# Clonality: A Package for Clonality testing

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#### 1 Overview

This document presents an overview of the Clonality package. This package can be used to test whether two tumors are clonal (metastases) or independent (double primaries) using their copy number or loss of heterozygosity (LOH) profiles. For LOH data it implements Concordant Mutations (CM) test (Begg et al., 2007) and Likelihood Ratio (LR) test (Ostrovnaya et al., 2008). For copy number profiles the package implements the methodology based on the likelihood ratio described in (Ostrovnaya et al., 2010).

## 2 Copy number profiles

We will show how to test independence of the copy number profiles from the same patient using breast cancer data. The BAC arrays of the pairs of lobular carcinoma in situ (LCIS) and invasive lobular carcinoma (ILC) were studied in (Hwang et al., 2004) and available at http://waldmanlab.org/Breast/Hwang.data.xls. We will load package gdata in order to read the excel file.

- > library(DNAcopy)
- > library(Clonality)
- > library(gdata)

We will read the dataset and remove rows or columns with too many NAs.

```
> data<-read.xls("http://waldmanlab.org/Breast/Hwang.data.xls")</pre>
> data<-data[!is.na(data[,2]),]</pre>
> data<-data[apply(is.na(data),1,sum)<=50,]</pre>
> data<-data[,apply(is.na(data),2,sum)<=1000]</pre>
> data[1:5,1:10]
       Clone Chromosome Position LCO2LCIS LCO3LCIS LCO4LCIS LCO6LCIS LCO7LCIS
2 RP11-82D16
                                 2
                                      0.139
                                                0.121
                                                          0.184
                                                                    0.164
                                                                            -0.151
3 RP11-62M23
                       1
                                 3
                                      0.097
                                               -0.009
                                                          0.160
                                                                   -0.024
                                                                            -0.136
4 RP11-11105
                       1
                                 4
                                      0.062
                                               -0.060
                                                          0.189
                                                                   0.050
                                                                            -0.086
5 RP11-51B4
                       1
                                 5
                                     -0.059
                                               -0.165
                                                          0.100
                                                                   -0.036
                                                                            -0.291
6 RP11-60J11
                                 6
                                      0.146
                                                                            -0.071
                       1
                                                0.054
                                                          0.174
                                                                   0.073
 LC12LCIS LC17LCIS
2
   -0.070
              0.180
   -0.104
3
               0.032
4
     0.012
             -0.013
```

Rows of data correspond to probes (genomic markers). The first column is probe name; the second column is the chromosome where the probe is located; the third column is probe's genomic position recorded as an index. All subsequent columns correspond to the samples and contain log-ratios.

Since there are no genomic locations in this dataset, we will download another dataset and map the genomic locations to the probes. If the genomic locations were known, we would not need this step and the column with the probe names.

```
> arrayinfo<-read.xls("http://waldmanlab.org/Colon/nakao.data.xls") #needed to extract get
> data$Position<-arrayinfo$Mb[match(toupper(as.character(data[,1])),toupper(as.character(ar.
> data<-data[!is.na(data$Position),]
```

Now we will remove repeated genomic locations:

```
> length(unique(paste(data$Chromosome, data$Position))) #there are repeated genomic location [1] 1740
```

- > data<-data[c(TRUE,data\$Position[-1]!=data\$Position[-1864]),] #discard probes with repeated > data<-data[data\$Chromosome<=22,] #getting rid of X and Y chromosomes
- > dim(data)

5

-0.277

0.027

-0.134

0.068

[1] 1696 51

As the final step of data preparation, we have to create a CNA (copy number array) object as described DNAcopy. To save computational time, we only take the first three patients. (As a result, gain/loss frequencies used for analysis will be very imprecise and the reference distribution will have very few comparisons.)

> dataCNA < -CNA(as.matrix(data[,c(4:6,28:30)]), maploc=data\$Position, chrom=data\$Chromosome, satisfies the context of the c

Our methodology allows at most one genomic change per chromosome arm, estimated by the one-step Circular Binary Segmentation (CBS) algorithm ((Venkatraman and Olshen, 2007)).

