

INTRODUCTION TO RCPP:

FROM SIMPLE EXAMPLES TO MACHINE LEARNING

PRE-CONFERENCE TUTORIAL

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Ketchum Trading; Debian and R Projects

VERY BROAD OUTLINE

Overview

- · Why?
- · How?

INTRODUCTION: WHY?

THREE KEY REASONS

- · Speed, Performance, ...
- \cdot Do things you could not do before
- $\cdot\,$ Easy to extend R this way

SIMPLE EXAMPLE

R Version of 'is this number odd or even'

```
isOdd_r <- function(num = 10L) {
    result = (num %% 2L == 1L)
    return(result)
}
isOdd_r(42L)</pre>
```

```
## [1] FALSE
```

SIMPLE EXAMPLE (CONT.)

C++ Version of 'is this number odd or even'

```
bool isOdd_cpp(int num = 10) {
  bool result = (num % 2 == 1);
  return result;
}
```

Free-standing code, not yet executable...

SIMPLE EXAMPLE (CONT.)

Rcpp Version of 'is this number odd or even'

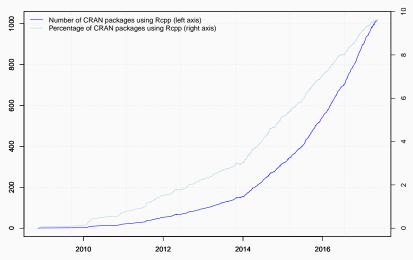
```
Rcpp::cppFunction("
bool isOdd_cpp(int num = 10) {
   bool result = (num % 2 == 1);
   return result;
}")
isOdd_cpp(42L)
```

[1] FALSE

AN ASIDE

GROWTH

Growth of Rcpp usage on CRAN



USERS ON CORE REPOSITORIES

Rcpp is currently used by

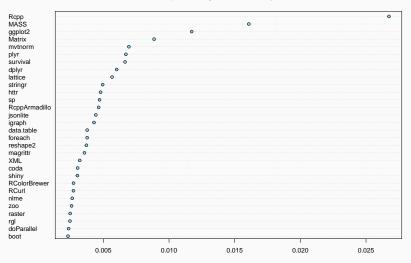
- 1022 CRAN packages (with 350 added since last year)
- 91 BioConductor packages
- · an unknown (but "large") number of GitHub projects

```
suppressMessages(library(utils))
library(pagerank) # cf github.com/andrie/pagerank

cran <- "http://cloud.r-project.org"
pr <- compute_pagerank(cran)
round(100*pr[1:5], 3)</pre>
```

```
## Rcpp MASS ggplot2 Matrix mvtnorm
## 2.675 1.608 1.172 0.887 0.694
```

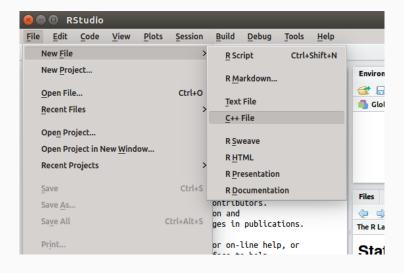
Top 30 of Page Rank as of May 2017



INTRODUCTION: How?

JUMPING RIGHT IN

RStudio makes starting very easy:



A FIRST EXAMPLE: CONT'ED

The following file gets created:

```
#include <Rcpp.h>
using namespace Rcpp;
// This is a simple example of exporting a C++ function to R. You can
// source this function into an R session using the Rcpp::sourceCpp
// function (or via the Source button on the editor toolbar). ...
// [[Rcpp::export]]
NumericVector timesTwo(NumericVector x) {
  return x * 2;
// You can include R code blocks in C++ files processed with sourceCpp
// (useful for testing and development). The R code will be automatically
// run after the compilation.
/*** R
timesTwo(42)
*/
```

A FIRST EXAMPLE: CONT'ED

So what just happened?

