

VAR-Seq project report template: Some Descriptive Title

Project ID: VARseq_PI_Name_Organism_Mar2015

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1 Introduction

This report describes the analysis of an VAR-Seq project from Dr. First Last's lab which studies the gene expression changes of ... in *organism* The experimental design is as follows...

2 Sample definitions and environment settings

2.1 Environment settings and input data

Typically, the user wants to record here the sources and versions of the reference genome sequence along with the corresponding annotations. In the provided sample data set all data inputs are stored in a data subdirectory and all results will be written to a separate results directory, while the `systemPipeVARseq.Rnw` script and the `targets` file are expected be located in the parent directory. The R session is expected to run from this parent directory.

To run this sample report, mini sample FASTQ and reference genome files can be downloaded from [here](#). The chosen data set [SRP010938](#) contains 18 paired-end (PE) read sets from *Arabidopsis thaliana* [Howard et al. \(2013\)](#). To minimize processing time during testing, each FASTQ file has been subsetting to 90,000-100,000 randomly sampled PE reads that map to the first 100,000 nucleotides of each chromosome of the *A. thaliana* genome. The corresponding reference genome sequence (FASTA) and its GFF annotation files (provided in the same download) have been truncated accordingly. This way the entire test sample data set is less than 200MB in storage space. A PE read set has been chosen for this test data set for flexibility, because it can be used for testing both types of analysis routines requiring either SE (single end) reads or PE reads.

2.2 Required packages and resources

The `systemPipeR` package needs to be loaded to perform the analysis steps shown in this report ([Girke, 2014](#)).

```
> library(systemPipeR)
```

If applicable load custom functions not provided by `systemPipeR`

```
> source("systemPipeVARseq_Fct.R")
```

2.3 Experiment definition provided by targets file

The `targets` file defines all FASTQ files and sample comparisons of the analysis workflow.

```
> targetspath <- system.file("extdata", "targetsPE.txt", package="systemPipeR")
> targets <- read.delim(targetspath, comment.char = "#")[,1:5]
> targets
```

	FileName1	FileName2	SampleName	Factor	SampleLong
1	./data/SRR446027_1.fastq	./data/SRR446027_2.fastq	M1A	M1	Mock.1h.A
2	./data/SRR446028_1.fastq	./data/SRR446028_2.fastq	M1B	M1	Mock.1h.B
3	./data/SRR446029_1.fastq	./data/SRR446029_2.fastq	A1A	A1	Avr.1h.A
4	./data/SRR446030_1.fastq	./data/SRR446030_2.fastq	A1B	A1	Avr.1h.B
5	./data/SRR446031_1.fastq	./data/SRR446031_2.fastq	V1A	V1	Vir.1h.A
6	./data/SRR446032_1.fastq	./data/SRR446032_2.fastq	V1B	V1	Vir.1h.B

7	./data/SRR446033_1.fastq	./data/SRR446033_2.fastq	M6A	M6	Mock.6h.A
8	./data/SRR446034_1.fastq	./data/SRR446034_2.fastq	M6B	M6	Mock.6h.B
9	./data/SRR446035_1.fastq	./data/SRR446035_2.fastq	A6A	A6	Avr.6h.A
10	./data/SRR446036_1.fastq	./data/SRR446036_2.fastq	A6B	A6	Avr.6h.B
11	./data/SRR446037_1.fastq	./data/SRR446037_2.fastq	V6A	V6	Vir.6h.A
12	./data/SRR446038_1.fastq	./data/SRR446038_2.fastq	V6B	V6	Vir.6h.B
13	./data/SRR446039_1.fastq	./data/SRR446039_2.fastq	M12A	M12	Mock.12h.A
14	./data/SRR446040_1.fastq	./data/SRR446040_2.fastq	M12B	M12	Mock.12h.B
15	./data/SRR446041_1.fastq	./data/SRR446041_2.fastq	A12A	A12	Avr.12h.A
16	./data/SRR446042_1.fastq	./data/SRR446042_2.fastq	A12B	A12	Avr.12h.B
17	./data/SRR446043_1.fastq	./data/SRR446043_2.fastq	V12A	V12	Vir.12h.A
18	./data/SRR446044_1.fastq	./data/SRR446044_2.fastq	V12B	V12	Vir.12h.B

