# A Machine Profile Summary

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# Providing a characterization of a particular computing environment for R calculations

If we want to compare results computed on different machines, we need a way to provide measures and identifiers of the particular computing environment at hand. This write-up attempts to summarize some possibilities available in the R statistical software and language. Note that one of our interests is to have tools that work in R across different platforms. That is, we want to be able to call an R function to get machine and computing environment characterizations regardless of the particular platform which we are currently using. Moreover, we want to have a succinct output of such information.

# Some possible information desired

There are multiple ways to get information of the type useful for characterizing the computing infrastructure used for a particular test or timing. However, we need to ensure that identifiers of particular pieces of information are unique, and that common identifiers really do specify identical, or at least equivalent, information. Otherwise, new names for some pieces of our set of capability information are needed.

# Temporal information

When we wish to analyze performance results, it is extremely helpful to have information on when the calculations were performed. Most computations should have a time/date stamp that is unlikely to be confused with any other, at least in combination with a machine name and/or other tags. Here is one suggestion based on the R Sys.time() function.

```
#get a timestamp
tsstr<- format(Sys.time(), "%Y%m%d%H%M") # tsstr == time stamp string in form YYYYmmddHHMM
cat("Date and Time stamp:",tsstr,"\n")</pre>
```

## Date and Time stamp: 202201181239

# Identifier for the computing software environment

Some of the important information elements concerning the computing environment that we need for reporting tests are

• a machine name. While most systems (and Linux in particular) offer to let the user provide a machine name, there are generally defaults that many people accept, and these are often uninformative and may be non-unique. We note that VirtualBox with a Windows 10 guest machine gave the name DESKTOP-HF4CKVA. Where this name was generated we are not sure. Settings / System allows "rename this PC."

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- operating system and version. For Linux systems, we note that besides the distribution variant, e.g., Linux Mint 20.3 MATE 64 bit, it is necessary to add the Linux kernel identifier, e.g., 5.13.0-25-generic, from the uname command.
- compiler or interpreter version. For our needs, it is obvious that the R version will be important. However, if any code is compiled or linked to libraries, we would like to know that. In particular, versions of compilers (e.g., gfortran, gcc) or BLAS or LAPACK libraries used will affect performance. Linux generally displays the BLAS and LAPACK filenames in sessionInfo(), but to get the version information, one needs to dig deeper, for example, using operating system commands. For the present, we will omit such extra information unless we can extract it easily within R.
- specific variations, if any, that should be noted. Here we may want to note if a machine is running background tasks, or if some special steps have been taken to speed up or slow down the operation e.g., overclocking.

#### Hardware information

Some of the factors relating to hardware that could be important are:

- cpu (i.e., processor)
- number of cores
- operating cycle speed (this is sometimes specific to the cpu specified )
- RAM size total
- RAM size available, possibly indicating swap space and usage
- RAM speed (which could be variable)
- GPU (i.e, additional processors) information.

#### Software information

Clearly it is important to identify which software packages are active, and the R sessionInfo() command is useful for this. However, there may be additional features that should be mentioned if the user has modified the software in any way.

The use of tools to carry out computations in parallel, using multiple cores or GPUs, is clearly an issue. In our opinion, such possibilities are not well supported by common R functions, or indeed in other computing languages, though we would very much like to hear of tools to provide such information.

#### R tools for machine information

#### sessionInfo()

```
si <- sessionInfo()</pre>
si <- as.vector(si)</pre>
si
## R version 4.1.2 (2021-11-01)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Linux Mint 20.3
##
## Matrix products: default
           /usr/lib/x86_64-linux-gnu/openblas-pthread/libblas.so.3
## LAPACK: /usr/lib/x86_64-linux-gnu/openblas-pthread/liblapack.so.3
##
## locale:
## [1] LC CTYPE=en CA.UTF-8
                                    LC NUMERIC=C
## [3] LC_TIME=en_CA.UTF-8
                                    LC_COLLATE=en_CA.UTF-8
```

```
[5] LC MONETARY=en CA.UTF-8
                                   LC_MESSAGES=en_CA.UTF-8
   [7] LC_PAPER=en_CA.UTF-8
##
                                   LC NAME=C
   [9] LC ADDRESS=C
                                   LC TELEPHONE=C
## [11] LC_MEASUREMENT=en_CA.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats
                 graphics grDevices utils
                                               datasets methods
                                                                    base
##
## loaded via a namespace (and not attached):
   [1] compiler_4.1.2 magrittr_2.0.1 fastmap_1.1.0
                                                         tools_4.1.2
   [5] htmltools_0.5.2 yaml_2.2.1
                                        stringi_1.7.5
                                                        rmarkdown_2.11
   [9] knitr_1.36
                                        xfun_0.28
                                                         digest_0.6.28
                        stringr_1.4.0
## [13] rlang_0.4.12
                        evaluate_0.14
```

#### benchmarkme

An important tool for acquiring machine information is the CRAN package benchmarkme (Gillespie (2022)). We found a minor bug in the get\_ram() function of this package for some Windows 10 versions which is now corrected (??check it is uploaded). ??Do we wish to comment on the "GB" issue?

