jackknife — Jackknife estimation

Description

`jackknife` performs jackknife estimation of the specified statistics (or expressions) for a Stata command or a user-written program. Statistics are jackknifed by estimating the command once for each observation or cluster in the dataset, leaving the associated observation or cluster out of the calculations. `jackknife` is designed for use with nonestimation commands, functions of coefficients, or user-written programs. To jackknife coefficients, we recommend using the `vce(jackknife)` option when allowed by the estimation command.

Quick start

Jackknife estimate of the standard deviation of `v1` returned by `summarize` in `r(sd)`

```
jackknife sd=r(sd), rclass: summarize v1
```

Jackknife estimate of the statistic `r(mystat)` returned by rclass program `myprog1` that returns the sample size in `r(N)`

```
jackknife stat=r(mystat), rclass: myprog1 v1
```

Same as above, and save the results from each replication in `mydata.dta`

```
jackknife stat=r(mystat), rclass saving(mydata): myprog1 v1
```

Jackknife estimate of a difference in coefficients estimated by `regress`

```
jackknife diff=(_b[x2]-_b[x1]): regress y x1 x2 x3
```

Jackknife estimate of the statistic `e(mystat)` returned by eclass program `myprog2` that returns the sample size in `e(N)`

```
jackknife stat=e(mystat), eclass: myprog2 y x1 x2 x3
```

Jackknife estimates of coefficients stored in `e(b)` by `myprog2`

```
jackknife _b, eclass: myprog2 y x1 x2 x3
```

Add variables containing the pseudovalues of the coefficients to the dataset

```
jackknife _b, eclass keep: myprog2 y x1 x2 x3
```

Menu

Statistics > Resampling > Jackknife estimation

Syntax

```
jackknife exp_list [ , options eform_option ] : command
```

options	Description
Main	
`eclass`	number of observations used is stored in `e(N)`
`rclass`	number of observations used is stored in `r(N)`
`n(`*exp*`)`	specify *exp* that evaluates to the number of observations used
Options	
`cluster(`*varlist*`)`	variables identifying sample clusters
`idcluster(`*newvar*`)`	create new cluster ID variable
`saving(`*filename*`, ...)`	save results to *filename*; save statistics in double precision; save results to *filename* every # replications
`keep`	keep pseudovalues
`mse`	use MSE formula for variance estimation
Reporting	
`level(#)`	set confidence level; default is `level(95)`
`notable`	suppress table of results
`noheader`	suppress table header
`nolegend`	suppress table legend
`verbose`	display the full table legend
`nodots`	suppress replication dots
`dots(#)`	display dots every # replications
`noisily`	display any output from *command*
`trace`	trace *command*
`title(`*text*`)`	use *text* as title for jackknife results
display_options	control columns and column formats, row spacing, line width, display of omitted variables and base and empty cells, and factor-variable labeling
eform_option	display coefficient table in exponentiated form
Advanced	
`nodrop`	do not drop observations
`reject(`*exp*`)`	identify invalid results
`coeflegend`	display legend instead of statistics

command is any command that follows standard Stata syntax. All weight types supported by *command* are allowed except `aweights`; see **[U] 11.1.6 weight**.

`collect` and `svy` are allowed; see **[U] 11.1.10 Prefix commands**.

`coeflegend` does not appear in the dialog box.

See **[U] 20 Estimation and postestimation commands** for more capabilities of estimation commands.

exp_list contains	(*name*: *elist*)
	elist
	eexp
elist contains	*newvar* = (*exp*)
	(*exp*)
eexp is	*specname*
	[*eqno*]*specname*
specname is	_b
	_b[]
	_se
	_se[]
eqno is	# #
	name

exp is a standard Stata expression; see **[U] 13 Functions and expressions**.

Distinguish between `[]`, which are to be typed, and [ ], which indicate optional arguments.

Options

Main

`eclass`, `rclass`, and `n(`*exp*`)` specify where *command* stores the number of observations on which it based the calculated results. We strongly advise you to specify one of these options.

`eclass` specifies that *command* store the number of observations in `e(N)`.

`rclass` specifies that *command* store the number of observations in `r(N)`.

`n(`*exp*`)` specifies an expression that evaluates to the number of observations used. Specifying `n(r(N))` is equivalent to specifying the `rclass` option. Specifying `n(e(N))` is equivalent to specifying the `eclass` option. If *command* stores the number of observations in `r(N1)`, specify `n(r(N1))`.

