Introduction to NGS Analysis

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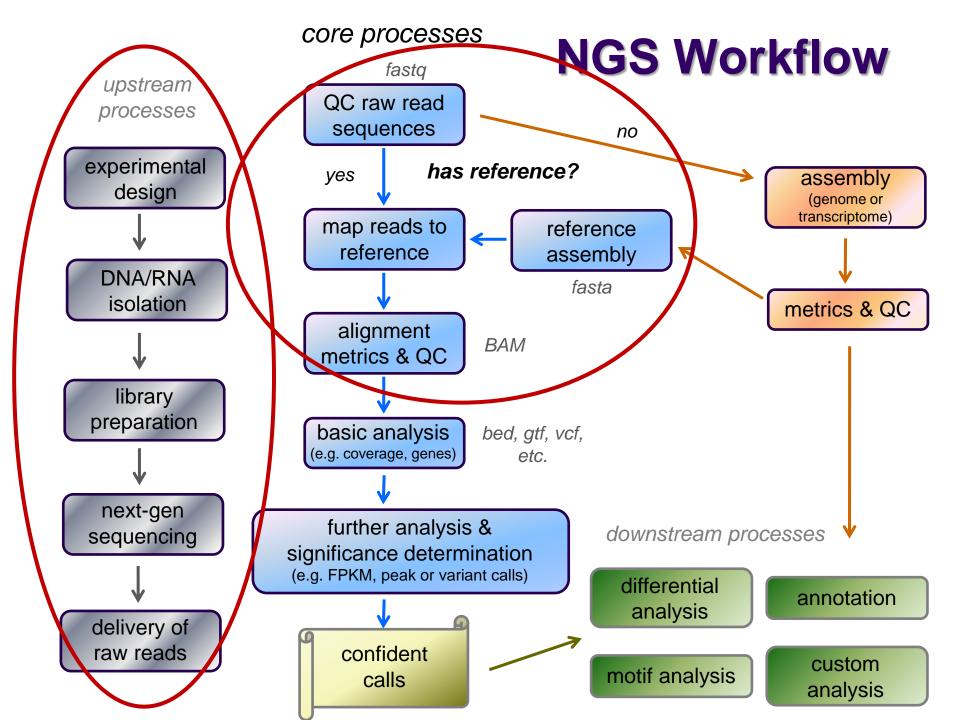
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Outline

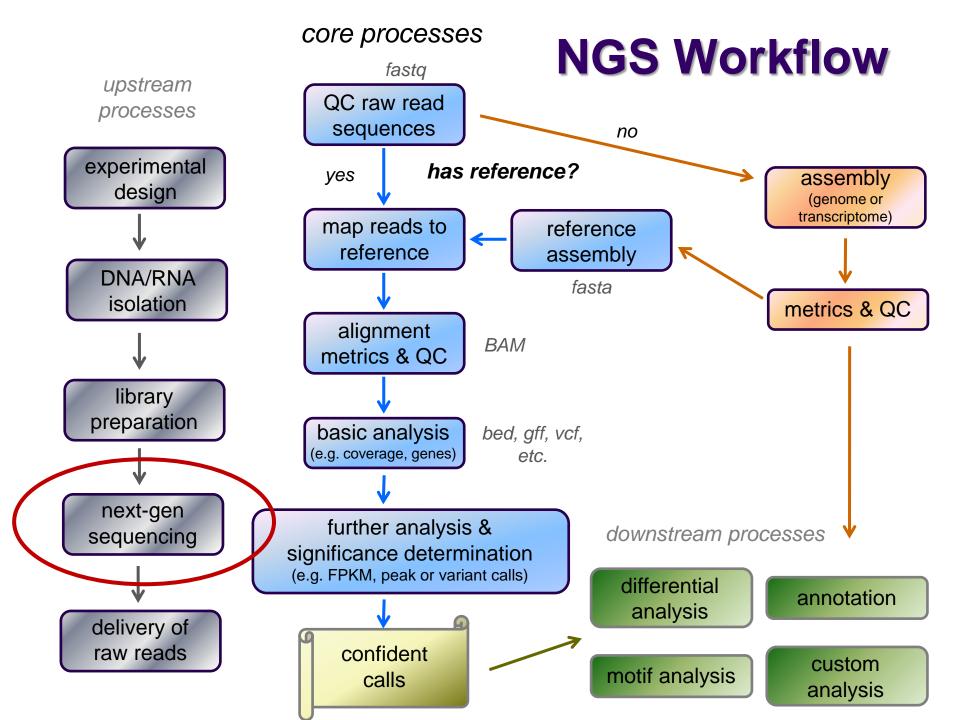


- 1. Overview of sequencing technologies
- NGS concepts and terminology
- 3. The FASTQ format and raw data QC & preparation
- 4. Alignment to a reference

Part 1: Overview of Sequencing Technologies

- High-throughput ("next gen") sequencing
- Illumina short-read sequencing
- Long read sequencing





"Next Generation" sequencing



Massively parallel

 simultaneously sequence "library" of millions of different DNA fragments

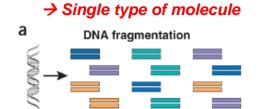
PCR colony clusters generated

- individual template DNA fragments titrated onto a flowcell to achieve inter-fragment separation
- PCR "bridge amplification" creates
 clusters of identical molecules

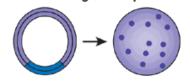
Sequencing by synthesis

- fluorescently-labeled dNTs added
- incorporation generates signal
- flowcell image captured after each cycle
- images computationally converted to base calls (including a quality score)
- results in 30 300 base "reads"
 - much shorter than Sanger sequencing

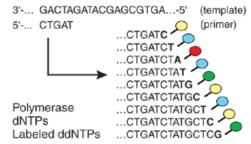
Sanger



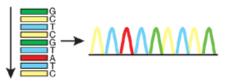
In vivo cloning and amplification



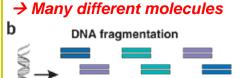
Cycle sequencing



Electrophorsesis (1 read/capillary)



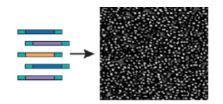
NGS



In vitro adaptor ligation (sequencing adapter)



polony = PCR colony cluster
Generation of polony array



Cyclic array sequencing (>10⁶ reads/array)







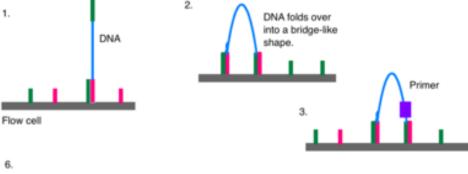
What is base 1? What is base 2? What is base 3?

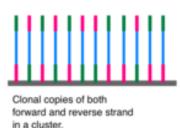
Shendure et al, Nature Biotechnology. 2008.

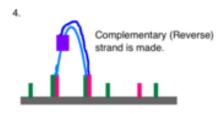
Illumina sequencing

- 1. Library preparation
- 2. Cluster generation via bridge amplification
- 3. Sequencing by synthesis
- 4. Image capture
- Convert to base calls

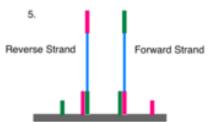
Short Illumina video (https://tinyurl.com/hvnmwjb)





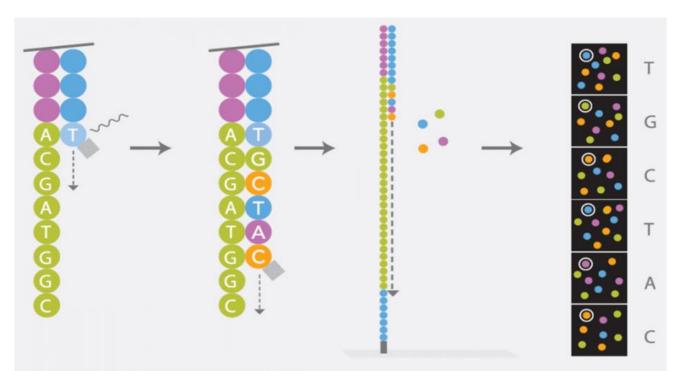


- Note
 - 2 PCR amplifications performed
 - during library preparation
 - 2. during cluster generation
 - amplification always introduces bias!



