

SOFTWARE

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An R package that automatically collects and archives details for reproducible computing

Zhifa Liu and Stan Pounds*

Abstract

Background: It is scientifically and ethically imperative that the results of statistical analysis of biomedical research data be computationally reproducible in the sense that the reported results can be easily recapitulated from the study data. Some statistical analyses are computationally a function of many data files, program files, and other details that are updated or corrected over time. In many applications, it is infeasible to manually maintain an accurate and complete record of all these details about a particular analysis.

Results: Therefore, we developed the `rctrack` package that automatically collects and archives *read only* copies of program files, data files, and other details needed to computationally reproduce an analysis.

Conclusions: The `rctrack` package uses the `trace` function to temporarily embed detail collection procedures into functions that read files, write files, or generate random numbers so that no special modifications of the primary R program are necessary. At the conclusion of the analysis, `rctrack` uses these details to automatically generate a *read only* archive of data files, program files, result files, and other details needed to recapitulate the analysis results. Information about this archive may be included as an appendix of a report generated by `Sweave` or `knitr`. Here, we describe the usage, implementation, and other features of the `rctrack` package. The `rctrack` package is freely available from <http://www.stjuderesearch.org/site/depts/biostats/rctrack> under the GPL license.

Background

The ability to reproduce research results is a cornerstone of the scientific method. Research results that are not reproducible are generally considered invalid by the scientific community. The result of a laboratory investigation must be independently reproduced by others to be considered valid. Similarly, the validity of a clinical research finding is established via recapitulation in multiple distinct clinical research cohorts.

The exciting advances in data collection biotechnologies (microarrays, sequencing, proteomics, etc) present the biomedical research community with special challenges related to data interpretation and reproducibility of research results. The interpretation of massive data sets requires the use of sophisticated computational and statistical analysis methods. The data sets themselves evolve as new technologies are introduced, additional tissue samples become available, clinical follow-up is updated, etc.

A specific analysis result may be a function of dozens of data files, program files, and other details. Without carefully tracking details about how those files are used to generate a specific result, it can be challenging if not impossible to describe how any particular research result can be recapitulated from the study data. This challenge represents a crisis for reproducibility as a pillar of the scientific method. Consequently, some journals are developing policies to encourage computational reproducibility [1] and federal authority to audit the computational reproducibility of research results is being expanded [2]. Most recently, the National Cancer Institute issued guidelines that strongly encourage investigators to maintain high standards of computational reproducibility in their ‘omics’ studies [3,4].

Also, some recent mishaps in clinical cancer genomics research have shown that there is an ethical obligation to ensure that computational results are fully reproducible. The first indication of problems in a series of clinical trials was the inability to computationally reproduce the results of the supporting scientific publications from the publicly available data sets [5,6]. Investigative

*Correspondence: stanley.pounds@stjude.org

Department of Biostatistics, St Jude Children’s Research Hospital,
Memphis TN 38105, USA

follow-up of the analysis result discrepancies showed that the supporting science was fundamentally flawed [7]. Subsequently, the publications have been retracted and the clinical trials closed [8]. Thus, it is ethically imperative that data analysis results be computationally reproducible before they are used to direct therapy in a clinical trial [2,9].

Several computational tools have been developed to facilitate reproducible computing [10]. For example, *Sweave* [11] uses R (www.r-project.org) to compute statistical results and inserts them into a *LAT_EX* (<http://www.latex-project.org/>) typesetting program that is subsequently converted into a PDF report. To enable this functionality, *Sweave* defines a special syntax to switch between R code and *LAT_EX* code within the same file. In this way, the *Sweave* program file directly documents the top-level R code used to generate the corresponding PDF report. Similarly, the R packages *knitr* [12] and *lazyWeave* [13] enable one to computationally insert results determined with R into a Wiki mark-up file and an open office document file, respectively. These literate programming [14] systems internally document exactly how the results in the report file were produced. These literate programming environments significantly enhance the scientific community's ability to perform reproducible computing.

However, literate programming systems fail to address other challenges to reproducible computing. As previously mentioned, one analysis result may be a function of a very large number of input data files (such as microarray image files, sequence alignment files, etc) and computer program files (such as the primary analysis program, some locally developed R routines, specific versions of R packages, etc). These files must be collected and archived to ensure that the result may be reproduced at a later date. Manually reading through a program to identify and collect all those files is very tedious and cumbersome. Thus, there is a need to computationally collect and archive those files.

