

mvreg — Multivariate regression

Syntax Remarks and examples Also see	Menu Stored results	Description Methods and formulas	Options Reference
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Syntax

```
mvreg depvars = indepvars [if] [in] [weight] [, options]
```

<i>options</i>	Description
<hr/>	
Model	
<u>noconstant</u>	suppress constant term
Reporting	
<u>level(#)</u>	set confidence level; default is level(95)
<u>corr</u>	report correlation matrix
<u>display_options</u>	control column formats, row spacing, line width, display of omitted variables and base and empty cells, and factor-variable labeling
<u>noheader</u>	suppress header table from above coefficient table
<u>notable</u>	suppress coefficient table
<u>coeflegend</u>	display legend instead of statistics

indepvars may contain factor variables; see [\[U\] 11.4.3 Factor variables](#).

depvars and *indepvars* may contain time-series operators; see [\[U\] 11.4.4 Time-series varlists](#).

bootstrap, *by*, *jackknife*, *mi estimate*, *rolling*, and *statsby* are allowed; see [\[U\] 11.1.10 Prefix commands](#).

Weights are not allowed with the *bootstrap* prefix; see [\[R\] bootstrap](#).

*aweight*s are not allowed with the *jackknife* prefix; see [\[R\] jackknife](#).

*aweight*s and *fweight*s are allowed; see [\[U\] 11.1.6 weight](#).

noheader, *notable*, and *coeflegend* do not appear in the dialog box.

See [\[U\] 20 Estimation and postestimation commands](#) for more capabilities of estimation commands.

Menu

Statistics > Multivariate analysis > MANOVA, multivariate regression, and related > Multivariate regression

Description

`mvreg` fits multivariate regression models.

Options

Model

`noconstant` suppresses the constant term (intercept) in the model.

Reporting

`level(#)` specifies the confidence level, as a percentage, for confidence intervals. The default is `level(95)` or as set by `set level`; see [U] 20.7 **Specifying the width of confidence intervals**.

`corr` displays the correlation matrix of the residuals between the equations.

display_options: `noomitted`, `vsquish`, `noemptycells`, `baselevels`, `allbaselevels`, `nofvlabel`, `fvwrap(#)`, `fvwrapon(style)`, `cformat(%fmt)`, `pformat(%fmt)`, `sformat(%fmt)`, and `nostretch`; see [R] **estimation options**.

The following options are available with `mvreg` but are not shown in the dialog box:

`noheader` suppresses display of the table reporting F statistics, R -squared, and root mean squared error above the coefficient table.

`notable` suppresses display of the coefficient table.

`coeflegend`; see [R] **estimation options**.

Remarks and examples

[stata.com](http://www.stata.com)

Multivariate regression differs from multiple regression in that *several* dependent variables are jointly regressed on the same independent variables. Multivariate regression is related to Zellner's seemingly unrelated regression (see [R] **sureg**), but because the same set of independent variables is used for each dependent variable, the syntax is simpler, and the calculations are faster.

The individual coefficients and standard errors produced by `mvreg` are identical to those that would be produced by `regress` estimating each equation separately. The difference is that `mvreg`, being a joint estimator, also estimates the between-equation covariances, so you can test coefficients across equations and, in fact, the `test` syntax makes such tests more convenient.

► Example 1

Using the automobile data, we fit a multivariate regression for space variables (`headroom`, `trunk`, and `turn`) in terms of a set of other variables, including three performance variables (`displacement`, `gear_ratio`, and `mpg`):

```
. use http://www.stata-press.com/data/r13/auto
(1978 Automobile Data)
```

```
. mvreg headroom trunk turn = price mpg displ gear_ratio length weight
```

Equation	Obs	Parms	RMSE	"R-sq"	F	P
headroom	74	7	.7390205	0.2996	4.777213	0.0004
trunk	74	7	3.052314	0.5326	12.7265	0.0000
turn	74	7	2.132377	0.7844	40.62042	0.0000

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
headroom						
price	-.0000528	.000038	-1.39	0.168	-.0001286	.0000229
mpg	-.0093774	.0260463	-0.36	0.720	-.061366	.0426112
displacement	.0031025	.0024999	1.24	0.219	-.0018873	.0080922
gear_ratio	.2108071	.3539588	0.60	0.553	-.4956976	.9173119
length	.015886	.012944	1.23	0.224	-.0099504	.0417223
weight	-.0000868	.0004724	-0.18	0.855	-.0010296	.0008561
_cons	-.4525117	2.170073	-0.21	0.835	-4.783995	3.878972
trunk						
price	.0000445	.0001567	0.28	0.778	-.0002684	.0003573
mpg	-.0220919	.1075767	-0.21	0.838	-.2368159	.1926322
displacement	.0032118	.0103251	0.31	0.757	-.0173971	.0238207
gear_ratio	-.2271321	1.461926	-0.16	0.877	-3.145149	2.690885
length	.170811	.0534615	3.20	0.002	.0641014	.2775206
weight	-.0015944	.001951	-0.82	0.417	-.0054885	.0022997
_cons	-13.28253	8.962868	-1.48	0.143	-31.17249	4.607429
turn						
price	-.0002647	.0001095	-2.42	0.018	-.0004833	-.0000462
mpg	-.0492948	.0751542	-0.66	0.514	-.1993031	.1007136
displacement	.0036977	.0072132	0.51	0.610	-.0106999	.0180953
gear_ratio	-.1048432	1.021316	-0.10	0.919	-2.143399	1.933712
length	.072128	.0373487	1.93	0.058	-.0024204	.1466764
weight	.0027059	.001363	1.99	0.051	-.0000145	.0054264
_cons	20.19157	6.261549	3.22	0.002	7.693467	32.68968