Illumina sequencing

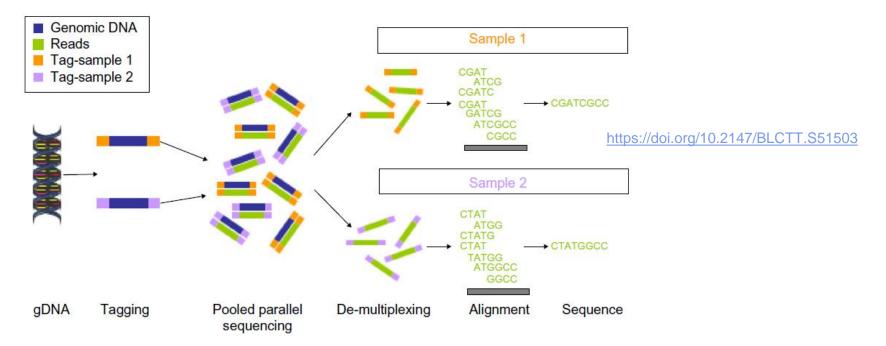
- Library preparation
- 2. Cluster generation via bridge amplification
- 3. Sequencing by synthesis
- 4. Image capture
- 5. Convert to base calls



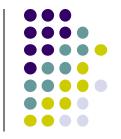
Multiplexing



- Illumina sequencers have one or more flowcell "lanes", each of which can generate *millions* of reads
 - ~20M reads/lane for MiSeq, ~10G reads/lane for NovaSeq
- When less than a full flowcell lane is needed, multiple samples with different barcodes (a.k.a. indexes) can be run on the same lane
 - 6-8 bp *library barcode* attached to DNA library fragments
 - data from sequencer must be demultiplexed to determine which reads belong to which library



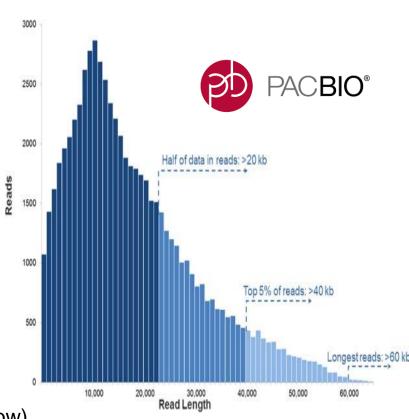
Long read sequencing



- Short read technology limitations
 - 30 300 base reads (150 typical)
 - PCR amplification bias
 - short reads are difficult to assemble
 - e.g., too short to span a long repeat region
- Newer "single molecule" sequencing
 - sequences single molecules, not clusters
 - allows for *much* longer reads (multi-Kb!)
 - no signal wash-out due to lack of synchronization among cluster molecules

but:

- weaker signal leads to high error rate
 - 10% vs <1% for Illumina (but improving now)
 - and fewer reads may be generated
 - ~ 1 million vs 10s to 100s of millions w/short reads

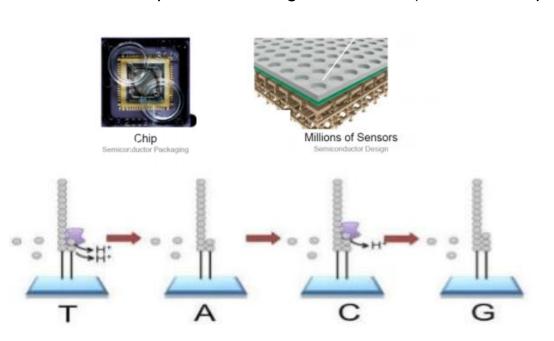


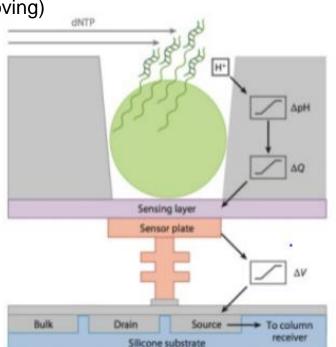
Long read sequencing



- Oxford Nanopore ION technology systems (e.g. MinION)
 - https://nanoporetech.com/
 - DNA "spaghetti's" through tiny protein pores
 - Addition of different bases produces different pH changes
 - measured as different changes in electrical conductivity
 - MinION is hand-held, starter kit costs ~\$1,000 including reagents!

inexpensive, but high error rates (~10%, but improving)





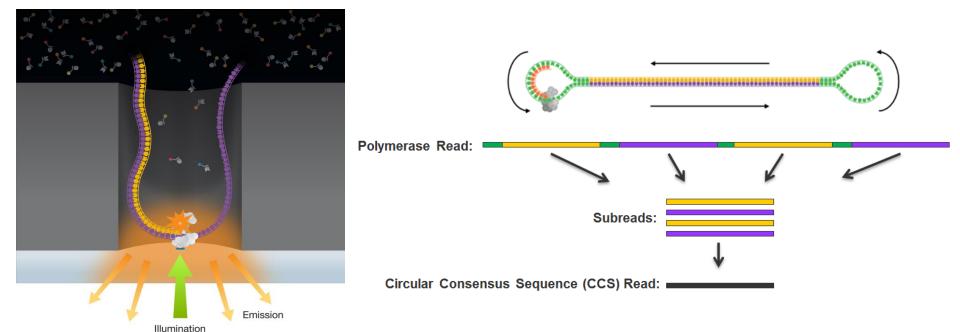
Long read sequencing



PacBio SMRT system



- http://www.pacb.com/smrt-science/smrt-sequencing/
- Sequencing by synthessis in Zero-Mode Waveguide (ZMW) wells
- DNA is circularized then repeatedly sequenced to achieve "consensus"
 - reduces error rate (~1-2%), but equipment quite expensive
- Now the preferred technology for assembly of large eukaryotic genomes
 - especially polyploid species (e.g. many plants)





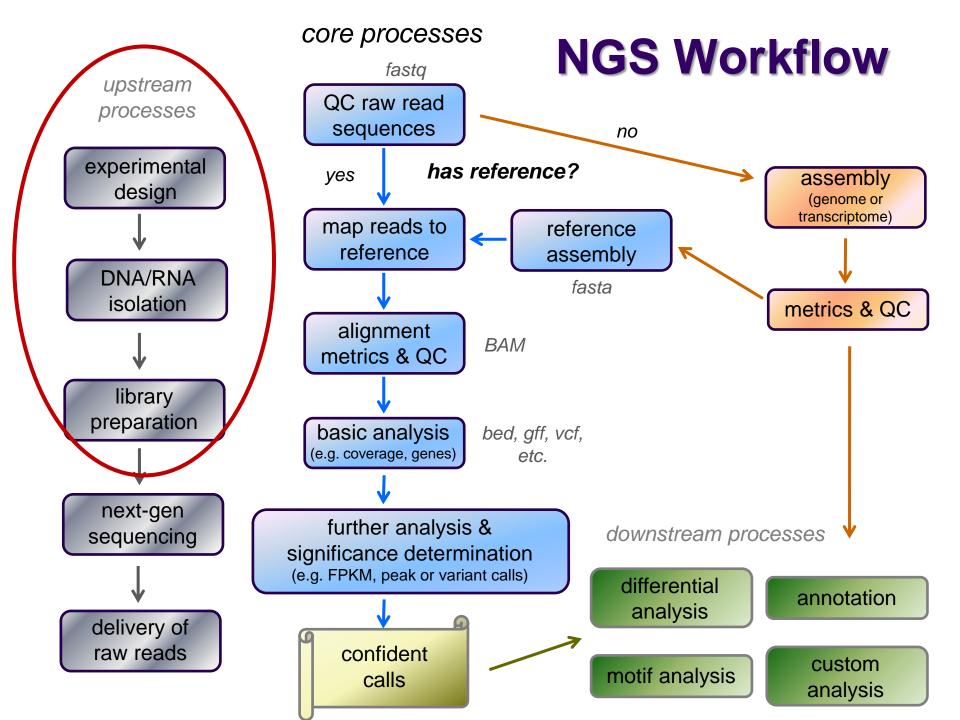


- "Next Generation" sequencing operates on a library of millions of different DNA fragments vs a single purified molecular species
- Illumina platforms are dominant for short reads (30-300 base)
 - Use sequencing by synthesis on clusters of identical molecules (clones); de-synchronization of signal limits read length
 - Since instruments are so high throughput, multiple barcoded sample libraries are pooled for sequencing, then demultiplexed
- Single molecule technologies produce long reads (multi-Kilobase)
 - Very low signal from single molecule readout presents accuracy challenges
 - Oxford Nanopore has low-cost models, but with high error rates
 - PacBio has lower error rates but cost of both equipment & sequencing are high

