442	26
Residual	11790.2539	22	535.920634	F(3, 22) = 54.78
Total	99870.0867	25	3994.80347	Prob > F = 0.0000
				R-squared = 0.8819
				Adj R-squared = 0.8658
				Root MSE = 23.15

sav	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
inc	.0803319	.0144968	5.54	0.000	.0502673 .1103964
dummy	152.4786	33.08237	4.61	0.000	83.86992 221.0872
d_inc	-.0654694	.0159824	-4.10	0.000	-.098615 -.0323239
_cons	1.016115	20.16483	0.05	0.960	-40.80319 42.83542

```
. test (dummy=0) (d_inc=0)
```

```
( 1) dummy = 0
```

```
( 2) d_inc = 0
```

```
F( 2, 22) = 10.69
Prob > F = 0.0006
```

Dummy Variable Regression Models

ANOVA Model

In the study of seasonal effect (Monday and Friday effects), ANOVA model is estimated using both models with and without intercept term:

Model with Intercept term: $Y_t = \beta_1 + \beta_2 D_{2t} + \beta_3 D_{3t} + \beta_4 D_{4t} + \beta_5 D_{5t} + u_t$

Model without Intercept term: $Y_t = \beta_1 D_{1t} + \beta_2 D_{2t} + \beta_3 D_{3t} + \beta_4 D_{4t} + \beta_5 D_{5t} + u_t$

where: Y_t = Return on asset at time t .

D_{1t} = 1 on Monday and = 0 otherwise.

D_{2t} = 1 on Tuesday and = 0 otherwise.

D_{3t} = 1 on Wednesday and = 0 otherwise.

D_{4t} = 1 on Thursday and = 0 otherwise.

D_{5t} = 1 on Friday and = 0 otherwise.

```
. reg y d2 d3 d4 d5
```

Source	SS	df	MS	Number of obs = 995		
Model	.00681682	4	.001704205	F(4, 990) = 7.60		
Residual	.221894717	990	.000224136	Prob > F = 0.0000		
Total	.228711537	994	.000230092	R-squared = 0.0298		
				Adj R-squared = 0.0259		
				Root MSE = .01497		

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
d2	.0044637	.0015009	2.97	0.003	.0015184	.0074089
d3	.0044351	.0015009	2.96	0.003	.0014898	.0073803
d4	.0040519	.0015009	2.70	0.007	.0011067	.0069972
d5	.0082592	.0015009	5.50	0.000	.005314	.0112045
_cons	-.0040211	.0010613	-3.79	0.000	-.0061037	-.0019385

```
. reg y d1 d2 d3 d4 d5, noconstant
```

Source	SS	df	MS	Number of obs = 995		
Model	.006865358	5	.001373072	F(5, 990) = 6.13		
Residual	.221894717	990	.000224136	Prob > F = 0.0000		
Total	.228760075	995	.00022991	R-squared = 0.0300		
				Adj R-squared = 0.0251		
				Root MSE = .01497		

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
d1	-.0040211	.0010613	-3.79	0.000	-.0061037	-.0019385
d2	.0004426	.0010613	0.42	0.677	-.0016401	.0025252
d3	.000414	.0010613	0.39	0.697	-.0016687	.0024966
d4	.0000308	.0010613	0.03	0.977	-.0020518	.0021134
d5	.0042381	.0010613	3.99	0.000	.0021555	.0063207

Interpretations of the estimated results of the two models are as follows:

Model	With intercept term	Without intercept
Monday	$\hat{Y}_t = \hat{\beta}_1 = -0.004021$	$\hat{Y}_t = \hat{\beta}_1 = -0.004021$
Tuesday	$\hat{Y}_t = \hat{\beta}_1 + \hat{\beta}_2 = -0.004021 + 0.004464 = 0.000443$	$\hat{Y}_t = \hat{\beta}_2 = 0.000443$
Wednesday	$\hat{Y}_t = \hat{\beta}_1 + \hat{\beta}_3 = -0.004021 + 0.004435 = 0.000414$	$\hat{Y}_t = \hat{\beta}_3 = 0.000414$
Thursday	$\hat{Y}_t = \hat{\beta}_1 + \hat{\beta}_4 = -0.004021 + 0.004052 = 0.000031$	$\hat{Y}_t = \hat{\beta}_4 = 0.000031$
Friday	$\hat{Y}_t = \hat{\beta}_1 + \hat{\beta}_5 = -0.004021 + 0.008259 = 0.004238$	$\hat{Y}_t = \hat{\beta}_5 = 0.004238$

