

# Interfacing an existing software library to R: The example of “RNiftyReg”

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LondonR meeting, 2011-06-15

## Health warning

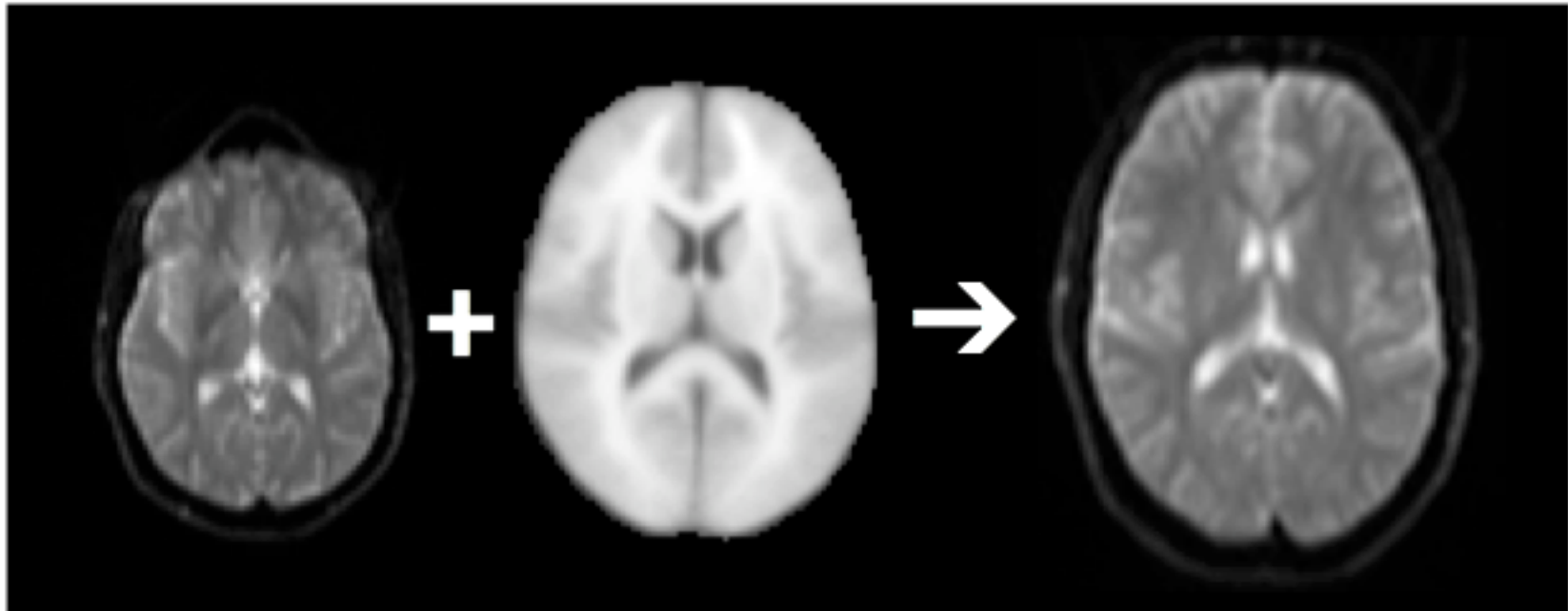
- I will only talk about interfacing **C/C++** code
- I try to follow R best practices, but I may be doing something “wrong”!
- What follows is my perspective, and your milage may vary
- Always a good idea to consult *Writing R Extensions* if in doubt (<http://cran.r-project.org/doc/manuals/R-exts.html>)

## A bit of context

- Image **registration**, or coalignment, is the process of finding an optimal **transformation** to align one (2D or 3D) image with another
- A key process in medical image processing pipelines, as well as microscopy, etc.
- The process is an **optimisation** over some space of transformations, followed by **application** of the final transformation and **regridding** (including interpolation)
- Typical 3D magnetic resonance image has **~1M voxels** (3D pixels) so optimisation needs to be done cleverly to keep run time reasonable
- Registration is a common step, so needs to be as **fast** as possible
- Interpreted R code is not going to cut it

## Illustration

- Typically need to allow for (at least) **rotation** and **translation** between source and target image
- A **scale** and **shear** are additionally allowed by affine transformations



## The starting point

- **Nifty Reg**, an open-source registration toolkit developed at UCL (M. Modat *et al.*; <http://sourceforge.net/projects/niftyreg/>; BSD licence)
- Written in C/C++
- Pretty clean separation between “work” functions and user interface code

