## R, C++ and Rcpp

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useR! 2014 Invited Talk UCLA, 2 July 2014

# Acknowledgements

Rcpp would not be where it today without

Rcpp Core which is comprised of JJ Allaire, Doug Bates,
John Chambers, Dirk Eddelbuettel, Romain
Francois and Kevin Ushey

Rcpp contributors listed in the THANKS file Rcpp and R users in the larger R ecosystem

## Outline



## A tweet about Rcpp from IQSS / Harvard



#### R excels via

- interactive data analysis
- exploratory work and visualization
- prototyping new methods
- high-level lingua franca
- vast array of packages

## R and C++

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- interactive data analysis
- exploratory work and visualization
- prototyping new methods
- high-level lingua franca
- vast array of packages

#### C++ excels via

- native / compiled speed at low overhead
- well-known language with multiple paradigms
- facilities for production deployment of code
- vast array of libraries in C / C++



### R and C++

### R excels via

- interactive data analysis
- exploratory work and visualization
- prototyping new methods
- high-level lingua franca
- vast array of packages

### ease and clarity

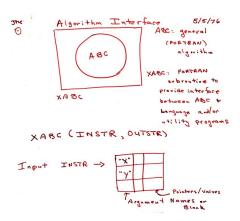
### C++ excels via

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- facilities for production deployment of code
- vast array of libraries in C / C++

### speed



# R and C++ Interface Vision



# R and C++ Interface Vision

- Use trusted numerical libraries (mostly/exclusively written in Fortran)
- Provide environment which statistician could use more easily
- Enable interactive and iterative data exploration
- Pass "objects" seamlessly between both components
- Make it extensible for research into statistical methods
- C.f. John Chambers (2008) regarding "Mission" and "Directive"

# Another tweet about Rcpp





Love that my reaction almost every time I rewrite R code in Rcpp is "holy shit that's fast" thanks @eddelbuettel &

@romain\_francois #rstats



9:08 PM - 18 Oct 2013

## **Outline**

- (Pre-)History
  - Cxx
  - Interface

# CRAN Package cxx (from 2000-2001)

Class definition in header

```
//X.hh
class X {
public:
   X (); \sim X ();
};
class Y {
public:
   \mathbf{Y} (); \sim \mathbf{Y} ();
};
```

## CRAN Package cxx (from 2000/2001)

Class constructor implementation in source file (could be in header)

```
//X.cc
#include <iostream>
#include "X.hh"
using namespace std;
static Y y;
X::X() { cout << "constructor X" << endl; }
X::\sim X() { cout << "destructor X" << endl; }
Y::Y() { cout << "constructor Y" << endl; }
Y::\sim Y()  { cout << "destructor Y" << endl; }
```

# CRAN Package cxx (from 2000/2001)

Glue function with C interface

```
// X_main.cc:
#include "X.hh"
extern "C" {
void X main () {
  X x;
```

# CRAN Package cxx (from 2000/2001)

R function calling glue function (and note .C() not .Call())

```
cxxtest <- function() {
   invisible(.C("X_main", PACKAGE = "cxx"))
}</pre>
```

- It still compiles and runs after minor adjustments:
  - C++ side:
    - include wants iostream not iostream.h
    - Namespaces: add using namespace std;
  - R side:
    - DESCRIPTION needs Maintainer (added in 0.0-3)
    - NAMESPACE needed
- A slightly revised / updated variant is in Writing R Extensions – but no corresponding package is maintained

Lessons

- Pretty sweet to compile 14-year old package
- Only somewhat minor adjustments requirement
- Stable interfaces matter a lot.
- Stick with standards whenever possible.
- Provides "existence proof" about R and C++.
- Also shows how C++ can provide state via classes with constructor and destructor,

# Rcpp Playground: Defined by .Call

### R side

### C++ side

```
extern "C"

SEXP nameOfFunction(SEXP a, SEXP b,

SEXP c, ...,)
```

## Outline



# So why C++?

Scott Meyers: "View C++ as a federation of languages"

- C provides a rich inheritance and interoperability as Unix, Windows, ... are all build on C.
- Object-Oriented C++ just to provide endless discussions about exactly what OO is or should be.
  - The STL which is a specific template library which is powerful but has its own conventions.
- Templated C++ which is mighty powerful; template meta programming unequalled in other languages.
  - C++1x adds enough to be called a fifth part of this language federation.