Part 2: NGS Concepts & Terminology

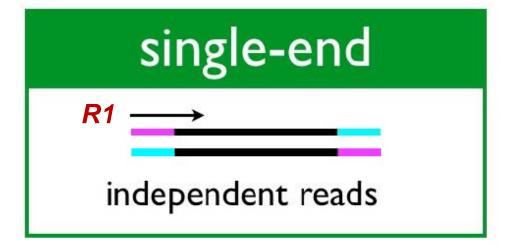
- Sequencing terminology
- Experiment types & library complexity
- Sequence duplication issues

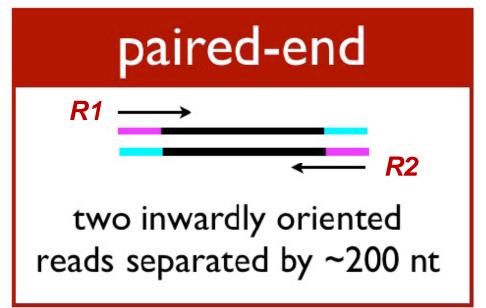




Read types

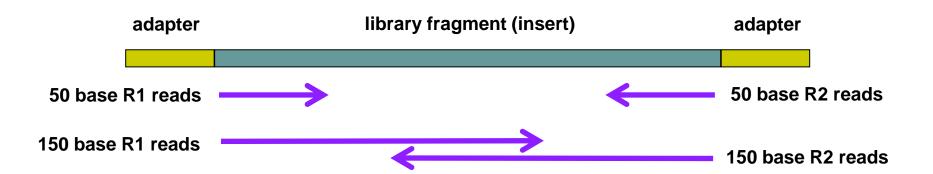






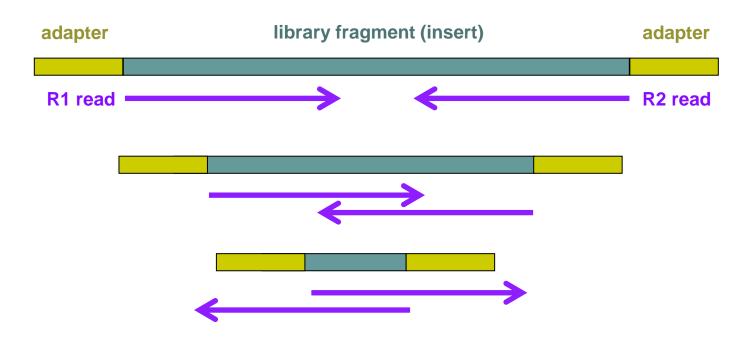
Reads and Fragments

- With paired-end sequencing, keep in mind the distinction between
 - the library fragment from your library that was sequenced
 - also called inserts
 - the sequence reads (R1s & R2s) you receive
- An R1 and its associated R2 form a read pair
 - a readout of part (or all) of the fragment molecule

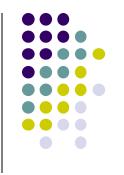


Library fragment distribution

- What is fixed size in your sequencing library:
 - the adapter region (including all barcodes)
 - the read length (e.g. 50, 100, 150)
- But the insert fragments are of variable length
 - due to random shearing during library preparation
 - bioanalyzer provides an estimate of the library's fragment distribution







Library complexity (diversity)
is a measure of the number of
distinct molecular species in the library.

Many different molecules → high complexity

Few different molecules → low complexity

The number of different molecules in a library depends on *enrichment* performed during *library construction*.

Popular Experiment Types

- Whole Genome sequencing (WGS)
 - applications: genome assembly
 - library: all genomic DNA (no enrichment)
 - complexity: high (fragments must cover the entire genome)
- Exome sequencing (WXS)
 - applications: polymorphism/SNP detection; genotyping
 - library: DNA from eukaryotic exons (uses special kits)
 - complexity: high/med (only ~5% of eukaryotic genome is in exons)
- RNA-seq
 - applications: differential gene expression between 2 or more conditions
 - library: extracted RNA converted to cDNA
 - complexity: med/high (only a subset of genes are expressed in any given tissue)
- Amplicon panels (targeted sequencing)
 - applications: genetic screening panels; metagenomics; mutagenesis
 - library: DNA from a set of PCR-amplified regions using custom primers
 - complexity: very low (only a few hundred-to-thousand different library molecules)

Туре	Library construction	Applications	Complexity
Whole genome (WGS)	extract genomic DNA & fragment	 Genome assembly Variant detection, genotyping	high
Bisulfite sequencing	 bisulfite treatment converts C → U but not 5meC 	Methylation profiling (CpG)	high
RAD-seq, ddRAD	 restriction-enzyme digest DNA & fragment 	 Variant detection (SNPs) Population genetics, QTL mapping	high
Exome (wxs)	 capture DNA from exons only (manufacturer kits) 	Variant detection, genotyping	high- medium
ATAC-seq	 high-activity transposase cuts DNA ligates adapters 	 Profile nucleosome-free regions ("open chromatin") 	medium- high
RNA-seq, Tag-seq	extract RNA & fragmentconvert to cDNA	 Differential gene or isoform expression Transcriptome assembly (RNA-seq only) 	medium, medium-low for Tag-seq
Transposon seq (Tn-seq)	 create library of transposon- mutated genomic DNA amplify mutants via Tn-PCR 	Characterize genotype/phenotype relationships w/high sensitivity	medium
ChIP-seq	 cross-link proteins to DNA pull-down proteins of interest w/ specific antibody, reverse cross-links 	 Genome-wide binding profiles of transcription factors, epigenetic marks & other proteins 	medium (but variable)
GRO-seq	• isolate actively-transcribed RNA	Characterize transcriptional dynamics	medium-low
RIP-seq	• like ChIP-seq, but with RNA	Characterize protein-bound RNAs	low-medium
miRNA-seq	• isolate 15-25bp RNA band	miRNA profiling	low
Amplicons	• amplify 1-1000+ genes/regions	• genotyping, metagenomics, mutagenesis	low

Library complexity is primarily a function of experiment type



Less enrichment for specific sequences

higher complexity

lower

complexity

genomic

bisulfite-seq

exon capture

RNA-seq

ChIP-seq

amplicons

More enrichment for specific sequences Higher diversity of library molecules

Lower sequence duplication expected More sequencing depth required

... as well as...

- genome size & sequencing depth
- library construction skill & luck!

