Random Jungle, version 2.0.0

A powerful Random Forests (TM) implementation Edition 2.0.0, 12 March 2013

by Daniel F. Schwarz et al. (Developer) to version 1.2.365 by Jochen Kruppa (Maintainer/Developer) to version 2.0.0 (randomjungle@googlegroups.com)



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1 Introduction and preliminaries

This first chapter explains what Random Jungle is, where Random Jungle comes from, how to read and use this documentation, how to call the Random Jungle program, and how to report bugs about it. It concludes by giving tips for reading the remainder of the manual.

Okay, I have no time to read, how do I get started? Go to the Chapter 5 [Application], page 15, where you get all the needed information on general Random Forest (TM) tasks. If you not found what you need, read the options list Section 2.1 [Operation modes], page 3 and finally write an e-mail to randomjungle@googlegroups.com.

The following chapters then detail all the features of the Random Jungle .

1.1 Introduction to Random Jungle

Random Jungle is an implementation of Random Forests (TM). It is supposed to analyse high dimensional data. In genetics, it can be used for analysing big Genome Wide Association (GWA) data. Random Forests (TM) is a powerful machine learning method. Most interesting features are variable selection, missing value imputation, classifier creation, generalization error estimation and sample proximities between pairs of cases.

Random Jungle is mostly compatible with the Linux and Windows (TM). Also, there exists compatibilities with Solaris.

1.2 Historical references

First ideas where seeded in 2006 during GAW15 workshop in Florida. In the beginning of 2008, Daniel F. Schwarz released 0.5.0 and 0.5.1 which was a "way pre-release" and fast, but lagged of features and documentation. However, the williams award was won at the International Genetic Epidemiology Society (IGES) converence 2008 in St. Louis with release 0.5.2. In late 2008 and 2009, intensive work improved the platform compatibility, added documentation and raised the number of features of Random Jungle.

1.3 Problems and bugs

If you have problems with Random Jungle or think you've found a bug, please report it. Before reporting a bug, make sure you've actually found a real bug. Carefully reread the documentation and see if it really says you can do what you're trying to do. If it's not clear whether you should be able to do something or not, report that too; it's a bug in the documentation!

Before reporting a bug or trying to fix it yourself, try to isolate it to the smallest possible input file that reproduces the problem. Then send us the input file and the exact results Random Jungle gave you. Also say what you expected to occur; this will help us decide whether the problem was really in the documentation.

Once you've got a precise problem, send e-mail to randomjungle@googlegroups.com. Please include the version number of Random Jungle you are using. You can get this information with the command rjungle --version or rjungle -Z.

Non-bug suggestions are always welcome as well. If you have questions about things that are unclear in the documentation or are just obscure features, please report them too.

1.4 Using this manual

This manual contains a number of examples of Random Jungle input and output, and a simple notation is used to distinguish input, output and error messages from Random Jungle. Examples are set out from the normal text, and shown in a fixed width font, like this

```
This is an example of an example!
```

To distinguish input from output, all output from Random Jungle is prefixed by the string ' \Rightarrow ', and all error messages by the string 'error '. When showing how command line options affect matters, the command line is shown with a prompt '\$ like this', otherwise, you can assume that a simple rjungle invocation will work. Thus:

```
$ command line to invoke rjungle
Example of input line
⇒Output line from rjungle
[error] and an error message
```

The sequence 'D' in an example indicates the end of the input file. The sequence 'NL' refers to the newline character. The majority of these examples are self-contained, and you can run them with similar results by invoking <code>rjungle-d</code>. In fact, the testsuite that is bundled in the Random Jungle package consists of the examples in this document! Some of the examples assume that your current directory is located where you unpacked the installation, so if you plan on following along, you may find it helpful to do this now:

```
$ cd randomjungle-2.0.0
```

2 Invoking Random Jungle

The format of the Random Jungle command is:

```
rjungle [option...]
or
rjunglesparse [option...]
```

All options begin with '-', or if long option names are used, with '--'. On some platforms long options might not work. A long option name need not be written completely, any unambiguous prefix is sufficient. POSIX requires Random Jungle to recognize arguments intermixed with files, even when POSIXLY_CORRECT is set in the environment. Most options take effect at startup regardless of their position, but some are documented below as taking effect after any files that occurred earlier in the command line. The argument '--' is a marker to denote the end of options.