- · We defined a simple C++ function
- It operates on a numeric vector argument
- We asked Rcpp to 'source it' for us
- · Behind the scenes Rcpp creates a wrapper
- · Rcpp then compiles, links, and loads the wrapper
- The function is available in R under its C++ name

ANOTHER EXAMPLE: FOCUS ON SPEED

Consider a function defined as

$$f(n)$$
 such that
$$\begin{cases} n & \text{when } n < 2 \\ f(n-1) + f(n-2) & \text{when } n \geq 2 \end{cases}$$

AN INTRODUCTORY EXAMPLE: SIMPLE R IMPLEMENTATION

R implementation and use:

```
f <- function(n) {
    if (n < 2) return(n)
    return(f(n-1) + f(n-2))
}
## Using it on first 11 arguments
sapply(0:10, f)</pre>
```

```
## [1] 0 1 1 2 3 5 8 13 21 34 55
```

AN INTRODUCTORY EXAMPLE: TIMING R IMPLEMENTATION

Timing:

```
library(rbenchmark)
benchmark(f(10), f(15), f(20))[,1:4]
```

```
## test replications elapsed relative

## 1 f(10) 100 0.010 1.0

## 2 f(15) 100 0.110 11.0

## 3 f(20) 100 1.164 116.4
```

AN INTRODUCTORY EXAMPLE: C++ IMPLEMENTATION

```
int g(int n) {
    if (n < 2) return(n);
    return(g(n-1) + g(n-2));
}</pre>
```

deployed as

```
Rcpp::cppFunction('int g(int n) {
    if (n < 2) return(n);
    return(g(n-1) + g(n-2)); }')
## Using it on first 11 arguments
sapply(0:10, g)
## [1] 0 1 1 2 3 5 8 13 21 34 55</pre>
```

AN INTRODUCTORY EXAMPLE: COMPARING TIMING

Timing:

```
library(rbenchmark)
benchmark(f(20), g(20))[,1:4]
```

```
## test replications elapsed relative
## 1 f(20) 100 1.164 232.8
## 2 g(20) 100 0.005 1.0
```

A nice gain of a few orders of magnitude.

SOME BACKGROUND

R Type mapping

Standard R types (integer, numeric, list, function, ... and compound objects) are mapped to corresponding C++ types using extensive template meta-programming – it just works:

```
library(Rcpp)
cppFunction("NumericVector la(NumericVector x){
  return log(abs(x));
}")
la(seq(-5, 5, by=2))
```

Also note: vectorized C++!

Use of std::vector<double> and STL algorithms:

```
#include <Rcpp.h>
using namespace Rcpp;
inline double f(double x) { return ::log(::fabs(x)); }
// [[Rcpp::export]]
std::vector<double> logabs2(std::vector<double> x) {
 std::transform(x.begin(), x.end(), x.begin(), f);
 return x;
```

STL Type mapping

Used via

```
library(Rcpp)
sourceCpp("code/logabs2.cpp")
logabs2(seq(-5, 5, by=2))
```

TYPE MAPPING IS SEAMLESS

Simple outer product of a col.~vector (using RcppArmadillo):

```
library(Rcpp)
cppFunction("arma::mat v(arma::colvec a) {
          return a*a.t();}",
          depends="RcppArmadillo")
v(1:3)
```

```
## [,1] [,2] [,3]
## [1,] 1 2 3
## [2,] 2 4 6
## [3,] 3 6 9
```

Uses implicit conversion via as<> and wrap – cf package vignette Rcpp-extending.

C++11: LAMBDAS, AUTO, AND MUCH MORE

```
We can simplify the log(abs(...)) example further:
#include <Rcpp.h>
// [[Rcpp::plugins(cpp11)]]
using namespace Rcpp;
// [[Rcpp::export]]
std::vector<double> logabs3(std::vector<double> x) {
   std::transform(x.begin(), x.end(), x.begin(),
                  [](double x) {
                    return ::log(::fabs(x));
                  } ):
   return x;
```

USAGE

BASIC USAGE: EVALCPP()

[1] 1.79769e+308

evalCpp() evaluates a single C++ expression. Includes and dependencies can be declared.

This allows us to quickly check C++ constructs.

```
library(Rcpp)
evalCpp("2 + 2")  # simple test

## [1] 4

evalCpp("std::numeric_limits<double>::max()")
```

BASIC USAGE: CPPFUNCTION()

cppFunction() creates, compiles and links a C++ file, and creates
an R function to access it.

```
cppFunction("
    int exampleCpp11() {
        auto x = 10;
        return x;
}", plugins=c("cpp11"))
exampleCpp11() # same identifier as C++ function
```

BASIC USAGE: SOURCECPP()

sourceCpp() is the actual workhorse behind evalCpp()
andcppFunction(). It is described in more detail in the package
vignette Rcpp-attributes.

sourceCpp() builds on and extends cxxfunction() from package
inline, but provides even more ease-of-use, control and helpers freeing us from boilerplate scaffolding.

