# SomatiCA: identifying, characterizing and quantifying somatic copy number aberrations from cancer genome sequencing

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# 1 Introduction

This guide provides a tour of the Bioconductor package SomatiCA, a R package that is capable of identifying, characterizing and quantifying somatic CNAs from cancer whole genome sequencing. It is especially designed for somatic copy number analyses taking into account: (i) an unknown fraction of normal cells (admixture rate) that are nearly always intermixed with cancer cells; and (ii) the heterogeneity of cancer cell population owing to ongoing subclonal evolution. The package implements a pipeline for characterizing somatic copy number aberrations based on different statistical methods: segmentation on Lesser Allele Frequency (LAF), Maximum Likelihood Estimation of somatic ratio (read depth ratio of tumor/ normal), admixture rate estimation by a Bayesian finite mixture model and subclonality characterization based on hypothesis testing. It is especially suitable for studies designed to understand tumor evolution. It currently works for cancer sample with control.

This manual is composed of basic test run on each module with a simulated dataset. The main purpose is to help users familiar with basic functions and the data format.

#### 1.1 User Guide

A user guide with detailed analysis on real data, is available as part of the online documentation. To reach the User's Guide, install the SomaticCNA package and load it into an R session by library(SomatiCA). In R for Windows, the User's Guide will then be available from the drop-down menu called "Vignettes". In other operating systems, type

```
> library(SomatiCA)
> SomatiCAUsersGuide()
```

[1] "/tmp/RtmpUKS2xe/Rinst1854231a810a/SomatiCA/doc/SomatiCAUserGuide.pdf"

at the R prompt to open the User's Guide in a pdf viewer.

# 2 Overview of capabilities

### 2.1 Input

Let us simulate a simple dataset with 1450 snps on 2 chromosome.

```
> set.seed(1)
> rawLAF <- c(rnorm(300, 0.2, 0.05), rnorm(300, 0.4, 0.05),</pre>
              rnorm(200, 0.3, 0.05), rnorm(200, 0.2, 0.05),
+
              rnorm(200, 0.3, 0.05), rnorm(250, 0.4, 0.05))
+
> rawLAF <- ifelse(rawLAF>0.5, 1-rawLAF, rawLAF)
> germLAF <- c(rnorm(800+650, 0.4, 0.05))</pre>
> germLAF <- ifelse(germLAF>0.5, 1-germLAF, germLAF)
> reads1 <- c(rpois(300, 25), rpois(300, 50), rpois(200, 60),
              rpois(200, 25), rpois(200, 40), rpois(250, 50))
+
> reads2 <- rpois(800+650,50)</pre>
> chr <- c(rep("chr1", 800), rep("chr2", 650))</pre>
> position <- c(seq(1, 16000000, by=20000), seq(1, 13000000, by=20000))
> zygo <- rep("het", 800+650)
> data <- data.frame(chr, as.integer(position), as.character(zygo), as.integer(re
>
```

The input of **SomatiCA** is formatted as a GRanges instance as defined in **Genomi**cRanges.

```
> colnames(data) <- c("seqnames", "start", "zygosity", "tCount",
+ "LAF", "tCountN", "germLAF")
> input <- SomatiCAFormat(data, missing = F, verbose = T)
> input
```

GRanges	with	1450	ranges	$\operatorname{and}$	5	metadata	a column	IS:
	seq	names				ranges	strand	
	~	<rle></rle>			<i< td=""><td>Ranges&gt;</td><td><rle></rle></td><td></td></i<>	Ranges>	<rle></rle>	
[1]	]	chr1		[	1,	1]	*	
[2]	]	chr1		[2000	)1,	20001]	*	
[3]	]	chr1		[4000	)1,	40001]	*	
[4]	]	chr1		[6000	)1,	60001]	*	
[5]	]	chr1		[8000	)1,	80001]	*	
•••						• • •		•••
[1446]	]	chr2	[129000	001,	12	900001]	*	