With short options, options that do not take arguments may be combined into a single command line argument with subsequent options, options with mandatory arguments may be provided either as a single command line argument or as two arguments, and options with optional arguments must be provided as a single argument. In other words, rjungle - QPDfoo -d a -df is equivalent to rjungle -Q -P -D foo -d -df -- ./a, although the latter form is considered canonical.

We strictly recommend to use long options. At first some options are not availabe in the short form and second the validation and understanding is much harder if the short option is used. Instead use some wrapper functions like described in the example section below.

With long options, options with mandatory arguments may be provided with an equal sign ('=') in a single argument, or as two arguments, and options with optional arguments must be provided as a single argument. In other words, rjungle --def foo --debug a is equivalent to rjungle --define=foo --debug= -- ./a, although the latter form is considered canonical (not to mention more robust, in case a future version of Random Jungle introduces an option named '--default').

Random Jungle understands the following options, grouped by functionality.

2.1 Command line options for operation modes

Several options control the overall operation of rjungle:

-h

--help Print a help summary on standard output, then immediately exit rjungle without reading any input files or performing any other actions.

--version

Print the version number of the program on standard output, then immediately exit rjungle without reading any input files or performing any other actions.

-f FILENAME

--file=FILENAME

FILENAME of input file with data. Input data has to been numerical. The default FILENAME is input.dat(.gz). In R (http://www.r-project.org), save data to file using write.table as follows:

- > make.your.data.in.R
- > quit()
- \$ rjungle -f input.dat [...]

In plink (pngu.mgh.harvard.edu/~purcell/plink/), save data to raw file using the recodeA option, set the ped file option and char memory mode option in rjungle as follows:

\$ plink --file yourDataFile --recodeA
\$ rjungle -f yourDataFile.raw -p -M 2 [...]

Avoid missing values in data. See See Chapter 3 [Input data], page 11, for more details.

-o FILEPREFIXNAME

--outprefix=FILEPREFIXNAME

FILEPREFIXNAME of output files is the first part of output files (i.e. rjungle.importance, rjungle.prediction, ...). The default FILEPREFIXNAME is rjungle. Use for example -o my_analysis_no123.

-е *CHAR*

--delimeter=CHAR

Set the delimeter in your input file to CHAR. Default is a space.

−w

--write=ID

Save Random Jungle model for a later prediction by -P.

$$= 0$$
 not

= 1 [ID] to a gzipped XML file FILEPREFIXNAME.jungle.xml.gz (Not supported yet)

=
$$2$$
 to a XML raw file FILEPREFIXNAME.jungle.xml

=
$$3$$
 [ID] probability estimation to a XML raw file FILEPREFIXNAME.jungle.xml

DEFAULT is 0.

-P FILENAME

--predict=FILENAME

Predict test data with model (forest) saved by -w2 given by the file FILE-NAME (test data file is given with option -f). Predictions will be written to FILEPREFIXNAME.prediction.

--probability

Writes probability predictions to file FILEPREFIXNAME.prediction. Choose -w3 for the training of the model!

--deterministic

Starts a deterministic forest with variables specified in -C.

FILEPREFIXNAME.prediction. Choose -w3 for the training of the model!

-y ID

--treetype=ID

Choose base classifier by setting *ID* There are serveral treetypes but only CART is fully supported. Explanatory or exposure variables will be named: input variables. Explained or response variable will be named: output variable. If you want to use CART like Random Forest (TM) does choose one of three possible values as follows:

= 1 or 5

CART, y (output variable): nominal, x (input variable(s)): numeric, ID = 1 is recommended for less different values in the input variables (i.e. GWA SNP data or integer data). ID = 5 is recommended for more different values in the input variables (i.e. many floating point numbers). Like original Breiman/Cutler/Friedman algorithm.

- = 2 [ID] CART, y (output variable): nominal, x (input variable(s)): nominal,
- = 3 [ID] CART regression trees, y (output variable): numeric, x (input variable(s)): numeric,
- = 4 [ID] CART regression trees, CART regression trees, y (output variable): numeric, x (input variable(s)): nominal,

DEFAULT is 1.