3 Read preprocessing

3.1 FASTQ quality report

The following `seeFastq` and `seeFastqPlot` functions generate and plot a series of useful quality statistics for a set of FASTQ files including per cycle quality box plots, base proportions, base-level quality trends, relative k-mer diversity, length and occurrence distribution of reads, number of reads above quality cutoffs and mean quality distribution. The results are written to a PDF file named `fastqReport.pdf`.

```
> args <- systemArgs(sysma="bwa.param", mytargets="targets.txt")
> fqlist <- seeFastq(fastq=infile1(args), batchsize=100000, klength=8)
> pdf("./results/fastqReport.pdf", height=18, width=4*length(fqlist))
> seeFastqPlot(fqlist)
> dev.off()
```

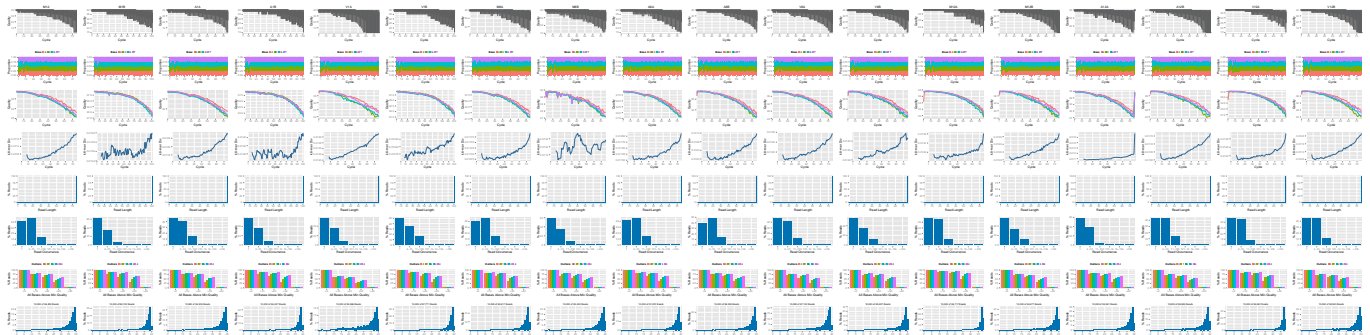


Figure 1: QC report for 18 FASTQ files.

4 Alignments

4.1 Read mapping with BWA

The NGS reads of this project will be aligned against the reference genome sequence using BWA (Li, 2013; Li and Durbin, 2009). The parameter settings of the aligner are defined in the `bwa.param` file.

```
> args <- systemArgs(sysma="bwa.param", mytargets="targets.txt")
> sysargs(args)[1] # Command-line parameters for first FASTQ file
```

Run alignments sequentially (e.g. on a single machine)

```
> bampaths <- runCommandline(args=args)
```

Alternatively, submit alignment jobs to compute cluster, here using 72 CPU cores (18 qsub processes each with 4 CPU cores).

```
> moduleload(modules(args))
> system("bwa index -a bwtsv ./data/tair10.fasta")
> resources <- list(walltime="20:00:00", nodes=paste0("1:ppn=", cores(args)), memory="10gb")
> reg <- clusterRun(args, conffile=".BatchJobs.R", template="torque.tmpl", Njobs=18, runid="01",
+                   resourceList=resources)
```

Check whether all BAM files have been created

```
> file.exists(outpaths(args))
```

4.2 Read and alignment stats

The following provides an overview of the number of reads in each sample and how many of them aligned to the reference.

```
> read_statsDF <- alignStats(args=args)
> write.table(read_statsDF, "results/alignStats.xls", row.names=FALSE, quote=FALSE, sep="\t")
> read.delim("results/alignStats.xls")
```

4.3 Create symbolic links for viewing BAM files in IGV

The symLink2bam function creates symbolic links to view the BAM alignment files in a genome browser such as IGV. The corresponding URLs are written to a file with a path specified under urlfile, here [IGVurl.txt](#).

```
> symLink2bam(sysargs=args, htmlldir=c("~/html/", "somedir/"),
+             urlbase="http://biocluster.ucr.edu/~tgirke/",
+             urlfile="./results/IGVurl.txt")
```

5 Variant calling

The following performs variant calling with GATK and BCFtools in parallel mode on a compute cluster ([McKenna et al., 2010](#); [Li, 2011](#)). Alternatively, runCommandline(args=args) will run them sequentially on a single machine.

