# rhipe Documentation

Release 0.61

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# **CONTENTS**

1	Setting up RHIPE 1.1 Requirements	<b>3</b> 3 3			
2	The rhlapply Command 2.1 Introduction 2.2 Return Value 2.3 Function Usage	<b>5</b> 5 5			
3	The rhmr Command 3.1 Introduction 3.2 Return Value 3.3 Function 3.4 RHIPE Options 3.5 Status, Counters and Writing Output 3.6 Side Effect files 3.7 Mapreduce Options 3.8 IMPORTANT  Miscellaneous Commands	7 7 7 8 9 9 9			
•	4.1 Introduction 4.2 Running Mapreduce 4.3 Serialization 4.4 HDFS Related 4.5 Map Files	11 11 12 12 14			
5	Using RHIPE on EC2 5.1 Introduction	<b>17</b> 17			
6	Examples         6.1 rhlapply          6.2 rhmr	19 19 20			
7	FAQ	21			
8	Protobuffer and R				
9	Datatypes				

Mainpage

CONTENTS 1

2 CONTENTS

**CHAPTER** 

**ONE** 

# **SETTING UP RHIPE**

# 1.1 Requirements

1. Protobuffers

RHIPE uses Google's Protobuf library for serialization. This(the C/C++ libraries) must be installed on *all* machines (master/workers). Get Protobuffers from http://code.google.com/p/protobuf/. RHIPE already has the protobuf jar file inside it.

**Non Standard Locations** If installing protobuf to a non standard location, update the PKG\_CONFIG\_PATH variable, e.g

```
export PKG_CONFIG_PATH=$PKG_CONFIG_PATH:$CUSTROOT/lib/pkgconfig/
```

- 2. *R* , tested on 2.8
- 3. rJava The R package needs rJava.

Tested on RHEL Linux, Mac OS 10.5.5 (Leopard). Does not work on Snow Leopard

# 1.2 Installation

#### Rhipe requires the following environment variables

```
HADOOP-location of Hadoop installation
HADOOP_LIB=location of lib jar files included with Hadoop, usually
$HADOOP/lib
HADOOP_CONF_DIR=location of Hadoop conf folder, (usually $HADOOP/conf)
```

#### On every machine

```
R CMD INSTALL Rhipe_VERSION.tar.gz
```

#### To load it

library(Rhipe)

# THE RHLAPPLY COMMAND

### 2.1 Introduction

rhapply applies a user defined function to the elements of a given R list or the function can be run over the set of numbers from 1 to n. In the former case the list is written to a sequence file, whose length is the default setting of rhwrite.

Running a hundreds of thousadands of seperate trials can be terribly inefficient, instead consider grouping them, i.e set mapred.max.tasks to a value much smaller than the length of the list.

### 2.2 Return Value

rhlapply returns a list, the names of which is equal to the names of the input list (if given).

# 2.3 Function Usage

Description follows

11 The list object, optional. Applies fun to ll[[i]]. If instead ll is a numeric, applies fun to each element of seq(1, ll). If not given, must provide a value for ifolder

**fun** A function that takes only one argument.

**ifolder** If 11 is null, provide a source here. Also change the value of inout [1] to either text or sequence.

**readIn** The results are stored in a temporary sequence file on the DFS which is deleted. Should the results be returned in a list? Default is TRUE. For large number of output key-values (e.g 1MM) set this to FALSE, using the default options to rhread is extremely slow.

**ofolder** If given the results are written to this folder and not deleted. If not, they are written to temporary folder, read back in (assuming readIn is TRUE) and deleted.

N The number of task to create, i.e the mapred.map.tasks and is passes onto the rhwrite function

mapred Options passed onto rhmr

setup And expression that is called before running func. Called once per JVM.

aggr A function (default is NULL) to aggregate results. If NULL (default), every list element is written to disk. This can be difficult to read back into R (especially when one has nearly 1MN trials, R has to combine a list of 1MN elements!). aggr is a function that takes one argument a list of values, each value being the result apply the user function to an element of the input list. E.g. if fun returns a data frame, one could write

```
aggr=function(x) do.call("rbind",x)
and the result of rhlapply will be one big data frame.
doLocal Default is F. Sent to rhread
... passed onto RHMR.
```

#### 2.3.1 **RETURN**

An object that is passed onto rhex.