If you specify no options, `jackknife` will assume `eclass` or `rclass`, depending on which of `e(N)` and `r(N)` is not missing (in that order). If both `e(N)` and `r(N)` are missing, `jackknife` assumes that all observations in the dataset contribute to the calculated result. If that assumption is incorrect, the reported standard errors will be incorrect. For instance, say that you specify

```
. jackknife coef=_b[x2]: myreg y x1 x2 x3
```

where `myreg` uses `e(n)` instead of `e(N)` to identify the number of observations used in calculations. Further assume that observation 42 in the dataset has `x3` equal to missing. The 42nd observation plays no role in obtaining the estimates, but `jackknife` has no way of knowing that and will use the wrong *N*. If, on the other hand, you specify

```
. jackknife coef=_b[x2], n(e(n)): myreg y x1 x2 x3
```

`jackknife` will notice that observation 42 plays no role. The `n(e(n))` option is specified because `myreg` is an estimation command but it stores the number of observations used in `e(n)` (instead of the standard `e(N)`). When `jackknife` runs the regression omitting the 42nd observation, `jackknife` will observe that `e(n)` has the same value as when `jackknife` previously ran the regression using all the observations. Thus `jackknife` will know that `myreg` did not use the observation.

Options

cluster(*varlist*) specifies the variables identifying sample clusters. If cluster() is specified, one cluster is left out of each call to *command*, instead of 1 observation.

idcluster(*newvar*) creates a new variable containing a unique integer identifier for each resampled cluster, starting at 1 and leading up to the number of clusters. This option may be specified only when the cluster() option is specified. idcluster() helps identify the cluster to which a pseudovalue belongs.

saving(*filename*[, *suboptions*]) creates a Stata data file (.dta file) consisting of (for each statistic in *exp_list*) a variable containing the replicates.

double specifies that the results for each replication be saved as doubles, meaning 8-byte reals. By default, they are saved as floats, meaning 4-byte reals. This option may be used without the saving() option to compute the variance estimates by using double precision.

every(#) specifies that results be written to disk every #th replication. every() should be specified only in conjunction with saving() when *command* takes a long time for each replication. This option will allow recovery of partial results should some other software crash your computer. See [P] **postfile**.

replace specifies that *filename* be overwritten if it exists. This option does not appear in the dialog box.

keep specifies that new variables be added to the dataset containing the pseudovalues of the requested statistics. For instance, if you typed

```
. jackknife coef=_b[x2], eclass keep: regress y x1 x2 x3
```

new variable coef would be added to the dataset containing the pseudovalues for _b[x2]. Let b be the value of _b[x2] when all observations are used to fit the model, and let $b(j)$ be the value when the jth observation is omitted. The pseudovalues are defined as

$$\text{pseudovalue}_j = N\{b - b(j)\} + b(j)$$

where N is the number of observations used to produce b.

When the cluster() option is specified, each cluster is given at most one nonmissing pseudovalue. The keep option implies the nodrop option.

mse specifies that jackknife compute the variance by using deviations of the replicates from the observed value of the statistics based on the entire dataset. By default, jackknife computes the variance by using deviations of the pseudovalues from their mean.

Reporting

level(#); see [R] **Estimation options**.

notable suppresses the display of the table of results.

noheader suppresses the display of the table header. This option implies nolegend.

nolegend suppresses the display of the table legend. The table legend identifies the rows of the table with the expressions they represent.

verbose specifies that the full table legend be displayed. By default, coefficients and standard errors are not displayed.

nodots and dots(#) specify whether to display replication dots. By default, one dot character is displayed for each successful replication. An "x" is displayed if *command* returns an error or if any value in *exp_list* is missing. You can also control whether dots are displayed using set dots; see [R] **set**.

nodots suppresses display of the replication dots.

dots(#) displays dots every # replications. dots(0) is a synonym for nodots.

noisily specifies that any output from *command* be displayed. This option implies the nodots option.

trace causes a trace of the execution of *command* to be displayed. This option implies the noisily option.

title(*text*) specifies a title to be displayed above the table of jackknife results; the default title is Jackknife results or what is produced in e(title) by an estimation command.

