

# Rcpp by Examples

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Sydney Users of R Forum (suRf)

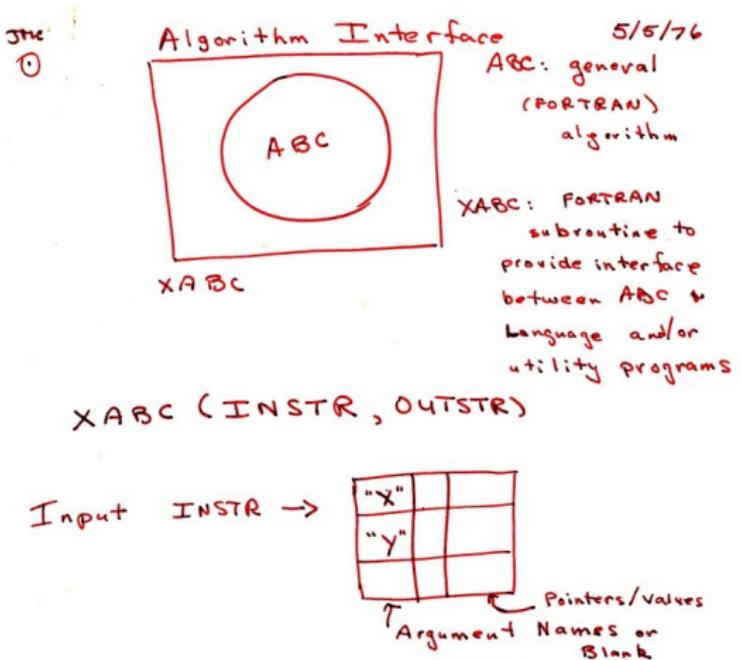
Sydney, Australia

10 July 2013

# Outline

- 1 Introduction
- 2 Usage
- 3 Sugar
- 4 Examples from the Rcpp Gallery
- 5 RInside
- 6 More

# A “vision” from Bell Labs from 1976



Source: John Chambers' talk at Stanford in October 2010; personal correspondence.

# An Introductory Example

Consider a function defined as

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

# 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)  
  
## [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  0.034     1.00
## 2 f(15)          100  0.394    11.59
## 3 f(20)          100  4.384   128.94
```

# An Introductory Example: C++ Implementation

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

Deployed as:

```
library(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
```

# An Introductory Example: Comparing timing

Timing:

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

##      test replications elapsed relative
## 1 f(20)          100   4.286    612.3
## 2 g(20)          100   0.007     1.0
```

A nice 600-fold gain.

# 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 logabs(NumericVector x) {
        return log(abs(x));
    }
")
logabs(seq(-5, 5, by=2))

## [1] 1.609 1.099 0.000 0.000 1.099 1.609
```

Also note: vectorized C++!

# Type mapping also with C++ STL types

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;
}
```

# Type mapping also with C++ STL types

Used via

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

## [1] 1.609 1.099 0.000 0.000 1.099 1.609
```

# Type mapping is seamless

Simple outer product of a column vector (using Armadillo / RcppArmadillo):

```
cppFunction("arma::mat v(arma::colvec a) {return  
a*a.t();}", depends="RcppArmadillo")  
v(1:5)  
  
##      [,1] [,2] [,3] [,4] [,5]  
## [1,]     1     2     3     4     5  
## [2,]     2     4     6     8    10  
## [3,]     3     6     9    12    15  
## [4,]     4     8    12    16    20  
## [5,]     5    10    15    20    25
```

This uses implicit conversion via `as<>` and `wrap - cf` package vignette Rcpp-extending.

# Basic Usage: FastLm using RcppArmadillo

## A simple but speedy reimplementation of `lm()`

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

#include <RcppArmadillo.h>

using namespace Rcpp;

// [[Rcpp::export]]
List fastLm(NumericVector yr, NumericMatrix Xr) {

    int n = Xr.nrow(), k = Xr.ncol();

    arma::mat X(Xr.begin(), n, k, false);
    arma::colvec y(yr.begin(), yr.size(), false);

    arma::colvec coef = arma::solve(X, y);
    arma::colvec resid = y - X*coef;

    double sig2 = arma::as_scalar(arma::trans(resid)*resid/(n-k));
    arma::colvec stderrest = arma::sqrt(
        sig2 * arma::diagvec(arma::inv(arma::trans(X)*X)) );

    return List::create(Named("coefficients") = coef,
                        Named("stderr") = stderrest);
}
```

