

# ETC4500/ETC5450

## Advanced R programming

Week 11: Rewriting R code in C++



# Outline

- 1 Motivation
- 2 The first steps with Rcpp
- 3 Some stats with RcppArmadillo
- 4 An R package with compiled code

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# About me

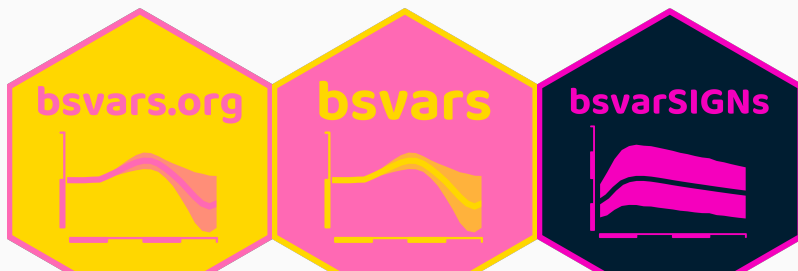
Tomasz Woźniak

- 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 arts
- I am nice!

# About me

Tomasz Woźniak

- **R** enthusiast and specialised user for 17 years
- associate editor of the R Journal
- author of **R** packages **bsvars** and **bsvarSIGNs**
- more info at [bsvars.org](http://bsvars.org)



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# Motivations

- Compiled code written in **C++** runs much faster than interpreted code in **R**
- 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 Eddelbuettel et al. facilitating the application of **C++** in **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++** in **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`

# Learning resources

- This session!
- vignettes: for packages **Rcpp** and **RcppArmadillo**
- online resources:
  - ▶ **Armadillo** library documentation
  - ▶ RcppGallery
  - ▶ [stackoverflow.com tag:rcpp](https://stackoverflow.com/tag/rcpp)
- François, R., *Optimizing R Code with Rcpp* on datacamp
- Tsuda, M., *Rcpp for everyone*
- Eddebuettel, 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	-0.173	1	FALSE	nice
2	0.105	2	TRUE	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 = seq_len(n);
  LogicalVector y = v > 0;
  CharacterVector z(n, "nice");
  return List::create(_["v"] = v, _["x"] = x, _["y"] = y, _["z"] = z);
}

/** R
nicetry(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

```
List nicetry (                                // output type and function name
    int n                                     // input type and name
) {
    NumericVector v = rnorm(n);               // define a numeric vector and fill it
    IntegerVector x = seq_len(n);             // define an integer vector as a sequence
    LogicalVector y = v > 0;                  // define a logical vector
    CharacterVector z(n, "nice");             // define a character vector
    // return a list with the created vectors
    return List::create(_["v"] = v, _["x"] = x, _["y"] = y, _["z"] = z);
}
```

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

## Your turn!

Develop a **C++** function that creates a  $T \times 3$  matrix with:

- an integer  $T$  as the only argument
- a constant term column: `NumericVector i(n, 1.0);`
- a linear trend  $t - \bar{t}$  column
- a quadratic trend  $(t - \bar{t})^2$  column

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

Hints:

- specify individual columns as `NumericVector` and put

# 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 `nice1ist.cpp` file in R

## ■ R script using the function from `nice1ist.cpp`:

```
Rcpp::sourceCpp("nice1ist.cpp")  
nice1ist(3)
```

```
$p  
[1] 0.331 -0.568 1.182
```

```
$s  
[1] 0.110 0.323 1.398
```

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

🔥 Your turn!

Consider a Gaussian random walk:

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

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

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

Hint: use functions `rnorm` and `cumsum`.

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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: IG2 distribution

Sampling random draws from an inverted gamma 2 distribution.

A positive random variable  $\sigma^2$  following an inverted gamma 2 distribution with positive scale  $s$  and shape  $\nu$  parameters is denoted by:

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

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



# Some stats with RcppArmadillo: IG2 distribution

## 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)
*/
```

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

🔥 Your turn!

Consider a Gaussian random walk:

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

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

- has an integer  $T$  as the only argument
- returns a  $T$ -vector of type `vec` with Gaussian random walk

Get some help [HERE](#).

# Some stats with RcppArmadillo: linear regression

## Contents of a nice1r.cpp file:

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

// [[Rcpp::export]]
vec nice1r (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)
nice1r(y, x)
*/
```

# Some stats with RcppArmadillo: linear regression

## Your turn!

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

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

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

Get some help [HERE](#).

# Some stats with RcppArmadillo: Simulation smoother

🔥 Additional resources!

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

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# An R package with compiled code

Run the following code in **R**:

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

Note: this function has a different effect if package **pkgKitten** is installed.

# An R package with compiled code

## ■ DESCRIPTION includes necessary dependencies

Imports: Rcpp (>= 1.0.14)

LinkingTo: Rcpp, RcppArmadillo

## ■ NAMESPACE includes dynamic library definition and inports

useDynLib(nicepackage, .registration=TRUE)

importFrom(Rcpp, evalCpp)



# An R package with compiled code

- **C++** code lives in `src/`
  - ▶ `src/Makevars` files specify compilation flags
  - ▶ `src/Makevars.win` files specify compilation flags for Windows
  - ▶ analyse sample `src/*.cpp` file
  - ▶ files `src/RcppExports.cpp` and `R/RcppExports.R` are generated automatically by running `Rcpp::compileAttributes()`
  - ▶ analyse **R** wrappers to **C++** functions in `R/RcppExports.R`

# An R package with compiled code

 Your turn!

Create an **R** package with compiled code following the steps from repository `donotdespair/15steps2nicepackage`

- Read the README file
- download file `nicepackage.R`
- follow the instructions in **R**

# What's next?

- Keep programming in **C++** for **R** applications
- Reach out for help
- Read the documentation of the **C++** libraries you're about to use
- Study the **Rcpp** family of packages
- Study **openMP** to facilitate parallel computing

**Rewrite all your code in Rcpp!**

**Nice!**