-t SIZE

--ntree=SIZE

SIZE is the number of trees in jungle. If SIZE=0 then the size will be set automatically depending on mtry and variable size (experimental feature). DE-FAULT is 500.

-m SIZE

--mtry=SIZE

SIZE of randomly choosen variable sets. At each node building step, a variable will be selected out of the set, that serves the biggest information gain. The

bigger SIZE is set, the higher computing time might be. The bigger SIZE is set, the more similar trees in jungle will be. High noised data sets should processed with a big SIZE. Default is square root of number of input variables.

-x NUM

--missingcode=NUM

Missings should always coded as NA or NUM in your data. The program takes NUM as a internal representation of a missing value. DEFAULT: -99.

-i *ID*

--impmeasure=ID

Variable selection: Choose an method for estimating variable importance as follows:

- = 1 [ID] Intrinsic Importrance (i.e. GINI-Index).
- = 2 [ID] Permutation Importance by Breiman, Cutler (observed in Fortran code).
- = 3 [ID] Permutation Importance by Liaw, Wiener (in R-package RandomForest).
- = 4 [ID] Permutation Importance, raw values, no normalization.
- = 5 [ID] Permutation Importance by Meng et. al

The results will be written to file FILEPREFIXNAME.importance. You can not turn off variable importance output. DEFAULT is 1.

- -B ID Variable selection/ model optimization: Choose an method for estimating variable importance as follows:
 - = 0 [ID] No backward elimination.
 - = 1 [ID] Backward elimination. Discard 50% at each step (slow). Stop if number of variables shrinked to STOPSIZE, see option -j.
 - = 2 [ID] Backward elimination. Discard 33% at each step (slow). Stop if number of variables shrinked to STOPSIZE, see option -j.
 - = 3 [ID] Backward elimination. Discard only negative values at each step (slow/recommended). (Shown at IGES2007 by Inke R. Konig).

DEFAULT is 0.

-j STOPSIZE

--nimpvar=STOPSIZE

Only necessary if --impmeasure = 2,3,5,6 or 7. How many variable should remain. The lesser STOPSIZE is, the reliable the result might be. The smaller SIZE is, the higher computing time will be. DEFAULT is 100.

-₩

--verbose

Print some nice information to screen. Otherwise put information to file FILEPREFIXNAME.verbose (DEFAULT).

-u

--downsampling

Choose randomly samples without replacement. DEFAULT: switched off.

-M ID

--memmode=ID

Usage of the heap memory (RAM) as follows:

= 1 [ID]

Single precision floating point (Normal).

= 2 [ID] Char (small). CHAR normally fits in one byte. DATA CELL VALUE HAS TO BE AN INTEGER IN [-127..127].

If you want to use very small data coding, i.e. for SNP analysis, give rjunglesparse a try! DEFAULT is 0.

-C FILE

--colselection=FILE

Only use selected columns listed in FILE. Example content of FILE:

var1 var20 var1000 var300

DEFAULT is take all variables.

-D NAME

$--{\tt depvarname} = {\tt NAME}$

Output variable name in the data SET! If *NAME* is empty then the rjungle switches to unsupervised mode.

-s

--sampleproximities

It computes proximities between pairs of cases that can be used in clustering, locating outliers, or (by scaling) give interesting views of the data. Can be used as the distance matrix for Multidimensional Scaling (MDS). The results will be written to file FILEPREFIXNAME.samproximity. DEFAULT: switched off

-z

--seeed Seed of random number generators.

-U

--nthreads=NUM

Maximally use *NUM* threads (CPUs) for parallel processing. Limit for *NUM* is number of CPUs in computer. DEFAULT: Number of CPUs in computer.

-p

--pedfile

Input file has got ped format (i.e. plink output with recodeA). DEFAULT: switched off.

-I NUM

--impute=NUM

Impute missings in input data using Random Forest(TM)'s imputation algorithm. The number of iterations is given by *NUM*. For imputing continuous data, use option -A (--impcont) as well. For more information, have a look at http://www.stat.berkeley.edu/~breiman/RandomForests/cc_home.htm.

Do not try to impute untyped SNPs (Schwarz et al. 2009, BMC Proceedings, 3, S65) if case-control-status is missing. Try a different program like:

IMPUTE, MACH, PLINK, ... DEFAULT: switched off.