#### 2.3.2 IMPORTANT

The object passed to rhex has variable called  $rhipe\_command$  which is the command of the program that Hadoop sends information to. In case the client machine's (machine from which commands are being sent) R installation is different from the tasktrackers' R installation the RHIPE command runner wont be found. For example suppose my cluster is linux and my client is OS X, then the  $rhipe\_command$  variable will reflect the location of the rhipe command runner on OS X and not that of the tasktrackers(Linux) R distribution.

There are two ways to fix this a) after z <- rhlapply(...) change  $r[[1]][[1]] \text{ prhipe\_command}$  to the value it should be on the tasktrackers.

oı

b) set the environment variable RHIPECOMMAND on each of tasktrackers. RHIPE java client will read this first before reading the above variable.

# THE RHMR COMMAND

### 3.1 Introduction

The rhmr command runs a general mapreduce program using user supplied map and reduce commands.

# 3.2 Return Value

In general a set of files on the Hadoop Distributed File System. It can be of Text Format or a Sequence file format. In case of the latter, the key and values can be any R data structure.

# 3.3 Function

```
rhmr <- function(map, reduce=NULL,</pre>
             combiner=F,
2
             setup=NULL,
3
             cleanup=NULL,
             ofolder='',
             ifolder='',
             inout=c("text", "text"),
             mapred=NULL,
             shared=c(),
             jarfiles=c(),
10
             copyFiles=F,
11
             partitioner=NULL,
12
             opts=rhoptions(),jobname="")
```

map A map expression, not a function. The map expression can expect a list of keys in map.keys and list of values in map.values.

reduce Can be null if only a map job. If not, reduce should be an expression with three attributes

pre Called for a new key, but no values have been read. The key is present in reduce.key.

reduce Called for reducing the incoming values. The values are in a list called reduce.values

post Called when all the values have been sent.

**combiner** Uses a combiner if TRUE. If so, then reduce.values present in the reduce\$reduce expression will be a *subset* of values. The reducer algorithm should be able process input emitted from map *or* reduce.

setup An expression that can be called to setup the environment. Called once for every task. It can be a list of two attributes map and reduce which are expressions to be run in the map and reduce stage. If a single expression then that is run for both map and reduce

cleanup Same as for setup, run when all work for a task is complete.

**ifolder** A folder or file to be processed. Can be a vector of strings.

**ofolder** The folder to store output in. Side effects will be copied here.

inout \

A vector of input type and output type.

text indicates Text Format. Use mapred.field.separator to seperate the elements of a vector.

**sequence** is a sequence format. Outputs in this form /can/ be used as an input.

**binary** is a simple binary format consisting of key-length, key data, value-length, value data where the lengths are integers in network order. Though *much* faster than sequence in terms of reading in data, it *cannot* be used an input to a map reduce operation.

map Only as OutputFormat! That is, map can only be the second element of inout. If so, the output part files will be directories, each containing a data and an index file. If the reducer writes the same key as the one received then using the function rhgetkey, specifying the get and the output folder part files, one can use the output as a hash table (do keep the keys small then). However, if the keys are changed before being written (using rhcollect), the order is lost and even though one can still use the individual part file as a Map file reader, the part file containing the key needs to be known (as opposed to just specifying the directory of part files). To remedy this just run a identity map job converting map input to map output (see rhM2M and rhS2M). Map Output formats can be used an input format. Use the function rhmap.sq on a directory of map part files e.g rhmap.sq("/tmp/out/p\*), this will return a vector of paths pointing to the data files in each of the part folders (the folders also contain index files, which can't be used as sequence file input to Hadoop, so these have to be filtered).

**shared** A vector of files on the HDFS that will be copied to the working directory of the R program. These files can then be loaded as easily as load (filename) (removed leading path)

jarfiles Copy jar files if required. Experimental, probably doesn't work.

copyFiles For side effects to be copied back to the DFS, set this to TRUE, otherwise they wont be copied.

mapred Set Hadoop options here and RHIPE options.

partitioner A list with two names: lims and type. type can be one of string, numeric, integer and lims must be positive. The vector lims is used as a partitioner, that is if c("a", "b", "c") is the key and lims=c(1,2) the first two elements will be used for partitioning. If a partitioner is used, all the emitted keys must of be the same type. If lims is of length 1, it will partitioned on that element.