If the data had many more than 15,000 markers, most outstanding, and likely a short change would be picked up, which would not be representative of the chromosome pattern. To avoid this, one can use the following function:

> dataAve<- ave.adj.probes(dataCNA,2)

Total number of markers after averaging is 842

Here we have averaged every two consecutive marker. For this dataset, though, averaging is not necessary.

The chromosomes should be split into arms before the clonality analysis since it increases the number of independent genomic units.

- > dataCNA\$maploc<-dataCNA\$maploc\*1000 #transforming maploc to Kb scale
- > dataCNA\$chrom<- splitChromosomes(dataCNA\$chrom,dataCNA\$maploc) #splits the chromosomes is

chrom

chr01p	chr01q	chr02p	chr02q	chr03p	chr03q	chr04p	chr04q	chr05p	chr05q	chr06p
49	60	18	34	38	38	25	113	17	71	33
chr06q	chr07p	${\tt chr07q}$	chr08p	chr08q	chr09p	chr09q	chr10p	${\tt chr10q}$	chr11p	chr11q
35	59	92	44	72	34	63	29	78	54	81
chr12p	chr12a	chr13a	chr14a	chr15a	chr16p	chr16a	chr17p	chr17a	chr18p	chr18a
1				J J	P	JIII 1 0 9	P			
16	•	38	-	-	-	-	16	-	-	30
16	60	38	-	60	19	-	-	-	-	

Next we have to create a vector of patient labels that matches the samples.

> ptlist<-substr(names(dataCNA)[-c(1,2)],1,4)

Finally, we can run the clonality analysis:

> results<-clonality.analysis(dataCNA, ptlist, nmad = 1.25, reference = TRUE, allpairs =

Calculating LR...

Calculating reference LR: %completed 17, 33, 50, 67, 83, 100,

The main information is in the output LR:

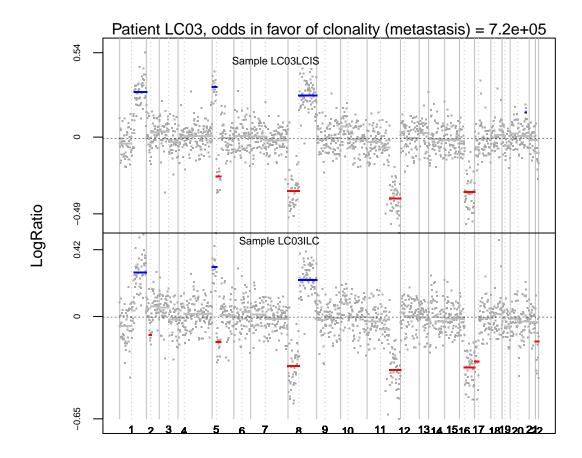
#### > results\$LR

```
Sample1 Sample2
                           LR1
                                        LR2 GGorLL NN GL GNorLN
1 LC02LCIS LC02ILC 0.003301258 3.829228e-03
                                                       0
                                                              19
                                                  3 17
2 LC03LCIS LC03ILC 76.398788106 7.233995e+05
                                                               4
                                                  7 28 0
3 LC04LCIS LC04ILC 3.822789581 9.791084e+06
                                                  6 25 0
                                    IndividualComparisons LR2pvalue
1
                                              chr01q 1.16 0.8333333
2
                  chr01q 7.82; chr05q 34.84; chr11q 34.75 0.0000000
3 chr01p 24.25; chr01q 18.87; chr10q 32.11; chr11q 174.27 0.0000000
```

The likelihood ratios LR2 for sample LC02 is much smaller than 1, therefore these tumors are independent. Patients LC03 and LC04 have LR2 much higher than one, and we can conclude that their tumors are clonal. The reference distribution for LR2 under the hypothesis of independence is constructed by pairing tumors from different patients that are independent by default. The p-value column reflects the percentiles of a particular patient's LR2 in the reference distribution: clonal tumors would have small p-values.