-k NUM

--maxtreedepth=NUM

This is a stop criterium/tunning parameter. Tree growing will stop, when the tree exceeds a depth of *NUM*. DEFAULT: switched off.

-1 NUM

--targetpartitionsize=NUM

This is a stop criterium/tunning parameter. Tree growing will stop, when a partion falls below a size of *NUM* samples. DEFAULT: switched off.

-K NUM

--condimp=NUM

Perform conditional importance if option -i > 1. NUM is the pearson's cor. coef. cutoff. The smaller NUM, the bigger a conditional importance permutation group will be created. (=> More accurate, but slower) Requires: 0 <= NUM <= 1 NUM < 0 => switched off DEFAULT: switched off.

-W

Adjust class weights of unbalanced datasets automatically. DEFAULT: switched off.

--oobset Outputs the oobset of forest in file *.oob (each row == one tree) DEFAULT: switched off.

--classweights=STRING

Sets the weights of classes E.G.: "1.23;22;" \dots first class gets weight 1.23 \dots second class gets weight 22 DEFAULT: switched off

2.2 Experimental command line options for operation modes

The following options are more or less experimental without a deep validation process. Be cautious about using one of those.

```
-d-g-n-r-c-a-b-w-V-S-A-G-q-E-X-Q
```

Those options are experimental. Be cautious about using one of those.

-d --depvar=*POS*

Output variable at column POS in the data SET. POS=[0-...] DEFAULT is 0 and switched off.

-g --depvarcol=*POS*

Output variable is the variable POS in the data FILE. POS=[0-...] DEFAULT is 0 and switched off.

--varnamesrow=POS

Variable names at row POS in data set. POS=[0-...] DEFAULT is 0 and switched off.

-r --nrow=SIZE

Number (SIZE) of samples. SIZE=[1-...]. SIZE = 0: use all samples in data file DEFAULT is 0.

-c --ncol=*SIZE*

Number (SIZE) of input variables (features). SIZE=[1-...]. SIZE=0: use all variables in data file DEFAULT is 0.

--skiprow=SIZE

Skip SIZE rows (samples) before reading data from file. SIZE=[1-...] DEFAULT is 0.

-b --skipcol=*SIZE*

Skip SIZE columns (samples) before reading data from file. SIZE=[1-...] DE-FAULT is 0.

-S --summary

Print a summary of all trees in jungle. DEFAULT: switched off.

-A --impcont

Impute data is continuous. DEFAULT: switched off (categorical data).

-G --gwa Impute data with GWA settings. DEFAULT: switched off.

-q

--tunemtry=THETA

Tune mtry parameter. Step size = THETA. E.g.: Quarter stepping (-q 0.25): $100 \text{ predictor variables (ncol} = 100) => \text{mtry} = 25, 50, 75. DEFAULT: switched off.}$

-E

--extractdata

Extract data to file. DEFAULT: switched off.

-X

--plugin=FILE

Use Random Jungle generator from given plugin: FILE. Totally experimental. Please contact the maintainer for more and detailed information on the plugin process. So far this part is not fully supported. DEFAULT: switched off (Empty String).

-Q

--pluginpar=STRING

Additional plugin parameters. As decribed above this part is totally experimental and not under maintance. DEFAULT: switched off (Empty String).

3 Input data

This chapter describes various input file types of rjungle.

Remember that the Random Forest (TM) approach can not handle missing values. Random Jungle has a implemented imputation algorithm but it should be used carefully. Please consider well known literature for your special problem or remove all missings from the data (not recommended). We demonstrate some wrapper functions in the example section below.

3.1 Input the whole data

Random Jungle analyses data given in a file and name of response variable, see option -f and -D. The file format is matrix like and is as follows:

names [Variable]

The first line of the input file contains the variable names. The variable name must not contain quotes or space charaters. The variables are separated by space charaters.

[Data]

Each following line represents one sample (observation). Every single sample needs to have one numerical value for each variable. The values are separated by space characters and are ordered corresponding to the variable names (of course).