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

eform_option causes the coefficient table to be displayed in exponentiated form; see [R] ***eform_option***. *command* determines which *eform_option* is allowed (eform(*string*) and eform are always allowed).

command determines which of the following are allowed (eform(*string*) and eform are always allowed):

eform_option	Description
eform(*string*)	use *string* for the column title
eform	exponentiated coefficient, *string* is exp(b)
hr	hazard ratio, *string* is Haz. ratio
shr	subhazard ratio, *string* is SHR
irr	incidence-rate ratio, *string* is IRR
or	odds ratio, *string* is Odds ratio
rrr	relative-risk ratio, *string* is RRR

Advanced

nodrop prevents observations outside e(sample) and the if and in qualifiers from being dropped before the data are resampled.

reject(*exp*) identifies an expression that indicates when results should be rejected. When *exp* is true, the resulting values are reset to missing values.

The following option is available with jackknife but is not shown in the dialog box:

coeflegend; see [R] **Estimation options**.

Remarks and examples

Remarks are presented under the following headings:

Introduction

Although the jackknife—developed in the late 1940s and early 1950s—is of largely historical interest today, it is still useful in searching for overly influential observations. This feature is often forgotten. In any case, the jackknife is

- an alternative, first-order unbiased estimator for a statistic;
- a data-dependent way to calculate the standard error of the statistic and to obtain significance levels and confidence intervals; and
- a way of producing measures called pseudovalues for each observation, reflecting the observation's influence on the overall statistic.

The idea behind the simplest form of the jackknife—the one implemented here—is to repeatedly calculate the statistic in question, each time omitting just one of the dataset's observations. Assume that our statistic of interest is the sample mean. Let y_j be the jth observation of our data on some measurement y, where $j = 1, \ldots, N$ and N is the sample size. If $\overline{y}$ is the sample mean of y using the entire dataset and $\overline{y}_{(j)}$ is the mean when the jth observation is omitted, then

$$\overline{y} = \frac{(N-1)\,\overline{y}_{(j)} + y_j}{N}$$

Solving for y_j, we obtain

$$y_j = N\overline{y} - (N-1)\,\overline{y}_{(j)}$$

These are the pseudovalues that `jackknife` calculates. To move this discussion beyond the sample mean, let $\hat{\theta}$ be the value of our statistic (not necessarily the sample mean) using the entire dataset, and let $\hat{\theta}_{(j)}$ be the computed value of our statistic with the jth observation omitted. The pseudovalue for the jth observation is

$$\hat{\theta}^*_j = N\hat{\theta} - (N-1)\,\hat{\theta}_{(j)}$$

The mean of the pseudovalues is the alternative, first-order unbiased estimator mentioned above, and the standard error of the mean of the pseudovalues is an estimator for the standard error of $\hat{\theta}$ (Tukey 1958).

The jackknife estimate of variance has been largely replaced by the bootstrap estimate (see [R] **bootstrap**), which is widely viewed as more efficient and robust. The use of jackknife pseudovalues to detect outliers is too often forgotten and is something the bootstrap does not provide. See Mosteller and Tukey (1977, 133–163) and Mooney and Duval (1993, 22–27) for more information.

Using jackknife

Typing

```
. jackknife exp_list: command
```

executes *command* once for each observation in the dataset, leaving the associated observation out of the calculations that make up *exp_list*.

command defines the statistical command to be executed. Most Stata commands and user-written programs can be used with `jackknife`, as long as they follow standard Stata syntax and allow the `if` qualifier; see **[U] 11 Language syntax**. The `by` prefix may not be part of *command*.

exp_list specifies the statistics to be collected from the execution of *command*. If *command* changes the contents in `e(b)`, *exp_list* is optional and defaults to `_b`.

When the `cluster()` option is given, clusters are omitted instead of observations, and N is the number of clusters instead of the sample size.

▷ Example 1

As our first example, we will show that the jackknife standard error of the sample mean is equivalent to the standard error of the sample mean computed using the classical formula in the `ci means` command. We use the `double` option to compute the standard errors with the same precision as the `ci means` command.

```
. use https://www.stata-press.com/data/r19/auto
(1978 automobile data)
. jackknife r(mean), double: summarize mpg
(running summarize on estimation sample)
Jackknife replications (74): .........10.........20.........30.........40......
> ...50.........60.........70.... done
Jackknife results                                     Number of obs = 74
                                                      Replications  = 74

      Command: summarize mpg
        _jk_1: r(mean)
          n(): r(N)
```

	Coefficient	Jackknife std. err.	t	P>|t|	[95% conf.	interval]
_jk_1	21.2973	.6725511	31.67	0.000	19.9569	22.63769

```
. ci means mpg
```