[1447]	chr2 [12	2920001, 12	2920001]	*
[1448]	chr2 [12	2940001, 12	2940001]	*
[1449]	chr2 [12	2960001, 12	2960001]	*
[1450]	chr2 [12	2980001, 12	2980001]	*
	zygosity	tCount	LAF	tCountN
	<character></character>	<integer></integer>	<numeric></numeric>	<integer></integer>
[1]	het	28	0.1686773	51
[2]	het	24	0.2091822	40
[3]	het	25	0.1582186	53
[4]	het	18	0.2797640	58
[5]	het	33	0.2164754	49
[1446]	het	44	0.3354413	50
[1447]	het	45	0.4453190	46
[1448]	het	49	0.2602019	43
[1449]	het	48	0.4116780	51
[1450]	het	44	0.3865482	40
	germLAF			
	<numeric></numeric>			
[1]	0.3670727			
[2]	0.4031673			
[3]	0.4027238			
[4]	0.4123505			
[5]	0.4062018			
•••				
[1446]	0.3966479			
[1447]	0.4220027			
[1448]	0.3239019			
[1449]	0.4054811			
[1450]	0.4094522			
seqlen	gths:			
chr1	chr2			
NA	NA			

### 2.2 Segmentation

Given input in the format of a GRanges object, larsCBSsegment() segments each chromosome with LAF of heterozygous sites on that chromosomes. larsCBSsegment() extracts hetergozygous sites by 'grep' any site with 'het' in the column of zygosity. For Complete Genomic data, sites with zygosity of 'het-ref' and 'het-alt' will be extracted. If genotype calling results are obtained from other platforms, a transformation is needed. For example, the 0/1/2 coding from VCF file will be needed to transform to 'het'/'hom' coding. Users can name 'zygosity' in their own way but keep in mind that only names containing 'het' will be used as heterozygous sites for segmentations in SomatiCA, such as 'het', 'het-ref', 'heter', 'heter1' etc.

larsCBSsegment() firstly calls a function denoise() to smooth the outliers. Then it segment each chromosome with CBS followed by a model selection procedure. The default model selection criteria is Bayesian Information Criterion (BIC). Users can set rss=T to apply BIC plus a minimum cut-off for change in residue sum of squares (RSS) between neighboring change points. collapse.k is an option to average LAF on each k SNPs.

> seg <- larsCBSsegment(input, collapse.k = 0, ncores = 1, verbose = T, rss = FAL

This is a toy example without much noise. Consider to use rss=T to select change points from sequencing data. Output of larsCBSsegment() includes two part: segmentation results and heterozygous sites used for segmentation (denoised).

> seg

\$segment							
GRanges	with 6	ranges and	2 metadata	a column	s	:	
se	qnames		ranges	strand	I	medLAF	
	<rle></rle>		<iranges></iranges>	<rle></rle>	I	<numeric></numeric>	
[1]	chr1	[ 0,	5900001]	*	I	0.1981	
[2]	chr1	[ 5900001,	11900001]	*	I	0.3976	
[3]	chr1	[11900001,	15980001]	*	I	0.2961	
[4]	chr2	[ 0,	3900001]	*	Ι	0.2068	
[5]	chr2	[ 3900001,	7920001]	*	Ι	0.2936	
[6]	chr2	[ 7920001,	12980001]	*	I	0.3977	
:	medgLAF	,					
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[1]	0.4009	)					
[2]	0.3964	-					
[3]	0.3908	3					
[4]	0.4037	,					
[5]	0.3977	,					
[6]	0.4036	5					
seqlen	gths:						
chr1	chr2						
NA	NA						
\$hetsite	s						