A key feature are the plugins and dependency options: other packages can provide a plugin to supply require compile-time parameters (cf RcppArmadillo, RcppEigen, RcppGSL).

BASIC USAGE: PACKAGES

Package are the standard unit of R code organization.

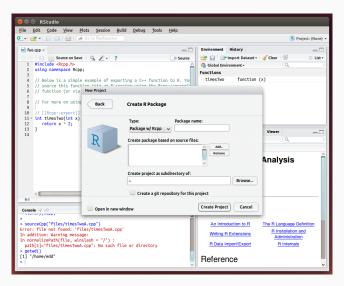
Creating packages with Rcpp is easy; an empty one to work from can be created by Rcpp.package.skeleton()

The vignette Rcpp-packages has fuller details.

As of mid May 2017, there are 1022 packages on CRAN which use Rcpp, and a further 91 on BioConductor — with working, tested, and reviewed examples.

PACKAGES AND RCPP

Best way to organize R code with Rcpp is via a package:



Rcpp.package.skeleton() and its derivatives. e.g.
RcppArmadillo.package.skeleton() create working packages.

```
// another simple example: outer product of a vector,
// returning a matrix
//
// [[Rcpp::export]]
arma::mat rcpparma_outerproduct(const arma::colvec & x) {
    arma::mat m = x * x.t();
    return m;
// and the inner product returns a scalar
//
// [[Rcpp::export]]
double rcpparma_innerproduct(const arma::colvec & x) {
    double v = arma::as scalar(x.t() * x);
    return v;
```

PACKAGES AND RCPP

Two (or three) ways to link to external libraries

- Full copies: Do what RcppMLPACK (v1) does and embed a full copy; larger build time, harder to update, self-contained
- With linking of libraries: Do what RcppGSL or RcppMLPACK (v2)
 do and use hooks in the package startup to store compiler and
 linker flags which are passed to environment variables
- With C++ template headers only: Do what RcppArmadillo and other do and just point to the headers

More details in extra vignettes.

MACHINE LEARNING

OVERVIEW

Among the 1000+ CRAN packages using Rcpp, several wrap Machine Learning libraries.

Here are three:

- · RcppShark based on Shark
- · RcppMLPACK based on MLPACK
- · dlib based on DLib

MLPACK

High-level:

- · Written by Ryan Curtin et al, Georgia Tech
- · Uses Armadillo, and like Armadillo, "feels right"
- Qiang Kou created 'RcppMLPACK v1', it is on CRAN

MLPACK 'v2'

High-level:

- A few of us are trying to update RcppMLPACK to 'v2'
- · Instead of embedding, amd external library is used
- · This makes deployment a little tricker on Windows and macOS

MLPACK

List of Algorithms:

- · Collaborative filtering (with many decomposition techniques)
- Decision stumps (one-level decision trees)
- Density estimation trees
- · Euclidean minimum spanning tree calculation
- · Gaussian mixture models
- · Hidden Markov models
- · Kernel Principal Components Analysis (optionally with sampling)
- · k-Means clustering (with several accelerated algorithms)
- Least-angle regression (LARS/LASSO)
- Linear regression (simple least-squares)
- · Local coordinate coding
- · Locality-sensitive hashing for approximate nearest neighbor search
- Logistic regression
- Max-kernel search
- · Naive Bayes classifier
- · Nearest neighbor search with dual-tree algorithms
- Neighborhood components analysis
- Non-negative matrix factorization
- Perceptrons
- · Principal components analysis (PCA)
- · RADICAL (independent components analysis)
- · Range search with dual-tree algorithms
- Rank-approximate nearest neighbor search
- Sparse coding with dictionary learning

RCPPMLPACK: K-MEANS EXAMPLE

```
#include "RcppMLPACK.h"
using namespace mlpack::kmeans;
using namespace Rcpp;
// [[Rcpp::depends(RcppMLPACK)]]
// [[Rcpp::export]]
List cppKmeans(const arma::mat& data, const int& clusters) {
    arma::Col<size t> assignments;
    KMeans<> k; // Initialize with the default arguments.
    k.Cluster(data, clusters, assignments);
    return List::create(Named("clusters") = clusters,
                        Named("result") = assignments);
```

RCPPMLPACK: K-MEANS EXAMPLE

Timing

Table 1: Benchmarking result

test	replications	elapsed	relative	user.self	sys.self
mlKmeans(t(wine), 3)	100	0.028	1.000	0.028	0.000
kmeans(wine, 3)	100	0.947	33.821	0.484	0.424

Table taken 'as is' from RcppMLPACK vignette.