Lower diversity of library molecules

Higher sequence duplication expected Less sequencing depth required

Sequence Duplication



- The set of sequences you receive can contain exact duplicates
- Duplication can arise from:
 - sequencing of species enriched in your library (biological good!)
 - each read comes from a different DNA molecule
 - sequencing of artifacts (technical bad!)
 - differentially amplified PCR species (PCR duplicates)
 - recall that 2 PCR amplifications are performed w/Illumina sequencing
 - cannot tell which using standard sequencing methods!
- Different experiment types have different expected duplication

Expected sequence duplication is primarily a function of experiment type



Less enrichment for specific sequences

higher complexity

lower

complexity

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More enrichment for specific sequences Higher diversity of library molecules

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Lower diversity of library molecules

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Less sequencing depth required

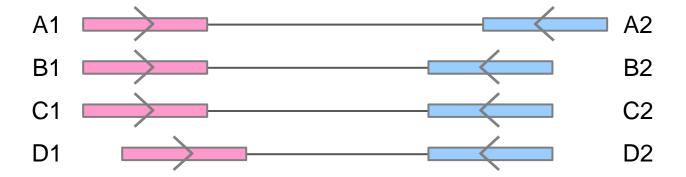
Single end vs Paired end

- Single End (SE) reads are less expensive
- Paired End (PE) reads:
 - provide more bases around a locus
 - e.g. for analysis of polymorphisms
 - actual fragment sizes can be easily determined
 - helps distinguish the true complexity of a library
 - by clarifying which fragments are duplicates (vs sequence duplicates)
 - but PE reads are more expensive and larger
 - more storage space and processing time required
- General guidelines
 - use PE for high location accuracy and/or base-level sensitivity
 - use SE for lower-complexity experiment types



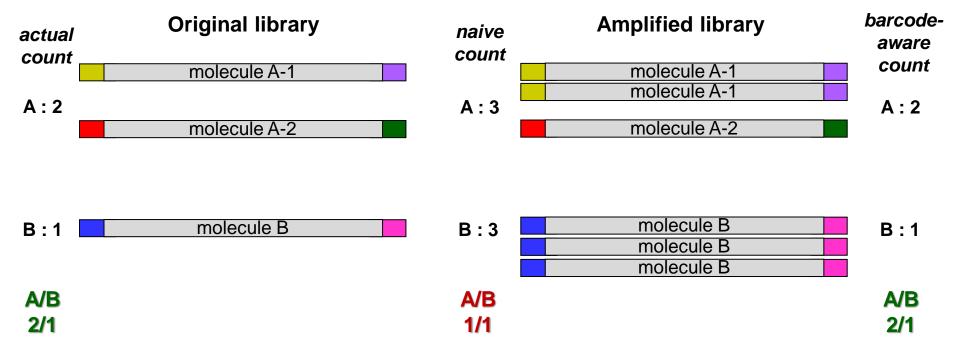
Read vs Fragment duplication

- Consider the 4 "aligned" fragments below
 - 4 R1 reads (pink), 4 R2 reads (blue)
- Duplication when only 1 end considered
 - A1, B1, C1 have identical start locations, D1 different
 - 2 unique + 2 duplicates = 50% duplication rate
 - B2, C2, D2 have identical start locations, A2 different
 - 2 unique + 2 duplicates = 50% duplication rate
- Duplication when both ends considered
 - only fragments B and C are duplicates (same external locations)
 - 3 unique + 1 duplicate = 25% duplication rate



Molecular Barcoding

- Resolves ambiguity between biological and technical (PCR amplification) duplicates
 - adds secondary, internal barcodes to pre-PCR molecules
 - a.k.a <u>UMIs</u> (<u>U</u>nique <u>M</u>olecular <u>I</u>ndexes)
 - combination of barcodes + insert sequence provides accurate quantification
 - but requires specialized pre- and post-processing



Single Cell sequencing



- Standard sequencing library starts with *millions* of cells
 - will be in different states unless synchronized
 - a heterogeneous "ensemble" with (possibly) high cell-to-cell variability
- Single cell sequencing technologies aim to capture this variability
 - e.g: cells in different tissue layers/regions or different areas of a tumor
 - essentially a very sophisticated library preparation technique
- Typical protocol (RNA-seq)
 - isolate a few thousand cells (varying methods)
 - 2. the single-cell platform partitions each cell into an emulsion droplet
 - e.g. 10x Genomics (https://www.10xgenomics.com/solutions/single-cell/)
 - a different barcode is added to the RNA in each cell
 - 4. resulting library submitted for standard Illumina short-read sequencing
 - 5. custom downstream analysis links results to their cell (barcode) of origin

Part 2 summary

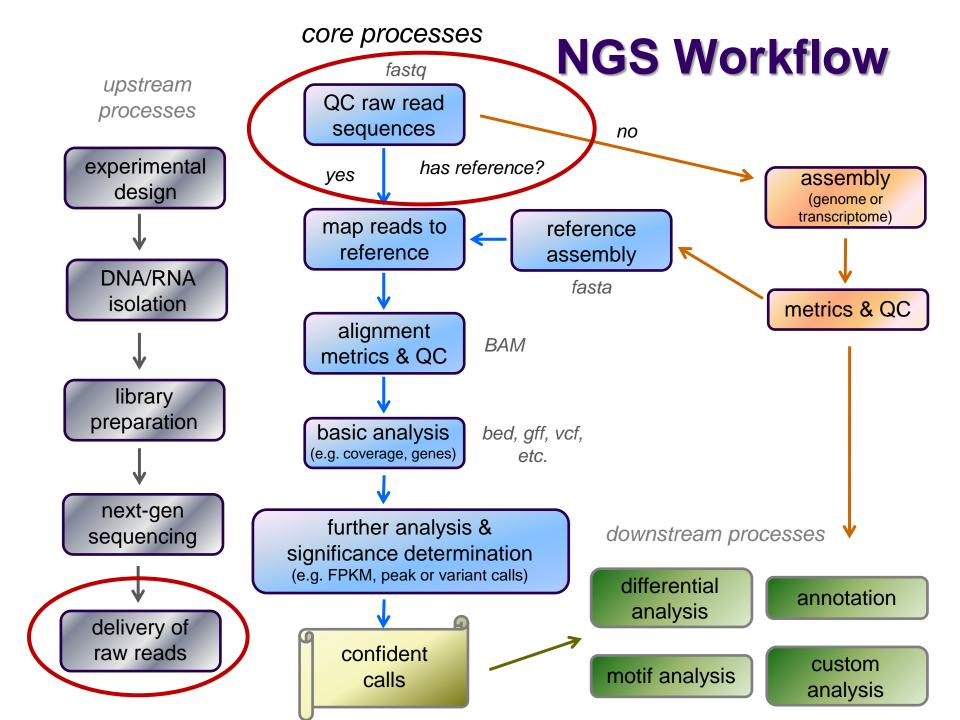


- Read terminology:
 - Single-end (SE) reads sequence from one end only (R1)
 - Paired-end (PE) reads sequence inwardly from both ends (R1/R2 pair)
 - Adapters (including primers & library barcodes) are fixed size & added to both ends
 - DNA library fragments have a distribution of insert sizes between adapters (~200 bases)
- Library complexity describes the number of different molecular species
 - Many distinct molecules → high complexity; Few → low complexity
 - Primarily a function of the experiment type's enrichment profile
 - Less enrichment → higher complexity; More enrichment → lower complexity
 - Popular experiment types have different expected library complexity
 - Whole genome sequencing → high complexity; Amplicons → low complexity
- Expected sequence duplication is also a function of library complexity
 - Also due to desired enrichment (biological/good) or PCR duplicates (techincal/bad)
 - Only addition of Unique Molecular Indexes (UMIs) can properly distinguish
- Single cell sequencing is designed to capture cell-to-cell variability
 - Essentially a sophisticated library prep technique followed by short-read sequencing

Part 3: The FASTQ format, Data QC & preparation

- FASTA and FASTQ formats
- QC of raw sequences with FastQC tool
- Dealing with adapters





FASTQ format

- Text format for storing sequence and quality data
 - http://en.wikipedia.org/wiki/FASTQ_format
- 4 lines per sequence:
 - @read name (plus extra information after a space)
 - R1 and R2 reads have the same read name
 - called base sequence (ACGTN)
 always 5' to 3'; usually excludes 5' adapter
 - 3. +optional information
 - 4. base quality scores encoded as text characters
- FASTQ representation of a single, 50 base R2 sequence

```
@HWI-ST1097:97:D0WW0ACXX:4:1101:2007:2085 2:N:0:ACTTGA
ATTCTCCAAGATTTGGCAAATGATGAGTACAATTATATGCCCCAATTTACA
+
?@@?DD;?;FF?HHBB+:ABECGHDHDCF4?FGIGACFDFH;FHEIIIB9?
```





- Base quality probabilities expressed as Phred scores
 - Phred scores are log scaled, higher = better
 - Quality $20 = 1.0e^{-2} = 1/100 \text{ errors}$, $30 = 1.0e^{-3} = 1/1000 \text{ errors}$ Probability of Error = $10^{-0/10}$
- Integer Phred score converted to Ascii text (add 33)

https://www.asciitable.com/

```
Quality character !"#$%&'()*+,-./0123456789:;<=>?@ABCDEFGHIJ

ASCII Value 33 53 63 73

Base Quality (Q) 0 10 20 30 40
```

?@@?DD;?;FF?HHBB+:ABECGHDHDCF4?FGIGACFDFH;FHEIIIB9?

