



Introducing ShortRead

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Plant genomes

- Arabidopsis thaliana is a small flowering plant
- First plant to have its genome sequenced
- Genome size 135 megabase pairs
 (Mbp)





Sequencing companies

















fastq vs fasta

fastq

```
@ unique sequence identifier
raw sequence string
+ optional id
quality encoding per sequence letter
```

• fastq, fq

fasta

> unique sequence identifier
raw sequence string

• fasta, fa, seq



fasta

```
library(ShortRead)
# read fasta
fasample <- readFasta(dirPath = "data/", pattern = "fasta")
# print fasample
print(fasample)

class: ShortRead
length: 500 reads; width: 50 cycles

# methods accessors
methods(class = "ShortRead")
# Write a ShortRead object
writeFasta(fasample, file = "data/sample.fasta")</pre>
```



fastq

```
library(ShortRead)
# read fastq
fqsample <- readFastq(dirPath = "data/", pattern = "fastq")
# print fqsample
fqsample

class: ShortReadQ
length: 500 reads; width: 50 cycles

# methods accessors
methods(class = "ShortReadQ")
# Write a ShortRead object
writeFastq(fqsample, file = "data/sample.fastq.gz")</pre>
```



fastq sample

```
library(ShortRead)

# set the seed to draw the same read sequences every time
set.seed(123)

# Subsample of 500 bases
sampler <- FastqSampler("data/SRR1971253.fastq", 500)

# save the yield of 500 read sequences
sample_small <- yield(sampler)

# Class ShortReadQ
class(sample_small)

# length 500 reads
length(sample_small)</pre>
```





You are ready!





Sequence quality

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Quality scores - Phred table

Quality value	Chance is wrong	Accuracy (%)	
10	1 in 10	90	
20	1 in 100	99	
30	1 in 1 000	99.9	
40	1 in 10 000	99.99	
50	1 in 100 000	99.999	



Encoding - Phred +33

```
# quality encoding
encoding(quality(fqsample))
```

Encoding characters and their scores

```
! " # $ % & ' ( ) * + , - . # encoding
0 1 2 3 4 5 6 7 8 9 10 11 12 13 # score

/ 0 1 2 3 4 5 6 7 8 9 : ; < # encoding
14 15 16 17 18 19 20 21 22 23 24 25 26 27 # score

= > ? @ A B C D E F G H I # encoding
28 29 30 31 32 33 34 35 36 37 38 39 40 # score
```



fastq quality



Exploring quality encoding and scores

```
library(ShortRead)
sread(fqsample)[1]
[1]
       50 GTCCCATTTACCTCTGACTCTTTTGATGCTGCAATTGCTGCTCATATACT
# Quality is represented with ASCII characters
quality(fqsample)[1]
[1]
       50 ?@@DDDDDHDFDHE>AHFEGFIIEBGDBHH<3FEBEEEEGGIGIIGHGHC
## PhredQuality instance
pq <- PhredQuality(quality(fqsample))</pre>
# transform encoding into scores
qs <- as(pq, "IntegerList")</pre>
qs # print scores
30 31 31 35 35 35 35 35 39 35 37 35 39 36 29 32 39 37 36 38
37 40 40 36 33 38 35 33 39 39 27 18 37 36 33 36 36 36 36 38
38 40 38 40 40 38 39 38 39 34
```



Quality assessment

```
library(ShortRead)
# Quality assessment
qaSummary <- qa(fqsample, lane = 1)  # optional lane
# class: ShortReadQQA(10)
# Names accessible with the quality assessment summary
names (qaSummary)
    "readCounts" "baseCalls"
                                                  "readQualityScore"
 [4] "baseQuality" "alignQuality"
                                                  "frequentSequences"
                                                  "perTile"
 [7] "sequenceDistribution" "perCycle"
[10] "adapterContamination"
# QA elements are accessed with qa[["name"]]
# Get a HTML report
browseURL(report(qaSummary))
```



Alphabet by cycle

```
library(ShortRead)
# sequences alphabet
alphabet(sread(fullSample))
# [1] A,C,G,T,M,R,W,S,Y,K,V,H,D,B,N,-,+,.
abc <- alphabetByCycle(sread(fullSample))</pre>
# each observation is a letter and each variable is a cycle
# first, select the four first rows nucleotides A, C, G, T
# then, transpose
nucByCycle <- t(abc[1:4,])</pre>
nucByCycle <- nucByCycle %>%
  as.tibble() %>% # convert to tibble
 mutate(cycle = 1:50) # add cycle numbers
nucByCycle
```

```
A C G T cycle
16839 16335 16740 10878 1
13056 13327 12064 22389 2
13666 15617 13198 18355 3
14723 15439 14239 16435 4
```





Are you excited?





Match and filter

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Duplicate sequences

- Biological sequence duplicates occur in nature
- Amplification from the steps in library preparation (PCR)
- Sequencing the sample more than once

Remove duplicates or at least mark them

Whole genome sequencing or exome sequencing

Mark duplicates using a threshold

RNA-seq and ChIP-seq



srduplicated

```
library(ShortRead)
# Counting duplicates TRUE is the number of duplicates
table(srduplicated(dfqsample))
FALSE
      TRUE
500
      500
# Cleaning reads from duplicates x[fun(x)]
cleanReads <- mydReads[srduplicated(mydReads) == FALSE]</pre>
# Counting duplicates
table(srduplicated(cleanReads))
FALSE
500
```



Creating your own filters

srFilter to filter based on a condition x[fun(x)]

Filter example



nFilter



idFilter and polynFilter

```
# filter to remove poly-A regions
myFilterPolyA <- polynFilter(threshold = 10, nuc = c("A"))
# will return the sequences that have a maximum number of 10 consecutive A's
# use the filter for subsetting
filtered[myFilterPolyA(filtered)]</pre>
```





Let's practice using filters!