Variable	Obs	Mean	Std. err.	[95% conf.	interval]
mpg	74	21.2973	.6725511	19.9569	22.63769

◁

Jackknifed standard deviation

▷ Example 2

Mosteller and Tukey (1977, 139–140) request a 95% confidence interval for the standard deviation of the 11 values:

0.1, 0.1, 0.1, 0.4, 0.5, 1.0, 1.1, 1.3, 1.9, 1.9, 4.7

Stata's `summarize` command calculates the mean and standard deviation and stores them as `r(mean)` and `r(sd)`. To obtain the jackknifed standard deviation of the 11 values and save the pseudovalues as a new variable, `sd`, we would type

```
. clear
. input x
            x
  1. 0.1
  2. 0.1
  3. 0.1
  4. 0.4
  5. 0.5
  6. 1.0
  7. 1.1
  8. 1.3
  9. 1.9
 10. 1.9
 11. 4.7
 12. end
. jackknife sd=r(sd), rclass keep: summarize x
(running summarize on estimation sample)
Jackknife replications (11): .........10. done
Jackknife results                                     Number of obs = 11
                                                      Replications  = 11

      Command: summarize x
           sd: r(sd)
         n(): r(N)
```

	Coefficient	Jackknife std. err.	t	P>|t|	[95% conf.	interval]
sd	1.343469	.624405	2.15	0.057	-.047792	2.73473

Interpreting the output, the standard deviation reported by `summarize mpg` is 1.34. The jackknife standard error is 0.62. The 95% confidence interval for the standard deviation is −0.048 to 2.73.

By specifying keep, jackknife creates in our dataset a new variable, sd, for the pseudovalues.

```
. list, sep(4)
```

	x	sd
1.	.1	1.139977
2.	.1	1.139977
3.	.1	1.139977
4.	.4	.8893147
5.	.5	.824267
6.	1	.632489
7.	1.1	.6203189
8.	1.3	.6218889
9.	1.9	.835419
10.	1.9	.835419
11.	4.7	7.703949

The jackknife estimate is the average of the sd variable, so sd contains the individual values of our statistic. We can see that the last observation is substantially larger than the others. The last observation is certainly an outlier, but whether that reflects the considerable information it contains or indicates that it should be excluded from analysis depends on the context of the problem. Mosteller and Tukey created the data by sampling from an exponential distribution, so the observation is informative.

◁

▷ Example 3

Let's repeat the example above using the automobile dataset, obtaining the standard error of the standard deviation of mpg.

```
. use https://www.stata-press.com/data/r19/auto, clear
(1978 automobile data)
. jackknife sd=r(sd), rclass keep: summarize mpg
(running summarize on estimation sample)
Jackknife replications (74): .........10.........20.........30.........40......
> ...50.........60.........70.... done
Jackknife results                                     Number of obs = 74
                                                      Replications  = 74
      Command: summarize mpg
           sd: r(sd)
           n(): r(N)
```

	Coefficient	Jackknife std. err.	t	P>|t|	[95% conf.	interval]
sd	5.785503	.6072509	9.53	0.000	4.575254	6.995753

Let's look at sd more carefully:

```
. summarize sd, detail
                    pseudovalues: r(sd)
-------------------------------------------------------------
      Percentiles      Smallest
 1%     2.870471       2.870471
 5%     2.870471       2.870471
10%     2.906255       2.870471       Obs                  74
25%     3.328489       2.870471       Sum of wgt.          74

50%     3.948335                      Mean           5.817374
                       Largest        Std. dev.       5.22377
75%     6.844418       17.34316
90%     9.597018        19.7617       Variance       27.28777
95%     17.34316        19.7617       Skewness       4.07202
99%     38.60905       38.60905       Kurtosis       23.37823
. list make mpg sd if sd > 30

     +-------------------------------+
     | make        mpg          sd   |
     |-------------------------------|
71.  | VW Diesel    41    38.60905   |
     +-------------------------------+
```

Here the VW Diesel is the only diesel car in our dataset.