RCPPMLPACK: LINEAR REGRESSION EXAMPLE

```
#include <RcppMLPACK.h>
                                    // MLPACK, Rcpp and RcppArmadillo
// particular algorithm used here
#include <mlpack/methods/linear regression/linear regression.hpp>
// [[Rcpp::export]]
Rcpp::List linearRegression(arma::mat& matX,
                            arma::vec& vecY,
                            const double lambda = 0.0.
                            const bool intercept = true) {
    matX = matX.t();
    mlpack::regression::LinearRegression lr(matX. vecY. lambda. intercept):
    arma::vec parameters = lr.Parameters();
    arma::vec fittedValues(vecY.n elem):
    lr.Predict(matX, fittedValues);
    return Rcpp::List::create(Rcpp::Named("parameters") = parameters,
                              Rcpp::Named("fitted") = fittedValues);
```

RCPPMLPACK: LINEAR REGRESSION EXAMPLE

[1] TRUF

```
suppressMessages(library(utils))
library(RcppMLPACK)
data("trees", package="datasets")
X <- with(trees, cbind(log(Girth), log(Height)))</pre>
y <- with(trees, log(Volume))</pre>
lmfit < - lm(v \sim X)
# summary(fitted(lmfit))
mlfit <- with(trees, linearRegression(X, y))</pre>
# summary(mlfit)
all.equal(unname(fitted(lmfit)), c(mlfit[["fitted"]]))
```

RCPPMLPACK: LOGISTIC REGRESSION EXAMPLE

```
#include <RcppMLPACK.h>
                                   // MLPACK, Rcpp and RcppArmadillo
#include <mlpack/methods/logistic regression/logistic regression.hpp> // algo use here
// [[Rcpp::export]]
Rcpp::List logisticRegression(const arma::mat& train. const arma::irowyec& labels.
                              const Rcpp::Nullable<Rcpp::NumericMatrix>& test = R NilValue) {
    // MLPACK wants Row<size t> which is an unsigned representation that R does not have
    arma::Row<size t> labelsur, resultsur;
    // TODO: check that all values are non-negative
    labelsur = arma::conv to<arma::Row<size t>>::from(labels);
    // Initialize with the default arguments. TODO: support more arguments>
    mlpack::regression::LogisticRegression<> lrc(train, labelsur);
    arma::vec parameters = lrc.Parameters();
    Rcpp::List return val:
    if (test.isNotNull()) {
        arma::mat test2 = Rcpp::as<arma::mat>(test):
       lrc.Classifv(test2. resultsur):
       arma::vec results = arma::conv to<arma::vec>::from(resultsur);
        return val = Rcpp::List::create(Rcpp::Named("parameters") = parameters,
                                        Rcpp::Named("results") = results):
    } else {
        return val = Rcpp::List::create(Rcpp::Named("parameters") = parameters);
    return return val:
```

RCPPMLPACK: LINEAR REGRESSION EXAMPLE

```
suppressMessages(library(utils))
library(RcppMLPACK)
example(logisticRegression)
##
## lgstcR> data(trainSet)
##
## lgstcR> mat <- t(trainSet[, -5]) ## train data, transpose and removing class labels
##
## lgstcR> lab <- trainSet[, 5]  ## class labels for train set
##
## lgstcR> logisticRegression(mat, lab)
## $parameters
## [1] -11.0819909 13.9022481 0.8034972 -9.3485217 -13.0869968
##
##
## lgstcR> testMat <- t(testSet[, -5]) ## test data
##
## lgstcR> logisticRegression(mat, lab, testMat)
## $parameters
## [1] -11.0819909 13.9022481 0.8034972 -9.3485217 -13.0869968
##
## $results
## [1] 0 0 0 1 1 1 1
```