For the purpose of testing only Monday and Friday effect, the model can be stated as:

Model to test Monday and Friday effect: $Y_t = \beta_0 + \beta_1 D_{1t} + \beta_5 D_{5t} + u_t$

```
. reg y d1 d5
```

Source	SS	df	MS			
Model	.006795783	2	.003397892	Number of obs =	995	
Residual	.221915754	992	.000223705	F(2, 992) =	15.19	
Total	.228711537	994	.000230092	Prob > F =	0.0000	
				R-squared =	0.0297	
				Adj R-squared =	0.0278	
				Root MSE =	.01496	

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
d1	-.0043169	.0012243	-3.53	0.000	-.0067194	-.0019144
d5	.0039423	.0012243	3.22	0.001	.0015399	.0063448
_cons	.0002958	.0006121	0.48	0.629	-.0009055	.001497

Interpretation of the estimated result of the model is as follow:

Model	
Tuesday Wednesday Thursday	$\hat{Y}_t = \hat{\beta}_0 = 0.000296$
Monday	$\hat{Y}_t = \hat{\beta}_0 + \hat{\beta}_1 = 0.000296 - 0.004317 = -0.004021$
Friday	$\hat{Y}_t = \hat{\beta}_0 + \hat{\beta}_5 = 0.000296 - 0.003942 = 0.004238$

ANCOVA Model

If the model also includes quantitative independent variable (X_t), the model can be stated as:

Model with intercept dummy: $Y_t = \beta_0 + \beta_1 X_{1t} + \gamma_0 D_{1t} + \lambda_0 D_{5t} + u_t$

This model can be interpreted as:

Model for Tuesday, Wednesday and Friday: $Y_t = \beta_0 + \beta_1 X_{1t} + u_t$

Model for Monday: $Y_t = (\beta_0 + \gamma_0) + \beta_1 X_{1t} + u_t$

Model for Friday: $Y_t = (\beta_0 + \lambda_0) + \beta_1 X_{1t} + u_t$

```
. reg y x1 d1 d5
```

Source	SS	df	MS	Number of obs = 995		
Model	.007808227	3	.002602742	F(3, 991)	= 11.68	
Residual	.220903311	991	.000222909	Prob > F	= 0.0000	
				R-squared	= 0.0341	
				Adj R-squared	= 0.0312	
Total	.228711537	994	.000230092	Root MSE	= .01493	

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	1.90e-06	8.91e-07	2.13	0.033	1.50e-07	3.65e-06
d1	-.0043131	.0012221	-3.53	0.000	-.0067113	-.0019149
d5	.0039385	.0012221	3.22	0.001	.0015403	.0063368
_cons	-.0708331	.0333809	-2.12	0.034	-.1363383	-.0053278

Model with intercept and slope dummy variables:

$$Y_t = \beta_0 + \gamma_0 D_{1t} + \lambda_0 D_{5t} + \beta_1 X_{1t} + \gamma_1 D_{1t} X_{1t} + \lambda_1 D_{5t} X_{1t} + u_t$$

This model can be interpreted as:

Model for Tuesday, Wednesday and Friday: $Y_t = \beta_0 + \beta_1 X_{1t} + u_t$

Model for Monday: $Y_t = (\beta_0 + \gamma_0) + (\beta_1 + \gamma_1) X_{1t} + u_t$

Model for Friday: $Y_t = (\beta_0 + \lambda_0) + (\beta_1 + \lambda_1) X_{1t} + u_t$

```
. g x1d1 = x1*d1
```

```
. g x1d5 = x1*d5
```

```
. reg y d1 d5 x1 x1d1 x1d5
```

Source	SS	df	MS	Number of obs = 995		
Model	.008022894	5	.001604579	F(5, 989)	= 7.19	
Residual	.220688644	989	.000223143	Prob > F	= 0.0000	
				R-squared	= 0.0351	
				Adj R-squared	= 0.0302	
Total	.228711537	994	.000230092	Root MSE	= .01494	