# Other key aspects

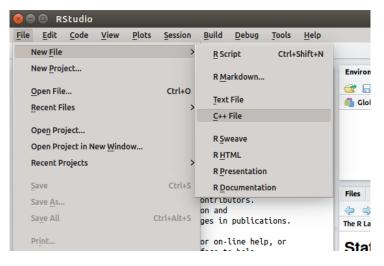
Slightly paraphrasing from talks by Stroustrup and Sutter last fall

- There should no additional layer (or intermediate language) between C++ and the machine.
- Performance is unsurpassed by other languages.
- "You don't pay for what you don't need": While the language has surely many features, what is not used does not affect performance or resource use.
- C interface / extension major reason for wide and early adoption.
- Very widely used, countless libraries, extensive tools.

### Outline

- 4 Rcpp
  - How
  - Example: Gibbs
  - Example: cumsum
  - Example: VAR(1)
  - Packages

# How do we use Rcpp? RStudio makes it very easy: Single File



RStudio example cont'ed

### The following file gets created:

```
#include <Rcpp.h>
using namespace Rcpp;
// Below is a simple example of exporting a C++ function to R.
// You can source this function into an R session using the
int timesTwo(int x) {
   return x * 2;
```

Rcpp Attributes: evalCpp, cppFunction, sourceCpp

```
## evaluate a C++ expression, retrieve result
evalCpp("2 + 2")
## [1] 4
## create ad-hoc R function 'square'
cppFunction('int square(int x) { return x*x;}')
square (7L)
## [1] 49
## or source an entire file (including R code)
#sourceCpp("code/squareWithRCall.cpp")
```

MCMC and Gibbs Samplers

Markov chain Monte Carlo, and the Gibbs sampler in particular are popular to simulate posterior densities.

A set of posts by Darren Wilkinson spawned a cottage industry of comparisons among languages, dialects, variants, ...

We will briefly revisit it here too.

Gibbs Sampler Setup

The example by Darren shows a simple MCMC Gibbs sampler of this bivariate density::

$$f(x, y) = kx^2 \exp(-xy^2 - y^2 + 2y - 4x)$$

with conditional distributions

$$f(x|y) \sim \text{Gamma}(3, y^2 + 4)$$
  
 $f(y|x) \sim N\left(\frac{1}{1+x}, \frac{1}{2(1+x)}\right)$ 

i.e. we need repeated RNG draws from both a Gamma and a Gaussian distribution.

Gibbs Sampler: R Code

```
## The actual Gibbs Sampler
Rgibbs <- function(N,thin) {
    mat <- matrix(0,ncol=2,nrow=N)</pre>
    x < -0
    v <- 0
    for (i in 1:N) {
         for (j in 1:thin) {
             x \leftarrow \mathbf{rgamma}(1,3,y*y+4)
             y <- rnorm(1,1/(x+1),1/sqrt(2*(x+1)))
         mat[i,] \leftarrow c(x,y)
    mat
library(compiler) ## to byte-compile
RCgibbs <- cmpfun (Rgibbs)
```

# Gibbs Sampler: C++ Code How do we use Rcpp?

```
#include <Rcpp.h>
using namespace Rcpp;
NumericMatrix RcppGibbs (int n, int thn) {
    NumericMatrix mat(n, 2);
    double x=0, v=0:
    for (int i=0; i<n; i++) {
        for (int j=0; j<thn; j++) {</pre>
            x = R::rgamma(3.0, 1.0/(y*y+4));
            y = R::rnorm(1.0/(x+1), 1.0/sqrt(2*x+2));
        mat(i,0) = x;
        mat(i,1) = v;
    return mat;
```

# Gibbs Sampler: Benchmark

How do we use Rcpp?

```
source("code/gibbs.R")
sourceCpp ("code/gibbs.cpp")
library(rbenchmark)
benchmark (Rgibbs (2000, 200),
          RCqibbs (2000, 200),
          RcppGibbs (2000, 200),
          replications=10,
          order="relative") [, \mathbf{c}(1,3:4)]
##
                     test elapsed relative
  3 RcppGibbs (2000, 200) 1.727 1.00
## 2 RCgibbs(2000, 200) 40.911 23.69
## 1 Rgibbs(2000, 200) 44.858 25.98
```

Motivation (Pre-)History C++ Rcpp Doc Growth ... How Ex:Gibbs Ex:cumsum Ex:VAR Packages