Here, an example of an input file indat.data:

```
responseVar inputVar1 inputVar2 inputVar3 inputVar4 0 1.2 3.4 5.6 7.8 0 1.1 3.3 5.5 7.7 0 2.2 4.4 6.6 8.8 1 1.0 3.0 5.0 7.0 1 1.0 3.0 5.0 7.0 1 2.0 4.0 6.0 8.0
```

The corresponding call would be \$ rjungle --file indat.data -D responseVar [...]

If you choose the ped file option -p then the input file format is ped like. The file must have variables FID, IID, PAT, MAT, SEX, PHENOTYPE and at least three variables (SNPs). Here, an example of an input file:

```
FID IID PAT MAT SEX PHENOTYPE rs1 rs2 rs3 rs4
1000 NA2001 0 0 2 0 0 1 0 1
1000 NA2002 0 0 2 0 0 2 0 1
1001 NA2003 0 0 1 0 1 2 0 1
1001 NA2004 0 0 1 0 1 1 1 1
1002 NA2005 0 0 2 1 1 1 1 1
1004 NA2006 0 0 2 1 1 2 1 1
1005 NA2007 0 0 2 1 2 2 2 2
1006 NA2008 0 0 1 1 2 2 2 2
1007 NA2008 0 0 1 1 2 2 2 2
```

3.2 Restricted analysis

The rjungle can be run also with just a subset of all variables which are given in the input file (option $\neg f$). See description of option $\neg C$, for more details.

4 Output data

This chapter describes various output file types of rjungle.

4.1 The standard output files

If rjungle is executed it will always produce three files: Log file, Confusion file, and Importance file. You can run an example in your randomjungle-2.0.0 directory and see it.

```
$ rjungle -f your_data_file.txt -D response_variable_name -v
$ ls
rjungle.importance rjungle.log rjungle.prediction test.ped
Or without PED file as follows:
$ rjungle -f your_data_file.txt -D response_variable_name -v
```

4.2 Log file

The log file contains all options / parameter of the rjungle execution. This is useful to reproduce results.

4.3 Confusion file

The rjungle always evaluates the classifier. The accuracy on training and test data. Test data is called oob data which is collected during growing process (similar to cross validation's test sets). The results are shown in so-called confusion matrices in file FILEPREFIX-NAME.confusion. Columns represent the predicted values and rows represent the real values.

When using option -y1 a file FILEPREFIXNAME.confusion2 will be created. The file contains class specific error rates.

4.4 Prediction file

The rjungle is able to perform new data to a saved classifier (option -w2 and -P). The predicted results will be saved to file FILEPREFIXNAME.prediction.

4.5 Importance file

Random Jungle estimates the importance of variables (option -i). The results are saved to file FILEPREFIXNAME.importance and/or FILEPREFIXNAME.importance2. The first file contains four columns but most interesting columns are varname (variable name) and value (importance value). The higher the value, the more important is the variable with name varname. This list is sorted ascending. So, look at the tail of file to see the most important variables. The second file contains various permutation importance informations.

4.6 Verbose file

We recommend to use always -v in each Random Jungle run. Especially new users should use this option to see what Random Jungle is doing and if all wanted options are active. In addition a timer is given with a processing information.

If Random Jungle is invoked from command line without verbose option it is very schtum and do not output anything to screen. It writes all information occurring during a run to file FILEPREFIXNAME.verbose. Nevertheless, if you want a takly rjungle which puts all process information to screen then use option (-v)

```
$ rjungle -f your_data_file.txt -D PHENOTYPE -v
```

```
Start: ...
+----+
   RandomJungle | .....
                              +----+
Output to:
rjungle.*
loading data...
Read 9 row(s) and 10 column(s).
Use 9 row(s) and 6 column(s).
dependent variable name: PHENOTYPE
Growing jungle...
Number of variables: 6 mtry = 2
1 thread(s) growing 500 tree(s)
Growing time estimate: ~0 sec.
Generating and collecting output data...
Compiling trees.
Writing accuracy information...
calculating confusion matrix...
Elapsed time: 0 sec
Finished: ...
```

\$

5 Some nice examples

This chapter describes various examples of working with rjungle. We concentrate here on examples included into the R project (http://www.r-project.org/) and connected packages. Why do we use Random Jungle invoking R? First, a normal project has some data handling issues. It is much more easier to do with R, like impute missing values or renaming and removing columns. Second, if the Random Jungle code is included into a R script it can be validated and easy repeated.