Some tools have been developed to support reproducible computing by automatically archiving files. CDE software (www.pgbovine.net/cde.html) automatically generates an archive folder that replicates the entire directory structure for every file used to execute a specific Linux command so that the identical command can be executed on another Linux machine without any conflicts [15]. The technical completeness of the CDE archive is very impressive and CDE is very helpful for many Linux users who wish to ensure the computational reproducibility of their work.

Nevertheless, CDE has several limitations and drawbacks. Obviously, CDE is not helpful for people who use operating systems other than Linux and Unix. Also, the archive folder of CDE is excessively redundant for some

settings. For example, the CDE archive includes a copy of many installation files for every software program (R, MatLab, SAS, Java, etc) used to perform an analysis. As CDE is used to document many analyses for multiple projects by a group of users that all have access to the same programs (such as a department of biostatistics or computational biology), the redundancy of such program files in CDE archives will begin to unnecessarily strain the storage resources of the computing infrastructure. This redundancy is amplified when the same large data files (such as genotype data files for genome-wide association studies) are present in multiple CDE archives. Additionally, collecting a large number of copies of proprietary software files in many CDE archives may inadvertently pose some legal problems. Furthermore, although the CDE archive is complete from a technical computing perspective, it still does not necessarily contain all the information necessary to exactly recapitulate the results of statistical analysis procedures such as bootstrapping, permutation, and simulation that rely on random number generation [16]. The initial seed and the random number generator must be retained for such analyses to be fully reproducible and such information will not be stored in the CDE archive if the program files did not explicitly set the value for the initial seed. Finally, CDE does not automatically generate any file to help a reviewer to understand the computational relationships among files such as indicating that specific program files generated specific result files. Understanding such computational relationships is necessary to critically evaluate the appropriateness of the methods and the scientific validity of the results.

We have developed the package **rctrack** (Additional file 1, also available from our website listed below) to automatically archive data files, code files, and other details to support reproducible computing for R software that is widely used for statistical analysis of biomedical data and is freely available for the Unix/Linux, Windows, and MacOS operating systems. The **rctrack** package may be used in conjunction with literate programming R packages (*Sweave*, *knitr*) and does not require the user to modify previously written R programs. It automatically documents details regarding random number generation and a sketch of the R function call stack at the time that data, code, or result files are accessed or generated. These details are saved in an archive that includes additional files that have been automatically archived according to default or user-specified parameters to control the contents and size of the archive. The files in this archive can be used to develop a custom R package or other set of files to help others recapitulate the analysis results at a later date. Finally, the **rctrack** package also provides functions to audit one archive and to compare two archives at a later date.

Implementation

Figure 1 illustrates the implementation of the rctrack package. The function begin.rctrack sets the initial value of the seed for random number generation, creates a dedicated memory environment `rc.env` to store details that are subsequently collected, and issues a series of trace statements that temporarily *embed* a set of detail-collection procedures into the *definitions* of R functions that access files for reading/writing, generate random numbers, and issue system calls. The detail-collection procedures are then executed *as part of* the functions that access files, generate random numbers, initialize the graphic devices, and issue system calls until the `end.rctrack` command is executed. Thus, the detail collection is performed automatically without any special modification of the R program.