# Well-known packages using Rcpp

`Amelia` by Gary King et al: Multiple Imputation from cross-section, time-series or both; uses Rcpp and RcppArmadillo

`forecast` by Rob Hyndman et al: Time-series forecasting including state space and automated ARIMA modeling; uses Rcpp and RcppArmadillo

`RStan` by Andrew Gelman et al: Rcpp helps with automatic model parsing / generation for MCMC / Bayesian modeling

`rugarch` by Alexios Ghalanos: Sophisticated financial time series models using Rcpp and RcppArmadillo

`bigviz` by Hadley Wickham: High-performance visualization of datasets in the 10-100 million observations range

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# Basic Usage: evalCpp

`evalCpp()` evaluates a single C++ expression. Includes and dependencies can be declared.  
This allows us to quickly check C++ constructs.

```
evalCpp( "std::numeric_limits<double>::max()" )  
## [1] 1.798e+308
```

# Basic Usage: `cppFunction()`

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

```
cppFunction("  
    int useCpp11() {  
        auto x = 10;  
        return x;  
}", plugins=c("cpp11"))  
useCpp11() # same identifier as C++ function  
  
## [1] 10
```

# Basic Usage: `sourceCpp()`

`sourceCpp()` is the actual workhorse behind `evalCpp()` and `cppFunction()`. 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`).

We are also starting to provide other compiler features via plugins. A first plugin to enable C++11 support was added in `Rcpp` 0.10.3.

# Basic Usage: Packages

Packages are *the* standard unit of R code organization.

Creating packages with Rcpp is easy; an minimal one to extend from can be created by calling `Rcpp.package.skeleton()`

The vignette [Rcpp-package](#) has fuller details.

As of July 2013, there are 125 packages on CRAN which use Rcpp, and a further 10 on BioConductor — with working, tested, and reviewed examples.

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# Syntactic 'sugar': Simulating $\pi$ in R

Basic idea: for point  $(x, y)$ , compute distance to origin. Do so repeatedly, and the ratio of points below one to number N of simulations will approach  $\pi/4$  as we fill the area of one quarter of the unit circle.

```
piR <- function(N) {  
  x <- runif(N)  
  y <- runif(N)  
  d <- sqrt(x^2 + y^2)  
  return(4 * sum(d <= 1.0) / N)  
}  
  
set.seed(5)  
sapply(10^(3:6), piR)  
  
## [1] 3.156 3.155 3.139 3.141
```

# Syntactic 'sugar': Simulating $\pi$ in C++

The neat thing about *Rcpp sugar* is that it enables us to write C++ code that looks almost as compact.

```
#include <Rcpp.h>
using namespace Rcpp;

// [[Rcpp::export]]
double piSugar(const int N) {
    RNGScope scope; // ensure RNG gets set/reset
    NumericVector x = runif(N);
    NumericVector y = runif(N);
    NumericVector d = sqrt(x*x + y*y);
    return 4.0 * sum(d <= 1.0) / N;
}
```

Apart from RNG set/reset, the code is essentially identical.