**jobname** the jobname, if not given, then current date and time is the job title.

# 3.4 RHIPE Options

**rhipe\_stream\_buffer** The size of the STDIN buffer used to write data to the R process(in bytes) *default:* 10\*1024 bytes

mapred.textoutputformat.separator The text that seperates the key from value when inout [2] equals text. default: Tab

mapred.field.separator The text that seperates fields when inout [2] equals text. default: Space

rhipe\_reduce\_buff\_size The maximum length of reduce.values default: 10,000

rhipe\_map\_buff\_size The maximum length of map.values (and map.keys) default: 10,000

# 3.5 Status, Counters and Writing Output

#### 3.5.1 Status

To update the status use rhstatus which takes a single string e.g rhstatus("Nice") This will also indicate progress.

#### 3.5.2 Counter

To update the counter C in the group G with a number N, user rhcounter (G, C, N) where C and G are strings and N is a number. However, C and G can be atomic vectors and they will be converted to strings. Previously a "," in C or G would upset Hadoop, but not with version 0.52 onwards. The values will be returned to the R session. Output ^^^^^ To output data use rhcollect (KEY, VALUE) where KEY and VALUE are R objects that can be serialized by rhsz (see the misc page). If one needs to send across complex R objects e.g the KEY is a function, do something like rhcollect (serialize (KEY, NULL), VALUE)

# 3.6 Side Effect files

Files written to tmp/ (no leading slash!) e.g pdf ("tmp/x.pdf") will be copied to the output folder.

# 3.7 Mapreduce Options

Many mapreduce configuration variables are stored in the environment. To get the value use Sys.getenv(), e.g in the map stage, to find out the name of the current input file, use Sys.getenv('mapred.input.file').

### 3.8 IMPORTANT

The object passed to rhex has variable called  $rhipe\_command$  which is the command of the program that Hadoop sends information to. In case the client machine's (machine from which commands are being sent) R installation is different from the tasktrackers' R installation the RHIPE command runner wont be found. For example suppose my cluster is linux and my client is OS X, then the  $rhipe\_command$  variable will reflect the location of the rhipe command runner on OS X and not that of the tasktrackers(Linux) R distribution.

There are two ways to fix this a) after z <- rhmr(...) change r[[1]] \$rhipe\_command to the value it should be on the tasktrackers.

or

b) set the environment variable RHIPECOMMAND on each of tasktrackers. RHIPE java client will read this first before reading the above variable.

**CHAPTER** 

**FOUR** 

# MISCELLANEOUS COMMANDS

### 4.1 Introduction

This is a list of supporting functions for reading, writing sequence files and manipulating files on the Hadoop Distributed File System (HDFS).

# 4.2 Running Mapreduce

#### 4.2.1 rhex

Once an object is created using rhmr and rhlapply, it must be sent to the Hadoop system. The function rhex does this

```
rhex <- function(o, async=FALSE, mapred)</pre>
```

Where o is the object that rhmr or rhlapply returns. mapred is a list of the same shape as in rhmr and rhlapply. Values in this over-ride those passed in rhmr '' (and ''rhlapply). If async is FALSE, the function returns when the job has finished running. The value returned is a list of Hadoop Counters (e.g bytes sent, bytes written, time taken etc).

If async is TRUE, the function returns immediately. In this case, the value returned can be printed (i.e just type the returned value at the REPL) or passed to rhstatus to monitor the job.

### 4.2.2 rhjoin

```
rhjoin <- function(o,ignore.stderr=TRUE)</pre>
```

where o is returned from rhex with async=TRUE. The function returns when the job is complete and the return value is the same as rhex when async is FALSE (i.e counters and the result(failure/success) of the job). If ignore.stderr is FALSE, the progress is displayed on the screen(exactly like rhex).

#### 4.2.3 rhstatus

```
rhstatus <- function(o)</pre>
```

where o is returned from rhex with async=TRUE (or a Hadoop job id (e.g "job\_20091031\_0001"). This will return list of counters and the progress status of the job(number of maps complete, % map complete etc).

#### 4.2.4 print

This a generic function for printing objects returned from rhex when async=TRUE. The default returns start time, job name and job id, and job state, map/reduce progress. For more verbosity, type print (o, verbose=2) which returns a list of counters too (like rhstatus).