◁

Collecting multiple statistics

▷ Example 4

jackknife is not limited to collecting just one statistic. For instance, we can use summarize, detail and then obtain the jackknife estimate of the standard deviation and skewness. summarize, detail stores the standard deviation in r(sd) and the skewness in r(skewness), so we might type

```
. use https://www.stata-press.com/data/r19/auto, clear
(1978 automobile data)
. jackknife sd=r(sd) skew=r(skewness), rclass: summarize mpg, detail
(running summarize on estimation sample)
Jackknife replications (74): .........10.........20.........30.........40......
> ...50.........60.........70.... done
Jackknife results                                     Number of obs = 74
                                                      Replications  = 74
      Command: summarize mpg, detail
           sd: r(sd)
         skew: r(skewness)
           n(): r(N)
------------------------------------------------------------------------------
             |              Jackknife
             | Coefficient  std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
          sd |    5.785503   .6072509     9.53   0.000     4.575254    6.995753
        skew |    .9487176   .3367242     2.82   0.006     .2776272    1.619808
------------------------------------------------------------------------------
```

◁

Collecting coefficients

▷ Example 5

`jackknife` can also collect coefficients from estimation commands. For instance, using `auto.dta`, we might wish to obtain the jackknife standard errors of the coefficients from a regression in which we model the mileage of a car by its weight and trunk space. To do this, we could refer to the coefficients as `_b[weight]`, `_b[trunk]`, `_se[weight]`, and `_se[trunk]` in the *exp_list*, or we could simply use the extended expressions `_b`. In fact, `jackknife` assumes `_b` by default when used with estimation commands.

```
. use https://www.stata-press.com/data/r19/auto
(1978 automobile data)

. jackknife: regress mpg weight trunk
(running regress on estimation sample)

Jackknife replications (74): .........10.........20.........30.........40......
> ...50.........60.........70.... done

Linear regression                               Number of obs     =        74
                                                Replications      =        74
                                                F(2, 73)          =     78.10
                                                Prob > F          =    0.0000
                                                R-squared         =    0.6543
                                                Adj R-squared     =    0.6446
                                                Root MSE          =    3.4492

------------------------------------------------------------------------------
             |               Jackknife
         mpg |      Coef.   std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
      weight |  -.0056527   .0010216    -5.53   0.000    -.0076887   -.0036167
       trunk |   -.096229   .1486236    -0.65   0.519    -.3924354    .1999773
       _cons |   39.68913   1.873324    21.19   0.000      35.9556    43.42266
------------------------------------------------------------------------------
```

If you are going to use `jackknife` to estimate standard errors of model coefficients, we recommend using the `vce(jackknife)` option when it is allowed with the estimation command; see [R] ***vce_option***.

```
. regress mpg weight trunk, vce(jackknife, nodots)
Linear regression                               Number of obs     =        74
                                                Replications      =        74
                                                F(2, 73)          =     78.10
                                                Prob > F          =    0.0000
                                                R-squared         =    0.6543
                                                Adj R-squared     =    0.6446
                                                Root MSE          =    3.4492

------------------------------------------------------------------------------
             |               Jackknife
         mpg |      Coef.   std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
      weight |  -.0056527   .0010216    -5.53   0.000    -.0076887   -.0036167
       trunk |   -.096229   .1486236    -0.65   0.519    -.3924354    .1999773
       _cons |   39.68913   1.873324    21.19   0.000      35.9556    43.42266
------------------------------------------------------------------------------
```

◁

❑ Technical note

When the jackknife prefix is used with a user-defined program and when the expression list is _b, jackknife calls

```
set coeftabresults off
```

before entering the replication loop to prevent Stata from performing unnecessary calculations. This means that, provided option noisily is not specified, estimation commands will not build or post the coefficient table matrix r(table).

If your program calls an estimation command and needs r(table) to exist to perform properly, then your program will need to call

```
set coeftabresults on
```

before calling other estimation commands.

❑

John Wilder Tukey (1915–2000) was born in Massachusetts. He studied chemistry at Brown and mathematics at Princeton and afterward worked at both Princeton and Bell Labs, as well as being involved in a great many government projects, consultancies, and committees. He made outstanding contributions to several areas of statistics, including time series, multiple comparisons, robust statistics, and exploratory data analysis. Tukey was extraordinarily energetic and inventive, not least in his use of terminology: he is credited with inventing the terms bit and software, in addition to ANOVA, boxplot, data analysis, hat matrix, jackknife, stem-and-leaf plot, trimming, and winsorizing, among many others. Tukey's direct and indirect impacts mark him as one of the greatest statisticians of all time.

