# BCGES short courses, session 1 Picard tools, CNV analysis

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## 1 Two more advanced ideas to consider

#### 1.1 samtools over the web to download BAM files

A powerful feature of BAM/SAM and samtools is the ability to use the view feature with indexes over the web, to only pull the chunks of interest. I will provide an example of this using the 1,000 Genomes data. To do that, one should first download the list of whole genome alignments from the 1,000 Genomes to identify where the files of interest are located on the server.

```
wget -0 1KG_alignment.index ftp://ftp.1000genomes.ebi.ac.uk/vol1/ftp/alignment.index
```

**Exercise:** Download data from a BAM file generated by the ENCODE project that contains ChIP-Seq information about *GATA2* binding sites 1 Mb around the gene *IL18RAP*. Just to help the search look for the data located here.

index=//hgdownload.cse.ucsc.edu/goldenPath/hg19/encodeDCC/wgEncodeUchicagoTfbs/files.txt
wget -0 encode\_index.tab ftp:\$index

#### 1.2 Linux piping to limit input/output usage

The lines of code below cannot be run as such but are just meant to give an example of the sort of things that multiple lines of code put together can do. In this example the only thing written to the disk is the sorted BAM file. No other intermediate file is written to disk.

```
novoalign -c 11 -o SAM -F STDFQ -f fasta1.fq fasta2.fq -d $reference | samtools view - -u -S -b | novosort - -t /scratch0/ -c 1 -m 3G -i -o output_sorted.bam
```

Exercise: Understand what the various options mean, at least for samtools which is freely available.

# 2 Some examples PICARD tools

#### 2.1 Marking duplicates

## 2.2 Collect summary statistics

And here is some example code to compute summary statistics on a BAM file:

```
BAM=../data/BAM_files/HG00130.mapped.ILLUMINA.bwa.GBR.exome.20130415.bam
java -Xmx4g -jar picard/CollectAlignmentSummaryMetrics.jar \
INPUT=$BAM \
OUTPUT=insert_size.txt
```

## 2.3 Convert a BAM file to fastq

Now something quite a bit more challenging: try the following script to convert a BAM file to fastq: Here is a first attempt

```
BAM=../data/BAM_files/HG00130.mapped.ILLUMINA.bwa.GBR.exome.20130415.bam
java -Xmx4g -jar picard/SamToFastq.jar \
INPUT=$BAM \
FASTQ=read1.fq SECOND_END_FASTQ=read2.fq
```

The issue is that some reads do not have a mate, and that creates issues with the FASTQ files. We need a fix to deal with that error message. We can try the following, which should remove non-mapped reads:

```
BAM=../data/BAM_files/HG00130.mapped.ILLUMINA.bwa.GBR.exome.20130415.bam samtools view -f 0x0001 -f 0x0002 -b -o only_paired.bam $BAM

java -Xmx4g -jar picard/SamToFastq.jar INPUT=only_paired.bam FASTQ=read1.fq SECOND_END_FASTQ=read2.f
```

But it fails again. With only 2 unpaired mates this time, so things are getting better. We can find out what these are:

Now let us remove these problematic reads, but we have to do it manually. I could not find a better way to do this than what is shown below:

```
samtools view -h only_paired.bam | awk '{if (($1 != "SRR707198.24062323") && ($1  != "SRR707198.20187117")) print}' > paired_fixed samtools view -b -S -o paired_fixed.bam paired_fixed.sam
```

The second line above takes a SAM file input (option -S) and returns a BAM file (option -b). The resulting file can, at last, be converted into a FASTQ format:

```
java -Xmx4g -jar picard/SamToFastq.jar INPUT=paired_fixed.bam FASTQ=read1.fq SECOND_END_FASTQ=read2
```