#### ?? Need to comment on

- uses proc.time() for timings
- has single core and parallel timings
- ?? does it do GPUs
- provides far too much for most needs i.e., a succinct summary of how capable a computing system should be

```
library(benchmarkmeData)
library(benchmarkme)
ls(package:benchmarkmeData)
## Warning in ls(package:benchmarkmeData): 'package:benchmarkmeData' converted to
## character string
## [1] "get_datatable_past" "is_blas_optimize"
                                                   "make data set"
## [4] "move files"
                             "past results"
                                                   "past results v2"
                             "select_results"
## [7] "plot_past"
                                                   "summarise results"
ls(package:benchmarkme)
## Warning in ls(package:benchmarkme): 'package:benchmarkme' converted to character
## string
##
    [1] "benchmark_io"
                                       "benchmark_matrix_cal"
##
    [3] "benchmark_matrix_fun"
                                       "benchmark_prog"
        "benchmark_std"
                                       "bm_matrix_cal_cross_product"
##
    [5]
##
    [7]
        "bm_matrix_cal_lm"
                                       "bm_matrix_cal_manip"
##
   [9] "bm_matrix_cal_power"
                                       "bm_matrix_cal_sort"
## [11] "bm_matrix_fun_cholesky"
                                       "bm_matrix_fun_determinant"
  [13] "bm_matrix_fun_eigen"
                                       "bm_matrix_fun_fft"
## [15] "bm_matrix_fun_inverse"
                                       "bm_parallel"
## [17] "bm_prog_escoufier"
                                       "bm_prog_fib"
## [19] "bm_prog_gcd"
                                       "bm_prog_hilbert"
## [21] "bm_prog_toeplitz"
                                       "bm read"
## [23] "bm_write"
                                       "create_bundle"
## [25] "get_available_benchmarks"
                                       "get_byte_compiler"
                                       "get_linear_algebra"
## [27] "get_cpu"
```

```
## [29] "get_platform_info"
                                      "get r version"
## [31] "get_ram"
                                      "get_sys_details"
## [33] "is blas optimize"
                                      "plot past"
## [35] "rank_results"
                                      "sample_results"
## [37] "upload_results"
lsf.str("package:benchmarkme")
## benchmark_io : function (runs = 3, size = c(5, 50), tmpdir = tempdir(), verbose = TRUE,
##
      cores = OL)
## benchmark_matrix_cal : function (runs = 3, verbose = TRUE, cores = 0L)
## benchmark_matrix_fun : function (runs = 3, verbose = TRUE, cores = 0L)
## benchmark_prog : function (runs = 3, verbose = TRUE, cores = 0L)
## benchmark_std : function (runs = 3, verbose = TRUE, cores = 0L)
## bm_matrix_cal_cross_product : function (runs = 3, verbose = TRUE)
## bm_matrix_cal_lm : function (runs = 3, verbose = TRUE)
## bm_matrix_cal_manip : function (runs = 3, verbose = TRUE)
## bm_matrix_cal_power : function (runs = 3, verbose = TRUE)
## bm_matrix_cal_sort : function (runs = 3, verbose = TRUE)
## bm_matrix_fun_cholesky : function (runs = 3, verbose = TRUE)
## bm_matrix_fun_determinant : function (runs = 3, verbose = TRUE)
## bm_matrix_fun_eigen : function (runs = 3, verbose = TRUE)
## bm_matrix_fun_fft : function (runs = 3, verbose = TRUE)
## bm_matrix_fun_inverse : function (runs = 3, verbose = TRUE)
## bm_parallel : function (bm, runs, verbose, cores, ...)
## bm_prog_escoufier : function (runs = 3, verbose = TRUE)
## bm_prog_fib : function (runs = 3, verbose = TRUE)
## bm prog gcd : function (runs = 3, verbose = TRUE)
## bm_prog_hilbert : function (runs = 3, verbose = TRUE)
## bm_prog_toeplitz : function (runs = 3, verbose = TRUE)
## bm_read : function (runs = 3, size = c(5, 50), tmpdir = tempdir(), verbose = TRUE)
## bm_write : function (runs = 3, size = c(5, 50), tmpdir = tempdir(), verbose = TRUE)
## create_bundle : function (results, filename = NULL, args = NULL, id_prefix = "")
## get_available_benchmarks : function ()
## get_byte_compiler : function ()
## get_cpu : function ()
## get_linear_algebra : function ()
## get_platform_info : function ()
## get_r_version : function ()
## get_ram : function ()
## get_sys_details : function (sys_info = TRUE, platform_info = TRUE, r_version = TRUE, ram = TRUE,
       cpu = TRUE, byte_compiler = TRUE, linear_algebra = TRUE, locale = TRUE,
       installed_packages = TRUE, machine = TRUE)
##
## is_blas_optimize : function (results)
## plot_past : function (test_group, blas_optimize = NULL, cores = 0, log = "y")
## rank_results : function (results, blas_optimize = is_blas_optimize(results), verbose = TRUE)
## upload_results : function (results, url = "http://www.mas.ncl.ac.uk/~ncsg3/form.php", args = NULL,
       id_prefix = "")
lsf.str("package:benchmarkmeData")
## get_datatable_past : function (test_group, blas_optimize = NULL, cores = 0)
## is_blas_optimize : function (results)
## make_data_set : function (from)
## move_files : function (from, to)
```

```
## plot_past : function (test_group, blas_optimize = NULL, cores = 0, log = "y")
## select_results : function (test_group, results = NULL, blas_optimize = NULL, cores = 0)
## summarise_results : function (res)

## This next line takes a lot of time to run, so is commented out here
# benchmark_std()
get_byte_compiler()

## byte_optimize
## 2
gla<-get_linear_algebra()
gsd<-get_sys_details()</pre>
```