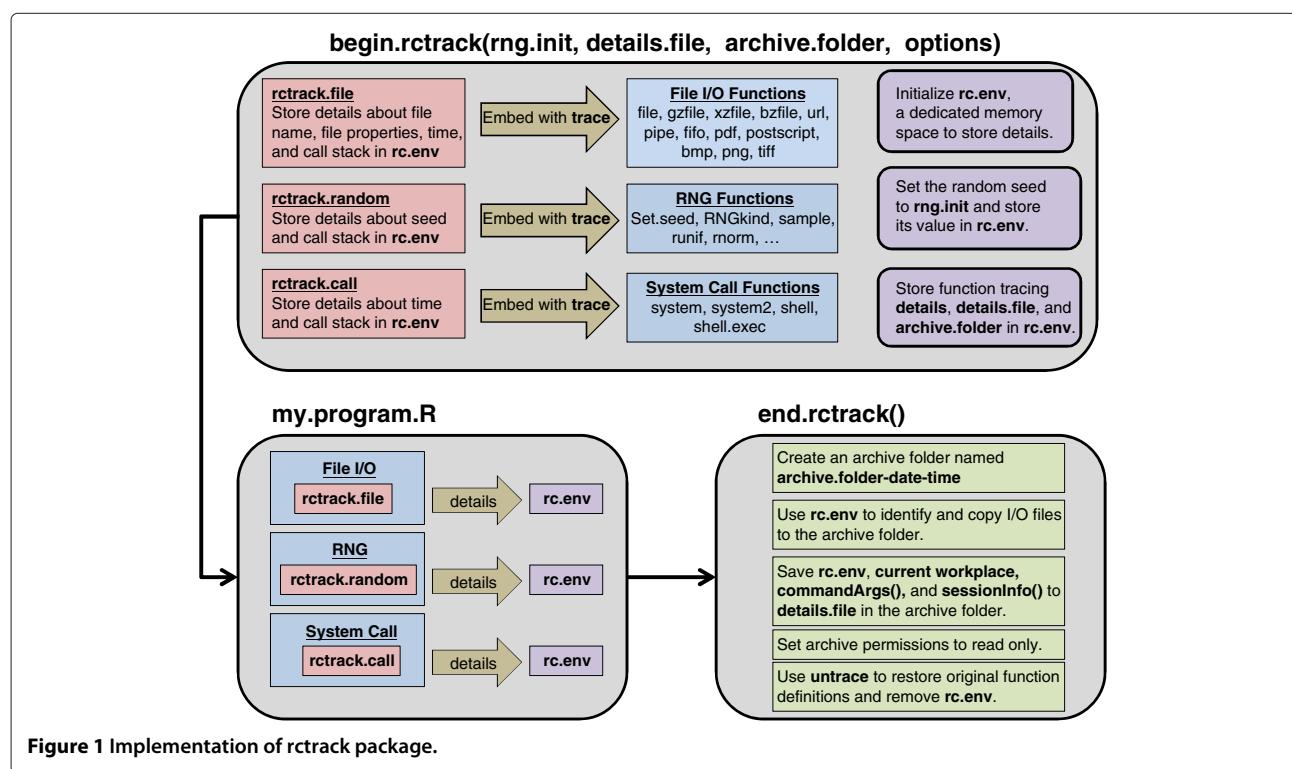
After the program execution is complete, the `end.rctrack` statement creates an archive directory with a time-stamp suffix in its name. The `end.rctrack` function then accesses the information in `rc.env` to identify and copy files to this archive directory. The `md5sum` function is used to compute the 32-byte MD5 checksum for each file and these results are recorded in `rc.env`. The `end.rctrack` saves the `R sessionInfo()`, `commandArgs()`, and the contents of `rc.env` to an Rdata file in the archive directory. Next, `end.rctrack` sets the archive directory permissions to *read only* to prevent the user from inadvertently

modifying the archive later. Finally, `end.rctrack` removes the `rc.env` environment and uses `untrace` to restore the original definitions of all the tracked functions.

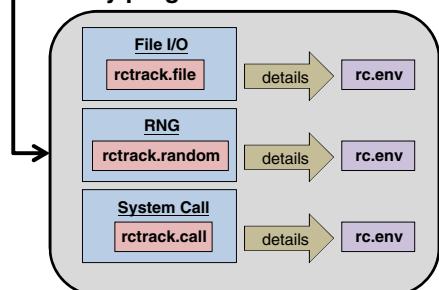
The rctrack package also includes tools to audit and compare archives at a later time. The `rc.check.archive` checks the existence, size, MD5 checksum, and modification times of files that are listed in the Rdata file in the archive and reports any discrepancies in those parameters. The `rc.compare.archives` function compares two archives generated by the rctrack package. The `rc.compare.archives` function compares the size, MD5 checksums, and modification times of files that are present in each of two archives and also lists files that are present in one archive folder but not the other archive folder. Finally, the `rc.check.zipfile` function checks the existence, size, MD5 checksum, and modification times of files in an rctrack zip file archive and compares them to the values of those file attributes that were recorded at the time the zip file archive was generated. These auditing functions are documented in the package.

