

**tolerance** — Use and specification of tolerances

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## Description

The results provided by many of the numerical routines in Mata depend on tolerances. Mata provides default tolerances, but those can be overridden.

## Syntax

*somefunction*(..., *real scalar tol*, ...)

where, concerning argument *tol*,

- optional    Argument *tol* is usually optional; not specifying *tol* is equivalent to specifying  $tol = 1$ .
- $tol > 0$     Specifying  $tol > 0$  specifies the amount by which the usual tolerance is to be multiplied:  $tol = 2$  means twice the usual tolerance;  $tol = 0.5$  means half the usual tolerance.
- $tol < 0$     Specifying  $tol < 0$  specifies the negative of the value to be used for the tolerance:  $tol = -1e-14$  means  $1e-14$  is to be used.
- $tol = 0$     Specifying  $tol = 0$  means all numbers are to be taken at face value, no matter how close to 0 they are. The single exception is when *tol* is applied to values that, mathematically, must be greater than or equal to zero. Then negative values (which arise from roundoff error) are treated as if they were zero.

The default tolerance is given by formula, such as

$$eta = 1e-14$$

or

$$eta = \text{epsilon}(1) \quad (\text{see [M-5] } \text{epsilon}())$$

or

$$eta = 1000 * \text{epsilon}(\text{trace}(\text{abs}(A)) / \text{rows}(A))$$

Specifying  $tol > 0$  specifies a value to be used to multiply *eta*. Specifying  $tol < 0$  specifies that  $-tol$  be used in place of *eta*. Specifying  $tol = 0$  specifies that *eta* be set to 0.

The formula for *eta* and how *eta* is used are found under *Remarks and examples*. For instance, the *Remarks and examples* might say that *A* is declared to be singular if any diagonal element of *U* of its LU decomposition is less than or equal to *eta*.

## Remarks and examples

Remarks are presented under the following headings:

*The problem*  
*Absolute versus relative tolerances*  
*Specifying tolerances*

### The problem

In many formulas, zero is a special number in that, when the number arises, sometimes the result cannot be calculated or, other times, something special needs to be done.

The problem is that zero—0.00000000000—seldom arises in numerical calculation. Because of roundoff error, what would be zero were the calculation performed in infinite precision in fact is  $1.03948\text{e-}15$ , or  $-4.4376\text{e-}16$ , etc.

If one behaves as if these small numbers are exactly what they seem to be ( $1.03948\text{e-}15$  is taken to mean  $1.03948\text{e-}15$  and not zero), some formulas produce wildly inaccurate results; see [M-5] `lusolve()` for an example.

Thus routines use *tolerances*—preset numbers—to determine when a number is small enough to be considered to be zero.

The problem with tolerances is determining what they ought to be.

### Absolute versus relative tolerances

Tolerances come in two varieties: absolute and relative.

An absolute tolerance is a fixed number that is used to make direct comparisons. If the tolerance for a particular routine were  $1\text{e-}14$ , then  $8.99\text{e-}15$  in some calculation would be considered to be close enough to zero to act as if it were, in fact, zero, and  $1.000001\text{e-}14$  would be considered a valid, nonzero number.

But is  $1\text{e-}14$  small? The number may look small to you, but whether  $1\text{e-}14$  is small depends on what is being measured and the units in which it is measured. If all the numbers in a certain problem were around  $1\text{e-}12$ , you might suspect that  $1\text{e-}14$  is a reasonable number.

That leads to relative measures of tolerance. Rather than treating, say, a predetermined quantity as being so small as to be zero, one specifies a value (for example,  $1\text{e-}14$ ) multiplied by something and uses that as the definition of small.

Consider the following matrix:

$$\begin{bmatrix} 5.5\text{e-}15 & 1.2\text{e-}16 \\ 1.3\text{e-}16 & 6.4\text{e-}15 \end{bmatrix}$$

What is the rank of the matrix? One way to answer that question would be to take the LU decomposition of the matrix and then count the number of diagonal elements of  $U$  that are greater than zero. Here, however, we will just look at the matrix.

The absolutist view is that the matrix is full of roundoff error and that the matrix is really indistinguishable from the matrix

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

The matrix has rank 0. The relativist view is that the matrix has rank 2 because, other than a scale factor of  $1e-16$ , the matrix is indistinguishable from

$$\begin{bmatrix} 55.0 & 1.2 \\ 1.3 & 64.0 \end{bmatrix}$$

There is no way this question can be answered until someone tells you how the matrix arose and the units in which it is measured.

Nevertheless, most Mata routines would (by default) adopt the relativist view: the matrix is of full rank. That is because most Mata routines are implemented using relative measures of tolerance, chosen because Mata routines are mostly used by people performing statistics, who tend to make calculations such as  $X'X$  and  $X'Z$  on data matrices, and those resulting matrices can contain very large numbers. Such a matrix might contain

$$\begin{bmatrix} 5.5e+14 & 1.2e+12 \\ 1.3e+13 & 2.4e+13 \end{bmatrix}$$

Given a matrix with such large elements, one is tempted to change one's view as to what is small. Calculate the rank of the following matrix:

$$\begin{bmatrix} 5.5e+14 & 1.2e+12 & 1.5e-04 \\ 1.3e+13 & 2.4e+13 & 2.8e-05 \\ 1.3e-04 & 2.4e-05 & 8.7e-05 \end{bmatrix}$$

This time, we will do the problem correctly: we will take the LU decomposition and count the number of nonzero entries along the diagonal of  $U$ . For the above matrix, the diagonal of  $U$  turns out to be (5.5e+14, 2.4e+13, 0.000087).

An absolutist would tell you that the matrix is of full rank; the smallest number along the diagonal of  $U$  is 0.000087 (8.7e-5), and that is still a respectable number, at least when compared with computer precision, which is about  $2.22e-16$  (see [\[M-5\] `epsilon\(\)`](#)).

Most Mata routines would tell you that the matrix has rank 2. Numbers such as 0.000087 may seem respectable when compared with machine precision, but 0.000087 is, relatively speaking, a very small number, being about  $4.6e-19$  relative to the average value of the diagonal elements.

## Specifying tolerances

Most Mata routines use relative tolerances, but there is no rule. You must read the documentation for the function you are using.

When the tolerance entry for a function directs you here, [\[M-1\] `tolerance`](#), then the tolerance works as summarized under *Syntax* above. Specify a positive number, and that number multiplies the default; specify a negative number, and the corresponding positive number is used in place of the default.

## Also see

[\[M-5\] `epsilon\(\)`](#) — Unit roundoff error (machine precision)

[\[M-5\] `solve\_tol\(\)`](#) — Tolerance used by solvers and inverters

[\[M-1\] `intro`](#) — Introduction and advice