We should have specified the `corr` option so that we would also see the correlations between the residuals of the equations. We can correct our omission because `mvreg`—like all estimation commands—typed without arguments redisplay results. The `noheader` and `notable` (read “notable”) options suppress redisplaying the output we have already seen:

```
. mvreg, notable noheader corr
```

```
Correlation matrix of residuals:
```

	headroom	trunk	turn
headroom	1.0000		
trunk	0.4986	1.0000	
turn	-0.1090	-0.0628	1.0000

```
Breusch-Pagan test of independence: chi2(3) = 19.566, Pr = 0.0002
```

The Breusch–Pagan test is significant, so the residuals of these three space variables are not independent of each other.

The three performance variables among our independent variables are `mpg`, `displacement`, and `gear_ratio`. We can jointly test the significance of these three variables in all the equations by typing

```

. test mpg displacement gear_ratio
( 1) [headroom]mpg = 0
( 2) [trunk]mpg = 0
( 3) [turn]mpg = 0
( 4) [headroom]displacement = 0
( 5) [trunk]displacement = 0
( 6) [turn]displacement = 0
( 7) [headroom]gear_ratio = 0
( 8) [trunk]gear_ratio = 0
( 9) [turn]gear_ratio = 0

      F( 9, 67) = 0.33
      Prob > F = 0.9622

```

These three variables are not, as a group, significant. We might have suspected this from their individual significance in the individual regressions, but this multivariate test provides an overall assessment with one p -value.

We can also perform a test for the joint significance of all three equations:

```

. test [headroom]
(output omitted)
. test [trunk], accum
(output omitted)
. test [turn], accum
( 1) [headroom]price = 0
( 2) [headroom]mpg = 0
( 3) [headroom]displacement = 0
( 4) [headroom]gear_ratio = 0
( 5) [headroom]length = 0
( 6) [headroom]weight = 0
( 7) [trunk]price = 0
( 8) [trunk]mpg = 0
( 9) [trunk]displacement = 0
(10) [trunk]gear_ratio = 0
(11) [trunk]length = 0
(12) [trunk]weight = 0
(13) [turn]price = 0
(14) [turn]mpg = 0
(15) [turn]displacement = 0
(16) [turn]gear_ratio = 0
(17) [turn]length = 0
(18) [turn]weight = 0

      F( 18, 67) = 19.34
      Prob > F = 0.0000

```

The set of variables as a whole is strongly significant. We might have suspected this, too, from the individual equations.



□ Technical note

The `mvreg` command provides a good way to deal with multiple comparisons. If we wanted to assess the effect of `length`, we might be dissuaded from interpreting any of its coefficients except that in the `trunk` equation. `[trunk]length`—the coefficient on `length` in the `trunk` equation—has a p -value of 0.002, but in the other two equations, it has p -values of only 0.224 and 0.058.

A conservative statistician might argue that there are 18 tests of significance in `mvreg`'s output (not counting those for the intercept), so p -values more than $0.05/18 = 0.0028$ should be declared

insignificant at the 5% level. A more aggressive but, in our opinion, reasonable approach would be to first note that the three equations are jointly significant, so we are justified in making some interpretation. Then we would work through the individual variables using `test`, possibly using $0.05/6 = 0.0083$ (6 because there are six independent variables) for the 5% significance level. For instance, examining `length`:

```
. test length
( 1) [headroom]length = 0
( 2) [trunk]length = 0
( 3) [turn]length = 0
      F( 3, 67) = 4.94
      Prob > F = 0.0037
```

The reported significance level of 0.0037 is less than 0.0083, so we will declare this variable significant. `[trunk]length` is certainly significant with its p -value of 0.002, but what about in the remaining two equations with p -values 0.224 and 0.058? We perform a joint test:

```
. test [headroom]length [turn]length
( 1) [headroom]length = 0
( 2) [turn]length = 0
      F( 2, 67) = 2.91
      Prob > F = 0.0613
```

At this point, reasonable statisticians could disagree. The 0.06 significance value suggests no interpretation, but these were the two least-significant values out of three, so we would expect the p -value to be a little high. Perhaps an equivocal statement is warranted: there seems to be an effect, but chance cannot be excluded.