### 4.3 Serialization

#### 4.3.1 rhsz

```
rhsz <- function(object)</pre>
```

Serializes a given R object. Currently the only objects that can be serialized are vectors of Raws, Numerics, Integers, Strings(including NA), Logical(including NA) and lists of these and lists of these. Attributes are copied to(e.g names attributes). It appears objects like matrices, factors also get serialized and unserialized successfully.

### 4.3.2 rhuz

```
rhuz <- function(object)</pre>
```

Unserializes a raw object returned from rhsz

# 4.4 HDFS Related

#### 4.4.1 rhload

```
rhload <- function(file,...)</pre>
```

Loads an R data set stored on the DFS.

#### **4.4.2** rhsave

```
rhsave <- function(..., file)</pre>
```

Saves the objects in . . . to file on the HDFS. All other options are passed onto the R function save

### 4.4.3 rhsave.image

```
rhsave.image <- function(..., file)</pre>
```

Same as R's save.image, except that the file goes to the HDFS.

# 4.4.4 rhcp

```
rhcp <- function(ifile, ofile)</pre>
```

Copies a ifile to ofile on the HDFS, i.e. both files must be present on the HDFS.

### 4.4.5 rhmv

```
rhmv <- function(ifile,ofile)</pre>
```

Moves if ile to of ile on the HDFS (and deletes if ile).

# 4.4.6 rhput

```
rhput <- function(src,dest,deleteDest=TRUE)</pre>
```

Copies the file in src to the dest on the HDFS, deleting destination if deleteDest is TRUE.

# 4.4.7 rhget

```
rhget <- function(src,dest)</pre>
```

Copies src''(on the HDFS) to ''dest on the local. If src is a directory and dest exists, src is copied inside dest''(i.e a folder inside ''dest). If not(i.e dest does not exist), src's contents is copied to a new folder called dest. If src is a file, and dest is a directory src is copied inside dest. If dest does not exist, it is copied to that file

Wildcards allowed

OVERWRITES!

#### 4.4.8 rhls

```
rhls <- function(dir,recur=FALSE)</pre>
```

Lists the path at dir. Wildcards allowed. Use recur (FALSE/TRUE) to not recurse or to recurse.

### 4.4.9 rhdel

```
rhdel <- function(dir)</pre>
```

Deletes file(s) at/in dir. Wildcards allowed.

#### 4.4.10 rhwrite

```
rhwrite <- function(lo,f,n=NULL,...)</pre>
```

Writes the list lo to the file f. n is the number of sequence files to split the list into. The default value of n is mapred.map.tasks \* mapred.tasktracker.map.tasks.maximum.

#### 4.4.11 rhread

```
rhread <- function(files, max=-1, type="sequence", verbose=T, mc=FALSE)</pre>
```

4.4. HDFS Related 13

Reads files(s) from files (which could be a directory). Wildcards allowed.

If verbose is True, information is displayed (useful when reading many files)

If max is positive, max key-value pairs will be read.

Set type to "map" if the directory files contains map folders.

Provided you have the multicore package, set mc to TRUE and the descrialization will occur in parallel. You have to load multicore beforehand.

# 4.4.12 rhmerge

```
rhmerge (inr, ou)
```

ing can have wildcards. Usually used to merge all files in a directory into one file ou on the local file system.

#### 4.4.13 rhreadBin

```
rhreadBin <- function(filename, max=as.integer(-1), bf=as.integer(0))</pre>
```

Reads data outputed in 'binary' form. max is the maximum number to read, -1 is all. bf is the read buffer, 0 implies the os specified default BUFSIZ

# 4.5 Map Files

#### 4.5.1 rhS2M

```
rhS2M <- function (files, ofile, dolocal = T, ignore.stderr = F, verbose = F)
```

Converts the sequence files specified by files and places them in destination ofile. If dolocal is True the conversion is done on the local machine, otherwise over the cluster (which is much faster for anything greater than hundreds of megabytes). If ignore.stderr is True, the mapreduce output is displayed on the R console. e.g

```
rhS2m("/tmp/so/p*","/tmp/so.map",dolocal=F)
```

#### 4.5.2 rhM2M

```
rhM2M <- function (files, ofile, dolocal = T, ignore.stderr = F, verbose = F)
```

Same as S2M, except it converts a group of Map files to Map files. Why? Consider a mapreduce job that outputs modified keys in the reduce part, i.e the reduce receives key K0 but emits f(K0), where f(K0) <> K0, the result of this the keys in the reduce output part files wont be sorted even though the K0 are sorted.

