



ETC4500/ETC5450 Advanced R programming

Week 11: Rewriting R code in C++



Outline

- 1 Motivation
- The first steps with Rcpp
- 3 Some stats with RcppArmadillo
- 4 Create an R package with compiled code in ten steps

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About me

Tomasz Woźniak

- Senior Lecturer in Econometrics at the unimelb
- Econometrician: Bayesian time series analyst
- Develops methods for applied macro research
- Loves cycling, yoga, books, volunteering, contemporary theatre, music, and art
- I am nice!

About me

Tomasz Woźniak

- **R** enthusiast and specialised user for 16 years
- **bsvars** package author (more coming up)



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Motivations

- Coding in C++ for R applications has always been possible
- It requires:
 - writing C++ code
 - compiling it, and
 - linking it to R
- Difficulties:
 - tedious object-oriented programming
 - necessity of assuring object compatibility
- Benefits are great, but the cost was too high

Motivations

- **Rcpp** is a family of packages by Dirk Eddelbuetel et al. facilitating the application of **C++** to **R**
- An interface for communication between **R** and **C++**
- Greatly simplifies the workflow
- Easier to benefit from the best of the two worlds:
 - C++ programs are pre-compiled assuring fast computations perfect for writing functions
 - R code is interpreted and dynamic: perfect for data analysis

Objectives for this session

- to facilitate working with C++ for R applications
- to perform a sequence of exercises
- to focus on:
 - basic programming structures
 - functional programming
 - object types: scalars, vectors, matrices, lists, etc.
 - linear algebra
 - statistical distributions

Materials for this session

- Lecture slides
- **C++** scripts:
 - ▶ nicetry.cpp
 - ▶ nicelr.cpp
 - nicelist.cpp
 - nicerig2.cpp
- **R** scripts:
 - nicepackage.R

learning resources

- This session!
- vignettes: for packages Rcpp and RcppArmadillo
- online resources:
 - Armadillo library documentation
 - RcppGallery
 - stackoverflow.com tag:rcpp
- François, R., Optimizing R Code with Rcpp on datacamp
- Tsuda, M., Rcpp for everyone
- Eddelbuettel, D., Seamless R and C++ Integration with Rcpp

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The first steps with Rcpp

Consider the following **C++** applications in **R**:

- Define a C++ function in an R script
 - promptly available for fast computations
- Develop a C++ function in a .cpp file
 - perfect for developing, testing, and benchmarking
- Use a function from a *.cpp file in **R** computations
 - perfect for elaborate projects
- Develop an R package using C++ code
 - perfect for sharing your work with the community

Define a C++ function in an R script

```
Rcpp::cppFunction('
  DataFrame nicetry (int n) {
    NumericVector v = rnorm(n);
    IntegerVector x = seq_len(n);
    LogicalVector y = v > 0;
    CharacterVector z(n, "nice");
    return DataFrame::create(_["v"] = v, _["x"] = x, _["y"] = y, _["z"] = z);
}
')
nicetry(2)
```

```
V X Y Z
1 0.945 1 TRUE nice
2 -0.664 2 FALSE nice
```

Develop a C++ function in a nicetry.cpp file

A *.cpp file sample contents:

```
#include <Rcpp.h>
using namespace Rcpp;
// [[Rcpp::export]]
List nicetry (int n) {
  NumericVector v = rnorm(n);
  IntegerVector x = sea len(n):
  Logical Vector y = v > 0;
  CharacterVector z(n, "nice");
  return List::create( ["v"] = v, ["x"] = x, ["v"] = v, ["z"] = z);
/*** R
nicetrv(2)
*/
```

Develop a C++ function in a nicetry.cpp file

The script includes:

Rcpp library and namespace declarations (skip: Rcpp::)

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

Rcpp marker to export the nicetry function to R

```
// [[Rcpp::export]]
```

sample R script

```
/*** R
nicetry(2)
*/
```

Develop a C++ function in a nicetry.cpp file

The script includes:

the function definition

Develop a C++ function in a .cpp file

- Your turn!
- Develop a C++ function that creates a Tx3 matrix with:
 - \blacksquare an integer \top as the only argument
 - a constant term column
 - \blacksquare a linear trend $t \bar{t}$ column
 - \blacksquare a quadratic trend $(t \bar{t})^2$ column

where t goes from 1 to T, and \bar{t} is the mean of sequence t.

Get some help HERE.

Use a function from a nicelist.cpp file in R

nicelist.cpp file contents:

```
#include <Rcpp.h>
using namespace Rcpp;
// [[Rcpp::export]]
List nicelist (int n) {
 NumericVector p = rnorm(n);
 NumericVector s(n):
 for (int i=0; i<n; i++) {
    s[i] = pow(p[i], 2);
 return List::create(_["p"] = p, _["s"] = s);
```

Use a function from a nicelist.cpp file in R

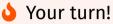
■ **R** script using the function from nicelist.cpp:

```
Rcpp::sourceCpp("nicelist.cpp")
nicelist(3)

$p
[1] -1.548 -0.554 -0.601

$s
[1] 2.397 0.307 0.362
```

Develop a C++ function in a .cpp file



Consider a Gaussian random walk:

$$y_t = y_{t-1} + \varepsilon_t, \qquad \varepsilon_t \sim N(0, 1), \qquad y_0 = 0$$

Develop a **C++** function that:

- has an integer T as the only argument
- returns a T-vector with Gaussian random walk

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Some stats with RcppArmadillo

- Data objects from Rcpp have limited functionality
- **Armadillo** is a **C++** library for linear algebra that
 - provides a rich set of functions
 - has a simple and intuitive syntax
 - includes fast linear algebra routines, and
 - fast random number generators
 - has fantastic documentation
- RcppArmadillo is a simplified interface with Armadillo
 - allows seamless integration with Rcpp
 - easily passes data between R and C++

