

# $\mu Toss$ Quick Start Guide

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## 1 Introduction

The  $\mu Toss$  packages allow the user to discover, apply and compare multiple testing procedures and multiple interval estimation procedures.

They include a corpus of functions implementing and integrating these procedures and a GUI. These are found in the *mutoss* and *mutossGUI* packages respectively.

## 2 $\mu Toss$ Rationale

The rationale behind the  $\mu Toss$  packages is two-fold.

It is aimed at allowing statisticians to discover, apply and compare standard and custom multiplicity controlling procedures. This is achieved by the *mutoss* package which includes:

- **Mutoss Class:** An S4 class object designed to hold the input and output of various multiple testing procedures and to serve as a standard input and output of future procedures.
- **Elementary Functions:** A mass of functions either implementing a procedure or calling an existing implementation. These are used by both the GUI and the command line interface.
- **Simulation Platform:** *simulation( )* and *gatherStatistics( )* functions allowing the mass generation of data, application of different procedures and analysis of results (documented separately).

$\mu Toss$  is also aimed at the researcher wishing to analyze new data or to reproduce published results. This is accomplished by the *mutossGUI* package.

At the time of release, the package has only undergone basic testing. This being the case, we recommend new data to be analyzed with standard software alongside $\mu Toss$ . This is planned to change in future releases.
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## 3 System Requirements

### 3.1 mutoss Package

The package will run on any machine running R with recommended version 2.8 and above.

### 3.2 mutossGUI package

On top of the mutoss package requirements, Java Run time Environment ver 5 and above is needed.

## 4 GUI Work flow

Download and install the *mutossGUI* package. The GUI should start automatically. Otherwise load it with

```
>mutossGUI()
```

### 4.1 Testing of Hypotheses

If you have already a vector of p-values start at step (5).

1. Load the raw data (assumed to be a *data.frame* object) using the **Data** button.
2. Specify the model type and explanatory variables using the **Model** button.  
For linear contrasts choose **Single endpoint in k groups**.  
For applying the one single model to many response variables choose **Multiple endpoints in k groups**.
3. Define model by choosing response and explanatory variables.
4. Define the hypotheses of interest by specifying the contrasts using the **Hypotheses** button.
5. Insert p-values using the **p-Value** button or calculate them following the previous steps.
6. Choose the error type to control using the **Error Rate** button.
7. Use the **Adjusted p-Values** to calculate the procedure specific adjusted p-values (you will be prompted for additional options when necessary) or choose **Rejected** to apply the procedure and reject hypotheses.
8. Visualize results by choosing the **Info** option in the **Adjusted p-Values** or **Rejected** buttons.

9. Save the output as an R object using the **File->Export MuToss Object to R** option.

Further analysis is now possible using the `compareMutoss` functions or other R functionality.

## 4.2 Interval Estimations

Steps 1-4 are identical to the hypothesis testing work flow.

1. Load the raw data (assumed to be a *data.frame* object) using the **Data** button.
2. Specify the model type and explanatory variables using the **Model** button.  
For linear contrasts choose **Single endpoint in k groups**.  
For applying the one single model to many response variables choose **Multiple endpoints in k groups**.
3. Define model by choosing response and explanatory variables.
4. Define the contrasts of interest by specifying the contrasts using the **Hypotheses** button.
5. Choose the error type to control using the **Error Rate** button.
6. Use the **Confidence Intervals** to compute confidence intervals on parameters of interest.
7. Visualize results by choosing the **Info** option in the **Confidence Intervals** button.
8. Save the output as an R object using the **File->Export MuToss Object to R** option.

Further analysis is now possible using R functionality.

## 5 Command Line Work Flow

Download and install the *mutoss* package to access the different procedures in the package (note *mutossGUI* is not needed for this purpose). A list can presented using

```
>help(package='mutoss')
```

To work with these elementary functions, just use them as any other R function. See in line help for further details.

To use these functions to read and write into Mutoss S4 class objects use the *mutoss.apply()* function. See the in line help of the function for further details.

## 6 Glossary

**Hypotheses-Testing-Procedures** The corpus of procedures for testing multiple statistical hypotheses.

**Interval-Estimating-Procedures** The corpus of procedures for constructing interval estimates for multiple parameters.

**p-Value-Procedures** The corpus of (multiple) hypotheses testing procedures which rely on the marginal p-values of each hypothesis (and do not require the original data and model). This procedures might possibly require additional information such as logical relations between procedures, a qualitative description of the probabilistic relation between test statistics etc.

**Data-Procedures** The corpus of (multiple) testing procedures which require the original response variables, the explanatory variables (model) and the parameters of interest (contrasts).  
These procedures can be seen as p-value-procedures with a specific relation between test-statistics which is derived from the model and the contrasts.

**Error-Type** The type of error a procedure aims to control. This can be a hypothesis testing error rate (FWER, FDR,...) or an interval estimation error rate (simultaneous coverage, false coverage rate,...).

**Error-Rate** The allowed rate of the *Error Type*.