Stored results

jackknife stores the following in e():

Scalars

e(N)	sample size
e(N_reps)	number of complete replications
e(N_misreps)	number of incomplete replications
e(N_clust)	number of clusters
e(k_eq)	number of equations in e(b)
e(k_extra)	number of extra equations
e(k_exp)	number of expressions
e(k_eexp)	number of extended expressions (_b or _se)
e(df_r)	degrees of freedom

Macros

e(cmdname)	command name from *command*
e(cmd)	same as e(cmdname) or jackknife
e(command)	*command*
e(cmdline)	command as typed
e(prefix)	jackknife
e(wtype)	weight type
e(wexp)	weight expression
e(title)	title in estimation output
e(cluster)	cluster variables
e(pseudo)	new variables containing pseudovalues
e(nfunction)	e(N), r(N), n() option, or empty

e(exp#)	expression for the #th statistic
e(mse)	from mse option
e(vce)	jackknife
e(vcetype)	title used to label Std. err.
e(properties)	b V

Matrices

e(b)	observed statistics
e(b_jk)	jackknife estimates
e(V)	jackknife variance–covariance matrix
e(V_modelbased)	model-based variance

In addition to the above, the following is stored in r():

Matrices

r(table)	matrix containing the coefficients with their standard errors, test statistics, p-values, and confidence intervals

Note that results stored in r() are updated when the command is replayed and will be replaced when any r-class command is run after the estimation command.

When *exp_list* is _b, jackknife will also carry forward most of the results already in e() from *command*.

Methods and formulas

Let $\hat{\theta}$ be the observed value of the statistic, that is, the value of the statistic calculated using the original dataset. Let $\hat{\theta}_{(j)}$ be the value of the statistic computed by leaving out the jth observation (or cluster); thus $j = 1, 2, \ldots, N$ identifies an individual observation (or cluster), and N is the total number of observations (or clusters). The jth pseudovalue is given by

$$\hat{\theta}^*_j = \hat{\theta}_{(j)} + N\{\hat{\theta} - \hat{\theta}_{(j)}\}$$

When the mse option is specified, the standard error is estimated as

$$\widehat{\text{se}} = \left\{ \frac{N-1}{N} \sum_{j=1}^{N} (\hat{\theta}_{(j)} - \hat{\theta})^2 \right\}^{1/2}$$

and the jackknife estimate is

$$\overline{\theta}_{(.)} = \frac{1}{N} \sum_{j=1}^{N} \hat{\theta}_{(j)}$$

Otherwise, the standard error is estimated as

$$\widehat{\text{se}} = \left\{ \frac{1}{N(N-1)} \sum_{j=1}^{N} (\hat{\theta}^*_j - \overline{\theta}^*)^2 \right\}^{1/2} \qquad \overline{\theta}^* = \frac{1}{N} \sum_{j=1}^{N} \hat{\theta}^*_j$$

where $\overline{\theta}^*$ is the jackknife estimate. The variance–covariance matrix is similarly computed.

References

Belotti, F., and F. Peracchi. 2020. Fast leave-one-out methods for inference, model selection, and diagnostic checking. *Stata Journal* 20: 785–804.

Brillinger, D. R. 2002. John W. Tukey: His life and professional contributions. *Annals of Statistics* 30: 1535–1575. https://doi.org/10.1214/aos/1043351246.

Canette, I. 2014. Using resampling methods to detect influential points. *The Stata Blog: Not Elsewhere Classified.* https://blog.stata.com/2014/05/08/using-resampling-methods-to-detect-influential-points/.

Mooney, C. Z., and R. D. Duval. 1993. *Bootstrapping: A Nonparametric Approach to Statistical Inference*. Newbury Park, CA: Sage.

Mosteller, C. F., and J. W. Tukey. 1977. *Data Analysis and Regression: A Second Course in Statistics*. Reading, MA: Addison–Wesley.

Overgaard, M., P. K. Andersen, and E. T. Parner. 2023. Pseudo-observations in a multistate setting. *Stata Journal* 23: 491–517.

Tukey, J. W. 1958. Bias and confidence in not-quite large samples. Abstract in *Annals of Mathematical Statistics* 29: 614. https://doi.org/10.1214/aoms/1177706647.

Also see

[R] **jackknife postestimation** — Postestimation tools for jackknife

[R] **bootstrap** — Bootstrap sampling and estimation

[R] **permute** — Permutation tests

[R] **simulate** — Monte Carlo simulations

[SVY] **svy jackknife** — Jackknife estimation for survey data

[U] 13.5 Accessing coefficients and standard errors

[U] 13.6 Accessing results from Stata commands

[U] 20 Estimation and postestimation commands