Main features

Table 1 lists optional arguments of `begin.rctrack` that govern the detail collection and archiving. We anticipate that the `rng.init`, `details.file`, and `archive.folder` options will be of greatest interest to most users. The `rng.init` argument dictates the



my.program.R



end.rctrack()

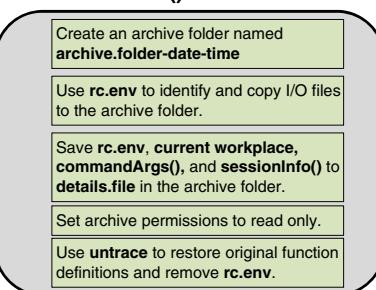


Figure 1 Implementation of rctrack package.

Table 1 Optional arguments of begin.rctrack

Argument	Value	Action
rng.init	a list with components seed, kind, and normal.kind	specifies how random number generation will be initialized by set.seed
details.file	name of Rdata file to store details	end.rctrack will save details to this file
archive.folder	character string name of the archive directory	end.rctrack will create a directory with name defined by appending a time stamp to archive.folder
maxsize.archive	a numeric with file size in bytes	end.rctrack will NOT copy INPUT files larger than this size to the archive directory
do.not.archive	a vector of character strings	files with extensions matching these strings will NOT be moved to the archive directory
skip.file.calls	a vector of character strings with function names	a list of high-level file access statements for which details will NOT be collected
skip.empty.description	a logical (TRUE/FALSE) that indicates whether to skip detail collection for empty file descriptions	if TRUE, then no details about file access events with empty descriptions (no file name) will be collected
rng.trace	a logical that indicates whether to collect details about every random number generation event	if TRUE, begin.rctrack will embed rctrack.random into each of the functions listed in the rng.functions argument so that a record of every call to those functions is retained.
rng.functions	a vector of character strings with the names of functions that generate random numbers	begin.rctrack will embed rctrack.random into each of these functions so that their use is documented in rc.env and details.file
print.trace	a logical that indicates whether to print messages about detail tracking	Messages will be issued if TRUE

random number generator to be used and the initial seed for random number generation. The default seed is 123456789 and the default generator is R's default generator. The default of details.file is "rc.details.Rdata". The details will be saved to this file when end.rctrack is called. The default of archive.folder is NULL. With these defaults, an archive folder with a name of the form YYYY-MM-DD-HH-MM-SS (year-month-day-hour-minute-second) will be created in the present working directory at the time that end.rctrack is called. Otherwise, the name of the archive folder will have the form archive.folder-YYYY-MM-DD-HH-MM-SS. The details.file and copies of other files will be saved in this archive directory when end.rctrack is called. We provide defaults for rng.init, details.file and archive.folder for user convenience but nevertheless recommend explicitly setting the values of these arguments in most applications.

There are other options that control the level and extent of detail collection and archiving. These options may be useful in certain settings. For example, the user can set up the maxsize.archive to control the maximum size of an INPUT file to be copied and frozen in the

archive folder. This option allows the user to avoid filling their disk space with multiple copies of very large data files, such as genotype data files for a genome-wide association study. The do.not.archive option also controls the automatic file archiving. This option specifies a character vector of file name extensions. Files with these extensions will not be automatically archived. This may be useful to prevent repeated archiving of individual genotype or expression signal files (such as an Affymetrix CEL file). The skip.file.calls option is a vector of character strings with function names. No details about files accessed through calls to these functions will be collected nor will files accessed through calls to these functions be archived. This option prevents the archive from being cluttered with many low-level files accessed by loading R packages with the library or require statements. Details about the usage of R packages are captured separately through saving the results of the sessionInfo() command. Likewise, the option skip.empty.description prevents the collection of many unnecessary details about some technical low-level calls to the file function that do not actually read or write files. The rng.trace option indicates

whether or not to retain a record of every random number generation event. The `rng.trace` option defaults to true. However, for some applications in some computing environments, users may want to set `rng.trace` to FALSE so that calculations which are intensive in their use of random number can be more rapidly completed. The `rng.functions` option accepts a character vector of names of functions that generate random numbers. This is the list of random number generating functions that will be traced (if `rng.trace` is TRUE) from the `begin.rctrack` statement until the `end.rctrack` statement. Finally, the `print.trace` option controls whether messages regarding the collection of reproducible computing details are issued while the program is running. Defaults have been provided for all of these options.

Sometimes, R may be used to issue system commands to perform some calculation with other software (MatLab, PLINK, etc). The `rctrack` package will note that such an action has occurred but is currently unable to collect details about the actions of those other softwares. In some cases, it may be possible to write a wrapper function in R that will add information about what the other software does to the collection of details collected by `rctrack` (such as the input files and output files of the external software program). In other cases, it can be useful to have a record indicating that these actions have occurred so that the user can manually identify and archive the necessary details.