# When would we use Rcpp? Cumulative Sum

### A basic looped version:

```
#include <Rcpp.h>
#include <numeric> // for std::partial sum
using namespace Rcpp;
NumericVector cumsum1 (NumericVector x) {
    double acc = 0:
    NumericVector res(x.size());
    for(int i = 0; i < x.size(); i++) {</pre>
         acc += x[i];
         res[i] = acc;
    return res;
```

# When would we use Rcpp? Cumulative Sum

#### An STL variant:

# When would we use Rcpp? Cumulative Sum

### Or just Rcpp sugar:

```
// [[Rcpp::export]]
NumericVector cumsum3(NumericVector x) {
    return cumsum(x); // compute + return result vector
}
```

### Of course, all results are the same.

# When would we use Rcpp?

Easy speed gain: VAR(1) Simulation

Let's consider a simple possible VAR(1) system of *k* variables.

For k = 2:

$$X_t = X_{t-1}B + E_t$$

where  $X_t$  is a row vector of length 2, B is a 2 by 2 matrix and  $E_t$  is a row of the error matrix of 2 columns.

# When do we use Rcpp?

Easy speedup:: VAR(1) Simulation

In R code, given both the coefficient and error matrices (revealing k and n):

```
rSim <- function(B,E) {
    n <- nrow(E); k <- ncol(E)
    X <- matrix(0, n, k)
    for (r in 2:n) {
        X[r,] = X[r-1,] %*% B + E[r,]
    }
    return(X)
}</pre>
```

# When do we use Rcpp?

Easy speed gain: VAR(1) Simulation

```
cppFunction('
arma::mat cppSim(const arma::mat& B,
                 const arma::mat& E) {
    int n = E.n_rows; int k = E.n_cols;
    arma::mat X = arma::zeros<arma::mat>(n,k);
    for (int r=1; r < n; r++) {
        X.row(r) = X.row(r-1) * B + E.row(r);
    return X;
}', depends="RcppArmadillo")
```

## When do we use Rcpp?

Easy speed gain: VAR(1) Simulation

```
#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
// [[Rcpp::export]]
arma::mat cppSim(const arma::mat& B,
                 const arma::mat& E) {
    int n = E.n_rows; int k = E.n_cols;
    arma::mat X = arma::zeros<arma::mat>(n,k);
    for (int r=1; r < n; r++) {
        X.row(r) = X.row(r-1) * B + E.row(r);
    return X;
```

## When do we use Rcpp?

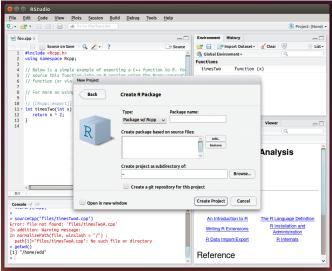
Easy speed gain: VAR(1) Simulation

```
a <- matrix(c(0.5,0.1,0.1,0.5),nrow=2)
e <- matrix(rnorm(10000),ncol=2)
all.equal(cppSim(a,e), rSim(a,e))
## [1] TRUE
benchmark (cppSim(a,e), rSim(a,e),
         order="relative")[,1:4]
##
            test replications elapsed relative
## 1 cppSim(a, e)
                        100 0.231 1.00
## 2 rSim(a, e)
                        100 2.644 11.45
```

Motivation (Pre-)History C++ Ropp Doc Growth ... How Ex:Gibbs Ex:cumsum Ex:VAR Packages

## How do we use Rcpp?

RStudio makes it very easy (using Rcpp.package.skeleton()



# How do we use Rcpp?

Key Features

As we just saw in the small example:

- No build system issues: We just use R
- No operating system dependencies
- No (manual or explicit) memory management
- No complicated C++ code (though we could if we wanted to)
- Easy transition from exploration ("one-liners") to deployment ("packages")

# How do we use Rcpp?

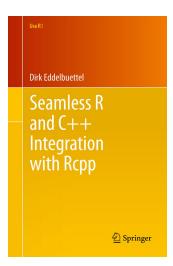
Mix and match between packages

- plugins allow us to turn on
  - C++11 as needed without imposing on build system
  - OpenMP to allow for widely-supported and understood parellism
- depends allow us to pull in package via
  - headers-only as eg Armadillo, Eigen and Boost
  - actual linking as eg RcppGSL
- registration for easier linking on the user-side