So, if the reducer emits K0, the output part files constitute a valid collection of sorted map files. If the reducer emits f(K0), this does not hold any more. Running rhM2M on this output produces another output in which the keys are now sorted (i.e we just run an identity mapreduce emitting f(K0), though now the input to the reducers are f(K0)).

To specify the input files, it is not enough to specify the directory containing the part files, because the part files are directories which contain a sequence file and a non sequence file. Specifying the list of directories to a mapreduce job will cause it to fail when it reads the non-map file.

```
Use rhmap.sqs.
```

# 4.5.3 rhgetkey

```
 \texttt{rhgetkey} \gets \textbf{function} \text{ (keys, paths, sequence=NULL, skip=0, ignore.stderr = T, verbose = F)} \\
```

Given a list of keys and vector of map directories (e.g /tmp/ou/mapoutput/p\*"), returns a list of key,values. If sequence is a string, the output key,values will be written to the sequence files on the DFS(the values will not be read into R). Set skip to larger(integr) values to prevent reading in all keys of the table - slower to find your key, but can search a much large database.

4.5. Map Files 15

**CHAPTER** 

**FIVE** 

# **USING RHIPE ON EC2**

# 5.1 Introduction

RHIPE also works on EC2 using Cloudera's scripts. Let me demonstrate

#### 5.1.1 Download

The Cloudera scripts can be found at http://archive.cloudera.com/docs/\_getting\_started.html

Follow the instructions to test your working EC2 installation.

# 5.1.2 Using RHIPE on EC2

1. You need to create an entry in your ~/.hadoop-ec2/ec2-clusters.cfg, e.g.

```
[test2]
ami=ami-6159bf08 # Fedora 32 bit instance
instance_type=c1.medium
key_name=saptarshiguha ## Your key name
availability_zone=us-east-1c
private_key=PATH_TO_PRIVATE_KEY
ssh_options= -i % (private_key) s -o StrictHostKeyChecking=no
user_data_file=the file you download in step 2
```

In particular, RHIPE only works with 32/64 bit Fedora instance types, so choose those AMIs.

- 2. Download this file( http://github.com/saptarshiguha/RHIPE/blob/master/code/hadoop-ec2-init-remote.sh ) and replace the file of the same name (it is in the Cloudera distribution). This file contains one extra shell function to install code RHIPE requires: R, Google's protobuf and RHIPE
- 3. Now start your cluster

```
python hadoop-ec2 launch-cluster --env REPO=testing --env HADOOP_VERSION=0.20 test2 3
```

The number (3) must be greater than 1.

4. Wait, till you it completely finishes booting up (the cloudera scripts tell you the url of the jobtracker). Login to the cluster

```
python hadoop-ec2 login test2
```

5. Start R, and try the following

```
library(Rhipe)
z <- rhlapply(10,runif)
## Runs on a local machine(i.e the master)
rhex(z,changes=list(mapred.job.tracker='local'))
library(Rhipe)
## Runs on the cluster
z <- rhlapply(10,runif)
rhex(z)</pre>
```

6. Consider the more involved problem of bootstrapping. See this question posed on the R-HPC mailing list (http://permalink.gmane.org/gmane.comp.lang.r.hpc/221). Using Rhipe( chunksize (see the posting) is 1000 per task which results in 100 tasks)

```
y <- iris[which(iris[,5] != "setosa"), c(1,5)]</pre>
   rhsave(y,file="/tmp/tmp.Rdata")
   ## The function 'f' depends on 'x' so we must save it
   ## using rhsave and then load it in the setup
   setup <- expression({</pre>
      load("tmp.Rdata")
8
     })
10
   f<- function(i) {
      ind <- sample(100, 100, replace=TRUE)</pre>
       result1 <- glm(y[ind,2]~y[ind,1], family=binomial(logit))
13
       return(structure(coefficients(result1), names=NULL))
14
15
16
   z <- rhlapply(100000L, f, shared="/tmp/tmp.Rdata", setup=setup,
17
                 mapred=list(mapred.map.tasks=100000L/1000
18
                    , mapred.reduce.tasks=5))
19
20
   q \leftarrow rhex(z)
21
  g1 <- do.call("rbind", lapply(g, function(r) r[[2]]))</pre>
   g2 <- cbind(unlist(lapply(g,function(r) r[[1]])),g1)</pre>
```

I used 3 c1.xlarge nodes(each \$0.68/hr). This took 2 minutes and 5 seconds to run and another minute to read the data back in.