Usage

The `rctrack` package is extremely simple to use; it only requires loading the library and issuing `begin.rctrack` to start collecting details and issuing `end.rctrack` to save the details to an `Rdata` file, create a read only archive of the files used in the data analysis, and discontinue detail collection. Figure 2 shows how to use the `rctrack` package to track the computing details of an R program.

In this example, R will collect details about every instance that `my.program.R` and all of its subprograms read or write a file, generate random numbers, or issue a system call. All these details, the starting value of the seed for random number generation, and the session information will be saved in a read-only file `mydir-date-time/rc.details.Rdata`, where `date-time` is a date-time stamp appended to the name

```
library(rctrack)
begin.rctrack(details.file="rc.details.Rdata",
              archive.folder="mydir")
source("my.program.R")
end.rctrack()
```

Figure 2 Use of `rctrack` package.

of the archive directory `mydir`. Additionally, read-only copies of all input files below a specified size and all output files (regardless of size) will be placed in the archive directory `mydir-date-time`. (The permissions are automatically set to read-only to prevent the user from inadvertently deleting or modifying the archived files; it does not prevent the deliberate modification of archived files because the read-only permissions can be modified later.) Thus, `mydir-date-time` provides a complete record of the calculations performed by `my.program.R` at the date and time indicated in the directory name. The date-time stamp serves to identify specific versions of the analysis. Most importantly, no modifications need to be made to `my.program.R` for these detail tracking and archiving operations to occur. Thus, `rctrack` provides a comprehensive solution for collecting and archiving reproducible computing details while minimizing user burden.

Results

Figure 3 provides an illustrative example of the contents of `my.program.R` of Figure 2. The example includes random number generation (`rnorm`), writing a tabular output file (`write.table`), reading a tabular input file (`read.table`), and generating a graphical output file (`jpeg`).

```
# Generate example data
n=50          # No. of Subjects
pt.id=1:n     # Subject identifier
x=rnorm(n)    # Random "X" variable
y=rnorm(n)    # Random "Y" variable
pt.data0=cbind.data.frame(id=pt.id, x=x, y=y)

# Write data to a file
pt.file="pt.data.txt"
write.table(pt.data0,pt.file,sep="\t",
            row.names=F,col.names=T)

# Import the data
pt.data1=read.table(pt.file,sep="\t",
                    header=T,as.is=T)

# Perform a Pearson correlation test
cor.res=cor.test(x,y,data=pt.data1)

# Generate a figure
fig.file="scatterplot.jpg"
jpeg(fig.file)
plot(pt.data1$x,
      pt.data1$y,
      xlab="x",ylab="y")
dev.off()
```

Figure 3 Contents of `my.program.R` from Figure 2.

Table 2 shows an abbreviated version of the contents of the `rc.file.details` object that was stored in the `rc.env` environment and then saved to `rc.details.Rdata`. The `rc.file.details` object is a `data.frame` that contains a sketch of the call stack at the time a file was accessed (in the columns `top.call`, `mid.call`, and `bot.call`), the time the traced file input/output command was called, the file description (usually a file name), and how the file was opened (read or write). The full version contains the entire function calls and the full file path in the description (Figure 4). In this example, the file details indicate that the files were accessed during a call to `Sweave` (to generate this paper), which subsequently called `read.table`, `write.table`, or `jpeg`.

Also, Table 3 shows an abbreviated version of the contents of the `random.details` object that was stored in the `rc.env` environment and then saved to the file `rc.details.Rdata`. Again, the `random.details` object is a `data.frame` that contains a sketch of the call stack at the time a random number generation operation occurred. The `random.details` objective also contains the starting value of the random seed and indicates what random number generator was used. The top row indicates that the seed was set by `begin.rctrack` and subsequent rows are records of how `rnorm` was used to generate the data. The full version contains complete function calls in the sketch of the call stack.

Finally, the file `rc.details.Rdata` contains additional information that may be useful for documenting and achieving reproducible computing. In addition to all the information collected in the special `rc.env` memory environment, it contains the R `sessionInfo()`, `Sys.info()`, and `memory.size()` at the time `end.rctrack` was called. The Figure 4 shows the complete contents of `rc.details.Rdata` for the example shown by Figures 2 and 3. Additional file 2 is the archive generated by this example; Additional file 3 provides the program files and gives detailed instructions for how to execute this example.