RCPPMLPACK: NEAREST NEIGHBORS EXAMPLE

```
#include "RcppMLPACK.h"
using namespace Rcpp;
using namespace mlpack:
                                  using namespace mlpack::neighbor:
using namespace mlpack::metric;
                                  using namespace mlpack::tree;
// [[Rcpp::depends(RcppMLPACK)]]
// [[Rcpp::export]]
List nn(const arma::mat& data, const int k) {
    // using a test from MLPACK 1.0.10 file src/mlpack/tests/allknn test.cpp
    CoverTree<LMetric<2>. FirstPointIsRoot.
              NeighborSearchStat<NearestNeighborSort> > tree =
       CoverTree<LMetric<2>. FirstPointIsRoot.
                  NeighborSearchStat<NearestNeighborSort> >(data):
    NeighborSearch<NearestNeighborSort, LMetric<2>,
                   CoverTree<LMetric<2>. FirstPointIsRoot.
                             NeighborSearchStat<NearestNeighborSort> > >
       coverTreeSearch(&tree, data, true);
    arma::Mat<size t> coverTreeNeighbors:
    arma::mat coverTreeDistances;
    coverTreeSearch.Search(k, coverTreeNeighbors, coverTreeDistances);
    return List::create(Named("clusters") = coverTreeNeighbors,
                        Named("result") = coverTreeDistances);
```

EXTRA: RCPP EXAMPLES

A basic looped version:

```
#include <Rcpp.h>
#include <numeric> // for std::partial_sum
using namespace Rcpp;
// [[Rcpp::export]]
NumericVector cumsum1(NumericVector x){
    double acc = 0:  // init an accumulator variable
    NumericVector res(x.size()); // init result vector
    for(int i = 0; i < x.size(); i++){</pre>
         acc += x[i];
        res[i] = acc;
    return res;
```

CUMULATIVE SUM: vector-cumulative-sum

An STL variant:

```
// [[Rcpp::export]]
NumericVector cumsum2(NumericVector x){
    // initialize the result vector
    NumericVector res(x.size());
    std::partial_sum(x.begin(), x.end(), res.begin());
    return res;
}
```

CUMULATIVE SUM: vector-cumulative-sum

Or just Rcpp sugar:

```
// [[Rcpp::export]]
NumericVector cumsum_sug(NumericVector x){
   return cumsum(x); // compute + return result vector
}
```

Of course, all results are the same.

R FUNCTION CALL FROM C++: r-function-from-c++

```
#include <Rcpp.h>
using namespace Rcpp;
// [[Rcpp::export]]
NumericVector callFunction(NumericVector x,
                            Function f) {
    Numeric Vector res = f(x);
    return res;
/*** R
callFunction(x, fivenum)
*/
```

Using Boost via BH: using-boost-with-bh

```
// [[Rcpp::depends(BH)]]
#include <Rcpp.h>
// One include file from Boost
#include <boost/date time/gregorian/gregorian types.hpp>
using namespace boost::gregorian;
// [[Rcpp::export]]
Rcpp::Date getIMMDate(int mon. int vear) {
    // compute third Wednesday of given month / year
    date d = nth day of the week in month(
                       nth day of the week in month::third,
                       Wednesday, mon).get date(year);
    date::ymd type ymd = d.year month day();
    return Rcpp::wrap(Rcpp::Date(ymd.year, ymd.month, ymd.day));
```

Using Boost via BH: using-boost-with-bh

```
#include <Rcpp.h>
#include <boost/foreach.hpp>
using namespace Rcpp;
// [[Rcpp::depends(BH)]]
// the C-style upper-case macro name is a bit ugly
#define foreach BOOST FOREACH
// [[Rcpp::export]]
NumericVector square( NumericVector x ) {
 // elem is a reference to each element in x
 // we can re-assign to these elements as well
  foreach( double& elem, x ) {
    elem = elem*elem;
  return x:
```

C++11 now has something similar in a smarter for loop.