4

GRanges	with 1450	ranges	$\operatorname{and}$	5	metada	ta d	columns	5:
	seqnames				range	s st	trand	
	<rle></rle>			<]	Ranges	> <	<rle></rle>	
[1]	chr1		[	1,	1	]	*	Ι
[2]	chr1		[2000	)1,	20001	]	*	Ι
[3]	chr1		[4000	)1,	40001	]	*	
[4]	chr1		[6000	)1,	60001	]	*	Ι
[5]	chr1		[8000	)1,	80001	]	*	Ι
		<b>F</b> 40000	201		•••	•	••••	•••
[1440]	cnr2	[129000	JUI,	12	2900001	] I	*	
	cnr2	[129200	JUI,	12	2920001	J	*	
[1448]	chr2	[129400	JUI,		2940001	J	*	
[1449]	chr2	[129600	JUI,		2960001	J	*	
[1450]	chr2	[129800	J01,	12	2980001	]	*	
	zygosi	ty 1	tCour	nt		LAF	tCou	intN
F / 7	<characte< td=""><td>er&gt; <int< td=""><td>teger</td><td>:&gt;</td><td><numer:< td=""><td>ic&gt;</td><td><integ< td=""><td>;er&gt;</td></integ<></td></numer:<></td></int<></td></characte<>	er> <int< td=""><td>teger</td><td>:&gt;</td><td><numer:< td=""><td>ic&gt;</td><td><integ< td=""><td>;er&gt;</td></integ<></td></numer:<></td></int<>	teger	:>	<numer:< td=""><td>ic&gt;</td><td><integ< td=""><td>;er&gt;</td></integ<></td></numer:<>	ic>	<integ< td=""><td>;er&gt;</td></integ<>	;er>
[1]	ł	net	2	28	0.1686	773		51
[2]	ł	net	2	24	0.2091	822		40
[3]	ł	net	2	25	0.1582	186		53
[4]	ł	net	1	8	0.2797	640		58
[5]	ł	net	3	33	0.2164	754		49
[1446]	ł	net		14	0.3354	 413		 50
[1447]	ł	net	4	15	0.4453	190		46
- [1448]	ł	net	4	19	0.2602	019		43
[1449]	ł	net	4	18	0.4116	780		51
[1450]	- }	net	4	4	0.3865	482		40
23	germI.AF	7	_					_ •
	<numeric></numeric>	>						
[1]	0.3670727	7						
[2]	0 4031673	2						
[3]	0 4027238	2						
[0] [4]	0 412350	5						
[5]	0.4062018	3						
[1446]	0.3966479	)						
[1447]	0.4220027	7						
[1448]	0.3239019	)						
[1449]	0.4054811	L						
[1450]	0.4094522	2						
seqler	gths:							



Figure 1: Segmentation on chr1 based on LAF.

chr1 chr2 NA NA

The segmentation result can be examined by plotSegment(). If multiple chromosomes are segmented, use k to plot segmentation on each chromosome respectively by their orders in the input.

```
> plotSegment(seg$segment, input, k = 1, smooth = F)
```

### 2.3 Estimation of somatic ratio

Somatic ratio is defined as the ratio of read depths between a tumor and its paired normal sample for a given segment. SomatiCA implements different methods to estimate somatic ratio. For the "mle" method somatic ratio is estimated by a maximum likelihood approach. For the "mean" method, somatic ratio is estimated by the ratio between mean of tumor sample and normal sample. For the "geometric", somatic ratio is estimated by geometric mean of somatic ratios of all sites in a given segment. To estimate somatic ratio, both segmentation and input with read depths are required.

```
> segmentwithRatio <- somaticRatio(seg$segment, input, method = "mle")
> segmentwithRatio
```

GRanges	s with 6	ranges and	3 metadata	a columr	າຮ	:	
5	seqnames		ranges	${\tt strand}$	Ι	medLAF	
	<rle></rle>		<iranges></iranges>	<rle></rle>	Ι	<numeric></numeric>	
[1]	chr1	[ 0,	5900001]	*	Ι	0.198	
[2]	chr1	[ 5900001,	11900001]	*		0.397	
[3]	chr1	[11900001,	15980001]	*		0.2956	
[4]	chr2	[ 0,	3900001]	*	Ι	0.2065	
[5]	chr2	[ 3900001,	7920001]	*		0.293	
[6]	chr2	[ 7920001,	12980001]	*		0.3986	
	medgLAF	' ratio					
•	<numeric></numeric>	<pre>   <numeric> </numeric></pre>					
[1]	0.4009	0.509					
[2]	0.3964	0.979					
[3]	0.3908	1.191					
[4]	0.4037	0.501					
[5]	0.3977	0.805					
[6]	0.4036	0.997					
seqle	engths:						
chr	1 chr2						
N	A NA						