We can view the genomewide plots of patient LC03 using:

<sup>&</sup>gt; genomewidePlots(results\$OneStepSeg, results\$ChromClass,ptlist , c("LCO3LCIS", "LCO3ILC"),

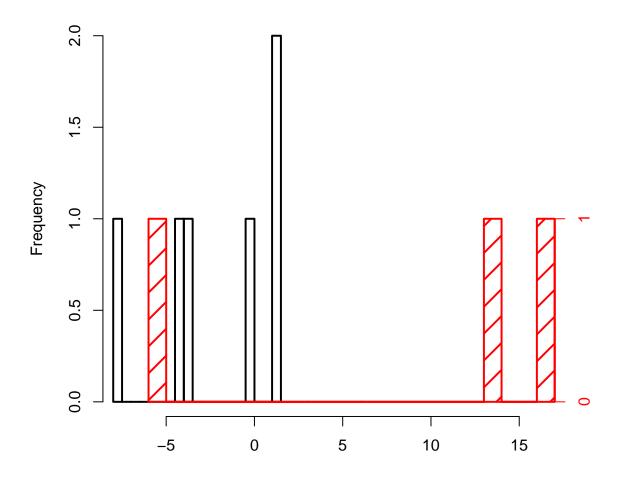


Patterns for each chromosome would be plotted by:

The overlap between the histograms of LR2 from original pairs of tumors and the reference distribution are produced by:

> histogramPlot(results\$LR[,4], results\$refLR[,4])

## Reference distribution of logLR (black), tested pairs (red)



## 2.1 Choice of segmentation algorithm

Note that the user can potentially specify the segmentation method to be used. Currently the default behavior of the clonality analysis function is to use the CBS algorithm to identify the most significant change in each chromosome arm. The internal function for this purpose is "oneseg" called as oneseg(x, alpha, nperm, sbdry)

There are 4 arguments to oneseg:

x: is the finite logratio data ordered by genomic position.

alpha: the significance level used by CBS.

nperm: the number of permutations for the reference distribution.

sbdry: early stopping boundary for declaring no change (calculated from alpha and nperm).

The output of this function is a vector of 3 numbers where the first is the number of change-points detected (must be 0, 1 or 2), and the second and the third numbers are the start and end of the left segment if there is only one change-point, and of the middle segment when there are 2 change-points.

The function allows the user to specify alternative alpha and nperm for 'oneseg' as a list using the segpar argument e.g. segpar=list(alpha=0.05, nperm=1000). Since sbdry is always calculated in clonality.analysis function from alpha and nperm it is not specified.

Alternate segmentation algorithm can be used. It requires the user to create a function that takes the ordered logratio from one chromosome arm as argument "x" as in oneseg. The name of this function should not be 'oneseg' and is passed through the 'segmethod' argument and all other necessary arguments that are needed passed as a list through 'segpar' argument.

## 3 LOH data

The LOH data has to be combined in a matrix where first column has marker names and the following columns have LOH calls for each sample. Here we simulate a dataset with 10 pairs of tumors and 20 markers. First pair of tumor is clonal, and the rest of them are independent. If the marker is heterozygous and there is no LOH, then it is denoted by 0. LOH at maternal or paternal alleles is marked by 1 or 2.

```
> set.seed(25)
> LOHtable<-cbind(1:20,matrix(sample(c(0,1,2),20*20,replace=TRUE),20))
> LOHtable[,3]<-LOHtable[,2]
> LOHtable[1,3]<-0</pre>
```

> LOHtable[,1:5]