The following example code is written in R. Therefore, we skiped the > in each line for a better copy&paste behaviour. R output instead is marked with an >. Shell commands are still marked by \$.

5.1 Rjungle in R

First of all the general framework for Random Jungle in R. We build up "paste" one command string rjungleCMD and send it to the system e.g. shell.

Question: Where do I get the "rjungleExe"? Visit the project webpage of Random Jungle (http://www.r-project.org/) and download the pre-compiled version of Random Jungle matching to your system.

Question: What mean "responseVar"? This is your variable of interest. Like the diabtes state or written in R formula: responseVar ~ Var1 + Var2.

5.2 Simple example in R

We want to grow a simple forest on the iris data. So, our input data contains 4 variables (Sepal.Length, Sepal.Width, Petal.Length, Petal.Width) and 1 response variable(Species). Caution, we write all results files into the working directionary of R! Change the rjungleOutFile for a better data handling.

```
## File handling for Random Jungle
rjungleExe <- file.path("/to/executable/rjungle")
rjungleInFile <- file.path("iris.dat")
rjungleOutFile <- file.path("iris")
## Data handling</pre>
```

Question: Where do I find the results? The results are written into the directionary of R or the specification of rjungleOutFile. Therefore, where your R script is running now. To get this information type getwd().

5.3 Simple example with ped file

Using Plink PED-files:

Question: What are PED-files? Visit http://pngu.mgh.harvard.edu/~purcell/plink/for more information.

Question: I have build up an PED-file but it does not work! Have you used the recodeA option? This is important for Random Jungle to handle PED-Files.

Question: It takes hours to analyse my data! Look at the sparse version of Random Jungle and other examples below.

5.4 Estimating variable importance

In the second example we used only a small data set. Know we want to analyse the Pima Indian Diabetes data set. In R the data is available in the package MASS. Firstly, lets estimate the fast GINI-Importance (i1) with 1000 trees.

```
library(MASS)

## File handling for Random Jungle
rjungleExe <- file.path("/to/executable/rjungle")
rjungleInFile <- file.path("pima.dat")
rjungleOutFile <- file.path("pima")

## Data handling</pre>
```

```
data(Pima.tr)
     sum(is.na(Pima.tr)) ## Is there any missing?
     write.table(Pima.tr, file = rjungleInFile, row.names = FALSE, quote =
     FALSE)
     ## Run Random Jungle
     rjungleGiniCMD <- paste(rjungleExe,</pre>
                             "-f", rjungleInFile,
                             "-v",
                                                 ## show processing
                             "-i1",
                                                 ## chose GINI-Importance
                                                 ## 1000 trees
                             "-t1000",
                                        ## response variable name
                             "-D type",
                             "-o", rjungleOutFile) ## out file path
     try(system(rjungleGiniCMD))
  Secondly, we now want to estimate the permutation importance (i2).
     ## Run Random Jungle
     rjunglePermCMD <- paste(rjungleExe,</pre>
                             "-f", rjungleInFile,
                             "-v",
                                     ## show processing
                             "-i2",
                                                 ## chose permutation-Importance
                             "-t1000", ## 1000 trees
"-D type", ## response variable name
                             "-o", rjungleOutFile) ## out file path
     try(system(rjunglePermCMD))
  Now, we got even a more reliable result.
  Finally, we want to use the conditional variable importance (--condimp). Caution it is
much more slower than the permutation importance alone!
     ## Run Random Jungle
     rjunglePermCMD <- paste(rjungleExe,</pre>
                             "-f", rjungleInFile,
                             "-v",
                                                 ## show processing
                             "-i3",
                                                 ## chose permutation-Importance
                             "-t1000", ## 1000 trees
"-D type", ## response variable name
                                                  ## 1000 trees
                             "-o", rjungleOutFile) ## out file path
     try(system(rjunglePermCMD))
  Now we can read in the wanted importance file.
     ## Read in the importance file
     rjungleGiniImportanceFile <- paste(rjungleOutFile, "importance", sep = "")
     rjunglePermImportanceFile <- paste(rjungleOutFile, "importance2", sep = "")
     importanceGini <- read.table(rjungleGiniImportanceFile)</pre>
     importancePerm <- read.table(rjunglePermImportanceFile)</pre>
```

Question: There are many importance values. Which one should I use? We recommend to use the Liaw score which is the same as in the R package randomForest. Again, look at

the options list (-i) and get more information on the used importance measures. Depending on your problem a different important measure might be feasible.