# Syntactic 'sugar': Simulating $\pi$

And by using the same RNG, so are the results.

```
sourceCpp("code/piSugar.cpp")
set.seed(42); a <- piR(1.0e7)
set.seed(42); b <- piSugar(1.0e7)
identical(a,b)

## [1] TRUE

print(c(a,b), digits=7)

## [1] 3.140899 3.140899
```

# Syntactic 'sugar': Simulating $\pi$

Here, the performance gain is less dramatic as the R code is already vectorised:

```
library(rbenchmark)
benchmark(piR(1.0e6), piSugar(1.0e6)) [,1:4]

##           test replications elapsed relative
## 1    piR(1e+06)          100  13.407    1.733
## 2 piSugar(1e+06)          100   7.738    1.000
```

More about Sugar is in the package vignette Rcpp-sugar.

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# From the Rcpp Gallery

The screenshot shows a web browser window for the Rcpp Gallery. The title bar says "Rcpp Gallery - Google Chrome". The address bar shows "Rcpp Gallery" and "gallery.rcpp.org". The page content is organized into sections:

- Featured Articles**: A list of 15 articles with titles like "Quick conversion of a list of lists into a data frame", "Passing user-supplied C++ functions", and "Using Rcpp to access the C API of xts". Each article has a link to its full post.
- Recently Published**: A list of 6 recent posts with dates and titles:
  - Apr 12, 2013 » Using the RcppArmadillo-based Implementation of R's sample() — Christian Gunning and Jonathan Olmsted
  - Apr 8, 2013 » Dynamic Wrapping and Recursion with Rcpp — Kevin Ushey
  - Mar 14, 2013 » Using bigmemory with Rcpp — Michael Kane
  - Mar 12, 2013 » Generating a multivariate gaussian distribution using RcppArmadillo — Ahmadou Dicko
  - Mar 1, 2013 » Using Rcpp with Boost.Regex for regular expression — Dirk Eddelbuettel
  - Feb 27, 2013 » Fast factor generation with Rcpp — Kevin Ushey

# Cumulative Sum

See <http://gallery.rcpp.org/articles/vector-cumulative-sum/>

A basic looped version:

```
#include <Rcpp.h>
#include <numeric>      // for std::partial_sum used below
using namespace Rcpp;

[[Rcpp::export]]
NumericVector cumsum1(NumericVector x) {
    // initialize an accumulator variable
    double acc = 0;

    // initialize the result vector
    NumericVector res(x.size());

    for(int i = 0; i < x.size(); i++) {
        acc += x[i];
        res[i] = acc;
    }
    return res;
}
```

# Cumulative Sum

See <http://gallery.rcpp.org/articles/vector-cumulative-sum/>

An STL variant:

```
// [[Rcpp::export]]
NumericVector cumsum2(NumericVector x) {

    // initialize the result vector
    NumericVector res(x.size());

    // use STL algorithm
    std::partial_sum(x.begin(), x.end(), res.begin());

    return res;
}
```

# Cumulative Sum

See <http://gallery.rcpp.org/articles/vector-cumulative-sum/>

Or just *Rcpp sugar* for one-line solution:

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

Of course, all results are the same.

```
cppFunction('NumericVector cumsum_sug(NumericVector
x) { return cumsum(x); }')
x <- 1:10
all.equal(cumsum_sug(x), cumsum(x))

## [1] TRUE
```

# Sugar head and tail

See <http://gallery.rcpp.org/articles/sugar-head-tail/>

The  $n$  largest values of a vector:

```
#include <Rcpp.h>
using namespace Rcpp;

// [[Rcpp::export]]
NumericVector top_n(NumericVector y, int n) {
    NumericVector x = clone(y);
    // sort x in ascending order
    std::sort(x.begin(), x.end());
    return tail(x, n);
}
```

# Sugar head and tail

See <http://gallery.rcpp.org/articles/sugar-head-tail/>

The  $n$  smallest:

```
#include <Rcpp.h>
using namespace Rcpp;

// [[Rcpp::export]]
NumericVector bottom_n(NumericVector y,
                       int n) {
  NumericVector x = clone(y);
  // sort x in ascending order
  std::sort(x.begin(), x.end());
  return head(x, n);
}
```