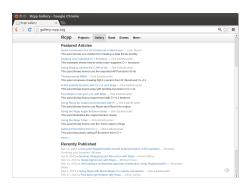
## Outline



## **Documentation**



### See gallery.rcpp.org

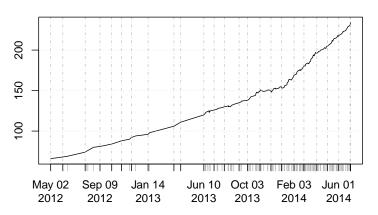


## Outline



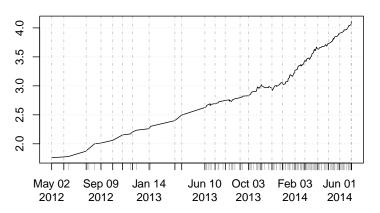
# Rcpp on CRAN: Now at 234 packages

### **Growth of CRAN packages using Rcpp**



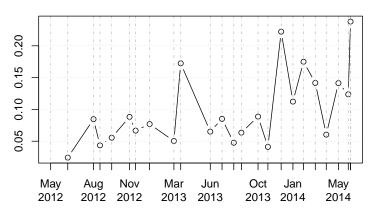
# Rcpp on CRAN: Or just over 4 per cent

### Rcpp packages as percentage of CRAN



# Rcpp on CRAN: 10+ per cent of new pkgs

### Rcpp (approx.) Percentage of New Packages by Month



# Growing pains?

From 0.6.0 in late 2008 to 0.11.0 in early 2014

- six major releases from 0.6.0 to 0.11.0
- changed build / deployment system in last release
- disruptive change which took effort to progagate
- we now use C bindings and registered functions
- hope that such breakage may be a thing of the past

# Growing pains?

At 250+ user packages across CRAN and BioConductor, we simply cannot make breaking changes.

We test against all direct dependencies.

CRAN tests even further.

Some hope that *in the not-so-distant future* we all may have something better...

## Outline



### Docker

- Intro
- Setup
- Commands
- Shells
- Dockerfiles
- Example

## Docker



#### What is Docker?

Docker is an open platform for developers and sysadmins to build, ship, and run distributed applications. Consisting of Docker Engine, a portable, lightweight runtime and packaging tool, and Docker Hub, a cloud service for sharing applications and automating workflows. Docker enables apps to be quickly assembled from components and eliminates the friction between development, QA, and production environments

Text from http://docker.com as of June 24, 2014.



## Docker



### Ok, seriously, what is Docker?

Docker is a very lightweight abstraction using recent Linux kernel features which lets us to run code in **cheap** (to launch) and **easy** (to build) units: **containers**.

We can share containers across OSs and into 'the cloud'.

It changes how we build and test R (and R packages).

- Most Unix variants have detailed (short) instructions on Docker website. Should just work.
- Or: sudo apt-get install docker.io (requires Ubuntu 14.04 or Debian testing)
- On Windows or OS X: Use boot2docker which installs an appliance for you (containing docker, git, virtualbox, ...).

Tested on Windows at work. Appears to Just works too.

First steps: Getting images

After installation, run for example

docker pull ubuntu

to pull a set of pre-built initial images.

Many such image sets are now being defined.

I will describe one for R.

First steps: Listing images

### This one call gets us

```
edd@max:~$
           docker images
REPOSITORY
            TAG
                     TMAGE ID
                                   CREATED
                                                VIRTUAL SIZE
ubunt.u
            10.04
                     3db9c44f4520
                                   9 weeks ago 183 MB
ubuntu
           lucid
                    3db9c44f4520
                                   9 weeks ago 183 MB
ubunt.u
            12.10
                     6006e6343fad
                                   3 weeks ago 172.2 MB
                    6006e6343fad
                                   3 weeks ago 172.2 MB
ubuntu
            quantal
ubuntu
            13.10
                     d2099a5ba6c5
                                   3 weeks ago 180.2 MB
ubunt.u
                     d2099a5ba6c5
                                   3 weeks ago 180.2 MB
            saucv
ubuntu
            14.04
                     5cf8fd909c6c
                                   3 weeks ago 274.3 MB
ubunt.u
            latest.
                     5cf8fd909c6c
                                   3 weeks ago 274.3 MB
ubunt.u
            trustv
                     5cf8fd909c6c
                                   3 weeks ago
                                                274.3 MB
ubuntu
            raring
                    7656cbf56a8c
                                   3 weeks ago 169.4 MB
ubunt.u
            13.04
                     7656cbf56a8c
                                   3 weeks ago 169.4 MB
            12.04
                     cc0067db4f11
                                   3 weeks ago
                                                210.1 MB
ubuntu
                     cc0067db4f11
                                   3 weeks ago
                                                210.1 MB
ubuntu
            precise
edd@max:~$
```