On 10 similar nodes, this took 1 minute and 2 seconds. There is a point where it won't become any faster.

On 20 nodes(with mapred.map.tasks=160), it takes 52 seconds (probably not worth the extra cost ...)

# **EXAMPLES**

# 6.1 rhlapply

# 6.1.1 Simple Example

Take a sample of 100 iid observations Xi from N(0,1). Compute the mean of the eight closest neighbours to X1. This is repeated 1,000,000 times.

```
nbrmean <- function(r) {
    d <- matrix(rnorm(200),ncol=2)
    orig <- d[1,]
    ds <- sort(apply(d,1,function(r) sqrt(sum((r-orig)^2)))[-1])[1:8]
    mean(ds)
}
trials <- 1000000

One Machine
trials is 1,000,000
system.time({r <- sapply(1:trials, nbrmean)})
user system elapsed
    1603.414    0.127 1603.789

Distributed, output to file
mapred <- list(mapred.map.tasks=1000)
r <- rhlapply(1000000, fun=nbrmean, ofolder="/test/one", mapred=mapred)
rhex(r)</pre>
```

Which took 7 minutes on a 4 core machine running 6 JVMs at once.

# 6.1.2 Using Shared Files and Side Effects

```
h=rhlapply(length(simlist)
,func=function(r){

## do something from data loaded from session.Rdata
pdf("tmp/a.pdf")

plot(animage)
dev.off()},

setup=expression({
    load("session.Rdata")
}),
```

```
hadoop=list(mapred.map.tasks=1000),
shared=("/tmp/session.Rdata")) ##session.Rdata created by rhsave(..., file="/tmp/session.Rdata")
```

Here session.Rdata is copied from HDFS to local temporary directories (making for faster reads). This is a useful idiom for loading code that the rhlapply function might depend on. For example, assuming the image is not *huge* 

```
rhsave.image("/tmp/myimage.Rdata")
rhlapply(N, function(r) {
   object <- dataset[[r]]
   G(object)
}, setup=expression({load("myimage.Rdata")})))</pre>
```

In the above example, I wish to apply the G to every element in dataset.

# 6.2 rhmr

#### 6.2.1 Word Count

Generate the words, 1 word every line

```
rhlapply(10000, function(r) paste(sample(letters[1:10],5),collapse=""),output.folder='/tmp/words')
```

Word count using the sequence file

Run it

#### 6.2.2 Subset a file

We can use this RHIPE to subset files. Setting mapred.reduce.tasks to 5 writes the subsetted data across 5 files (even though we haven't provided a reduce task)

```
1  m <- expression({
2     for(x in map.values) {
3         y <- strsplit(x," +")[[1]]
4     for(w in y) rhcollect(w,T)
5     }})
6     z <- rhmr(map=m,inout=c("text","binary"),
7         ifolder="X",ofolder='Y',mapred=list(mapred.reduce.tasks=5))
8     rhex(z)</pre>
```

### SEVEN

# FAQ

#### 1. Local Testing?

Easily enough. In rhmr or rhlapply, set mapred.job.tracker to 'local' in the mapred option of the respective command. This will use the local jobtracker to run your commands.

However keep in mind, shared.files will not work, i.e those files will not be copied to the working directory and side effect files will not be copied back.

#### 2. Speed?

Similar to Hadoop Streaming. The bottlenecks are writing and reading to STDIN pipes and R.

#### 3. What can RHIPE do?

Firstly, there are several R packages for parrallel computing. snow, 'snowfall' are packages for (mostly) embarrrasingly parallel computation and do not work with massive datasets. mapreduce implements the mapreduce algorithm on a single machine (which can be done with RHIPE by using a cluster of size 1).

RHIPE is a wrapper around Hadoop for the R user. So that he/she need not leave the R environment for writing, running mapreduce applications and computing wth massive datasets.