Multiple and parallel sequence quality assessment

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Rqc

library(Rqc)

- Uses Bioconductor packages that you have already used:
 - Biostrings, IRanges, methods, S4vectors
- New packages to discover in the following Bioconductor courses:
 - Rsamtools, GenomicAlignments, GenomicFiles, BiocParallel
- CRAN packages:
 - Knitr, dplyr, markdown, ggplot2, digest, shiny and Rccp



rqcQA

```
library(Rqc)

files <- # get the full path of the files you want to assess

qaRqc <- rqcQA(files)

# exploring qaRqc
class(qaRqc) # "list"
names(qaRqc) # name of the input files

# for each file
qaRqc[1]
# the class of the results is RqcResultSet</pre>
```



rqcQA arguments

```
library(Rqc)
# get the path of the files you want to assess
files <- "data/seq1.fq" "data/seq2.fq" "data/seq3.fq" "data/se4.fq"

qaRqc <- rqcQA(files, workers = 4))
# sample of sequences
set.seed(1111)

qaRqc_sample <- rqcQA(files, workers = 4, sample = TRUE, n = 500))
# paired-end files
pfiles <- "data/seq_11.fq" "data/seq1_2.fq" "data/seq2_1.fq" "data/seq2_2.fq"

qaRqc_paired <- rqcQA(pfiles, workers = 4, pair = c(1, 1, 2, 2)))</pre>
```



rqcReport and rqcResultSet

```
# create a report
reportFile <- rqcReport(qaRqc, templateFile = "myReport.Rmd")
browseURL(reportFile)

#The class of qaRqc is rqcResultSet
methods(class = "RqcResultSet")</pre>
```



perFileInformation

```
qaRqc <- rqcQA(files, workers = 4))
perFileInformation(qaRqc)</pre>
```

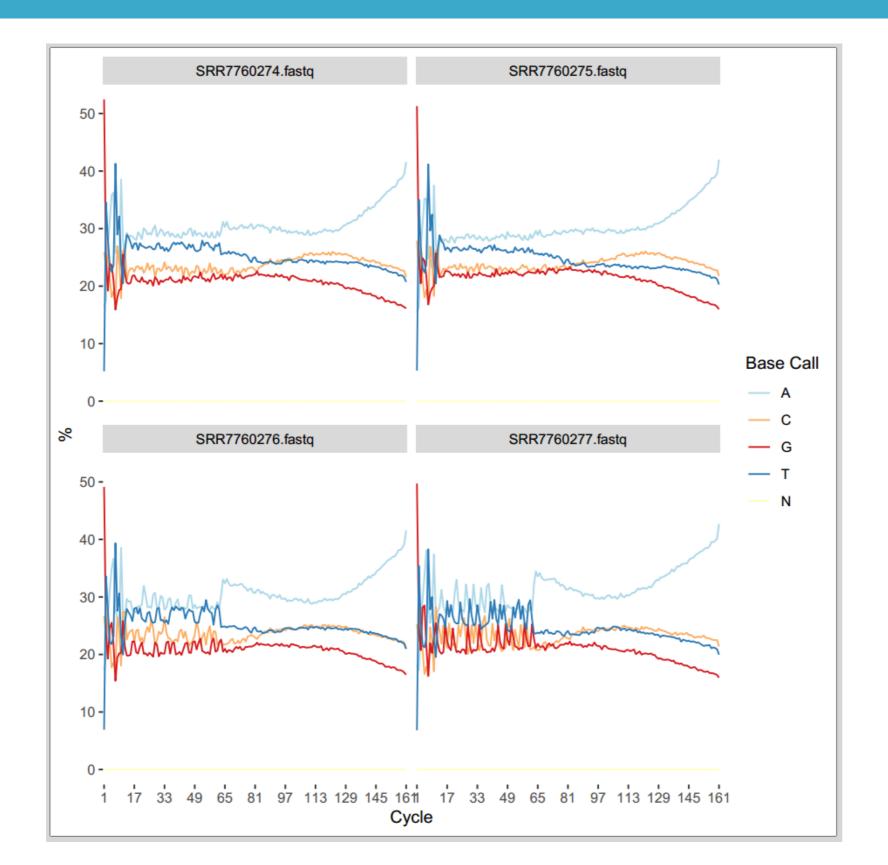
filename	pair	format	group	reads	total.reads	path
SRR7760274.fastq	1	FASTQ	None	1e+06	2404795	./data
SRR7760275.fastq	2	FASTQ	None	1e+06	1508139	./data
SRR7760276.fastq	3	FASTQ	None	1e+06	1950463	./data
SRR7760277.fastq	4	FASTQ	None	1e+06	2629588	./data



Plot functions

rqc Plot functions	rqc Plot functions		
rqcCycleAverageQualityPcaPlot()	rqcGroupCycleAverageQualityPlot()		
rqcCycleAverageQualityPlot()	rqcReadQualityBoxPlot()		
rqcCycleBaseCallsLinePlot()	rqcReadQualityPlot()		
rqcCycleBaseCallsPlot()	rqcReadWidthPlot()		
rqcCycleGCPlot()	rqcReadFrequencyPlot()		
rqcCycleQualityBoxPlot()	rqcCycleQualityPlot()		













You are ready!





Congratulations!

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You learned...

- Install packages from Bioconductor by using the BiocInstaller package.
- Techniques for reading, manipulating and filtering raw genomic data using BioStrings, GenomicRanges and ShortRead.
- To work with BSgenome and TxDb built-in datasets. Then used these to identify patterns by using matching functions.
- Check the quality of sequence files using ShortRead and Rqc.



You explored







Keep learning!