# Sugar head and tail

See <http://gallery.rcpp.org/articles/sugar-head-tail/>

## Example:

```
sourceCpp("code/sugar_head_tail.cpp")
set.seed(42)
x <- rnorm(5)
x

## [1] 1.3710 -0.5647  0.3631  0.6329  0.4043

top_n(x, 3)

## [1] 0.4043 0.6329 1.3710

bottom_n(x, 3)

## [1] -0.5647  0.3631  0.4043
```

# Using the R Rmath functions

See <http://gallery.rcpp.org/articles/using-rmath-functions/>

Besides all the vectorised d/p/q/r distribution functions available via *Rcpp sugar*, we can also access the R Rmath interface.

```
#include <Rcpp.h>
using namespace Rcpp;

// [[Rcpp::export]]
NumericVector mypnorm(NumericVector x) {
    int n = x.size();
    NumericVector y(n);

    for (int i=0; i<n; i++)
        y[i] = R:::pnorm(x[i], 0.0, 1.0, 1, 0);

    return y;
}
```

# Using the R Rmath functions

See <http://gallery.rcpp.org/articles/using-rmath-functions/>

A simple illustration:

```
sourceCpp("code/using-rmath-rng.cpp")
x <- seq(-2, 2)
mynorm(x)

## [1] 0.02275 0.15866 0.50000 0.84134 0.97725

pnorm(x)

## [1] 0.02275 0.15866 0.50000 0.84134 0.97725
```

The vectorised variants available via *Rcpp sugar* are even more convenient.

# Armadillo subsetting

See <http://gallery.rcpp.org/articles/armadillo-subsetting/>

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
using namespace arma;

// [[Rcpp::export]]
mat matrixSubset (mat M)  {

    // logical condition: where is transpose larger?
    umat a = trans(M) > M;
    mat N = conv_to<mat>::from(a);

    return N;
}
```

# Armadillo subsetting

See <http://gallery.rcpp.org/articles/armadillo-subsetting/>

```
M <- matrix(1:9, 3, 3)
```

```
M
```

```
##      [,1] [,2] [,3]
## [1,]     1     4     7
## [2,]     2     5     8
## [3,]     3     6     9
```

```
matrixSubset(M)
```

```
##      [,1] [,2] [,3]
## [1,]     0     0     0
## [2,]     1     0     0
## [3,]     1     1     0
```

# Armadillo subsetting

See <http://gallery.rcpp.org/articles/armadillo-subsetting/>

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
// [[Rcpp::export]]
arma::vec matrixSubset2(arma::mat M) {
    arma::mat Z = M * M.t();

    // extracting element matching a condition
    arma::vec v = Z.elem(arma::find(Z >= 100));
    return v;
}
```

## matrixSubset2 (M)

```
##      [,1]
## [1,] 108
## [2,] 108
## [3,] 126
```

# A C++11 example: auto

See <http://gallery.rcpp.org/articles/first-steps-with-C++11/>

```
#include <Rcpp.h>

// Enable C++11 via plugin (Rcpp >= 0.10.3)
// [[Rcpp::plugins(cpp11)]]

// [[Rcpp::export]]
int useAuto() {
    auto val = 42;    // val will be of type int
    return val;
}
```

# A C++11 example: Initialiser lists

See <http://gallery.rcpp.org/articles/first-steps-with-C++11/>

```
#include <Rcpp.h>

// Enable C++11 via plugin (Rcpp >= 0.10.3)
// [[Rcpp::plugins(cpp11)]]

using namespace std::vector;
using namespace std::string;

// [[Rcpp::export]]
vector<string> useInitLists() {
  vector<string> vec = {"larry", "curly", "moe"};
  return vec;
}
```

# A C++11 example: Loops over ranges

See <http://gallery.rcpp.org/articles/first-steps-with-C++11/>

```
#include <Rcpp.h>

// Enable C++11 via plugin (Rcpp >= 0.10.3)
// [[Rcpp::plugins(cpp11)]]

// [[Rcpp::export]]
int simpleProd(std::vector<int> vec) {
    int prod = 1;
    for (int &x : vec) { // loop over all vals of vec
        prod *= x;           // compute product
    }
    return prod;
}
```