# Calculating the machine precision

Until the IEEE 754 standard (??ref) became available in the 1980s, computers used multiple forms of floating point arithmetic. With some ingenuity, it is possible to detect many aspects of the floating point properties. This is a port of Mike Malcolm's ENVRON to R. ??ref. It computes the machine precision, radix and number of radix digits.

?? mention Kahan's Paranoia

```
envron<-function(){</pre>
D1<-as.numeric(1) # use variable to avoid constants when double precision invoked
E5 <- 10 #arbitrary scaling for additive equality tests
B9 <- 1E+35 # big number, not necessarily biggest possible
E6 <- 1 # initial value for radix
E9 <- 1 # initial value for machine precision
DO <- E6
repeat{
  E9 <- E9/2 #start of loop to decrease estimated machine precision
  DO <- E6+E9 #force storage of sum into a floating-point scalar
  if (DO <= E6) break
} # repeat reduction while (1+E9) > 1
E9 <- E9*2 # restore smallest E9 which gives (1+E9) > 1
repeat {
  E6 <- E6+1 # try different radix values
   DO <- E6+E9
   if (D0 <= E6) break
} # until a shift is observed
J1 <- 1 # initial count of radix digits in mantissa
E9 <- 1 # use radix for exact machine precision
repeat {
  E9 <- E9/E6 #loop while dividing by radix
  J1 <- J1+1 #increment counter
  DO <- D1+E9 #add tp 1
  if (DO<=D1) break #test and repeat until equality
}
E9 <- E9*E6 #recover last value of machine precision
J1 <- J1-1 #and adjust the number of digits
mpvals<-list(eps=E9, radix=E6, ndigits=J1)</pre>
}
mp<-envron()</pre>
E9<-mp$eps
E6<-mp$radix
```

```
J1<-mp$ndigits
cat("E9 = ",E9,"
                -- the machine precision, E9=MIN (X 1+X>1)\n")
## E9 = 2.220446e-16 -- the machine precision, E9=MIN (X 1+X>1)
cat( "E9*1E+16=",E9*1E+16,"\n")
## E9*1E+16= 2.220446
cat( "E6 = ",E6," -- the radix of arithmetic", "\n")
## E6 = 2 -- the radix of arithmetic
cat( "J1 = ",J1," -- the number of radix digits in mantissa of FP numbers", "\n")
## J1 = 53 -- the number of radix digits in mantissa of FP numbers
cat( "E6^{(-J1+1)} * 1E+16=",1E+16*(E6^{(-J1+1)),"\n")
## E6^(-J1+1) * 1E+16= 2.220446
cat(".Machine$double.eps=",.Machine$double.eps,"\n")
## .Machine$double.eps= 2.220446e-16
cat("From R .Machine:\n")
## From R .Machine:
cat("double.eps=", .Machine$double.eps,"\n",
    "double.base=", .Machine$double.base,"\n",
    "double.digits=", .Machine$double.digits,"\n")
## double.eps= 2.220446e-16
## double.base= 2
## double.digits= 53
```