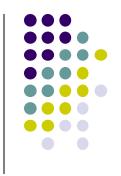
Raw sequence quality control



- Critical step! Garbage in = Garbage out
 - general sequence quality
 - base quality distributions
 - initial sequence duplication rate
 - trim 3' bases with poor quality?
 - important for de novo assembly
 - trim 3' adapter sequences?
 - important for RNA-seq
 - other contaminents?
 - biological rRNA in RNA-seq
 - technical samples sequenced w/other barcodes



FastQC

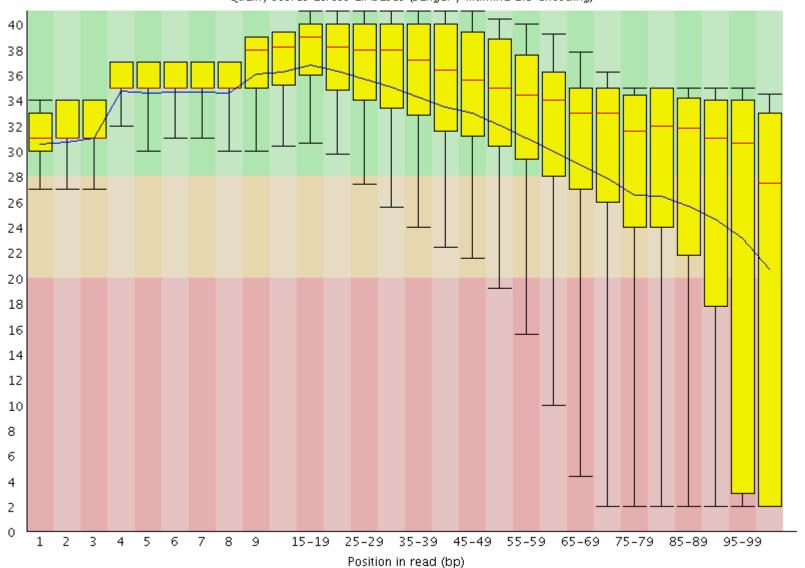


- Quality Assurance tool for FASTQ sequences http://www.bioinformatics.babraham.ac.uk
- Most useful reports:
 - Should I trim low quality bases?
 - Per-base sequence quality Report
 - 2. How complex is my sequence library?
 - Sequence duplication levels Report
 - 3. Do I need to remove adapter sequences?
 - Overrepresented sequences Report

1. FastQC Per-base sequence quality report



Quality scores across all bases (Sanger / Illumina 1.9 encoding)



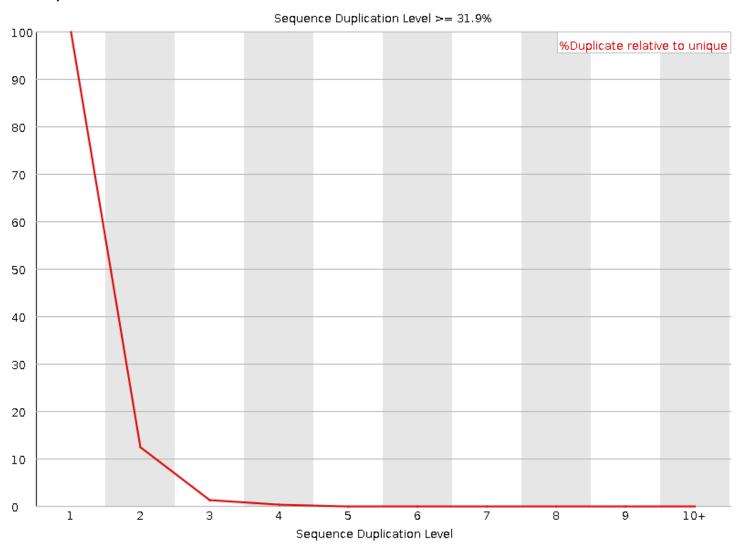
2. FastQC Sequence duplication report Yeast ChIP-seq



For every 100 unique sequences there are:

- ~12 sequences w/2 copies
- ~1-2 with 3 copies

Ok – Some duplication expected due to IP enrichment



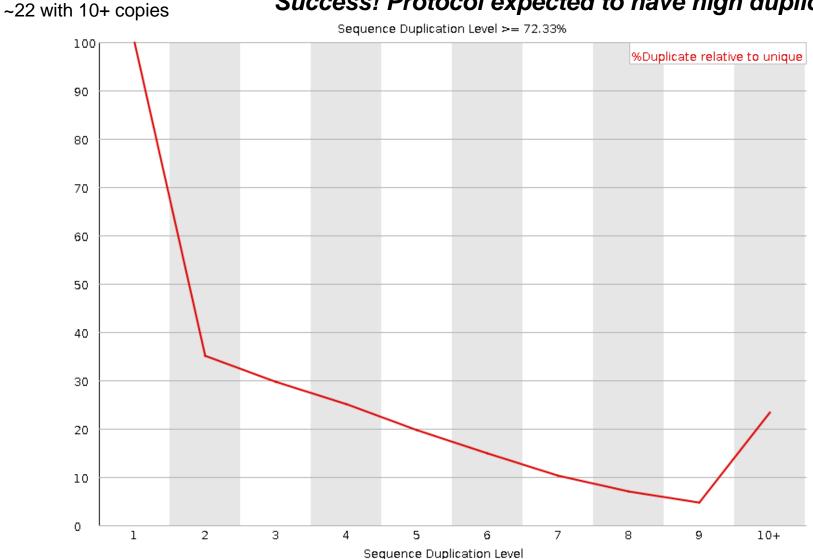
2. Sequence duplication report **Yeast ChIP-exo**



For every 100 unique sequences there are:

~35 sequences w/2 copies

Success! Protocol expected to have high duplication



Expected sequence duplication is primarily a function of experiment type



Less enrichment for specific sequences

higher complexity

genomic

bisulfite-seq

exon capture

RNA-seq

ChIP-seq

ChIP-exo

amplicons

lower complexity

Higher diversity of library molecules **Lower sequence duplication expected**More sequencing depth required

... as well as...

- genome size & sequencing depth
- library construction skill & luck!

Lower diversity of library molecules

Higher sequence duplication expected

Less sequencing depth required

More enrichment for specific sequences

3. FastQC Overrepresented sequences report



- FastQC knows Illumina adapter sequences
- Here ~9-10% of sequences contain adapters
 - calls for adapter removal or trimming

Sequence	Count	Percentage	Possible Source
AGATCGGAAGAGCACACGTCTGAACTCCAGTCACCTCAGAATCTCGTATG	60030	5.01369306977828	TruSeq Adapter, Index 1 (97% over 37bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACCTCAGAATCTCGTATGC	42955	3.5875926338884896	TruSeq Adapter, Index 1 (97% over 37bp)
CACACGTCTGAACTCCAGTCACCTCAGAATCTCGTATGCCGTCTTCTGCT	3574	0.29849973398946483	RNA PCR Primer, Index 40 (100% over 41bp)
CAGATCGGAAGAGCACACGTCTGAACTCCAGTCACCTCAGAATCTCGTAT	2519	0.2103863542024236	TruSeq Adapter, Index 1 (97% over 37bp)
GAGATCGGAAGACCACGTCTGAACTCCAGTCACCTCAGAATCTCGTAT	1251	0.10448325887543942	TruSeq Adapter, Index 1 (97% over 37bp)

3. Overrepresented sequences



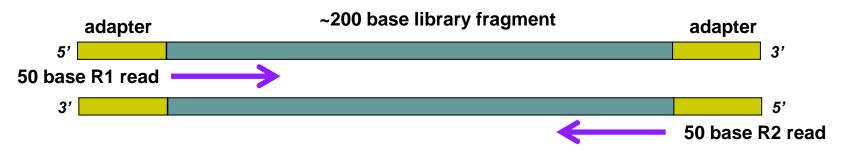
- Here nearly 1/3 of sequences some type of non-adapter contamination!
 - BLAST the sequence to identify it