□

Stored results

`mvreg` stores the following in `e()`:

Scalars

<code>e(N)</code>	number of observations
<code>e(k)</code>	number of parameters in each equation
<code>e(k_eq)</code>	number of equations in <code>e(b)</code>
<code>e(df_r)</code>	residual degrees of freedom
<code>e(chi2)</code>	Breusch–Pagan χ^2 (corr only)
<code>e(df_chi2)</code>	degrees of freedom for Breusch–Pagan χ^2 (corr only)
<code>e(rank)</code>	rank of <code>e(V)</code>

Macros

<code>e(cmd)</code>	<code>mvreg</code>
<code>e(cmdline)</code>	command as typed
<code>e(depvar)</code>	names of dependent variables
<code>e(eqnames)</code>	names of equations
<code>e(wtype)</code>	weight type
<code>e(wexp)</code>	weight expression
<code>e(r2)</code>	R -squared for each equation
<code>e(rmse)</code>	RMSE for each equation
<code>e(F)</code>	F statistic for each equation
<code>e(p_F)</code>	significance of F for each equation
<code>e(properties)</code>	<code>b V</code>
<code>e(estat_cmd)</code>	program used to implement <code>estat</code>
<code>e(predict)</code>	program used to implement <code>predict</code>
<code>e(marginsok)</code>	predictions allowed by <code>margins</code>
<code>e(marginsnotok)</code>	predictions disallowed by <code>margins</code>
<code>e(asbalanced)</code>	factor variables <code>fvset</code> as <code>asbalanced</code>
<code>e(asobserved)</code>	factor variables <code>fvset</code> as <code>asobserved</code>

Matrices

<code>e(b)</code>	coefficient vector
<code>e(Sigma)</code>	Σ matrix
<code>e(V)</code>	variance–covariance matrix of the estimators

Functions

<code>e(sample)</code>	marks estimation sample
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Methods and formulas

Given q equations and p independent variables (including the constant), the parameter estimates are given by the $p \times q$ matrix

$$\mathbf{B} = (\mathbf{X}'\mathbf{W}\mathbf{X})^{-1}\mathbf{X}'\mathbf{W}\mathbf{Y}$$

where \mathbf{Y} is an $n \times q$ matrix of dependent variables and \mathbf{X} is a $n \times p$ matrix of independent variables. \mathbf{W} is a weighting matrix equal to \mathbf{I} if no weights are specified. If weights are specified, let \mathbf{v} : $1 \times n$ be the specified weights. If `weight` frequency weights are specified, $\mathbf{W} = \text{diag}(\mathbf{v})$. If `aweight` analytic weights are specified, $\mathbf{W} = \text{diag}\{\mathbf{v}/(\mathbf{1}'\mathbf{v})(\mathbf{1}'\mathbf{1})\}$, meaning that the weights are normalized to sum to the number of observations.

The residual covariance matrix is

$$\mathbf{R} = \{\mathbf{Y}'\mathbf{W}\mathbf{Y} - \mathbf{B}'(\mathbf{X}'\mathbf{W}\mathbf{X})\mathbf{B}\}/(n - p)$$

The estimated covariance matrix of the estimates is $\mathbf{R} \otimes (\mathbf{X}'\mathbf{W}\mathbf{X})^{-1}$. These results are identical to those produced by `sureg` when the same list of independent variables is specified repeatedly; see [\[R\] sureg](#).

The [Breusch and Pagan \(1980\)](#) χ^2 statistic—a Lagrange multiplier statistic—is given by

$$\lambda = n \sum_{i=1}^q \sum_{j=1}^{i-1} r_{ij}^2$$

where r_{ij} is the estimated correlation between the residuals of the equations and n is the number of observations. It is distributed as χ^2 with $q(q-1)/2$ degrees of freedom.

Reference

Breusch, T. S., and A. R. Pagan. 1980. The Lagrange multiplier test and its applications to model specification in econometrics. *Review of Economic Studies* 47: 239–253.

Also see

- [MV] [mvreg postestimation](#) — Postestimation tools for mvreg
- [MV] [manova](#) — Multivariate analysis of variance and covariance
- [MI] [estimation](#) — Estimation commands for use with mi estimate
- [R] [nlsur](#) — Estimation of nonlinear systems of equations
- [R] [reg3](#) — Three-stage estimation for systems of simultaneous equations
- [R] [regress](#) — Linear regression
- [R] [regress postestimation](#) — Postestimation tools for regress
- [R] [sureg](#) — Zellner’s seemingly unrelated regression
- [SEM] [intro 5](#) — Tour of models
- [U] [20 Estimation and postestimation commands](#)