First steps: Running images

### Now we can execute commands in all these variants:

```
edd@max:~$ docker run -t ubuntu:12.04 head -1 /etc/motd
Welcome to Ubuntu 12.04.4 LTS (GNU/Linux 3.2.0-58-generic
x86 64)
edd@max:~$ docker run -t ubuntu:12.10 head -1 /etc/motd
Welcome to Ubuntu 12.10 (GNU/Linux 2.6.32-42-server x86 64)
edd@max:~$ docker run -t ubuntu:13.04 head -1 /etc/motd
Welcome to Ubuntu 13.04 (GNU/Linux 2.6.32-46-server x86_64)
edd@max:~$ docker run -t ubuntu:13.10 grep VERSION_ID
/etc/os-release
VERSION ID="13.10"
edd@max:~$ docker run -t ubuntu:14.04 grep VERSION ID
/etc/os-release
VERSION ID="14.04"
edd@max:~$
```

First steps: Running images interactively

#### Now we can execute commands in all these variants:

```
edd@max:~$ docker run -t -i ubuntu:12.04 /bin/bash root@fe2fe9a795ff:/$ whoami root root@fe2fe9a795ff:/$ exit exit edd@max:~$
```

First steps: Running images interactively

### We can mount directories for use by the Docker image

# Getting started with Docker Using a Dockerfile

installation:

Dockerfiles are 'recipes' (à la Travis.yml) which modify / create images. Below is 'add-r' which just adds an R

```
## start with the Debian testing
FROM debian:testing
MAINTAINER Dirk Eddelbuettel edd@debian.org
## Remain current
RUN apt-get update -gg
RUN apt-get dist-upgrade -v
RUN apt-get install -y --no-install-recommends \
           r-base r-base-dev r-recommended littler
RUN ln -s /usr/share/doc/littler/examples/install.r \
           /usr/local/bin/install.r
## use via:
## cat Dockerfile | docker build -t debian:testing-w-R -
```

I have started to create first sets of (partially nested) Dockerfiles creating images

add-r with basic R as a binary package
add-r-devel adding R-devel built from source
add-r-devel-san adding R-devel built from source and
enabling Address Sanitizer checks
add-r-devel-ubsan adding R-devel built from source and
enabling Undefined Behavior Sanitizer checks

which are downloadable from hub.docker.com under my handle eddelbuettel.

Test with R

which tests (fine) under the release version.

Test with R devel

[...]

which tests under the R-development version (which has been enabled for Sanitizer checks) and ...

Test with R devel

```
[...]
* checking tests ...
 Running 'simple.R'
 ERROR
Running the tests in 'tests/simple.R' failed.
Last 13 lines of output:
   Freed heap region:
                           fd
   Stack left redzone:
                           f1
   Stack mid redzone: f2
   Stack right redzone: f3
   Stack partial redzone: f4
   Stack after return:
                           f5
   Stack use after scope: f8
   Global redzone:
                          f9
   Global init order:
                           f6
   Poisoned by user:
                            f7
   Contiguous container OOB:fc
   ASan internal:
                            fе
 ==258==ABORTING
$
```

**Enormous Scope and Potential** 

- Docker provides hub.docker.com
- Which integrates with Github
- So we can upload/create/share/modify Dockerfiles
- Triggering rebuilds at Docker for updated images
- Which one can download, deploy as is, or modify
- Deploy locally or on via (growing) list of cloud providers incl. Google Compute Engine

in order to run R, or R-devel, or R-devel with Sanitizers, or R-devel with Profiling, or alternative R implementations ...

Upside for the R Community

- Linux as a Service: Getting identical, reliable, replicable test environments
- which we can contain different compilers: gcc-4.8, gcc-4.9, clang-3.4, clang-3.5, icc, ...
- as well as across configurations: enable profiling, sanitizers, valgrind, ...
- possibly using debian-r.debian.net and its
   6000+ Debian packages spanning CRAN and BioC
- without requiring anyone to install (or learn) Linux
- as the installations are lightweight and disposable
- yet can be used across Linux, Windows, OS X, ... and in the (Google etc) cloud



## Conclusion

### **Rcpp**

- mature after 5 1/2 years
- fairly widely used
- supported, tested
- still growing and improving
- speed of C++ and ease and clarity of R

#### **Docker**

- just had 1.0 release
- lots of momentum and interest
- perfect fit for our (batch, scripted) tests of R
- intend to work on some of this, help welcome