We also ran the simple example `my.program.R` in Figure 3 under CDE. The CDE archive was much larger than the `rctrack` archive. The `rctrack` archive contained 8 files in one folder with a total size of 96 KB. The CDE archive contained 249 files in 207 folders with a total

size of 173 MB, which is over 1,000 times larger than the `rctrack` archive. The bulk of the CDE archive contained environment files, shared libraries, R installation files, and R package files that are redundant for R users who already have those files or can easily obtain them. The exhaustive CDE archive does provide the advantage that others who do not have these R libraries can issue a single Linux command to recapitulate the calculation within random number generation variability. The `rctrack` provides sufficient information for others who already have R and the necessary libraries to recapitulate the calculation with some modest efforts (such as installing the packages and revising the file paths in the code file). The `rctrack` package also documents the initial random seed and random number generator so that the calculation can be exactly recapitulated. Furthermore, any Linux users who wish to have all the information captured by both CDE and `rctrack` may run `rctrack` under CDE.

Discussion

Reproducible computing is an essential component of biomedical research in the ‘big data’ era. Literate programming systems such as `Sweave` and `knitr` internally document specific calculations at the top-level program code level. However, to completely achieve complete permanent reproducibility, one must identify and archive *all* data files and other computational components of an analysis. The `rctrack` package provides a simple and effective computational approach to collect and archive the plethora of low-level details needed to achieve and document complete and permanent reproducibility of a statistical analysis. In particular, the analysis archive can be provided as Additional files of a published report to provide complete documentation for researchers, reviewers, supervisors, institutions, or regulatory agencies who are interested in recapitulating the analysis results or using the analysis methods in their own studies. A custom R package is an excellent format in which to distribute these supplementary materials; the `rctrack` package can help the user to create such a package by assembling all the necessary elements into a single archive file. This should greatly enhance the rigor and transparency of scientific discourse and expedite the evaluation and adoption of robust data analysis methodologies.