## 2.4 Refine segments

Two adjacent segments are merged if the difference in the somatic ratios is less than "threshold2", which is tunable in the implementation with its default value being 0.05, equivalent to 5% change in somatic copy-number without normal contamination. The MLEs of the somatic ratio for the refined segments are recalculated. This refinement procedure is applied repeatedly until no adjacent segments have somatic ratio difference less than T. "threshold1" is the threshold used to merge the segments based on median LAF.

```
> refined <- refineSegment(segmentwithRatio, input, method = "mle",
+ adjust = TRUE, threshold1 = 0 , threshold2 = 0.05)
```

GRange	es with 6	ranges and	3 metadata	a columi	ns	:
	seqnames		ranges	strand	I	medLAF
	<rle></rle>		<iranges></iranges>	<rle></rle>	Ι	<numeric></numeric>
[1]	chr1	[ 0,	5900001]	*		0.198
[2]	chr1	[ 5900001,	11900001]	*		0.397
[3]	chr1	[11900001,	15980001]	*		0.2956
[4]	chr2	[ 0,	3900001]	*	Ι	0.2065
[5]	chr2	[ 3900001,	7920001]	*		0.293
[6]	chr2	[ 7920001,	12980001]	*		0.3986
	medgLAF	ratio				
	<numeric></numeric>	<pre>&gt; <numeric></numeric></pre>				
[1]	0.4009	0.509				
[2]	0.3964	0.979				
[3]	0.3908	3 1.191				
[4]	0.4037	0.501				
[5]	0.3977	0.805				
[6]	0.4036	6 0.997				
seq	lengths:					
chi	r1 chr2					
1	IA NA					

#### 2.5 Estimation of admixture rate

The estimation of the admixture rate is accomplished by fitting the somatic copy number (somatic ratio<sup>\*</sup>2) of all segments with a Bayesian finite mixture model, with components centered at the discrete levels. Each segment was assigned with a discrete level based on corresponding posterior probability. Segments with ambiguous assignments will be classified as candidate subclonal events and excluded from admixture rate inference. The admixture rate will be estimated by an optimal solution contributed by explanation of tumor copy number with all remaining segments as integer level. *SomatiCA* implements a Markov Chain Monte Carlo with Metropolis Hasting algorithm to estimate posterior probabilities of the Bayesian finite mixture model. Option mcmc set the number of MCMC iteration, burnin set the number of MCMC iteration for burnin and p set the cutoff of posterior probability for ambiguous integer copy number assignments.

```
> ll <- admixtureRate(refined, mcmc = 5000, burnin = 1000, p = 0.01)
> admix <- ll$admix</pre>
```

copynumberCorrected could take a segmentation profile and an admixture rate to calculate the integer somatic copy number in tumor sample then characterize somatic events based on corrected somatic ratio. Each segment will be annotated with "=", "Loss", "Gain", "LOH", "neutral LOH" or "double deletion". Note that somatic copy number here is calculated from the ratio with the assumption that the control sample is diploid. This result will likely be modified in next step with the calculation of ploidy of each segment in the control sample.

```
> y <- copynumberCorrected(segmentwithRatio, admix)
> y
```

GRan	ges with	6 ranges ar	nd 5 metad	ata column	IS :		
	seqnames		ranges	strand	medLAF		
	<rle></rle>		<iranges></iranges>	<rle>  </rle>	<numeric></numeric>		
1	chr1	[ 0,	5900001]	*	0.1980		
2	chr1	[ 5900001,	11900001]	*	0.3970		
3	chr1	[11900001,	15980001]	*	0.2956		
4	chr2	[ 0,	3900001]	*	0.2065		
5	chr2	[ 3900001,	7920001]	*	0.2930		
6	chr2	[ 7920001,	12980001]	*	0.3986		
	medgLAF	ratio	somaCN	event	;		
	<numeric></numeric>	<numeric></numeric>	<array> &lt;</array>	character>			
1	0.4009	0.509	1	LOH	I		
2	0.3964	0.979	2	=	:		
3	0.3908	1.191	2	=	:		
4	0.4037	0.501	1	Loss	5		
5	0.3977	0.805	2	=	:		
6	0.4036	0.997	2	=	:		
	-						
se	seqlengths:						
C	NA NA						