4. The command runner, different client and tasktrackers.

The object passed to rhex has variable called  $rhipe\_command$  which is the command of the program that Hadoop sends information to. In case the client machine's (machine from which commands are being sent ) R installation is different from the tasktrackers' R installation the RHIPE command runner wont be found. For example suppose my cluster is linux and my client is OS X, then the  $rhipe\_command$  variable will reflect the location of the rhipe command runner on OS X and not that of the tasktrackers(Linux) R distribution.

There are two ways to fix this a) after z <- rhmr(...) change r[[1]] \$rhipe\_command to the value it should be on the tasktrackers. (in case of rhlapply, it should be r[[1]][[1]] \$rhipe\_command)

Ωŧ

b) set the environment variable RHIPECOMMAND on each of tasktrackers. RHIPE java client will read this first before reading the above variable.

done

#### 5. Data types

Stick to vectors of raws, character, logical, integer, complex and reals. For atomic vectors, don't use attributes (especially not the names attribute) *Stay away* from data.frames (These two(data.frames and named scalar vectors) are read and written successfully, but I'm not guaranteeing success)

In lists, the names are preserved.

Try and keep your objects simple (using types even more basic than R types :) ) and even on data sets, you find no object corruption, there can be on large data sets - \*\* if you use the advanced types such classes, data.frames etc \*\*

- 6. Key and Value Object Size: Are there limits? Yes, the serialized version of a key and object should be less than 64MB. I can fix this and will in future. For e.g. runif (8e6) is 61MB. Your keys and values should be less than this.
  - 7. java.lang.RuntimeException: RHMRMapRed.waitOutputThreads(): subprocess failed with code 141

This is because Hadoop broke the read/write pipe with the R code. To view the error, you'll need to go the job tracker website, click on one of the Failed attempts and see the error.

22 Chapter 7. FAQ

# PROTOBUFFER AND R

A package called rprotobuf which implements a simple serialization using Googles protocol buffers[1]. The package also includes some miscellaneous functions for writing/reading variable length encoded integers, and Base64 encoding/decoding related functions. The package can be downloaded from <a href="http://ml.stat.purdue.edu/rpackages/rprotobuf\_1.1.tar.gz">http://ml.stat.purdue.edu/rpackages/rprotobuf\_1.1.tar.gz</a>. It requires one to install libproto (Googles protobuffer library)

#### Requirements

Google's Protocol Buffer library. See [1].

#### **Brief Description**

The R objects that can be serialized are numerics, complex, integers, strings, logicals, raw, nulls and lists. Attributes of the aforementioned are preserved. NA is also preserved (for the above) As such, the objects include factors and matrices. The proto file can be found in the source.

Serialization/deserialization works perfectly for these types.

#### **Extras**

With version 1.1, rprotobuf will now serialize

- SYMSXP.
- LISTSXP
- CLOSXP
- ENVSXP ( no locking )
- PROMSXP
- LANGSXP
- DOTSXP
- S4SXP
- EXPRSXP

Serialization/deserialization(for these extras SEXP types) appear to work but I cannot prove that (one with a thorough knowledge of R internals needs to audit the code( src/message.cc )). They remain undocumented in the help pages.

#### Regrets

serialize.c (in R 2.9 sources) uses a hashtable to add references to previously added environments and symbols(instead of adding them again). This reduces the size of the serialized expresson. rprotobuf does not do any such thing. It ought to and in future it will.

### Download

 $\label{package} \textbf{Package} (with \ source): \ \texttt{http://ml.stat.purdue.edu/rpackages/rprotobuf\_1.1.tar.gz}$ 

[1] http://code.google.com/apis/protocolbuffers/docs/overview.html

Install
R CMD INSTALL rprotobuf\_1.1.tar.gz

**CHAPTER** 

NINE

# **DATATYPES**

I have tried to make RHIPE as flexible as possible with regards to data types exchanged. As such the following can be sent

- atomic vectors (raw, character, logical, complex, real, integer) (these include NA's)
- lists of the above and lists of lists. Names will be preserved.

So officially no matrices, data frames, time series objects etc. If there are needed, serialize them using serialize.

*Unofficially*, several attributes of are serialized, hence you can send matrices (the dim and dimnames attribute are involved) will be serialized. However, with regards to data frames the row.names attribute has caused me much grief. Keep it simple: scalar vectors (use as .vector to remove attributes) and lists (names are allowed).

I use data.frames too and it works. If you get a crash(error code 139,using version >= 0.53), email me. If in doubt, serialize.