

How to use the findSegments function to fit a piecewise constant curve

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May 22, 2005

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1 Introduction

The problem of segmenting a series of numbers into piecewise constant segments occurs in multiple application areas. Two examples are

- arrayCGH data, where the segments correspond to regions of copy number gain, loss, or no change.
- tiling microarray data for transcriptomics, where the segments correspond to transcripts. Here we assume that the probe effects (which lead to different fluorescence intensities even for the same mRNA abundance) have been normalized away, so that all probes for one transcript have the same fluorescence (in expectation).

To demonstrate and verify the correctness of the algorithm, let's generate simulated data:

```
> genData = function(lenx, nrcp, stddev = 0.1) {  
+   x = numeric(lenx)  
+   cp = c(1, sort(sample(1:floor(lenx/15), nrcp - 1) * 15),  
+         lenx + 1)
```

```

+   s = 0
+   for (j in 2:length(cp)) {
+     sel = cp[j - 1]:(cp[j] - 1)
+     s = (0.5 + runif(1)) * sign(rnorm(1)) + s
+     x[sel] <- rnorm(length(sel), mean = s, sd = stddev)
+   }
+   return(list(x = x, cp = cp[-1]))
+ }

> lenx = 1000
> nrcp = 10
> gd = genData(lenx, nrcp)
> plot(gd$x, pch = ".")
> abline(v = gd$cp, col = "red")

```

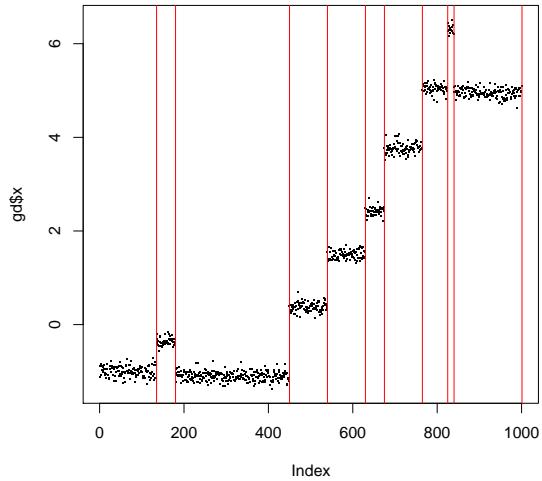


Figure 1: A simulated data example with 10 change points, shown with red vertical lines

The result is shown in Figure 1. We can use the function *findSegments* to reconstruct the changepoints from the x data alone.

```
> library(tilingArray)
```

```

> library(tilingArray)
> maxk = 500
> maxcp = 12
> seg = findSegments(gd$x, maxk = maxk, maxcp = maxcp)
> seg

$J
[1]      -Inf -0.414019  1.202624  1.792334  2.500388  2.879470  3.019222
[8]  3.480042  3.751572  4.642962  4.650403  4.662729

$th
[,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12]
[1,] 1001    0    0    0    0    0    0    0    0    0    0    0    0
[2,] 501   1001    0    0    0    0    0    0    0    0    0    0    0
[3,] 450   675  1001    0    0    0    0    0    0    0    0    0    0
[4,] 450   630   765  1001    0    0    0    0    0    0    0    0    0
[5,] 450   540   675   765  1001    0    0    0    0    0    0    0    0
[6,] 450   540   630   675   765  1001    0    0    0    0    0    0    0
[7,] 181   450   540   630   675   765  1001    0    0    0    0    0    0
[8,] 450   540   630   675   765   825   840  1001    0    0    0    0
[9,] 181   450   540   630   675   765   825   840  1001    0    0    0
[10,] 135   180   450   540   630   675   765   825   840  1001    0    0
[11,] 135   180   444   450   540   630   675   765   825   840  1001    0
[12,] 135   180   450   540   630   675   699   710   765   825   840  1001

attr("class")
[1] "segmentation" "list"

> gd$cp
[1] 135 180 450 540 630 675 765 825 840 1001

```

We see that the 10-th row of the matrix `seg$th` exactly reconstructs the change points `gd$cp` that were used in the simulation.

The parameters `maxcp` and `maxk` are the maximum number of segments and the maximum length per segment. The algorithm finds for each value of k from 1 to `maxcp` the best segmentation under the restriction that no individual segment be longer than `maxk`. In the original paper of Picard et al. [1] and in their software, `maxk` is implicitly set to the number of data points `length(x)`. I have introduced this parameter to reduce the algorithm's complexity. The complexity of Picard's software is `length(x)*length(x)`

in memory and `length(x)*length(x)*maxcp` in time, the complexity of the `findSegments` function is `length(x)*maxk` in memory and `length(x)*maxk*maxcp` in time. As I am envisaging applications with `length(x) ≈ 105` and `maxk ≈ 250`, the difference can be substantial.

2 How to choose the number of segments?

Need to assess goodness of fit (contained in `gd$J`) together with model complexity (i.e. the number of change points). Details will follow ...

3 Some more testing

Here is a little for-loop that generates data using random parameters and checks whether `findSegments` can reconstruct them. The purpose of this is for checking the validity of the code.

```
> set.seed(4711)
> for (i in 1:20) {
+   gd = genData(lenx, nrcp)
+   seg = findSegments(gd$x, maxk = maxk, maxcp = maxcp)
+   stopifnot(seg$th[nrcp, 1:nrcp] == gd$cp)
+ }
```

References

- [1] A statistical approach for CGH microarray data analysis.
 Franck Picard, Stephane Robin, Marc Lavielle, Christian Vaisse,
 Gilles Celeux, Jean-Jacques Daudin. Rapport de recherche
 No. 5139, Mars 2004, *Institut National de Recherche en In-
 formatique et en Automatique (INRIA)*, ISSN 0249-6399.
http://www.inapg.fr/ens_rech/mathinfo/recherche/mathematique/outil.html