# A simple C++ Lambda example

See <http://gallery.rcpp.org/articles/simple-lambda-func-c++11/>

```
#include <Rcpp.h>

using namespace Rcpp;

// Important: enable C++11 via plugin
// [[Rcpp::plugins(cpp11)]]

// [[Rcpp::export]]
std::vector<double>
transformEx(const std::vector<double>& x) {
  std::vector<double> y(x.size());
  std::transform(x.begin(), x.end(), y.begin(),
                [](double x) { return x*x; });
  return y;
}
```

# A simple C++ Lambda example

See <http://gallery.rcpp.org/articles/simple-lambda-func-c++11/>

An R example use of the 'lambda' function:

```
sourceCpp("code/lambda.cpp")
x <- c(1,2,3,4)
transformEx(x)

## [1] 1 4 9 16
```

# Using Boost via BH

See <http://gallery.rcpp.org/articles/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 year) {
    // 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

See <http://gallery.rcpp.org/articles/using-boost-with-bh/>

We can test this from R:

```
sourceCpp("code/boost-bh.cpp")
getIMMDate(9, 2013)

## [1] "2013-09-18"

getIMMDate(3, 2020)

## [1] "2020-03-18"

getIMMDate(12, 2030)

## [1] "2030-12-18"
```

# Using Exceptions

See <http://gallery.rcpp.org/articles/intro-to-exceptions/>

```
#include <Rcpp.h>

using namespace Rcpp;

<// [[Rcpp::export]]
double takeLog(double val) {
  try {
    if (val <= 0.0) {                                // log() not defined here
      throw std::range_error("Inadmissible value");
    }
    return log(val);
  } catch(std::exception &ex) {
    forward_exception_to_r(ex);
  } catch(...) {
    ::Rf_error("c++ exception (unknown reason)");
  }
  return NA_REAL;                               // not reached
}
```

# Using Exceptions

See <http://gallery.rcpp.org/articles/intro-to-exceptions/>

```
takeLog(exp(1))      # works

## [1] 1

takeLog(-1.0)        # throws exception

## Error: Inadmissible value

takeLog(exp(2))      # but carries on

## [1] 2
```

# Calling an R function from C++

See <http://gallery.rcpp.org/articles/r-function-from-c++/>

```
#include <Rcpp.h>

using namespace Rcpp;

// [[Rcpp::export]]
NumericVector callFunction(NumericVector x,
                           Function f) {
  NumericVector res = f(x);
  return res;
}
```

# Calling an R function from C++

See <http://gallery.rcpp.org/articles/r-function-from-cpp/>

R calling a C++ function supplying an R function to be called:

```
sourceCpp("code/r-from-cpp.cpp")
set.seed(42)
x <- rnorm(1e5)
callFunction(x, fivenum)

## [1] -4.043276 -0.682384 -0.002066  0.673325  4.328091

# result of course same as:
fivenum(x)

## [1] -4.043276 -0.682384 -0.002066  0.673325  4.328091
```

# Armadillo Eigenvalues

See <http://gallery.rcpp.org/articles/armadillo-eigenvalues/>

```
#include <RcppArmadillo.h>

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

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

# Armadillo Eigenvalues

See <http://gallery.rcpp.org/articles/armadillo-eigenvalues/>

```
set.seed(42)
X <- matrix(rnorm(4*4), 4, 4)
Z <- X %*% t(X)
getEigenValues(Z)

##          [,1]
## [1,]  0.3319
## [2,]  1.6856
## [3,]  2.4099
## [4,] 14.2100

# R gets the same results (in reverse)
# and also returns the eigenvectors.
```

# Multivariate Normal RNG Draw

<http://gallery.rcpp.org/articles/simulate-multivariate-normal/>

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]

using namespace Rcpp;

// [[Rcpp::export]]
arma::mat mvrnormArma(int n, arma::vec mu,
                      arma::mat sigma) {
  int ncols = sigma.n_cols;
  arma::mat Y = arma::randn(n, ncols);
  return arma::repmat(mu, 1, n).t() +
    Y * arma::chol(sigma);
}
```