5.5 Using R and rjungle seperately

We want to analyse data from R and do not want to start Random Jungle out of R. Instead we prepare the data in R and start Random Jungle from the shell.

```
$ R
...
data(iris)
iris$Species = as.integer(iris$Species) # convert factor to integer
write.table(iris, file = "iris.dat", row.names = FALSE, quote = FALSE)
quit("no")
$ rjungle -f iris.dat -v -D Species -o iris
$ cat iris.confusion
...
$ cat iris.importance
```

It is not recommended to use R and Random Jungle in this way.

5.6 Using plink and rjungle

We assume you have your prepared GWA data called gwadata.

```
## File handling for PLINK
plinkExe <- file.path("/to/executable/")</pre>
plinkInFile <- file.path("gwadata")</pre>
## File handling for Random Jungle
rjungleExe <- file.path("/to/executable/rjungle")</pre>
rjungleInFile <- file.path("pima.dat")
rjungleOutFile <- file.path("pima")
## Run PLINK
plinkCMD <- paste(plinkExe,</pre>
                   "--file", plinkInFile,
                   "--recodeA")
try(system(plinkCMD))
## Run Random Jungle
rjungleCMD <- paste(rjungleExe,
                     "-f", rjungleInFile,
                     "-v",
                                            ## show processing
                     "-i1",
                                            ## chose GINI-Importance
                     "-p",
                                           ## use PED-file
                     "-M2",
                                            ## memory mode
                     "-o", rjungleOutFile) ## out file path
try(system(rjungleCMD))
```

```
## Read in the importance file
  rjungleImportanceFile <- paste(rjungleOutFile, "importance", sep = "")
  importance <- read.table(rjungleImportanceFile)</pre>
or
  rjungleExe <- file.path("/to/executable/rjunglesparse")</pre>
  ## Run Random Jungle
  rjungleCMD <- paste(rjungleExe,
                       "-f", rjungleInFile,
                       "-v",
                                              ## show processing
                       "-i1".
                                            ## chose GINI-Importance
                       "-p",
                                             ## use PED-file
                       "-o", rjungleOutFile) ## out file path
  try(system(rjungleCMD))
  ## Read in the importance file
  rjungleImportanceFile <- paste(rjungleOutFile, "importance", sep = "")
  importance <- read.table(rjungleImportanceFile)</pre>
```

The rjunglesparse uses less memory than rjungle. But your data should only use the values 0,1,2 and 3(missing code).

Question: What is PLINK? Visit http://pngu.mgh.harvard.edu/~purcell/plink/ for more information.

Question: Where is the difference between both Rjungle runs? In the second one we use rjunglesparse.

5.7 Prediction (classification)

See examples above for the generation of the iris.dat data file.

Question: Is it not bad to use the same data for training and testing? Yes. You should avoid this. This example is only for demonstration.

5.8 Prediction (probability)

Here we present the probability estimation using Random Jungle. Please consider the project page for a overview of literature dealing with this issue (http://www.randomjungle.de/).

```
## File handling for Random Jungle
rjungleExe <- file.path("/to/executable/rjungle")</pre>
rjungleTrainFile <- file.path("iris.dat")</pre>
rjungleOutFile <- file.path("iris")
## Run Random Jungle
rjungleCMD <- paste(rjungleExe,</pre>
                    "-f", rjungleTrainFile,
                    "-w3",
                                         ## save trees for prediction
                    "-v",
                                        ## show processing
                    "-D Species", ## response variable name
                    "-o", rjungleOutFile) ## out file path
try(system(rjungleCMD))
## Run prediction
rjungleTestFile <- file.path("iris.dat")</pre>
rjungleXmlFile <- file.path("iris.jungle.xml")</pre>
rjunglePredCMD <- paste(rjungleExe,</pre>
                        "-f", rungleTestFile,
                        "--probability",
                        "-P", rjungleXmlFile, ## load saved trees for prediction
                        "-v",
                                  ## show processing
                        "-D Species", ## response variable name
                        "-o", rjungleOutFile) ## out file path
try(system(rjunglePredCMD))
## Get prediction
rjunglePredFile <- file.path("iris.prediction")</pre>
```

```
predRjungle <- read.table(rjunglePredFile)</pre>
```

Question: What is the difference to the classification prediction? I do not see it! You save the trees in a different format by using the option -w3 in the training step and predict the probabilities by using --probability.