**Exercise:** Can you find a better way to convert a BAM to fastq? I could not but I have not looked extensively.

# 3 Read depth analysis to detect CNVs

Run the example code provided in exomeDepth\_example.R. Familiarise yourself with the code, modify it to run with the sample UCLG\_94\_sorted\_unique.bam. You should be able to see a two exon heterozygous deletion located in the gene *GATA2*, which is causal in this individual.

```
library(ExomeDepth)
```

```
ExomeCount <- read.table('../data/exon_read_count.tab.gz', header = TRUE)</pre>
my.test <- ExomeCount$UCLG.186_sorted_unique.bam
all.bams <- grep(pattern = 'bam$', names(ExomeCount), value = TRUE)</pre>
my.ref.samples <- subset( all.bams, all.bams != 'UCLG.186_sorted_unique.bam')
my.reference.set <- as.matrix(ExomeCount[, my.ref.samples])</pre>
my.choice <- select.reference.set (test.counts = my.test,</pre>
                                    reference.counts = my.reference.set,
                                    bin.length = (ExomeCount$end - ExomeCount$start)/1000,
                                    n.bins.reduced = 10000)
## Warning: The data set contains at least one line with weight = 0.
##
## Warning: The data set contains at least one line with weight = 0.
## Warning: The data set contains at least one line with weight = 0.
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##
## Warning: The data set contains at least one line with weight = 0.
my.matrix <- as.matrix( ExomeCount[, my.choice$reference.choice, drop = FALSE])</pre>
my.reference.selected <- apply(X = my.matrix,
                               MAR = 1,
                               FUN = sum)
######## create the ExomeDepth object
all.exons <- new('ExomeDepth',</pre>
                 test = my.test,
                 reference = my.reference.selected,
                 formula = 'cbind(test, reference) ~ 1')
## Warning: The data set contains at least one line with weight = 0.
all.exons <- CallCNVs(x = all.exons,
                      transition.probability = 10^-4,
                      chromosome = ExomeCount$chromosome,
                      start = ExomeCount$start,
                      end = ExomeCount$end,
                      name = ExomeCount$exons)
############# Now annotate the CNV calls
data(Conrad.hg19)
all.exons <- AnnotateExtra(x = all.exons,
```

```
reference.annotation = Conrad.hg19.common.CNVs,
                          min.overlap = 0.5,
                          column.name = 'Conrad.hg19')
data(exons.hg19)
exons.hg19.GRanges <- GRanges (seqnames = exons.hg19$chromosome,
                             IRanges(start=exons.hg19$start,end=exons.hg19$end),
                             names = exons.hg19$name)
all.exons <- AnnotateExtra(x = all.exons,</pre>
                          reference.annotation = exons.hg19.GRanges,
                          min.overlap = 0.0001,
                          column.name = 'exons.hg19')
###### Take the final table, sort by significance (Bayes factor)
my.final.table <- all.exons@CNV.calls
my.final.table <- my.final.table[ order(my.final.table$BF, decreasing = TRUE),]
print(table(is.na(my.final.table$Conrad.hg19)))
## FALSE TRUE
## 33
         18
print(head(my.final.table))
     start.p end.p
                         type nexons
                                         start
                                                    end chromosome
## 51 66663 66693
                    deletion 31 151749334 151948710
      11958 11982
                    deletion 25 104202984 104293260
## 17
## 8
        2813 2821
                    deletion
                                   9 16905690 16915559
## 23 13791 13795
                    deletion
                                   5 143361979 143375274
                                                                  1
## 3 2273 2287 duplication 15 13109020 13165444 ## 4 2292 2295 deletion 4 13196272 13217250
                          id BF reads.expected reads.observed
## 51 chr4:151749334-151948710 913.0
                                            5348
## 17 chr1:104202984-104293260 129.0
                                             1025
                                                             196
                                            6259
## 8
     chr1:16905690-16915559 114.0
                                                            3302
## 23 chr1:143361979-143375274 72.4
                                             550
                                                            115
                                                            1255
## 3 chr1:13109020-13165444 66.0
                                              510
## 4
       chr1:13196272-13217250 48.0
                                             322
                                                             50
## reads.ratio
                       Conrad.hg19
## 51
          0.0015
## 17
          0.1910 CNVR266.3, CNVR266.5
          0.5280 CNVR93.3,CNVR93.6
## 8
## 23
          0.2090
                          CNVR330.2
## 3
          2.4600 CNVR73.1,CNVR73.2
## 4
          0.1550 CNVR73.2, CNVR73.3
## 51 LRBA_30,LRBA_29,LRBA_28,LRBA_27,LRBA_26,LRBA_25,LRBA_24,LRBA_23,LRBA_22,LRBA_21,LRBA_20,LRBA_1
## 17
## 8
```

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# 4 Use Rsamtools to identify reads characteristic of a deletion

Another strategy to detect CNVs consists of picking up reads that show an unusual pattern. For example, two pairs further apart than they normally should is potentially flagging a deletion. In this exercise I proposed to look at a well described 20 kb CNV located near the IRGM gene. We can start by looking at the location of the variant using the data from the Conrad et al paper, which are loaded into ExomeDepth.

We now want to look for individuals that carry this variant. Low coverage whole genome data for the 1KG sample HG00123 will do the job.

Exercise: Identify the BAM file for mapped reads for HG00123 and download the slice 5:150200000-150225000 using samtools view. How would you filter pairs of reads that span over the deletion in this sample?

```
BAM=//ftp.1000genomes.ebi.ac.uk/vol1/ftp/data/HG00123/alignment/HG00123.mapped.ILLUMINA.bwa.GBR.low_samtools view -b ftp:$BAM 5:150200000-150225000 > file.bam samtools view file.bam | awk '{if (($9 > 15000$) && ($9 < 25000)) print}' | less -S
```

(Optional) Exercise: Can you write a R script (using Rsamtools) to identify these reads?