Sequence	Count	Percentage	Possible Source
GAAGGTCACGGCGAGACGAGCCGTTTATCATTACGATAGGTGTCAAGTGG	5632816	32.03026785752871	No Hit
${\tt TATTCTGGTGTCCTAGGCGTAGAGGAACAACACCAATCCATCC$	494014	2.8091456822607364	No Hit
${\tt TCAAACGAGGAAAGGCTTACGGTGGATACCTAGGCACCCAGAGACGAGGA}$	446641	2.539765344040083	No Hit
${\tt TAAAACGACTCTCGGCAACGGATATCTCGGCTCTCGCATCGATGAAGAAC}$	179252	1.0192929387357474	No Hit
${\tt GAAGGTCACGGCGAGACGAGCCGTTTATCATTACGATAGGGGTCAAGTGG}$	171681	0.9762414422996221	No Hit
${\tt AACGACTCTCGGCAACGGATATCTCGGCTCTCGCATCGATGAAGAACGTA}$	143415	0.8155105483274229	No Hit
${\tt AGAACATGAAACCGTAAGCTCCCAAGCAGTGGGAGGAGCCCTGGGCTCTG}$	111584	0.6345077504066322	No Hit
${\tt AAAACGACTCTCGGCAACGGATATCTCGGCTCTCGCATCGATGAAGAACG}$	111255	0.6326369351474214	No Hit
${\tt ATTACGATAGGTGTCAAGTGGAAGTGCAGTGATGTATGCAGCTGAGGCAT}$	73682	0.41898300890326096	No Hit
GAAGGTCACGGCGAGACGAGCCGTTTATCATTACGATAGGTGTCAAGGGG	71661	0.4074908580252516	No Hit
GGATGCGATCATACCAGCACTAATGCACCGGATCCCATCAGAACTCCGCA	69548	0.3954755612388914	No Hit
${\tt ATATTCTGGTGTCCTAGGCGTAGAGGAACAACACCAATCCATCC$	54017	0.30716057099328803	No Hit

3' Adapter contamination



A. reads short compared to fragment size (no contamination)



B. Reads long compared to library fragment (3' adapter contamination)



The presence of the 3' adapter sequence in the read can cause problems during alignment, because it does not match the genome.

Dealing with 3' adapters



- Three main options:
 - 1. *Hard trim* all sequences by specific amount
 - e.g. trim 100 base reads to 50 bases
 - Pro: fast & easy to perform; trims low-quality 3' bases
 - Con: removes information (bases) you might want
 - 2. **Remove adapters** specifically
 - e.g. using specific tools (always needed for RNA-seq alignment)
 - Pro: removes adapter contamination without losing sequenced bases
 - Con: requires knowledge of insert fragment structure & adapters
 - Perform a *local alignment* (vs *global*)
 - e.g. bowtie2 --local or bwa mem
 - Pro: mitigates adapter contamination while retaining full query sequence
 - Con: limited aligner support

FASTQ trimming and adapter removal



Tools:

- cutadapt https://code.google.com/p/cutadapt/
- trimmomatic http://www.usadellab.org/cms/?page=trimmomatic
- FASTX-Toolkit http://hannonlab.cshl.edu/fastx_toolkit/

Features:

- hard-trim specific number of bases
- trimming of low quality bases
- specific trimming of adapters
- support for trimming paired end read sets (except FASTX-Toolkit)
- cutadapt has protocol for separating reads based on internal barcode

Local vs. Global alignment



- Global alignment
 - requires query sequence to map fully (end-to-end) to reference
- Local alignment
 - allows a subset of the query sequence to map to reference
 - "untemplated" 5' and 3' sequences will be "soft clipped" (ignored)

global (end-to-end) alignment of query

local (subsequence) alignment of query

CACAAGTACAATTATACAC

CTAGCTTATCGCCCTGAAGGACT

TACATACACAAGTACAATTATACACAGACATTAGTTCTTATCGCCCTGAAAATTCTCC

reference sequence

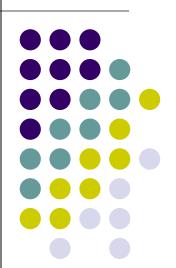
Part 3 summary

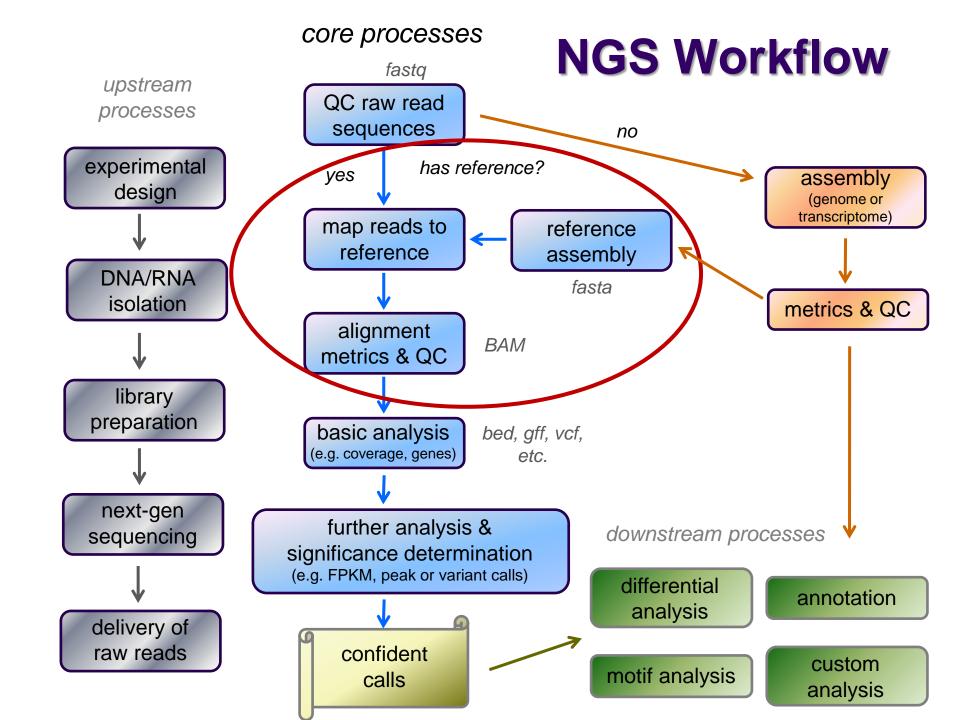


- Sequenced reads are delivered as FASTQ files
 - R1 read file(s) and R2 file(s) if paired end, with 4 lines per read:
 - read name; called sequence; optional info; ASCII-encoded quality scores
- The FastQC tool generates quality reports for each FASTQ file
 - 1. Per-base sequence quality → trim low quality bases?
 - Sequence duplication level → sequence complexity estimate
 - 3. Overrepresented sequences → adapter trimming needed?
- 3' adapter contamination can prevent reads from aligning
 - Dealing with adapter contamination:
 - 1. Hard trim a specific number of bases (simple, but information is lost)
 - Remove adapters specifically (required for RNA-seq analysis)
 - Tools: cutadapt, trimmomatic
 - Perform local alignment (versus standard global alignment)
 - Not suitable for RNA-seq reads since splice-junction information is missed

Part 4: Alignment to a reference assembly

- Alignment overview & concepts
- Preparing a reference genome
- Alignment workflow steps





Short Read Aligners



- Short read mappers determine placement of query sequences (your reads) against a known reference
 - BLAST
 - one query sequence (or a few)
 - many matches for each
 - short read aligners
 - many *millions* of query sequences
 - want only one "best" mapping (or a few) for each
- Many aligners available! Two of the most popular
 - bwa (Burrows Wheeler Aligner) by Heng Li <u>http://bio-bwa.sourceforge.net/</u>
 - bowtie2 part of the Johns Hopkins Tuxedo suite of tools http://bowtie-bio.sourceforge.net/bowtie2/manual.shtml
 - Given similar input parameters, they produce similar alignments
 - and both run relatively quickly