## Key tasks

- **Strip out** command-line interface and other unneeded code
- **Convert** R data structures to C data structures
- **Call** the “work” functions and convert the result back to the form R expects
- **Write** R wrapper function(s)
- **Write** auxiliary/convenience R functions
- **Create** Makevars, configure script, etc., if needed

## .C vs .Call/.External

- `.C` is generally simpler, avoids need to deal with R-specific functions and types (e.g. SEXP) in C
- `.C` duplicates its arguments by default; `.Call` and `.External` do not
- Handling R objects and functions with `.C` is clunky; `.Call` or `.External` is generally preferable
- `.Call` arguments should be considered read-only; `.C` arguments can be read/write
- I recommend `.C` unless you need to carry over large or complex R objects (medical images are both large and complex, so `RNiftyReg` uses `.Call`)

# Data structures

- In R, an image is an **S4 object** of class “nifti”, defined in the “oro.nifti” package (B. Whitcher *et al.*; <http://cran.r-project.org/web/packages/oro.nifti/>; BSD licence)
- In C, an image is a **struct**
- Conversion is tedious but important

```
nifti_image * s4_image_to_struct (SEXP object)
{
    int i;
    nifti_1_header header;

    header.sizeof_hdr = 348;

    for (i = 0; i < 8; i++)
        header.dim[i] = (short) INTEGER(GET_SLOT(object,
install("dim_")))[i];

    header.intent_p1 = (float) *(REAL(GET_SLOT(object,
install("intent_p1"))));
    ...
}
```



# Wrapper function

- Checks data types; deals with default cases; handles metadata and ensures that return value is the correct form

```
niftyreg <- function (source, target, targetMask = NULL, initAffine = NULL, scope =  
c("affine","rigid"), nLevels = 3, maxIterations = 5, useBlockPercentage = 50,  
finalInterpolation = 3, verbose = FALSE)  
{  
  if (!require("oro.nifti"))  
    stop("The \"oro.nifti\" package is required")  
  if (missing(source) || missing(target))  
    stop("Source and target images must be given")  
  ...  
  if (source@dim_[1] == target@dim_[1])  
  {  
    returnValue <- .Call("reg_aladin", .fixTypes(source), .fixTypes(target),  
scope, as.integer(nLevels), as.integer(maxIterations), as.integer  
(useBlockPercentage), as.integer(finalInterpolation), .fixTypes(targetMask),  
initAffine, as.integer(verbose), PACKAGE="RNiftyReg")  
  
    dim(returnValue[[1]]) <- dim(target)  
    dim(returnValue[[2]]) <- c(4,4)  
    attr(returnValue[[2]], "affineType") <- "niftyreg"  
    ...  
  }  
}
```

## Summary of process

- Source and target images are read from file or otherwise created using functions in the “oro.nifti” package, or similar
- These are passed to `niftyreg()`, which will do various checks to ensure that its parameters are sensible and then pass them to C
- In C, images are converted to the data type which the Nifty Reg functions expect
- The registration is performed
- A data vector containing the registered image data, and the transformation matrix, are passed back to R
- The `niftyreg()` function formats the result, applies the relevant metadata, and returns

## Making life easier

- Unless every scrap of speed is important, do type checking (and other “cheap” tasks) in the R wrapper function
- Don’t expose C parameters as R wrapper function arguments if they can be worked out (e.g. vector lengths)
- Remember that you can step through compiled code using

```
R -d gdb
```

(or other debugger, as appropriate), and then loading package and setting a breakpoint

- Look out for use of functions which will cause problems with R (e.g. `exit()`, which will quit the whole R process)
- Porting memory allocation, output and other functions to the R-specific equivalents (e.g. `R_alloc()`) is generally worthwhile

## RNiftyReg

- Provides 2D and 3D **rigid-body** and **affine** registration, including 3D-to-2D and 4D-to-3D
- Includes auxiliary functions for reading, writing and manipulating affine transformations
- Plans to interface to Nifty Reg's **nonlinear** registration algorithm at some point in the future
- Available on **CRAN**, or at <http://code.google.com/p/rniftyreg/>

## Conclusions

- Interfacing R to an existing software library requires some work, but needn't be too painful (initial release of RNiftyReg was days' work, not weeks', even with very little experience)
- (If the library changes a lot and/or separation between “work” functions and user interface isn't clear, it might be harder)
- Almost certainly faster than developing a new implementation from scratch in most cases
- Converting between data types can be time consuming, but for simple vectors it's much easier
- R really does provide a lot of facilities to help
- Remember, packages provide interfaces to languages other than C and Fortran