

Classification (Confusion) Matrices

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1 Introduction

Logistic regression might be considered to be a *classification algorithm*, as logistic regression provides predicted probabilities of an outcome. An important part of using any classification algorithm is evaluating the strength of the classification.

Classification matrices, sometimes (confusingly) called *confusion matrices*, provide a mechanism for evaluating many different statistical and machine learning methods.

2 Data

We use data from Hosmer, Lemeshow, and Sturdivant (2013) provided by Stata corporation.

```
use https://www.stata-press.com/data/r18/lbw
```

(Hosmer & Lemeshow data)

3 Describe The Data

```
describe // describe the data
```

Running /Users/agrogan/Desktop/GitHub/newstuff/categorical/classification/profile.do

> ...

Contains data from https://www.stata-press.com/data/r18/lbw.dta

Observations:	189	Hosmer & Lemeshow data
Variables:	11	15 Jan 2022 05:01

Variable name	Storage type	Display format	Value label	Variable label
id	int	%8.0g		Identification code
low	byte	%8.0g		Birthweight<2500g
age	byte	%8.0g		Age of mother
lwt	int	%8.0g		Weight at last menstrual period
race	byte	%8.0g	race	Race
smoke	byte	%9.0g	smoke	Smoked during pregnancy
ptl	byte	%8.0g		Premature labor history (count)
ht	byte	%8.0g		Has history of hypertension
ui	byte	%8.0g		Presence, uterine irritability
ftv	byte	%8.0g		Number of visits to physician during 1st trimester
bwt	int	%8.0g		Birthweight (grams)

Sorted by:

4 Use Logistic Regression To Predict Low Birthweight

```
logit low age lwt i.race smoke ptl ht ui, or // logistic regression
```

Running /Users/agrogan/Desktop/GitHub/newstuff/categorical/classification/profile.do

> ...

Iteration 0: Log likelihood = -117.336
Iteration 1: Log likelihood = -101.28644
Iteration 2: Log likelihood = -100.72617
Iteration 3: Log likelihood = -100.724
Iteration 4: Log likelihood = -100.724

Logistic regression

Number of obs = 189
LR chi2(8) = 33.22
Prob > chi2 = 0.0001
Pseudo R2 = 0.1416

Log likelihood = -100.724

	low	Odds ratio	Std. err.	z	P> z	[95% conf. interval]	

	age	.9732636	.0354759	-0.74	0.457	.9061578	1.045339
	lwt	.9849634	.0068217	-2.19	0.029	.9716834	.9984249
	race						
	Black	3.534767	1.860737	2.40	0.016	1.259736	9.918406
	Other	2.368079	1.039949	1.96	0.050	1.001356	5.600207
	smoke	2.517698	1.00916	2.30	0.021	1.147676	5.523162
	ptl	1.719161	.5952579	1.56	0.118	.8721455	3.388787
	ht	6.249602	4.322408	2.65	0.008	1.611152	24.24199
	ui	2.1351	.9808153	1.65	0.099	.8677528	5.2534
	_cons	1.586014	1.910496	0.38	0.702	.1496092	16.8134

Note: _cons estimates baseline odds.

5 Classification Matrix

The quantities of interest will often depend upon your discipline, and upon the specific research question.

However, the **overall accuracy (correctly classified)**, **sensitivity**, **specificity** and **positive predictive value** will often be of general interest.

```
estat classification // classification matrix
```

Running /Users/agrogan/Desktop/GitHub/newstuff/categorical/classification/profile.do

```
> ...
```

Logistic model for low

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

Classified + if predicted $\Pr(D) \geq .5$

True D defined as low != 0

Sensitivity	$\Pr(+ D)$	35.59%
Specificity	$\Pr(- \sim D)$	90.77%
Positive predictive value	$\Pr(D +)$	63.64%
Negative predictive value	$\Pr(\sim D -)$	75.64%
False + rate for true ~D	$\Pr(+ \sim D)$	9.23%
False - rate for true D	$\Pr(- D)$	64.41%
False + rate for classified +	$\Pr(\sim D +)$	36.36%
False - rate for classified -	$\Pr(D -)$	24.36%
Correctly classified		73.54%

References

Hosmer, David W, Stanley Lemeshow, and Rodney X Sturdivant. 2013. *Applied Logistic Regression*. *Applied Logistic Regression*. Third edition. Wiley.