Mapping vs Alignment



- Mapping determines one or more positions (a.k.a. seeds or hits)
 where a read shares a short sequence with the reference
- Alignment starts with the seed and determines how read bases are best matched, base-by-base, around the seed
- Mapping quality and alignment scores are both reported
 - High mapping quality ≠ High alignment score
 - mapping quality describes positioning
 - reflects the probability that the read is incorrectly mapped to the reported location

Read 2

- is a Phred score: P(incorrectly mapped) = 10^{-mappingQuality/10}
- alignment score describes fit
 - reflects the correspondence between the read and the reference sequence

• Maps to one location high mapping quality

 Has 2 mismatches low alignment score

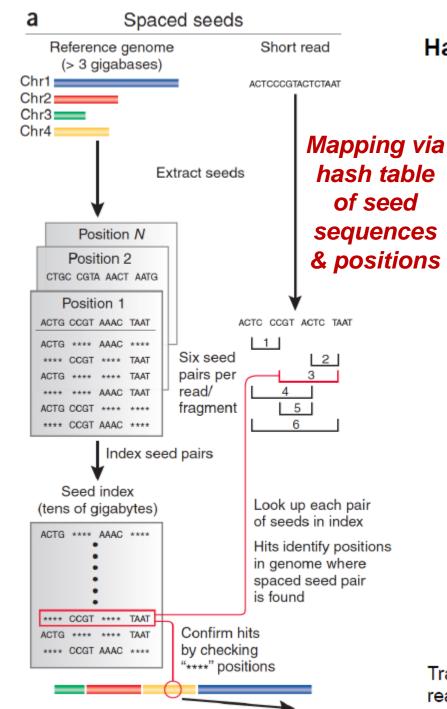
Read 1 GCGTAGTCTGCC || || || || ||| TAGCCTAGTGTGCCGC

ATCGGGAGATCC

r ATCGGGAGATCC ||||||||||| TTATCGGGAGATCCGC

- Maps to 2 locations low mapping quality
- Matches perfectly high alignment score

reference sequence



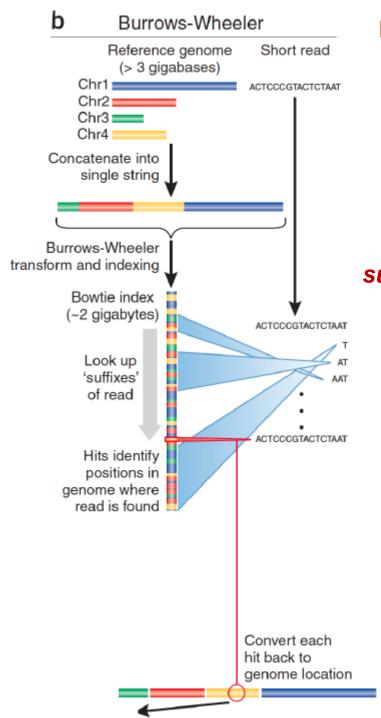
Hash table enables lookup of exact matches.

Reference Positions
2341, 2617264
134, 13311, 732661,
3452
234456673

Table is sorted and complete so you can jump immediately to matches.
(But this can take a lot of memory.)

May include N bases, skip positions, etc.

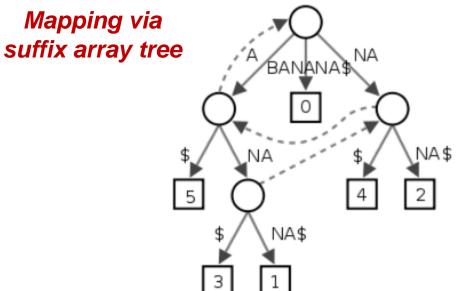
Trapnell, C. & Salzberg, S. L. How to map billions of short reads onto genomes. *Nature Biotech.* **27**, 455–457 (2009).



Burrows-Wheeler transform compresses sequence.

Input	SIX.MIXED.PIXIES.SIFT.SIXTY.PIXIE.DUST.BOXES
Output	TEXYDST.E.IXIXIXXSSMPPS.BE.S.EUSFXDIIOIIIT

Suffix tree enables fast lookup of subsequences.



http://en.wikipedia.org/wiki/Suffix_tree

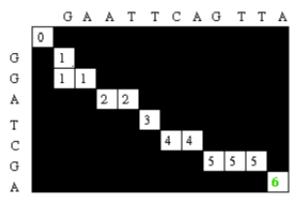
Exact matches at all positions below a node.

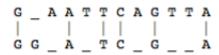
Trapnell, C. & Salzberg, S. L. How to map billions of short reads onto genomes. *Nature Biotech.* **27**, 455–457 (2009).

Alignment via dynamic programming

 Dynamic programming algorithm (Smith-Waterman | Needleman-Wunsch)

		G	Α	Α	T	Т	С	Α	G	T	T	A
	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1	1
G	0	1	1	1	1	1	1	1	2	2	2	2
Α	0	1	1	2	2	2	2	2	2	2	2	3
T	0	1	2	2	3	3	3	3	3	3	3	3
С	0	1	2	2	3	3	4	4	4	4	4	4
G	0	1	2	2	3	3	4	4	5	5	5	5
Α	0	1	2	3	3	3	4	5	5	5	5	6





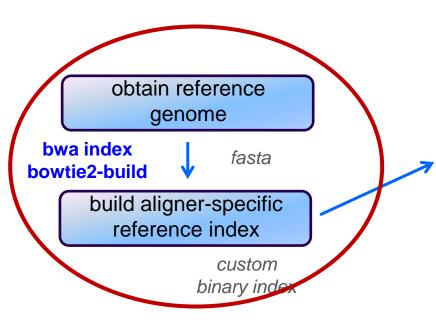
Alignment score = Σ

- match reward
- base mismatch penalty
- gap open penalty
- gap extension penalty

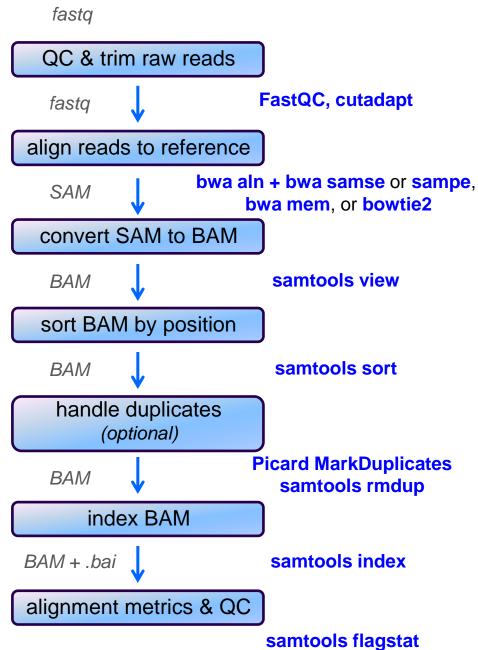
Reference sequence
ATTTGCGATCGGATGAAGACGAA
|||||||||||||||
ATTTGCGATCGGATGTTGACTTT

ATTTGCGATCGGATGAAGACG..AA
||||||||||||||||XX|||Xii||
ATTTGCGATCGGATGTTGACTTTAA

 rewards and penalties may be adjusted for quality scores of bases involved



Alignment Workflow



http://bio-bwa.sourceforge.net/bwa.shtml

http://bowtie-bio.sourceforge.net/bowtie2/manual.shtml

samtools flagstat samtools idxstat

Obtaining/building a reference

- What is a reference?
 - assembled genomes or transcriptomes
 - Ensembl, UCSC, for eukaryotes
 - NCBI RefSeq or GenBank for prokaryotes/microbes (prefer RefSeq)
 - any set of named DNA sequences
 - names are chromosome or gene names (technically referred to as contigs)
- Building a reference index (aligner-specific)
 - may take several hours to build
 - but you build each index once, use for multiple alignments
 - requires FASTA files (.fa, .fasta) containing DNA sequences
 - annotations (genome feature files, .gtf) may also be used to build the index, but will definitely be needed for downstream analysis