5.1 Variant calling with GATK

```
> writeTargetsout(x=args, file="targets_bam.txt")
> system("java -jar CreateSequenceDictionary.jar R=./data/tair10.fasta O=./data/tair10.dict")
> # system("java -jar /opt/picard/1.81/CreateSequenceDictionary.jar R=./data/tair10.fasta O=./data/tair10.dict")
> args <- systemArgs(sysma="gatk.param", mytargets="targets_bam.txt")
> resources <- list(walltime="20:00:00", nodes=paste0("1:ppn=", 1), memory="10gb")
> reg <- clusterRun(args, conffile=".BatchJobs.R", template="torque.tmpl", Njobs=18, runid="01",
+                   resourceList=resources)
> #unlink(outfile1(args), recursive = TRUE, force = TRUE)
```

5.2 Variant calling with BCFtools

```
> args <- systemArgs(sysma="sambcf.param", mytargets="targets_bam.txt")
> resources <- list(walltime="20:00:00", nodes=paste0("1:ppn=", 1), memory="10gb")
> reg <- clusterRun(args, conffile=".BatchJobs.R", template="torque.tmpl", Njobs=18, runid="01",
+                   resourceList=resources)
> #unlink(outfile1(args), recursive = TRUE, force = TRUE)
```

6 Filtering variants

6.1 Filtering variants from GATK

```
> args <- systemArgs(sysma="filter_gatk.param", mytargets="targets_gatk.txt")
> filter <- "totalDepth(vr) >= 2 & (altDepth(vr) / totalDepth(vr) >= 0.8) & rowSums(softFilterMatrix(vr))=0"
> # filter <- "totalDepth(vr) >= 20 & (altDepth(vr) / totalDepth(vr) >= 0.8) & rowSums(softFilterMatrix(vr))=0"
> filterVars(args, filter, varcaller="gatk", organism="A. thaliana")
> writeTargetsout(x=args, file="targets_gatk_filtered.txt")
```

6.2 Filtering variants from BCFtools

```
> args <- systemArgs(sysma="filter_sambcf.param", mytargets="targets_sambcf.txt")
> filter <- "rowSums(vr) >= 2 & (rowSums(vr[,3:4])/rowSums(vr[,1:4]) >= 0.8)"
> # filter <- "rowSums(vr) >= 20 & (rowSums(vr[,3:4])/rowSums(vr[,1:4]) >= 0.8)"
> filterVars(args, filter, varcaller="bcftools", organism="A. thaliana")
> writeTargetsout(x=args, file="targets_sambcf_filtered.txt")
```

7 Annotate filtered variants

7.1 Annotate filtered variants from GATK

```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_gatk_filtered.txt")
> txdb <- loadDb("./data/tair10.sqlite")
> fa <- FaFile(systemPipeR::reference(args))
> variantReport(args=args, txdb=txdb, fa=fa, organism="A. thaliana")
```

7.2 Annotate filtered variants from BCFtools

```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_sambcf_filtered.txt")
> txdb <- loadDb("./data/tair10.sqlite")
> fa <- FaFile(systemPipeR::reference(args))
> variantReport(args=args, txdb=txdb, fa=fa, organism="A. thaliana")
```

8 Combine annotation results among samples

8.1 Combine results from GATK

```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_gatk_filtered.txt")
> combineDF <- combineVarReports(args, filtercol=c(Consequence="nonsynonymous"))
> write.table(combineDF, "./results/combinedDF_nonsyn_gatk.xls", quote=FALSE, row.names=FALSE, sep="\t")
```

8.2 Combine results from BCFtools

```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_sambcf_filtered.txt")
> combineDF <- combineVarReports(args, filtercol=c(Consequence="nonsynonymous"))
> write.table(combineDF, "./results/combinedDF_nonsyn_sambcf.xls", quote=FALSE, row.names=FALSE, sep="\t")
```