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
d1	-.0541205	.086225	-0.63	0.530	-.2233254	.1150844
d5	.0576593	.0862319	0.67	0.504	-.1115592	.2268777
x1	1.92e-06	1.15e-06	1.67	0.096	-3.39e-07	4.18e-06
x1d1	1.33e-06	2.30e-06	0.58	0.564	-3.19e-06	5.85e-06
x1d5	-1.43e-06	2.30e-06	-0.62	0.533	-5.95e-06	3.08e-06
_cons	-.0716146	.0431142	-1.66	0.097	-.1562204	.0129911

Chow Test vs Dummy Variable Technique

If we would like to check whether the model should be separated into two sub-periods (2000-2001 and 2002-2004), the models can be stated as:

$$\text{Model 2000-2001: } Y_t = \beta_0 + \beta_1 X_{1t} + u_{1t}$$

$$\text{Model 2002-2004: } Y_t = \gamma_0 + \gamma_1 X_{1t} + u_{2t}$$

$$\text{Chow Test: } H_0: \beta_0 = \gamma_0 \text{ and } \beta_1 = \gamma_1$$

Dummy Variable Regression Model

Model with intercept and slope dummy:

$$Y_t = \beta_0 + \delta_0 D_{YEARt} + \beta_1 X_{1t} + \delta_1 D_{YEARt} X_{1t} + u_t$$

where: $D_{YEARt} = 1$ for 2002-2004 and $= 0$ for 2000-2001.

The model can be interpreted as:

$$\text{Model 2000-2001: } Y_t = \beta_0 + \beta_1 X_{1t} + u_{1t}$$

$$\text{Model 2002-2004: } Y_t = (\beta_0 + \delta_0) + (\beta_1 + \delta_1) X_{1t} + u_{2t}$$

$$H_{01}: \delta_0 = 0 \text{ and } \delta_1 = 0$$

```
. g x1dyear = x1*dyear
```

```
. reg y dyear x1 x1dyear
```

Source	SS	df	MS			
Model	.001849335	3	.000616445	Number of obs = 995		
Residual	.226862202	991	.000228923	F(3, 991) = 2.69		
Total	.228711537	994	.000230092	Prob > F = 0.0450		
				R-squared = 0.0081		
				Adj R-squared = 0.0051		
				Root MSE = .01513		

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
dyear	.2727136	.1521381	1.79	0.073	-.0258363	.5712635
x1	8.49e-06	3.60e-06	2.36	0.019	1.42e-06	.0000156
x1dyear	-7.38e-06	4.10e-06	-1.80	0.072	-.0000154	6.66e-07
_cons	-.3139704	.1329766	-2.36	0.018	-.5749185	-.0530223

```
. test dyear x1dyear
```

```
( 1) dyear = 0
( 2) x1dyear = 0
```

```
F( 2, 991) = 1.80
Prob > F = 0.1657
```

Distinction between the two methods is that while Chow test is to test whether all estimators are the same or not, dummy variables technique is to separately test equality of each estimator for different sub-period -- some estimators may be different while some may not.

Interaction Effect

Model with interaction effect dummy:

$$Y_t = \beta_0 + \gamma_0 D_{1t} + \lambda_0 D_{JANt} + \delta_0 D_{1t} D_{JANt} + \beta_1 X_{1t} + u_t$$

where: $D_{1t} = 1$ for Monday and $= 0$ otherwise.

$D_{JANt} = 1$ for January and $= 0$ otherwise.

The meaning of $D_{1t} D_{JANt}$

$D_{1t} D_{JANt} = 1$ for Monday in January and $= 0$ otherwise.

```
. g d1djan = d1*djan
```

```
. reg y d1 djan d1djan x1
```

Source	SS	df	MS			
Model	.006773371	4	.001693343	Number of obs =	995	
Residual	.221938166	990	.00022418	F(4, 990) =	7.55	
Total	.228711537	994	.000230092	Prob > F =	0.0000	
				R-squared =	0.0296	
				Adj R-squared =	0.0257	
				Root MSE =	.01497	

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
d1	-.0059289	.0012667	-4.68	0.000	-.0084147	-.0034432
djan	.0019612	.0017065	1.15	0.251	-.0013876	.0053099
d1djan	.004773	.0036262	1.32	0.188	-.0023429	.0118889
x1	2.08e-06	8.98e-07	2.32	0.021	3.21e-07	3.85e-06
_cons	-.0769402	.033645	-2.29	0.022	-.1429639	-.0109166