Isens - Graph sensitivity and specificity versus probability cutoff

Description	Quick start
Options	Remarks and examples
Reference	Also see

Menu Stored results Syntax Methods and formulas

# Description

lsens graphs sensitivity and specificity versus probability cutoff and optionally creates new variables containing these data.

lsens requires that the current estimation results be from logistic, logit, probit, or ivprobit; see [R] logistic, [R] logit, [R] probit, or [R] ivprobit.

# **Quick start**

Graph sensitivity and specificity versus a probability cutoff using current estimation results lsens

Generate variables v1, v2, and v3 to contain probability cutoffs, sensitivity, and specificity lsens, genprob(v1) gensens(v2) genspec(v3)

Add "My Title" to graph

lsens, genprob(v1) gensens(v2) genspec(v3) title(My Title)

## Menu

 $Statistics > Binary \ outcomes > Postestimation > Sensitivity/specificity \ plot$ 

## Syntax

lsens [depvar] [if] [in] [weight] [, options]

options	Description
Main	
all	graph all observations in the data
genprob( <i>varname</i> )	create variable containing probability cutoffs
gensens(varname)	create variable containing sensitivity
genspec(varname)	create variable containing specificity
replace	overwrite existing variables
$\underline{nog}$ raph	suppress the graph
Advanced	
<pre>beta(matname)</pre>	row vector containing model coefficients
Plot	
connect_options	affect rendition of the plotted points connected by lines
Add plots	
addplot( <i>plot</i> )	add other plots to the generated graph
Y axis, X axis, Titles, Legend, Ove	rall
twoway_options	any options other than by() documented in [G-3] twoway_options

collect is allowed; see [U] 11.1.10 Prefix commands.

fweights are allowed; see [U] 11.1.6 weight.

lsens is not appropriate with svy estimation results.

# 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.
- genprob(*varname*), gensens(*varname*), and genspec(*varname*) specify the names of new variables created to contain, respectively, the probability cutoffs and the corresponding sensitivity and specificity.
- replace requests that existing variables specified for genprob(), gensens(), or genspec() be overwritten.

nograph suppresses graphical output.

∫ Advanced ]

beta (*matname*) specifies a row vector containing model coefficients. The columns of the row vector must be labeled with the corresponding names of the independent variables in the data. The dependent variable *depvar* must be specified immediately after the command name. See *Models other than the last fitted model* later in this entry.

#### Plot

*connect\_options* affect the rendition of the plotted points connected by lines; see *connect\_options* in [G-2] graph twoway scatter.

Add plots

addplot (plot) provides a way to add other plots to the generated graph. See [G-3] addplot\_option.

Y axis, X axis, Titles, Legend, Overall

*twoway\_options* are any of the options documented in [G-3] *twoway\_options*, excluding by(). These include options for titling the graph (see [G-3] *title\_options*) and for saving the graph to disk (see [G-3] *saving\_option*).

## **Remarks and examples**

Remarks are presented under the following headings:

Introduction Models other than the last fitted model

#### Introduction

lsens plots sensitivity and specificity; it plots both sensitivity and specificity versus probability cutoff c. The graph is equivalent to what you would get from estat classification (see [R] estat classification) if you varied the cutoff probability c from 0 to 1.

#### Example 1

We illustrate lsens after logistic; see [R] logistic.



1sens optionally creates new variables containing the probability cutoff, sensitivity, and specificity.

. lsens, genprob(p) gensens(sens) genspec(spec) nograph

The variables created will have M + 2 distinct nonmissing values: one for each of the M covariate patterns, one for c = 0, and another for c = 1. Values are recorded for p = 0, for each of the observed predicted probabilities, and for p = 1. The total number of observations required to do this can be fewer than \_N, the same as \_N, or \_N + 1, or \_N + 2. If more observations are added, they are added at the end of the dataset and the values of the original variables are set to missing in the added observations. How the values added align with existing observations is irrelevant.

4

#### Technical note

logistic, logit, probit, or ivprobit and lsens keep track of the estimation sample. If you type, for instance, logistic ... if x==1, then when you type lsens, the statistics will be calculated on the x==1 subsample of the data automatically.

You should specify if or in with lsens only when you wish to produce graphs and calculate statistics for a set of observations other than the estimation sample.

If the logistic model was fit with fweights, lsens properly accounts for the weights in its calculations. You do not have to specify the weights when you run lsens. Weights should be specified with lsens only when you wish to use a different set of weights.

#### Models other than the last fitted model

By default, lsens uses the last model fit. You may also directly specify the model to lsens by inputting a vector of coefficients with the beta() option and passing the name of the dependent variable *depvar* to lsens.

#### Example 2

Suppose that someone publishes the following logistic model of low birthweight:

 $\Pr(\texttt{low} = 1) = F(-0.02 \text{ age} - 0.01 \text{ lwt} + 1.3 \text{ black} + 1.1 \text{ smoke} + 0.5 \text{ ptl} + 1.8 \text{ ht} + 0.8 \text{ ui} + 0.5)$ 

where F is the cumulative logistic distribution. These coefficients are not odds ratios; they are the equivalent of what logit produces.

We can see whether this model fits our data. First, we enter the coefficients as a row vector and label its columns with the names of the independent variables plus \_cons for the constant (see [P] matrix define and [P] matrix rownames).

. use https://www.stata-press.com/data/r19/lbw3, clear
(Hosmer & Lemeshow data)
. matrix input b = (-0.02, -.01, 1.3, 1.1, .5, 1.8, .8, .5)

- . materia input 5 (0.02, 101, 1.0, 1.1, 10, 1.0, 10, 10
- . matrix colnames b = age lwt black smoke ptl ht ui \_cons

We can use lroc (see [R] lroc) to examine the predictive ability of the model:

```
. lroc low, beta(b) nograph
Logistic model for low
Number of observations = 189
Area under ROC curve = 0.7275
```

The area under the curve indicates that this model does have some predictive power. We can obtain a graph of sensitivity and specificity as a function of the cutoff probability by typing



## Stored results

lsens stores the following in r():

Scalars r(N)

number of observations

### Methods and formulas

Let j index observations and c be the cutoff probability. 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  $p_j \ge c$  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.

### Reference

Hosmer, D. W., Jr., S. A. Lemeshow, and R. X. Sturdivant. 2013. Applied Logistic Regression. 3rd ed. Hoboken, NJ: Wiley.

## 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 covariates
- [R] lroc Compute area under ROC curve and graph the curve
- [R] estat classification Classification statistics and table
- [R] estat gof Pearson or Hosmer–Lemeshow goodness-of-fit test
- [R] roc Receiver operating characteristic (ROC) analysis
- [U] 20 Estimation and postestimation commands

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