9 Summary statistics of variants

9.1 Summary for GATK

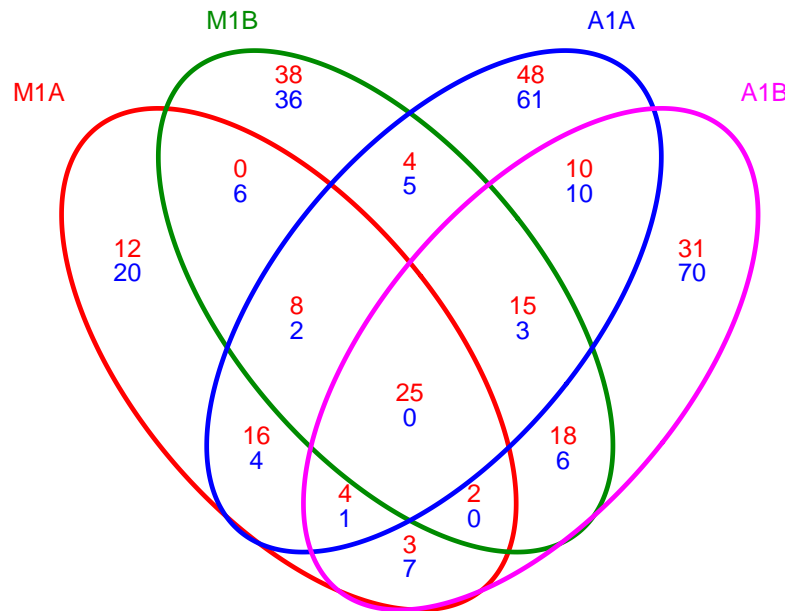
```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_gatk_filtered.txt")
> write.table(varSummary(args), "./results/variantStats_gatk.xls", quote=FALSE, col.names = NA, sep="\t")
```

9.2 Summary for BCFtools

```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_sambcf_filtered.txt")
> write.table(varSummary(args), "./results/variantStats_sambcf.xls", quote=FALSE, col.names = NA, sep="\t")
```

10 Venn diagram of variants

```
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_gatk_filtered.txt")
> varlist <- sapply(names(outpaths(args))[1:4], function(x) as.character(read.delim(outpaths(args)[x])$VAR))
> vennset_gatk <- overLapper(varlist, type="vennsets")
> args <- systemArgs(sysma="annotate_vars.param", mytargets="targets_sambcf_filtered.txt")
> varlist <- sapply(names(outpaths(args))[1:4], function(x) as.character(read.delim(outpaths(args)[x])$VAR))
> vennset_bcf <- overLapper(varlist, type="vennsets")
> pdf("./results/vennplot_var.pdf")
> vennPlot(list(vennset_gatk, vennset_bcf), mymain="", mysub="GATK: red; BCFtools: blue", colmode=2, ccol=2)
> dev.off()
```



GATK: red; BCFtools: blue

Figure 2: Venn Diagram for 4 samples from GATK and BCFtools.

11 Version Information

```
> toLatex(sessionInfo())
```

- R version 3.1.2 (2014-10-31), x86_64-unknown-linux-gnu
- Locale: C
- Base packages: base, datasets, grDevices, graphics, methods, parallel, stats, stats4, utils
- Other packages: AnnotationDbi 1.28.2, Biobase 2.26.0, BiocGenerics 0.12.1, BiocParallel 1.0.2, Biostrings 2.34.1, DBI 0.3.1, GenomInfoDb 1.2.4, GenomicAlignments 1.2.1, GenomicRanges 1.18.4, IRanges 2.0.1, RSQLite 1.0.0, Rsamtools 1.18.2, S4Vectors 0.4.0, ShortRead 1.24.0, XVector 0.6.0, systemPipeR 1.0.12
- Loaded via a namespace (and not attached): AnnotationForge 1.8.2, BBmisc 1.9, BatchJobs 1.5, BiocStyle 1.4.1, Category 2.32.0, GO.db 3.0.0, GOSTats 2.32.0, GSEABase 1.28.0, MASS 7.3-37, Matrix 1.1-5, RBGL 1.42.0, RColorBrewer 1.1-2, Rcpp 0.11.4, XML 3.98-1.1, annotate 1.44.0, base64enc 0.1-2, bitops 1.0-6, brew 1.0-6, checkmate 1.5.1, codetools 0.2-11, colorspace 1.2-6, digest 0.6.8, edgeR 3.8.5, fail 1.2, foreach 1.4.2, genefilter 1.48.1, ggplot2 1.0.0, graph 1.44.1, grid 3.1.2, gtable 0.1.2, hwriter 1.3.2, iterators 1.0.7, lattice 0.20-30, latticeExtra 0.6-26, limma 3.22.4, munsell 0.4.2, pheatmap 1.0.2, plyr 1.8.1, proto 0.3-10,

reshape2 1.4.1, rjson 0.2.15, scales 0.2.4, sendmailR 1.2-1, splines 3.1.2, stringr 0.6.2, survival 2.38-1, tools 3.1.2, xtable 1.7-4, zlibbioc 1.12.0

12 Funding

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13 References

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