	[,1]	[,2]	[,3]	[,4]	[,5]
[1,]	1	1	0	2	0
[2,]	2	2	2	0	0
[3,]	3	0	0	1	0
[4,]	4	2	2	2	2
[5,]	5	0	0	1	1
[6,]	6	2	2	0	2
[7,]	7	1	1	2	1
[8,]	8	1	1	2	2
[9,]	9	0	0	0	1
[10,]	10	0	0	2	0
[11,]	11	0	0	0	2
[12,]	12	1	1	2	0
[13,]	13	2	2	2	0
[14,]	14	1	1	1	2
[15,]	15	2	2	1	0

```
[16,]
                  0
                        0
                               0
          16
                                      1
[17,]
          17
                  1
                        1
                               0
                                      0
[18,]
                  2
                        2
                                      2
          18
                               1
                  1
                        1
                               0
                                      0
[19,]
          19
                  2
                        2
[20,]
          20
                               0
                                      0
```

> LOHclonality(LOHtable,rep(1:10,each=2),pfreq=NULL,noloh=0,loh1=1,loh2=2)

Tes	sting clo	onality :	for	pat	tie	ent	t 1	1, 2,	3,	4,	5,	6,	7,	8,	9,	10,	Done
	Sample1	Sample2	a	е	f	g	h	Ntot					CM	pval	lue	LRpv	alue
1	1	1	13	13	0	1	6	20	2.5	204	572	207	172	34e-	-08		0
2	2	2	3	6	4	6	4	20		0.6	633	257	174	3272	221		1
3	3	3	6	9	2	5	4	20		0	. 13	247	418	0050	031	0	.458
4	4	4	6	9	4	5	2	20		0.2	271	731	009	9409	983	0	.807
5	5	5	3	6	7	5	2	20		0.7	7680	026	723	9502	271		1
6	6	6	1	5	8	3	4	20		0.9	9640	059	678	1475	575	0	.442
7	7	7	6	12	3	4	1	20		0.6	6076	636	663	3207	756		1
8	8	8	5	11	4	2	3	20		0.5	585	520	481	5465	597	0	.719
9	9	9	4	7	6	5	2	20		0.5	5970	049	677	704:	141	0	.911
10	10	10	6	10	3	6	1	20		0.4	4249	944	369	1950	046	0	.794

First p-value is small, indicating clonality, for both CM and LR tests. The rest of the p-values are not significant.

Markers that are not informative (e.g. homozygous) in a particular tumor should be given NA instead of a call. Such markers will be dropped from the analysis of this specific patient.

#### 4 LOH data for 3 and more tumors

It is possible to test clonality of 3 or more tumors using Extended Concordant Mutations test, implemented in function 'ECMtesting'. The input LOH matrix can be in the same format as for 'LOHclonality' function: first column of a matrix contains marker names, subsequent columns are samples. For each patient all possible subsets of tumors are tested for clonality, with adjustment for multiple comparison performed using permutation MinP procedure.

Likelihood model can be extended for 3 or 4 tumors with function 'LRtesting3or4tumors'. The likelihood function depends on 2 parameters for 3 tumors, and 3 parameters for 4 tumors, allowing for non-symmetric relationship among tumors. Likelihood ratio test is computed and p-value is calculated using permutations.

Below are the details of the session information:

```
R version 3.0.2 Patched (2013-12-18 r64488) Platform: x86_64-unknown-linux-gnu (64-bit) locale:
```

[1] LC\_CTYPE=en\_US.UTF-8 LC\_NUMERIC=C

[3] LC\_TIME=en\_US.UTF-8 LC\_COLLATE=C

[5] LC\_MONETARY=en\_US.UTF-8 LC\_MESSAGES=en\_US.UTF-8

[7] LC\_PAPER=en\_US.UTF-8 LC\_NAME=C

[9] LC\_ADDRESS=C LC\_TELEPHONE=C

[11] LC\_MEASUREMENT=en\_US.UTF-8 LC\_IDENTIFICATION=C

#### attached base packages:

[1] stats graphics grDevices utils datasets methods base

### other attached packages:

[1] gdata\_2.13.2 Clonality\_1.10.1 DNAcopy\_1.36.0

loaded via a namespace (and not attached):

[1] gtools\_3.3.1 tools\_3.0.2

### References

Begg, C., Eng, K., and Hummer, A. (2007). Statistical tests for clonality. *Biometrics*, 63:522–530.

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