

estat classification — Classification statistics and table

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Syntax

```
estat classification [if] [in] [weight] [, options]
```

options

Description

Main

<code>all</code>	display summary statistics for all observations in the data
<code>cutoff(#)</code>	positive outcome threshold; default is <code>cutoff(0.5)</code>

`fweights` are allowed; see [U] 11.1.6 [weight](#).

`estat classification` is not appropriate after the `svy` prefix.

Menu for estat

Statistics > Postestimation > Reports and statistics

Description

`estat classification` reports various summary statistics, including the classification table.

`estat classification` requires that the current estimation results be from `logistic`, `logit`, `probit`, or `ivprobit`; see [R] [logistic](#), [R] [logit](#), [R] [probit](#), or [R] [ivprobit](#).

Options

Main

`all` requests that the statistic be computed for all observations in the data, ignoring any `if` or `in` restrictions specified by the estimation command.

`cutoff(#)` specifies the value for determining whether an observation has a predicted positive outcome. An observation is classified as positive if its predicted probability is $\geq \#$. The default is 0.5.

Remarks and examples

stata.com

`estat classification` presents the classification statistics and classification table after `logistic`, `logit`, `probit`, or `ivprobit`.

Statistics are produced either for the estimation sample (the default) or for any set of observations. When weights, `if`, or `in` is used with the estimation command, it is not necessary to repeat the qualifier when you want statistics computed for the estimation sample. Specify `if`, `in`, or the `all` option only when you want statistics computed for a set of observations other than the estimation sample. Specify weights only when you want to use a different set of weights.

▷ Example 1

We illustrate estat classification after logistic; see [R] [logistic](#).

```
. use http://www.stata-press.com/data/r13/lbw
(Hosmer & Lemeshow data)
```

```
. logistic low age lwt i.race smoke ptl ht ui
(output omitted)
```

```
. estat classification
```

```
Logistic model for low
```

Classified	True		Total
	D	~D	
+	21	12	33
-	38	118	156
Total	59	130	189

```
Classified + if predicted Pr(D) >= .5
```

```
True D defined as low != 0
```

Sensitivity	Pr(+ D)	35.59%
Specificity	Pr(- ~D)	90.77%
Positive predictive value	Pr(D +)	63.64%
Negative predictive value	Pr(~D -)	75.64%

False + rate for true ~D	Pr(+ ~D)	9.23%
False - rate for true D	Pr(- D)	64.41%
False + rate for classified +	Pr(~D +)	36.36%
False - rate for classified -	Pr(D -)	24.36%

Correctly classified	73.54%
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The overall rate of correct classification is estimated to be 73.54, with 90.77% of the normal weight group correctly classified (specificity) and only 35.59% of the low weight group correctly classified (sensitivity). Classification is sensitive to the relative sizes of each component group, and always favors classification into the larger group. This phenomenon is evident here.

By default, estat classification uses a cutoff of 0.5, although you can vary this with the cutoff() option. You can use the lsens command to review the potential cutoffs; see [R] [lsens](#).

◀

Stored results

estat classification stores the following in r():

Scalars

r(P_corr)	percent correctly classified
r(P_p1)	sensitivity
r(P_n0)	specificity
r(P_p0)	false-positive rate given true negative
r(P_n1)	false-negative rate given true positive
r(P_1p)	positive predictive value
r(P_0n)	negative predictive value
r(P_0p)	false-positive rate given classified positive
r(P_1n)	false-negative rate given classified negative

Methods and formulas

Let j index observations. Define c as the `cutoff()` specified by the user or, if not specified, as 0.5. Let p_j be the predicted probability of a positive outcome and y_j be the actual outcome, which we will treat as 0 or 1, although Stata treats it as 0 and non-0, excluding missing observations.

A prediction is classified as *positive* if $p_j \geq c$ and otherwise is classified as *negative*. The classification is *correct* if it is *positive* and $y_j = 1$ or if it is *negative* and $y_j = 0$.

Sensitivity is the fraction of $y_j = 1$ observations that are correctly classified. *Specificity* is the percentage of $y_j = 0$ observations that are correctly classified.

References

- Hosmer, D. W., Jr., S. A. Lemeshow, and R. X. Sturdivant. 2013. *Applied Logistic Regression*. 3rd ed. Hoboken, NJ: Wiley.
- Kohler, U., and F. Kreuter. 2012. *Data Analysis Using Stata*. 3rd ed. College Station, TX: Stata Press.

Also see

- [R] **logistic** — Logistic regression, reporting odds ratios
- [R] **logit** — Logistic regression, reporting coefficients
- [R] **probit** — Probit regression
- [R] **ivprobit** — Probit model with continuous endogenous regressors
- [R] **iroc** — Compute area under ROC curve and graph the curve
- [R] **lsens** — Graph sensitivity and specificity versus probability cutoff
- [R] **estat gof** — Pearson or Hosmer–Lemeshow goodness-of-fit test
- [R] **roc** — Receiver operating characteristic (ROC) analysis
- [U] **20 Estimation and postestimation commands**