# R-package "seqampl" A Short Introduction

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This document describes briefly how to use the R-package which implements the algorithm for "Sequential implementation of monte carlo tests with uniformly bounded resampling risk." based on Gandy [2009].

## 1 Installation

The installation is as for most R-packages that do not reside in CRAN. The general procedure is described in the Section 6 on "Add-on packages" in the R Manual on Istallation and Administration:

http://cran.r-project.org/doc/manuals/R-admin.html.

The following is merely an adaptation of those procedures to our package.

### 1.1 Linux/Unix

If you do not have write access to the package repository:

- 1. Download the package "simctest\_1.0-0.tar.gz" and place it into your home directory.
- 2. Issue the following commands:

```
echo ".libPaths(\"$HOME/Rlibrary\")" >$HOME/.Rprofile
R CMD INSTALL -L $HOME/Rlibrary simctest_1.0-0.tar.gz
```

3. You may now delete the file "simctest\_1.0-0.tar.gz".

## 2 Usage

Obviously, the pacakge is loaded by typing

> library(simctest)

This document can be accessed via

#### > vignette("simctest-intro")

Documentation of the most useful command can be obtained as follows:

#### > ? simctest

The following is an artificial example. By default the algorithm will report back after at most 10000 steps, work with a threshold of  $\alpha=0.05$  and use the spending sequence

$$\epsilon_n = 0.001 \frac{n}{1000 + n}.$$

A simple example of a test with true p-value 0.07.

> res <- simctest(function() runif(1) < 0.07)</pre>

> res

p.value: 0.06610067
Number of samples: 3298

One can also obtain a confidence interval (wrt the resampling procedure) of the computed p-value. By default a 95% confidence interval is computed.

> confint(res)

2.5 % 97.5 % p.value 0.05613075 0.07381973

#### 2.1 Behaviour at the Threshold

Next, consider an example where the true p-value is precisely equal to the threshold  $\alpha$ . Here, we will expect that the algorithm stops only with probability  $2\epsilon = 0.002$ . If the algorithm has not stopped after 10000 steps the algorithm will return.

> res <- simctest(function() runif(1) < 0.05)</pre>

> res

No decision reached.

Final estimate will be in [ 0.04159918 , 0.05952146 ]

Current estimate of the p.value: 0.0488

Number of samples: 10000

Note that a part of the output it the interval in which the final estimator will lie.

One can always take a few more steps

> res <- cont(res, 10000)

> res

No decision reached.

Final estimate will be in [ 0.04367672 , 0.056915 ]

Current estimate of the p.value: 0.0488

Number of samples: 20000

## 2.2 A simple bootstrap test

```
An example from [Davison and Hinkley, 1997, section 11.4, p. 534]:
> data(fir, package = "boot")
> fir.mle <- c(sum(fir$count), nrow(fir))</pre>
> fir.gen <- function(data, mle) {</pre>
      d <- data
      y <- sample(x = mle[2], size = mle[1], replace = TRUE)
      d$count <- tabulate(y, mle[2])</pre>
+ }
> fir.fun <- function(data) (nrow(data) - 1) * var(data$count)/mean(data$count)
> resampl <- function() {</pre>
      obs < fir.fun(fir.gen(data = fir, mle = fir.mle))</pre>
+ }
> obs <- fir.fun(fir)</pre>
> simctest(resampl)
p.value: 0.3809524
Number of samples: 21
2.3
      Computing the power of a test
> n <- 10
> system.time(replicate(1000, {
      obs \leftarrow mean(rnorm(n) + 0.01)
      simctest(function() mean(rnorm(n)) > obs, maxsteps = 1000)
+ }))
  user system elapsed
  8.947
        0.019
                   9.035
Compared with the naive approach:
> system.time(replicate(1000, {
      obs \leftarrow mean(rnorm(n) + 0.01)
      mean(replicate(1000, mean(rnorm(n)) > obs))
+ }))
   user system elapsed
          0.080 31.594
 30.610
  To reduce the overhead of computing the boundaries, they can be pre-
computed.
> alg <- getalgprecomp()</pre>
> system.time(replicate(1000, {
```

```
obs \leftarrow mean(rnorm(n) + 0.01)
      run(alg, function() mean(rnorm(n)) > obs, maxsteps = 1000)
+ }))
  user system elapsed
  7.414
         0.021
                 7.514
  For comparison purposes, the same without without pre-computation:
> alg <- getalgonthefly()</pre>
> system.time(replicate(1000, {
      obs \leftarrow mean(rnorm(n) + 0.01)
      run(alg, function() mean(rnorm(n)) > obs, maxsteps = 1000)
+ }))
  user system elapsed
  7.447
         0.009
                   7.811
```

## References

A.C. Davison and D.V. Hinkley. Bootstrap methods and their application. Cambridge University Press, 1997.

Axel Gandy. Sequential implementation of Monte Carlo tests with uniformly bounded resampling risk. To appear in JASA, 2009.