VECTOR SUBSETTING: subsetting

```
#include <Rcpp.h>
using namespace Rcpp;
// [[Rcpp::export]]
NumericVector positives(NumericVector x) {
   return x[x > 0];
// [[Rcpp::export]]
List first_three(List x) {
    IntegerVector idx = IntegerVector::create(0, 1, 2);
   return x[idx];
// [[Rcpp::export]]
List with_names(List x, CharacterVector y) {
    return x[y];
```

ARMADILLO EIGENVALUES: armadillo-eigenvalues

```
#include <RcppArmadillo.h>

// [[Rcpp::depends(RcppArmadillo)]]

// [[Rcpp::export]]
arma::vec getEigenValues(arma::mat M) {
    return arma::eig_sym(M);
}
```

ARMADILLO EIGENVALUES: armadillo-eigenvalues

```
sourceCpp("code/armaeigen.cpp")
set.seed(42)
X <- matrix(rnorm(4*4), 4, 4)</pre>
Z <- X %*% t(X)
getEigenValues(Z)
             [,1]
##
## [1,] 0.331887
## [2,] 1.685588
## [3,] 2.409920
## [4,] 14.210011
# R gets the same results (in reverse)
# and also returns the eigenvectors.
```

CREATE XTS FROM IN C++: creating-xts-from-c++

```
#include <Rcpp.h>
using namespace Rcpp;
NumericVector createXts(int sv. int ev) {
   IntegerVector ind = seg(sv, ev); // values
   NumericVector dv(ind);
                                   // date(time)s == reals
   dv = dv * 86400;
                                    // scaled to days
   dv.attr("tzone") = "UTC";
                                   // index has attributes
   dv.attr("tclass") = "Date";
   NumericVector xv(ind): // data has same index
   xv.attr("dim") = IntegerVector::create(ev-sv+1,1);
   xv.attr("index") = dv;
   CharacterVector cls = CharacterVector::create("xts","zoo");
   xv.attr("class") = cls;
   xv.attr(".indexCLASS") = "Date";
   // ... some more attributes ...
   return xv;
```

RCPPPARALLEL 1/3: parallel-matrix-transform

```
#include <Rcpp.h>
using namespace Rcpp;
#include <cmath>
#include <algorithm>
// [[Rcpp::export]]
NumericMatrix matrixSqrt(NumericMatrix orig) {
 // allocate the matrix we will return
  NumericMatrix mat(orig.nrow(), orig.ncol());
 // transform it
  std::transform(orig.begin(), orig.end(), mat.begin(), ::sqrt);
 // return the new matrix
  return mat;
```

RCPPPARALLEL 2/3: parallel-matrix-transform

```
// [[Rcpp::depends(RcppParallel)]]
#include <RcppParallel.h>
using namespace RcppParallel:
struct SquareRoot : public Worker {
   const RMatrix<double> input: // source matrix
   RMatrix<double> output; // destination matrix
   // initialize with source and destination
   SquareRoot(const NumericMatrix input, NumericMatrix output)
      : input(input), output(output) {}
   // take the square root of the range of elements requested
   void operator()(std::size t begin, std::size t end) {
      std::transform(input.begin() + begin,
                     input.begin() + end,
                     output.begin() + begin,
                     ::sart):
```

RCPPPARALLEL 3/3: parallel-matrix-transform

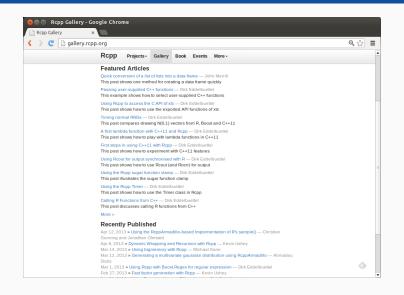
```
// [[Rcpp::export]]
NumericMatrix parallelMatrixSqrt(NumericMatrix x) {
 // allocate the output matrix
  NumericMatrix output(x.nrow(), x.ncol());
  // SquareRoot functor (pass input and output matrixes)
  SquareRoot squareRoot(x, output);
 // call parallelFor to do the work
  parallelFor(0, x.length(), squareRoot);
 // return the output matrix
  return output:
```

More

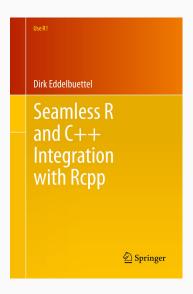
DOCUMENTATION

- The package comes with eight pdf vignettes, and numerous help pages.
- The introductory vignettes are now published (Rcpp and RcppEigen in J Stat Software, RcppArmadillo in Comp Stat & Data Anlys)
- The rcpp-devel list is *the* recommended resource, generally very helpful, and fairly low volume.
- · StackOverflow has a fair number of posts too.
- And a number of blog posts introduce/discuss features.

RCPP GALLERY



THE RCPP BOOK



On sale since June 2013.

Questions?

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