Some stats with RcppArmadillo: linear regression

Contents of a nicelr.cpp file:

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
using namespace arma;
// [[Rcpp::export]]
vec nicelr (vec y, mat x) {
  vec beta_hat = solve(x.t() * x, x.t() * y);
  return beta_hat;
/*** R
x = cbind(rep(1,5),1:5); y = x %*% c(1,2) + rnorm(5)
nicelr(y, x)
*/
```

Some stats with RcppArmadillo: linear regression



Extend the nicelr function to return also the covariance of $\hat{\beta}$:

$$\widehat{Cov}\left[\widehat{\beta}\right] = \widehat{\sigma}^2 \left(X'X\right)^{-1}, \qquad \widehat{\sigma}^2 = \frac{1}{T} \left(Y - \widehat{\beta}X\right)' \left(Y - \widehat{\beta}X\right)$$

- don't adjust the arguments
- return beta_hat and cov_beta_hat in a list

Get some help HERE.

Sampling random draws from an inverted gamma 2 distribution.

A positive random variable σ^2 following an inverted gamma 2 distribution with positive scale s and shape ν parameters is denoted by:

$$\sigma^2 \sim \mathsf{IG2}(\mathsf{s}, \nu)$$

- Generate random draw x from $\chi^2(\nu)$
- Return $\frac{s}{x}$

Contents of a nicerig2.cpp file:

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
using namespace arma;
// [[Rcpp::export]]
vec nicerig2 (const int n, const double s, const double nu) {
  vec rig2 = s / chi2rnd( nu, n );
  return rig2;
/*** R
nicerig2(2, 1, 1)
```

Normal-inverted gamma 2 distribution.

Random variables, an *N*-vector \mathbf{x} and a positive scalar, σ^2 , following the normal-inverted gamma 2 distribution with

- lacksquare an N-vector of the mean μ
- lacksquare a positive definite N imes N covariance matrix $oldsymbol{\Sigma}$
- a positive scale s
- lacksquare a positive shape u

Normal-inverted gamma 2 distribution.

$$(\mathbf{x}, \sigma^2) \sim \mathsf{NIG2}(\boldsymbol{\mu}, \boldsymbol{\Sigma}, \mathsf{s}, \nu)$$

$$p(\mathbf{x}, \sigma^2) = p(\mathbf{x} \mid \sigma^2) p(\sigma^2)$$

$$\sigma^2 \sim \mathsf{IG2}(\mathsf{s}, \nu)$$

$$\mathbf{x} \mid \sigma^2 \sim \mathsf{N}(\boldsymbol{\mu}, \sigma^2 \boldsymbol{\Sigma})$$

To generate *n* random draws from the normal-inverted gamma 2 distribution:

- Generate n independent random draws of $\sigma^{2(j)}$ from $IG2(s, \nu)$ for j = 1, ..., n
- For each j, generate the corresponding random draw of $\mathbf{x}^{(j)}$ from $N\left(\boldsymbol{\mu},\sigma^{2(j)}\boldsymbol{\Sigma}\right)$
- Return the collection of pairs $\left\{\mathbf{x}^{(j)}, \sigma^{2(j)}\right\}_{j=1}^{n}$



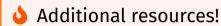
Your turn!

Complement the niceig2 function with another one that provides *n* random draws from the normal-inverted gamma 2 distribution.

- adjust the arguments
- return a list containing
 - an *n*-vector of σ^2 draws
 - ▶ an $n \times N$ matrix of **x** draws

Get some help HERE.

Some stats with RcppArmadillo: Simulation smoother



Have a look at my article on Simulation Smoother using RcppArmadillo at Rcpp Gallery.

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Step 1: create a package

Run the following code in **R**:

RcppArmadillo::RcppArmadillo.package.skeleton("nicepackage")

- **C++** code lives in src/
- DESCRIPTION includes necessary dependencies
- NAMESPACE **includes** useDynLib(nicepackage)
- R functions in R/ refer to C++ functions via .Call()
- File R/RcppExports.R contains all C++ functions exported to R

Step 2: create R project and open it

you know what to do

Step 3: some cleaning

remove files:

- Read-and-delete-me
- src/HelloWorld.cpp
- ► man/*

Step 4: include useful elements

- Set git usethis::use_git()
- 2 Set licencing usethis::use_gpl3_license()
- Set package doc usethis::use_package_doc()
- Set roxy roxygen2::roxygenise()
 - copy Encoding: UTC-8 to DESCRIPTION
 - remove the NAMESPACE file
 - run roxygen2::roxygenise() again

Step 5: edit the DESCRIPTION file

you know what to do!

Step 6: include compiled code

- copy the nicerig2.cpp file to src/ directory
- 2 edit it
- include the header file nicerig2.h in the src/ directory
- 4 edit it

Step 7: compile the code the first time

Run Rcpp::compileAttributes() to update the R/RcppExports.R file

Step 8: include R wrapper for C++ function

- **1** Create an R/nicerig2.R file
- Copy/paste the corresponding function from the R/RcppExports.R file
- Include argument check
- Include documentation
- Implement further adjustments

Step 9: some final touches

- Run Rcpp::compileAttributes()
- Run devtools::document() to generate documentation
- Run devtools::check() to check the package

TADA! Your package is ready!

Step 10: use the package for the first time

- Run devtools::load_all() to load the package
- Run hist(nicerig2(1e4), breaks = 100) to test the function
- Run ?nicerig2 and ?nicepackage to check the documentation

What's next?

Rewrite all your code in Rcpp! Nice!