Rcpp use case: segmentation by dynamic programming

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1 Motivation

source("dpseg.R")

Last week we have written a plain R implementation of segmentation by dynamic programming:

```
dpseg
function (y, K)
{
    n <- length(y)
    J <- getJ(y)
    dp <- getVandBkp(J, K)
    V <- dp$V
    bkp <- dp$bkp
    res.bkp <- backtrack(bkp)
    res.rse <- V[, n]
    list(bkpList = res.bkp, rse = res.rse, V = V)
}</pre>
```

This function makes use of three intermediate functions:

- getJ computes the n×n matrix J such that J[i,j] for i ≤ j is the Residual Squared Error (RSE) of the segmentation with only one segment (no breakpoint) between i and j;
- getVandBkp computes from J the matrices V (K + 1 × n) and bkp (K × n), where V[i,j] is the best RSE for segmenting intervals 1 to j with at most i-1 change points, and bkp[i, j] is the *last* bkp of the best segmentation of [1:j] in i segments;
- backtrack retrieves the optimal segmentation of [1,n] in k segments for all k from bkp.

The **dpseg** runs quickly for signals of length 1000. For a signal of length 10^4 , the segmentation in 10 segments takes approximately 4 minutes on a standard laptop. As the time complexity is quadratic, we expect a signal of length 10^5 to be segmented in approximately 400 minutes.

The goal of this section is to identify which part of the code takes the longest, and to use Rcpp (that is, C++ code interfaced with R) in order to speed up this part.

- Diagnostic: which part of the code takes time?
- Use Rcpp to obtain a faster implementation of this part of the code
- What is the relative gain in computing time?
- Incorporate this implementation into the whole dynamic programming function