Table 2 Abbreviated contents of rc.file.details

	top.call	mid.call	bot.call	call.time	description	open
1	Sweave	source	file	Thu Mar 27 14:52:14 2014	my.program.R	r
2	Sweave	write.table	file	Thu Mar 27 14:52:14 2014	pt.data.txt	w
3	Sweave	read.table	file	Thu Mar 27 14:52:14 2014	pt.data.txt	rt
4	Sweave	eval	jpeg	Thu Mar 27 14:52:14 2014	scatterplot.jpg	w

```
$maxsize.archive
[1] 1e+07

$rc.calls
NULL

$rc.message.list
[1] "utils::Sweave(\"rctrack-BMC-revision-20140327.Rnw\", encoding = \"ISO8859-1\") issued at Thu Mar 27 14:52:14 2014"
[2] "utils::Sweave(\"rctrack-BMC-revision-20140327.Rnw\", encoding = \"ISO8859-1\") issued at Thu Mar 27 14:52:14 2014"

$running.time
  user  system elapsed
 0.82   0.07  1.01

$rc.file.details
top.call mid.call
1 utils::Sweave("rctrack-BMC-revision-20140327.Rnw", encoding = "ISO8859-1") source("v:/StatTools/rctrack/manuscript/revision/my.program.R")
2 utils::Sweave("rctrack-BMC-revision-20140327.Rnw", encoding = "ISO8859-1") write.table(pt.data0, pt.file, sep = "\\t", row.names = F, col.names = T)
3 utils::Sweave("rctrack-BMC-revision-20140327.Rnw", encoding = "ISO8859-1") read.table(pt.file, sep = "\\t", header = T, as.is = T)
4 utils::Sweave("rctrack-BMC-revision-20140327.Rnw", encoding = "ISO8859-1") eval(expr, envir, enclos)

bot.call call.time description open size isdir mode mtime
1 file(flename, "", encoding = encoding) Thu Mar 27 14:52:14 2014 v:/StatTools/rctrack/manuscript/revision/my.program.R r 645 FALSE 666 2014-03-21 15:56:05
2 file(file, ifelse(append, "a", "w")) Thu Mar 27 14:52:14 2014 v:/StatTools/rctrack/manuscript/revision/pt.data.txt w 2014 FALSE 666 2014-03-27 14:52:14
3 file(file, "rt") Thu Mar 27 14:52:14 2014 v:/StatTools/rctrack/manuscript/revision/pt.data.txt rt 2014 FALSE 666 2014-03-27 14:52:14
4 jpeg(fig,file) Thu Mar 27 14:52:14 2014 v:/StatTools/rctrack/manuscript/revision/scatterplot.jpg w 12388 FALSE 666 2014-03-27 14:52:14

ctime atime exe exist checksum
1 2014-03-11 13:25:33 2014-03-11 13:25:33 no TRUE e98fb286af28475c38b0887be19004a3
2 2014-03-11 13:25:33 2014-03-11 13:25:33 no TRUE cf859bb327ada957dbcfa57f1b88ba29
3 2014-03-11 13:25:33 2014-03-11 13:25:33 no TRUE cf859bb327ada957dbcfa57f1b88ba29
4 2014-03-11 13:25:32 2014-03-11 13:25:32 no TRUE 94bb13c27daba2caf0d61f55da18535c

$skip.empty.description
[1] TRUE

$skip.file.calls
[1] "library" "require" "help"

$do.not.archive
NULL

$details.file
[1] "rc.details.Rdata"

$rc.random.details
top.call mid.call bot.call call.date seed kind normal.kind
1 begin.rctrack() begin.rctrack() begin.rctrack() Thu Mar 27 14:52:14 2014 123456789 NULL NULL
2 utils::Sweave("rctrack-BMC-revision-20140327.Rnw", encoding = "ISO8859-1") eval(expr, envir, enclos) rnorm(n) Thu Mar 27 14:52:14 2014 NA <NA> <NA>
3 utils::Sweave("rctrack-BMC-revision-20140327.Rnw", encoding = "ISO8859-1") eval(expr, envir, enclos) rnorm(n) Thu Mar 27 14:52:14 2014 NA <NA> <NA>

$archive.folder
[1] "v:/StatTools/rctrack/manuscript/revision/archive.folder-2014-03-27-14-52-14"

$rctrace.list
[1] "set.seed" "RNGkind" "runif" "rbeta" "rbinom" "rcauchy" "rchisq" "rexp" "rf" "rgamma" "rgeom" "rhyper" "rlnorm"
[14] "multinom" "rnbnom" "rnorm" "rpois" "rt" "runif" "rweibull" "sample" "sample.int" "file" "url" "gzfile" "bzfile"
[27] "xfile" "pipe" "fifo" "pdf" "postscript" "xfig" "pictex" "bitmap" "dev2bitmap" "bmp" "jpeg" "png" "tiff"
[40] "svg" "cairo_pdf" "cairo_ps" "system" "system2" "shell" "shell.exec"

$archive.files
orig.full.path archive.file.path in.archive open size isdir
1 v:/StatTools/rctrack/manuscript/revision/my.program.R v:/StatTools/rctrack/manuscript/revision/archive.folder-2014-03-27-14-52-14/my.program.R TRUE r 645 FALSE
2 v:/StatTools/rctrack/manuscript/revision/pt.data.txt v:/StatTools/rctrack/manuscript/revision/archive.folder-2014-03-27-14-52-14/pt.data.txt TRUE w 2014 FALSE
3 v:/StatTools/rctrack/manuscript/revision/pt.data.txt v:/StatTools/rctrack/manuscript/revision/archive.folder-2014-03-27-14-52-14/pt.data.txt TRUE rt 2014 FALSE
4 v:/StatTools/rctrack/manuscript/revision/scatterplot.jpg v:/StatTools/rctrack/manuscript/revision/archive.folder-2014-03-27-14-52-14/scatterplot.jpg TRUE w 12388 FALSE

mode mtime ctime atime exe exist checksum
1 666 2014-03-21 15:56:05 2014-03-11 13:25:33 2014-03-11 13:25:33 no TRUE e98fb286af28475c38b0887be19004a3
2 666 2014-03-27 14:52:14 2014-03-11 13:25:33 2014-03-11 13:25:33 no TRUE cf859bb327ada957dbcfa57f1b88ba29
3 666 2014-03-27 14:52:14 2014-03-11 13:25:33 2014-03-11 13:25:33 no TRUE cf859bb327ada957dbcfa57f1b88ba29
4 666 2014-03-27 14:52:14 2014-03-11 13:25:32 2014-03-11 13:25:32 no TRUE 94bb13c27daba2caf0d61f55da18535c

$session.info
R version 2.15.1 (2012-06-22)
Platform: x86_64-pc-mingw32/x64 (64-bit)

locale:
[1] LC_COLLATE=English_United States.1252 LC_CTYPE=English_United States.1252 LC_MONETARY=English_United States.1252 LC_NUMERIC=C
[5] LC_TIME=English_United States.1252

attached base packages:
[1] tools stats graphics grDevices utils datasets methods base

other attached packages:
[1] rctrack_1.2 R.utils_1.29.8 R.oo_1.17.0 R.methodsS3_1.6.1 knitr_1.4.1 markdown_0.6.3

loaded via a namespace (and not attached):
[1] digest_0.6.4 evaluate_0.5.1 formatR_0.10 stringr_0.6.2

$memory.size
[1] 29.53

$system.info
sysname release version nodename machine login
"Windows" "7 x64" "build 7601, Service Pack 1" "ZHIFA-PC" "x86-64" "zliu3"

$effective_user
user effective_user
"zliu3" "zliu3"

Figure 4 Appendix produced by rctrack.
```