#### 2.6 Subclonality characterization

SomatiCA estimates subclonality for each somatic copy number abberation. To do this, it first calculates allelic copy number nB and nA (segment level allelic copy number is estimated by median in that segment) in a control sample based on GC corrected read counts. SomatiCA tests whether copy number change in corresponding tumor sample can result in a change of exactly one copy of one allele. If the somatic ratio (corrected by admixture rate) in the corresponding tumor sample is greater than 1, SomatiCA tests for one copy gain, otherwise it tests for one copy loss. With null hypothesis that clonal

copy number ratio follows a normal distribution, p-value is calculated for each segment as the probability of obtaining a copy number ratio at least as extreme as the one that was actually observed. Segments with p-value less than 0.05 are classified as subclonal.

```
> data(GCcontent)
> segmentGCcorrected <- segmentGCbiasRemoval(y, input, GCcontent)
> segmentClonality <- subclonality(segmentGCcorrected, admix)</pre>
```

Users can further merge neighboring segments with same somatic copy number and events together using MergeSegment if necessary.

```
> merged <- MergeSegment(segmentClonality)</pre>
```

The subclonality and somatic copy number for a sample can be visualized by plotSubclonality.

```
> plotSubclonality(segmentClonality)
```

Then you can output all the clonal events or the subclonal events with high proportion of aberration.

```
> merged[elementMetadata(merged)[, "clonality"]=="clonal", ]
```

GRange	es with 2	ranges	and 6	metada	ta co	lumns:	
	seqnames	:	ranges	strand		somaCN	
	<rle></rle>	<ir:< td=""><td>anges&gt;</td><td><rle></rle></td><td>  <c]< td=""><td>haracter&gt;</td><td></td></c]<></td></ir:<>	anges>	<rle></rle>	<c]< td=""><td>haracter&gt;</td><td></td></c]<>	haracter>	
[1]	chr1	[0, 590	00001]	*		1	
[2]	chr2	[0, 390	00001]	*		1	
	eve	ent c	lonalit	y g	ermCN	subclonal(	CN
	<characte< td=""><td>er&gt; <ch< td=""><td>aracter</td><td>&gt; <int< td=""><td>eger&gt;</td><td><integer< td=""><td><u>;</u>&gt;</td></integer<></td></int<></td></ch<></td></characte<>	er> <ch< td=""><td>aracter</td><td>&gt; <int< td=""><td>eger&gt;</td><td><integer< td=""><td><u>;</u>&gt;</td></integer<></td></int<></td></ch<>	aracter	> <int< td=""><td>eger&gt;</td><td><integer< td=""><td><u>;</u>&gt;</td></integer<></td></int<>	eger>	<integer< td=""><td><u>;</u>&gt;</td></integer<>	<u>;</u> >
[1]	I	LOH	clona	l	2		1
[2]	Lo	DSS	clona	l	2		1
	subpercen	nt					
	<numerio< td=""><td>c&gt;</td><td></td><td></td><td></td><td></td><td></td></numerio<>	c>					
[1]		1					
[2]		1					
seql	lengths:						
chr	1 chr2						
Ν	IA NA						



Figure 2: An example of subclonality.

# 3 SomatiCA pipeline

Get familiar with how SomatiCA works? If so, call SomatiCApipe directly to run all the steps described above automatically. Multithread computing is still supported through ncores. We recommend to use verbose=T to print working messages and track the working progress.

# 4 Others

# 4.1 GC content

In case the precalculated GC content will be out of date or users may want to use smaller window size, we provide a function first downloading the .fa.gz of a given chromosome from UCSC genome browser and then calculating the GC content for a given window size (10,000 bp in the following example).

```
> chr <- paste("chr", c(1:22, "X"), sep="")
> url <- "http://hgdownload.soe.ucsc.edu/goldenPath/hg19/chromosomes/"
> GC <- foreach(i=chr, .combine=rbind)%dopar%{
+ return(GCcount(i, 10000, url))
+ }</pre>
```