

# Multiclasstesting Vignette

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## 1 The problem

This is an introduction to the `Multiclasstesting` package in R. Specificity, sensitivity, negative and positive predictive value are used in combination to quantify different aspects of the accuracy of a binary test, evaluating different proportions of correctly and incorrectly classified items, when compared to a known classification, considered the gold standard. In this context the *test* is the ensemble of all the operations performed to classify each item; *positive* and *negatives* label the items according to the two classes ( $c = 0, 1$ ) they belong to; *true* (T) and *false* (F) represent the ability of the test to classify coherently or not a given item in the test classification with respect to the gold standard classification. These concepts are usually formalized with the relationships in the left hand-side of Equations 1.

$$\begin{aligned} PPV &= TP / (TP + FP) = TP / P_t \\ NPV &= TN / (TN + FN) = TN / N_t \\ Se &= TP / (TP + FN) = TP / P_{gs} \\ Sp &= TN / (TN + FP) = TN / N_{gs} \end{aligned} \tag{1}$$

When the test classifies  $n > 2$  categories, these definitions become more complex to apply. In fact, the meaning of *positive* and *negative* is not relevant anymore, since there are now *positives*. Then, while the definition of *true* remains straightforward, as it indicates coherence between the classification of the test and the gold standard, the definition of *false* can be cumbersome, since there are  $n - 1$  ways to misclassify an item. To avoid confusion and ambiguities the actual values of all false can be identified by rewriting the problem in terms of a system of equation based on the relationships indicated in Table 1. Here  $P_t, N_t$  represent the total number of positive and negative items that can be found in the test ( $t$ ) categorization, and  $P_{gs}, N_{gs}$  in the gold standard ( $gs$ ) classification. The definitions can be generalized to  $n > 2$  classes changing the term negative and positive with the indices of the corresponding classes  $c = 0, 1, \dots, n$ , and having  $C_c$  to design the total number of positives for each given class. The system of equations obtained from the relationships in the rows and columns of Table 1 contains  $2 \cdot n$  equations (i.e.  $TP + FP = P_t$ ) and  $2 \cdot n$  unknown ( $x_{ij}$ ), thus it is completely specified. It is worth noticing, that with these general definitions, in case of 2-classes test, Se and Sp appear to be dual scores. Thus, when generalizing to  $n$ -classes it is possible to define the predictive ability of the test for each given class  $c \in 0, 1, \dots, n$  as  $PV_c = T_c / C_t$  and the Sensitivity/Specificity (now called S) for the same class  $c$  as  $S_c = T_c / C_{gs}$ . To clarify the situation it is extremely useful to rewrite the definitions as they are written on the right hand-side of Equation 1, namely:

For  $n$  classes this gives:

$$\begin{aligned} PPV &= \sum_c T_c / \sum_c C_{c,t}, c = 1, \dots, n \\ NPV &= T_0 / N_t = T_0 / C_{0,t} \\ Se &= \sum_c T_c / \sum_c C_{c,gs}, c = 1, \dots, n \\ Sp &= T_0 / N_{gs} = T_0 / C_{0,gs} \end{aligned} \tag{2}$$

This package is developed to estimate the performance of both binary and multiple test ( $n$ -ary is used to include both cases). The statistical scores described above are finally calculated as output of the function in this package.

## 2 MultiClassTesting usage

`nclasstest` is the only function in `Multiclasstesting`. It serves for the computation of the performance of  $n$ -class test. In binary case, the output includes the statistical scores, PPV, NPV, Se and Sp. In multiple

**Table 1. Classical definition and generalization to 3 classes for true, false, negatives, positives.**



classes case, the output consists of two parts. One is called `multi.performance`, indicating the details of the predictive value (PV) and Sensitivity/Specificity (S) for each class. The other, called `binary.performance`, is to summarize the PPV, NPV, Se and Sp for the classification operations, as described in Equation 2. This is useful when interested in the ability of the n-ary test to identify positives and negative, globally. A common example of application is the computation of the performances of a gene network algorithm reconstruction: when interested in the directed network each edge can belong to class 1 (direct interaction), class -1 (inverse interaction) or class 0 (no interaction). However, it may be interesting to know how the algorithm performs simply in terms of recognizing the existing connections (edges 1 AND -1) in this case the summary binary performances of the 3 class test are the correct way to compute PPV, NPV, Se, Sp.

For the binary case, R statement is ,

```
> library(Multiclasstesting)
> GS <- cbind(c(0, 1), c(0, 0), c(1, 1))
> T <- cbind(c(1, 1), c(1, 0), c(1, 1))
> Nclasstest(T, GS)
```

```
$binary.performance
      PPV NPV Se      Sp
[1,] 0.6   1   1 0.3333333
```

where, GS and T are the arguments to the function `Nclasstest`, representing the results from Gold Standard and test, respectively. They can be matrices or vectors with the elements labeling the category type, as 0 and 1 in the example above.

For multiple-class test,

```
> library(Multiclasstesting)
> GS <- cbind(c(0, -1, 1), c(0, 1, 0), c(1, 0, 1))
> T <- cbind(c(1, -1, 1), c(0, 1, -1), c(0, 1, 1))
> Nclasstest(T, GS)
```

```
$multi.performance
 class.type PV  S
1          -1 1.00 0.5
2           0 0.25 0.5
3           1 0.75 0.6
```

```
$binary.performance
      PPV NPV Se  Sp
[1,] 0.5714286 0.5 0.8 0.25
```