Table 3 Abbreviated contents of rc.random.details

	top.call	mid.call	bot.call	call.date	seed	kind	normal.kind
1	begin.rctrack	begin.rctrack	begin.rctrack	Mar 27 14:52:14	123456789	NULL	NULL
2	Sweave	eval	rnorm	Mar 27 14:52:14			
3	Sweave	eval	rnorm	Mar 27 14:52:14			

Unlike CDE, the `rctrack` package captures the initial seed for random number generation. Many statistical analysis procedures such as permutation, bootstrap, and Markov chain Monte Carlo simulation rely explicitly on random number generation. The `rctrack` package captures sufficient information to recapitulate random number generation with the `stats` package on a single processor and to recapitulate parallel random number generation as implemented in the `doRNG` package. The `rctrack` package does not capture all seeds in the parallel implementation in `doRNG` but still captures the initial seed which is sufficient information to recapitulate the random number series. The `rctrack` package only *monitors* random number generation, it does not attempt to *modify* or otherwise *control* random number generation. Thus, `rctrack` will not introduce any problems for random number generation that is properly implemented in a serial or parallel manner.

The current version of the `rctrack` package has some limitations that we wish to address in future versions. As previously mentioned, the user must modify the file paths in the archive to recapitulate the analysis. The `rctrack` package captures the complete original absolute file path and the archived file path for every file. We plan to use this information in a new version of the software that alleviates the user of the burden of manually modifying the file paths. We will accomplish this by using the information to computationally redirect file paths. The current version does not collect details or archive files for calculations that were performed by external software via a `system` call from R but instead simply notes that `system` calls to external software were performed. We plan to explore and develop ways to use the Linux `ptrace` command and analogous features in Windows to capture relevant information from external software calls made via the R `system` command. Also, the current version of `rctrack` does not combine information from multiple R sessions that run independently in parallel. We are currently developing routines that will combine information from such an array of R sessions.

Conclusion

The `rctrack` package was developed to minimize user burden so that it can be immediately useful in practice. The `rctrack` package only requires statements to initiate and terminate collection of details for reproducible computing; no additional programming effort is required. Furthermore, the `rctrack` package may be used in conjunction with the literate programming packages `Sweave`, `knitr`, and `lazyeave` so that the reproducible computing details may be incorporated as an appendix of the reports generated by those tools. Finally, the open source `rctrack` package provides a roadmap for advanced users to expand the capabilities to collect and archive

other details about their specific calculations. Therefore, `rctrack` will be very useful in the modern era of 'big data'.

Availability and requirements

Project name: `rctrack`

Project Homepage: <http://www.stjude.org/site/depts/biostats/rctrack>

Operating System: Platform independent

Other requirement: R 2.15.0 or higher

License: GPL

Additional files

Additional file 1: rctrack package. This is the `rctrack` package. Installation instructions are included in the file `example and instruction.rar`.

Additional file 2: Archive folder for the results. Archive folder generated by the `rctrack` package for the program `my.program.R` shown in Figure 3.

Additional file 3: Example and instruction. This file includes the instructions for installing the `rctrack` package and how to perform the example shown in Figure 2.

Competing interests

The authors above declare that they have no competing interests.

Authors' contributions

Dr. Pounds and Mr. Liu designed the study, developed and documented the software, and drafted the manuscript. All authors read and approved the final manuscript.

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