How to use BICseq

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Overview

The *BICseq* package provides the functionalities for copy number variation (CNV) detection in whole-genome sequencing data based on the Bayesian Information Criterion (BIC). The algorithm does not assume a parametric model on the read distribution on the genome, making it more robust than current available methods. It is also fast and able to handle high-coverage genome effectively.

In this vignette, we assume that the BAM files have been sorted and indexed. All reads in the BAM files will be used for the calculation of CNVs. If only the uniquely aligned reads are to be used, users are expected to provide the BAM files with such data. Work is in progress to add the option of excluding reads that are not uniquely aligned to the genome. To avoid false positives caused by mappability differences among reads of different lengths, we assume all reads in the BAM files are of the same length.

Main functions

To detect CNVs, a BAM file of a target sample (e. g. tumor tissue) and a BAM file of a reference sample (e. g. normal tissue) are needed. BICseq assumes that biases in the target sample and the reference sample are the same, and hence the reference sample can be used as a control to remove the biases in the target sample. Warning: Make sure that the target and the reference samples are processed with the same procedure (e.g. with the same number of PCR cycles); otherwise, the biases in the target sample and the reference sample may be different and this may result in false positives. We are working towards removing the biases in the target sample and the reference sample separately and calling CNVs based on the bias-removed data.

Suppose that the two BAM files are named sample.bam and reference.bam, respectively. One can create a BICseq object by passing the names of the two BAM files and other parameters:

```
> bicseq <- BICseq(sample = sample.bam, reference =reference.bam,
+ seqNames = c(1:22, "X", "Y"))</pre>
```

where *seqNames* is the vector of the reference sequence names defined in the headers of the BAM files.

The bicseq object can then be passed to the getBICseg function for the detection of genomic alterations:

```
> segs <- getBICseg(object = bicseq, bin = 100, lambda = 2, winSize = 200,
+ quant = 0.95, mult = 1)</pre>
```

where

- bin = the initial genomic bin size in base pair for read counts; BICseq allows the initial bin size to be 1 bp even for low coverage data, but a smaller bin size would need more memory and computational time.
- lambda = the penalty of the BIC; a larger lambda will give a smoother profile, i.e. a profile with less breakpoints.
- the following three parameters are for outlier removal before segmentation.
 - 1. winSize = the window size for outlier identification; given a genomic position s, the genomic window is chosen as (a, b) such that both (a, s) and (s, b) contain at least winSize/2 genomic positions that have at least one read aligned.
 - 2. quant =the probability of the read count quantile.
 - 3. mult = a positive number; a genomic position s is considered as an outlier if it has more than mult*quantile number of aligned reads, where quantile is the $quant^{th}$ quantile of the read counts calculated from the genomic window of s. If s is determined as an outlier, BICseq will set the read count at s as mult*quantile.

The lambda parameter is the major parameter of BICseq. A larger lambda gives CNV calls with fewer false discoveries, while a smaller lambda is more sensitive. To obtain optimal results, CNVs should be first filtered by a copy ratio threshold. For low coverage data (< 1X), a smaller lambda (e.g. 1 or 1.2) can achieve good sensitivity. To further reduce false discovery rate, one can apply a p-value threshold to remove the less significant CNVs. For medium coverage (2-5X), a larger lambda should work well with no additional p-value filtering. For high coverage (5X-30X), lambda=4 should give very confident call while still detecting many small CNVs. If the data is very noisy (e.g. two samples have different biases), one can choose very large lambda (e.g. 100), though this will result in low sensitivity in detecting small CNVs.

The plot function allows for viewing the data points and detected segmentations:

```
> plot(segs, sampleName = "Demo", save = FALSE,
+ plotBin=TRUE, chromOrder=c(1:22, "X", "Y"))
```

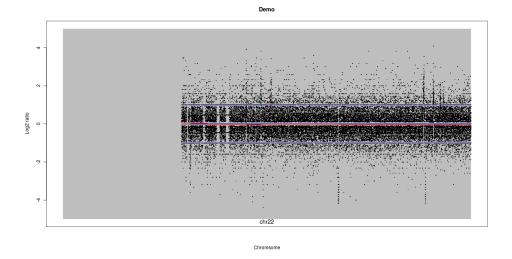


Figure 1: An example plot showing the genomic alterations. Black dots are log2 ratios of sample/reference read counts and red lines are the detected segmentations.

By setting save to TRUE, a temporary png file will be generated with the path to the file returned by the function. The option plotBin specifies whether or not the log2 copy ratios in the initial bins should be plotted. Since plotting the initial bins is slow for a large genome with a small initial bin size, one may turn off this feature by setting plotBin to FALSE. The plotting order of the chromosomes is determined by the option chromOrder. This option can contain only a subset of all chromosomes, in which case only the given chromosomes will be plotted.

The segs object contains both read counts in bins and the segmentation. The log2 copy ratios of the bins can be obtained by:

> bins <- BICseq:::getRatios(bin(segs), what = "bin")</pre>

One may get the segmentation statistics like pvalues, read counts, log2 copy ratios with the following:

> seg.summary <- BICseq:::getSummary(segs,correction=TRUE)</pre>

By setting *correction* to TRUE, the pvalues will be corrected by Bonferroni's method.