# Passing any R object with ease: Sparse Matrix

See <http://gallery.rcpp.org/articles/armadillo-sparse-matrix/>

Define a sparse matrix as an S4 object:

```
suppressMessages(library(Matrix))
i <- c(1,3:7); j <- c(2,9,6:9); x <- 6 * (1:6)
A <- sparseMatrix(i, j, x = x)
A

## 7 x 9 sparse Matrix of class "dgCMatrix"
##
## [1,] . 6 . . . . . .
## [2,] . . . . . . . .
## [3,] . . . . . . . 12
## [4,] . . . . . 18 . . .
## [5,] . . . . . . 24 . .
## [6,] . . . . . . . 30 .
## [7,] . . . . . . . . 36
```

# Passing any R object with ease: Sparse Matrix

See <http://gallery.rcpp.org/articles/armadillo-sparse-matrix/>

Access it in C++ with ease:

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
using namespace Rcpp;

// [[Rcpp::export]]
void accessSparse(S4 mat) {
    IntegerVector dims = mat.slot("Dim");
    IntegerVector i = mat.slot("i");
    IntegerVector p = mat.slot("p");
    NumericVector x = mat.slot("x");

    int nrow = dims[0], ncol = dims[1];
    arma::sp_mat res(nrow, ncol);
    // ... code now using Armadillo's sparse matrix
```

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# The first example

examples/standard/rinside\_sample0.cpp

```
// the embedded R via RInside
#include <RInside.h>

int main(int argc, char *argv[]) {

    // create an embedded R instance
    RInside R(argc, argv);

    // assign a char* (string) to 'txt'
    R["txt"] = "Hello, world!\n";

    // eval the init string, ignoring returns
    R.parseEvalQ("cat(txt)");

    exit(0);
}
```

# RInside in a nutshell

## Key aspects:

- RInside uses the embedding API of R
- An instance of R is launched by the RInside constructor
- It behaves just like a regular R process
- We submit commands as C++ strings which are parsed and evaluated
- Rcpp is used to easily get data in and out from the enclosing C++ program.

# Application example: Qt

RInside examples/qt/

The question is sometimes asked how to embed **RInside** in a larger program.

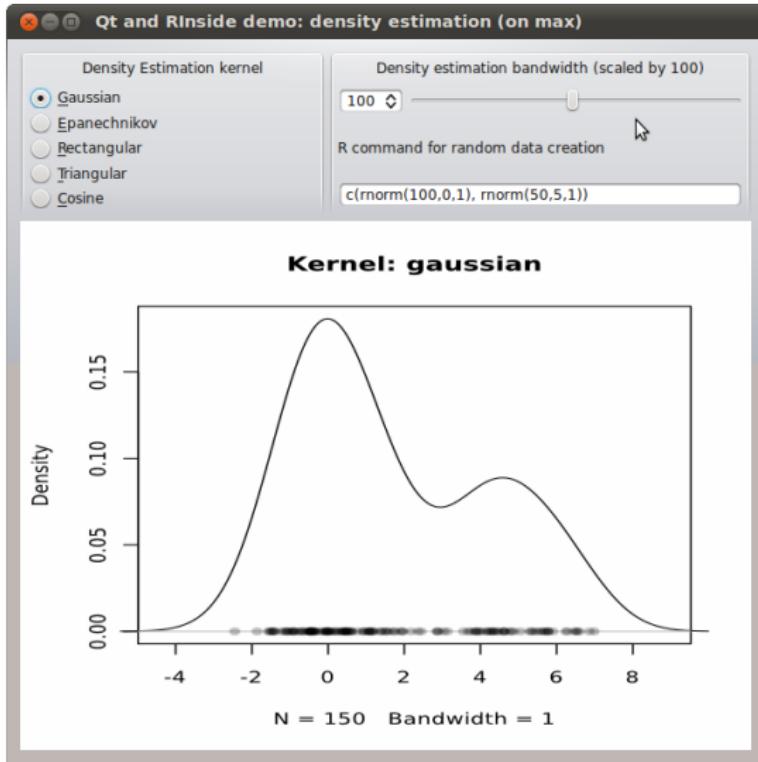
We have a nice example using **Qt**:

```
#include <QApplication>
#include "qtdensity.h"

int main(int argc, char *argv[]) {
    RInside R(argc, argv);           // embedded R inst.
    QApplication app(argc, argv);
    QtDensity qtdensity(R);         // passess by ref.
    return app.exec();
}
```

# Application example: Qt density slider

RInside examples/qt/



This uses standard **Qt** / GUI paradigms of

- radio buttons
- sliders
- textentry

all of which send values to the R process which provides a PNG image that is plotted.

# Application example: Wt

RInside examples/wt/

Given the desktop application with **Qt**, the question arises how to deliver something similar “over the web” — and **Wt** helps.

Witty WebApp With Rinside - Google Chrome  
dirk.eddelbuettel.com:8088

**Overview**

This example demonstrates some of the capabilities of the the [wt library](#), in combination with the [RInside](#) classes for embedding the [R](#) statistical language and environment.

It implements a standard GUI / application setting: drawing from a random distribution, and estimation a non-parametric density for which the user selects the kernel and bandwidth. [RInside](#) already contains an example of this using [Qt](#) to provide a standard application.

Here we show how to do the same in a web application which, thanks to the abstractions provided by the [wt](#), is rather straightforward.

**User Input for Density Estimation**

Density estimation scale factor (div. by 100)  
100

R Command for data generation

Gaussian  
 Epanechnikov  
 Rectangular  
 Triangular  
 Cosine

**Resulting R Chart**

Kernel: gaussian

Density

0.20  
0.15  
0.10  
0.05  
0.00

**Wt** is similar to **Qt** so the code needs only a few changes.

**Wt** takes care of all browser / app interactions and determines the most featureful deployment.

# Outline

- 1 Introduction
- 2 Usage
- 3 Sugar
- 4 Examples from the Rcpp Gallery
- 5 RInside
- 6 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 Anal.*).
- The rcpp-devel list is *the* recommended resource, generally very helpful, and fairly low volume.
- By now StackOverflow has a fair number of posts too.
- And a number of blog posts introduce/discuss features.

# Rcpp Gallery

The screenshot shows a web browser window for the Rcpp Gallery. The title bar says "Rcpp Gallery - Google Chrome". The address bar shows "Rcpp Gallery" and "gallery.rcpp.org". The main content area has a header with tabs: "Rcpp", "Projects", "Gallery" (which is selected), "Book", "Events", and "More". Below the tabs is a section titled "Featured Articles" which lists several posts. At the bottom of the article list is a link "More ». Below that is a section titled "Recently Published" with a list of recent posts. The right side of the browser window has scroll bars.

## Featured Articles

- [Quick conversion of a list of lists into a data frame](#) — John Merrill  
This post shows one method for creating a data frame quickly
- [Passing user-supplied C++ functions](#) — Dirk Eddelbuettel  
This example shows how to select user-supplied C++ functions
- [Using Rcpp to access the C API of xts](#) — Dirk Eddelbuettel  
This post shows how to use the exported API functions of xts
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This post compares drawing N(0,1) vectors from R, Boost and C++11
- [A first lambda function with C++11 and Rcpp](#) — Dirk Eddelbuettel  
This post shows how to play with lambda functions in C++11
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This post discusses calling R functions from C++

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# The Rcpp book

Use R!

Dirk Eddelbuettel

Seamless R  
and C++  
Integration  
with Rcpp

In print since June  
2013



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