#### Simple performance examples

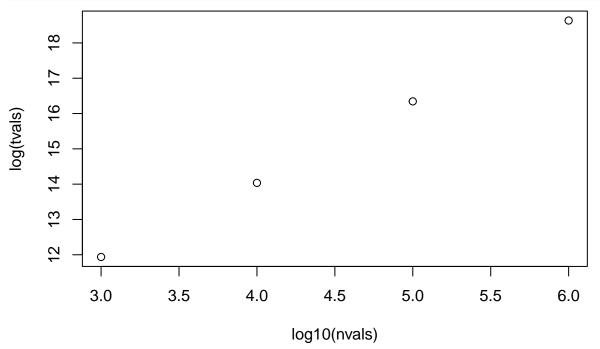
There are many performance examples used for testing the speed of computers. A well-known example is the Dongarra, Luszczek, and Petitet (2003) LINPACK test used for ranking supercomputers. These generally employ linear algebra calculations. The example below, by contrast, focuses on the special function computations. Variations on this simple performance test have been used by one of the authors for nearly half a century. Note the use of the microbenchmark timing package (Mersmann (2013)).

```
loopesc <- function(nn){
    ss<-0
    for (i in 1:nn) {
        xx<-exp(sin(cos(1.0*i)))
        ss<-ss+xx
    }
        xx
}
require("microbenchmark")

## Loading required package: microbenchmark

nvals<-c(1000, 10000, 100000, 1000000)
tvals<-rep(NA,length(nvals))
i<-0</pre>
```

```
for (nn in nvals) {
  i<-i+1
  cat(nn, "\n")
  tt<-microbenchmark(loopesc(nn), unit='us')</pre>
  print(tt)
  tvals[i] <-mean(tt$time)</pre>
}
## 1000
## Unit: microseconds
           expr min
                             lq
                                   mean median
                                                       uq
## loopesc(nn) 123.776 124.676 152.8587 124.8485 125.354 2899.152
## 10000
## Unit: microseconds
                               lq
           expr
                     min
                                      mean
                                             median
                                                          uq
  loopesc(nn) 1172.933 1239.608 1246.216 1242.347 1248.622 1339.766
## Unit: microseconds
           expr
                     min
                               lq
                                      mean
                                             median
                                                          uq
                                                                  max neval
## loopesc(nn) 12317.31 12439.25 12536.78 12484.01 12611.96 13439.64
## 1e+06
## Unit: microseconds
##
                               lq
                                                                  max neval
           expr
                     min
                                      mean
                                             median
                                                          uq
  loopesc(nn) 122013.5 123128.7 123814.4 123982.5 124287.1 127022.5
plot(log10(nvals),log(tvals))
```



# Sys.info()

Sys.info()

## sysname

```
##
                                                     "Linux"
##
                                                     release
                                       "5.13.0-25-generic"
##
##
                                                     version
##
   "#26~20.04.1-Ubuntu SMP Fri Jan 7 16:27:40 UTC 2022"
##
                                                   nodename
                                                       "M21"
##
##
                                                    machine
##
                                                    "x86 64"
##
                                                       login
##
                                                   "unknown"
##
                                                        user
##
                                                      "john"
                                             effective_user
##
##
                                                      "john"
```

Issues relating to random number generation

Issues relating to compilation of R

- LAPACK
- BLAS
- others?

# Tools for accessing and clearing environments and dataframes

?? should this be here – probably to list things in the workspace

```
sys.frame()
sys.frames()
sys.status()
sys.on.exit()
sys.parents()
sys.calls()
```

### Choices

For use in recording tests of R functions and packages for optimization and nonlinear least squares, it seems that the benchmarkme function get\_sys\_details() provides more than sufficient information for our needs.

From the above discussion, the following offers a possible compact solution. Users may wish to modify this to their own particular needs.

## M21:john-Linux-5.13.0-25-generic|Intel(R) Core(TM) i5-10400 CPU @ 2.90GHz|33474072576bytesRAM

Note that we are reading from the file MachID.R so that this file is usable elsewhere and the current document does not become out of sync with the working MachID.

# References

- Dongarra, J. J., P. Luszczek, and A. Petitet. 2003. "The LINPACK Benchmark: Past, Present, and Future." Concurrency and Computation: Practice and Experience 15 (9): 803–20.
- Gillespie, Colin. 2022. Benchmarkme: Crowd Sourced System Benchmarks. https://github.com/csgillespie/benchmarkme.
- Mersmann, Olaf. 2013. Microbenchmark: Sub Microsecond Accurate Timing Functions. http://CRAN.R-project.org/package=microbenchmark.