# Writing code with Rcpp and RcppEigen

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#### Introduction

Today, hope to answer the following:

- ▶ Why/when should I use C++ rather than R?
- ► How should I integrate R and C++?

Some examples and possibly useful tips.

# Why

#### R programs tend to be

- Fast and easy to write
- Easy to test and debug
- Convenient for reading in, cleaning, analyzing data

#### but can also be (when I write them)

- disorganized-programs may not have clearly delineated parts, copy and pasted code, object oriented programming forgotten
- slow, unless programmed in careful and sometimes unnatural ways ("vectorization")

Worse, R may still be slow no matter how carefully it is programmed

#### An alternative

C++ is a modern, widely used programming language

#### Pros

- "object-oriented"-may be easier to organize complicated programs through user-defined classes
- can generally write faster code than with R

#### Cons

- slower and harder to write than R
- Harder to test and debug

#### When does it matter?

R already calls fast functions for matrix multiplication, matrix decompositions, etc.

So basic linear algebra like

may not see any speed increase in C++.

On the other hand, iteration that cannot be vectorized can become much faster in C++/another compiled language.

# Example: Gibbs sampling

The Gibbs sampler is a Markov chain Monte Carlo (MCMC) sampling scheme. It involves lots of iterative updates on components of a vector. The iterations depend on each other and need to be done sequentially-hard (impossible?) to vectorize.

Timings from http://dirk.eddelbuettel.com/blog/2011/07/14/ for a bivariate Gibbs sampling example:

| test replications elapsed | relative user.self sys.self |
|---------------------------|-----------------------------|
| 4 GSLGibbs(N, thn)        | 10 7.845 1.000000           |
| 3 RcppGibbs(N, thn)       | 10 12.218 1.557425          |
| 2 RCgibbs(N, thn)         | 10 312.296 39.808286        |
| 1 Rgibbs(N, thn)          | 10 420.953 53.658764        |
|                           |                             |

## Rcpp and RcppEigen

- ▶ Rcpp: package that lets user pass data back and forth between R and C++ programs
- ▶ RcppEigen: provides access to Eigen C++ template library

```
#to install
install.packages("Rcpp")
install.packages("RcppEigen")
#may need to install Rtools as well
```

## Getting started: the sourceCpp() function

In the text file "firstProgram.cpp", type:

```
#include <Rcpp.h>

// [[Rcpp::export]]
void helloWorld(){
   Rcpp::Rcout<<"Hello world"<<std::endl;
}</pre>
```

In the R or RStudio console, type:

```
sourceCpp("C:/.../firstProgram.cpp")
helloWorld()
```

## Another speed comparison

In R

```
addUp_R<-function(n){
    val=0
    for (i in 1:n){
        val=val+i
    }
    return(val)
}</pre>
```

## Speed comparison continued

```
In C++
```

```
#include <Rcpp.h>

// [[Rcpp::export]]
int addUp_Cpp(int n){
   int start=0;
   for (int i=1;i<=n;i++){
      start=start+i;
   }
   return start;
}</pre>
```

## Speed comparison continued

#### Running

```
microbenchmark(addUp_R(10000))
microbenchmark(addUp_Cpp(10000))
```

on my computer gave a mean evaluation time of 307 microseconds for the R version and 3.16 microseconds for the C++ version, respectively

## Eigen

## What is Eigen?

- ▶ C++ template library for linear algebra
- includes useful functionality we'd rather not program ourselves
- efficient, actively maintained
- included with RcppEigen download

#### What does it do?

- implements matrices
- matrix multiplication, matrix addition
- matrix decompositions, elementwise operations on matrices
- gives us access to many linear algebra functions we are used to in R

## Example: compute quadratic form

## In the file "eigenExample.cpp", type

```
#include <RcppEigen.h>
//[[Rcpp::depends(RcppEigen)]]
//a function to compute x^tAy for vectors x,y
//and a matrix A
//[[Rcpp::export]]
double quadraticForm(Eigen::VectorXd vec1,
                     Eigen::MatrixXd A,
                     Eigen::VectorXd vec2){
  return vec1.adjoint()*A*vec2;
}
```

## Running quadraticForm()

## In the R console, type

```
x=c(1,2)
y=c(3,4)
A=matrix(c(5,6,7,8),ncol=2)
sourceCpp("eigenExample.cpp")
quadraticForm(x,A,y)
#for comparison:
t(x)%*%A%*%y
```

# Quadratic form speed comparison

Suppose x and y are length 1000 vectors, and A is a  $1000 \times 1000$  matrix. Which should be faster:

or

quadraticForm(x,A,y)

?

# Quadratic form speed comparison

I found averages of about 2 and 5 ms for the R and C++ functions, respectively. So the R code is not actually slower than the C++ code in this instance.

## Building a package

For larger projects, it may be useful to build a package to keep things organized. In the R console, the command

RcppEigen.package.skeleton("packageName")

will build a basic package with some example functions that can be deleted or modified.

Making the directory an R project in RStudio lets you build and compile the whole package at once.

## In practice

```
While RStudio does do code completion and help with compilation
for C++, I recommend (from experience) developing in a
dedicated IDE like Microsoft Visual Studio (for Windows) or
XCode (for Mac OS)
Example: the function "fail.cpp"
#include <RcppEigen.h>
//[[Rcpp::depends(RcppEigen)]]
//[[Rcpp::export]]
Eigen::VectorXd memory(int n){
  Eigen::VectorXd x;
  for (int i=0; i< n; i++){
    x(i)=1:
  return(x);
```

## Example continued

Running

```
sourceCpp("fail.cpp")
memory(100)
```

in the R console causes the R session to abort

#### Solution

The issue is much easier to debug in Visual Studio.

```
#include <RcppEigen.h>
//[[Rcpp::depends(RcppEigen)]]
//[[Rcpp::export]]
Eigen::VectorXd memory(int n){
  Eigen::VectorXd x;
  x.setZero(n); //problem: needed to
                //reserve space for x
  for (int i=0; i< n; i++){
    x(i)=1:
  return(x);
```

# Thank you

#### References

- Douglas Bates, Dirk Eddelbuettel (2013). Fast and Elegant Numerical Linear Algebra Using the RcppEigen Package. Journal of Statistical Software, 52(5), 1-24. URL http://www.jstatsoft.org/v52/i05/.
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