Question: What is the short option for --probability? There is no short option for --probability.

Question: Is it not bad to use the same data for training and testing? Yes. You should avoid this. This example is only for demonstration.

5.9 Deterministic Forest

If you want to force Random Jungle to select at each split always some variables than run a deterministic forest. In this example we would choose age and sex at each split among the other random selected variables in the data set.

5.10 Imputation

The following examples are describing the imputation by Random Jungle. Please be caution. You will manage to get rid of missings but it might procude false positives. Please consider the right literature matching to your problem to avoid missings. We cannot give you a general framework. Still, remove all missings, like na.omit().

SNP data (raw/fast imputation):

```
$ rjunglesparse -f mypedfile.raw -p -t1 -I1 -o example8_1_1
```

The imputation result was written to example 8_1_1.imputed.dat.gz and can be used for analysis. E.g.:

```
$ rjunglesparse -f example7_1_1.imputed.dat.gz -p ... -o example8_1_2 Imputing continuous data:
```

```
$ rjungle -f continuous.dat -A -D responseVar -I6 -o example8_2 or imputing categorical data:
```

```
$ rjungle -f cate.dat -D responseVar -I6 -o example8_3
or imputing data with no groups to classify on (unsupervised learning):
```

```
$ rjungle -f continuousAndNoGroups.dat -A -I5 -o example7_4
```

Question: I have SNP data are there other possibilities? Yes. Look for PLINK or IMPUTE.

5.11 Using plink and rjunglesparse

The rjunglesparse is the same program like rjungle, but you can use just a small set of values: 0,1,2 (and 3 as missing coding). Look at Section 5.6 [Example5], page 18 for a implementation in R. You might want to use rjunglesparse in conjunction with plink. The memory consumption of rjunglesparse is very small.

```
$ plink --file gwadata --recodeA
$ rjunglesparse -f plink.raw -p -v -o example8
...
$ tail example8.importance
```

5.12 Using MPI

For high speed parallel processing, Random Jungle could be used on computer clusters using MPI mode (Random Jungle has to be compiled for MPI!). The program was performed on huge data successfully using 150 processors (300 processes) in parallel on a high performance cluster. However, the MPI mode should be used by experienced users only. In MPI mode, performing permutation importance (option -i2, .., -i5) is allow exclusivly. Execute Random Jungle as follows (one possibility, look at your job system, it might differ):

Each process writes temp files to working directory (example9_mpi_id_*.*). Final results are written to usual files (example9.*).

Question: Where do I get the MPI pre-compiled versions? Please look at the project homepage (http://www.randomjungle.de/). If there is no MPI pre-compiled version, it is not possible at the moment.

Question: MPI does not work on the cluster etc.? Please contact your admin. Depending on your batch or job system there are many sources of errors.

Appendix A Indices of concepts and macros

A.1 Compiling and installing Random Jungle

There are pre-compiled versions on the internet site http://www.randomjungle.de/. Maybe, there exists a pre-compiled version for your platform (computer) and you do not need to compile it.

The source code is not freely available. Please write to <code>infoimbs@imbs-luebeck.de</code> and ask for a possible cooperation or a pre-compiled version for your system.

Random Jungle uses some GNU tools, libraries and other free open source code. The installation routine (./configure) prompt you to install packages which are missing on your machine. It is also recommended to compile the sources with an openMP supporting compiler (i.e. GNU gcc > 4.2). Invoke the following commands to compile rjungle:

```
./configure
make
make check
make install
```

For compiling Random Jungle with MPI (message passing interface) support, configure RJ using option --enable-mpi. The program is optimized for Open MPI (http://www.open-mpi.org/). The MPI mode should be used by experienced users only.

```
./configure --enable-mpi
```

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