>chrM Mitochondiral Chromosome

GATCACAGGTCTATCACCCTATTAACCACTCACGGGAGCTCTCCATGCAT TTGGTATTTTCGTCTGGGGGGTGTGCACGCGATAGCATTGCGAGACGCTG GAGCCGGAGCACCCTATGTCGCAGTATCTGTCTTTGATTCCTGCCTCATT

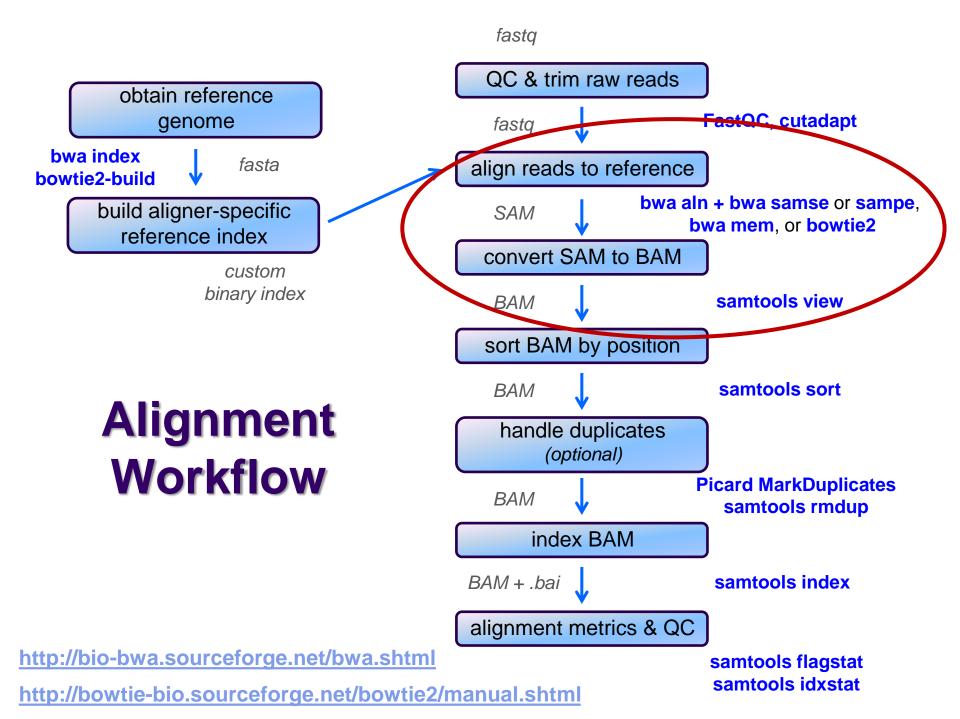
sequence name line

- always starts with >
- followed by a *name* and other (optional) descriptive information

one or more line(s) of sequence characters

never starts with >

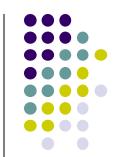
. . .



SAM / BAM file format

- Aligners take FASTQ as input, output alignments in Sequence Alignment Map (SAM) format
 - plain-text file format that describes how reads align to a reference
 - http://samtools.github.io/hts-specs/SAMv1.pdf (the "Bible")
 - and now https://github.com/samtools/hts-specs/blob/master/SAMtags.pdf
- SAM and BAM are two forms of the same data
 - BAM Binary Alignment Map
 - same data in a custom compressed (gzip'd) format
 - much smaller than SAM files
 - when sorted + indexed, support fast random access (SAM files do not)
- SAM file consists of
 - a header (includes reference sequence names and lengths)
 - alignment records, one for each sequence read
 - alignments for R1 and R2 reads have separate records
 - records have 11 fixed fields + extensible-format key:type:value tuples

SAM file format Fixed fields (tab-separated)



Col	Field	Type	Regexp/Range	Brief description
1	QNAME	String	[!-?A-~]{1,255}	Query template NAME read name from fastq
2	FLAG	Int	[0,2 ¹⁶ -1]	bitwise FLAGS
3	RNAME	String	* [!-()+-<>-~][!-~]*	Reference sequence NAME contig + start
4	POS	Int	[0,2 ²⁹ -1]	1-based leftmost mapping POSition = locus
5	MAPQ	Int	[0,2 ⁸ -1]	MAPping Quality
6	CIGAR	String	* ([0-9]+[MIDNSHPX=])+	CIGAR string use this to find end coordinate
7	RNEXT	String	* = [!-()+-<>-~][!-~]*	Ref. name of the mate/next segment
8	PNEXT	Int	[0,2 ²⁹ -1]	Position of the mate/next segment
9	TLEN	Int	[-2 ²⁹ +1,2 ²⁹ -1]	observed Template LENgth insert size, if paired
10	SEQ	String	* [A-Za-z=.]+	segment SEQuence
11	QUAL	String	[!-~]+	ASCII of Phred-scaled base QUALity+33

SRR030257.264529

99 NC_012967

1521

29 (34M2S)

1564

positive for plus strand reads

CTGGCCATTATCTCGGTGGTAGGACATGGCATGCCCAAAAAA;AA;AAAAAA??A%.;?&'3735',()0*,

AAAAAA,AA,AAAAAA::A/0.,:&3/33,(JU ,

XT:A:M NM:i:3 SM:i:29 AM:i:29 XM:i:3 XO:i:0 XG:i:0 MD:Z:23T0G4T4

SRR030257.2669090

147 NC 012967

1521

60 X 36M

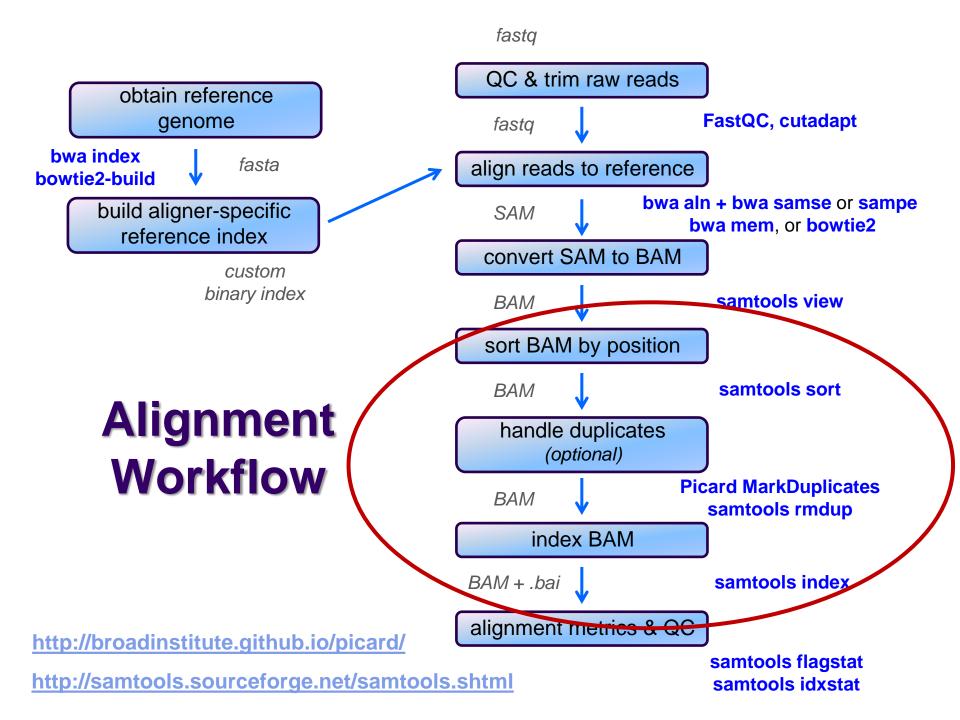
1458

-99

negative for minus strand reads

CTGGCCATTATCTCGGTGGTAGGTGATGGTATGCGC

XT:A:U NM:i:0 SM:i:37 AM:i:37 X0:i:1 X1:i:0 XM:i:0 XO:i:0 XG:i:0 MD:Z:36







- SAM created by aligner contains read records in name order
 - same order as read names in the input FASTQ file
 - R1, R2 have adjacent SAM records
 - SAM → BAM conversion does not change the name-sorted order
- Sorting BAM puts records in position (locus) order
 - sorting is very compute, I/O and memory intensive!
 - can take hours for large BAM
- Indexing a locus-sorted BAM allows fast random access
 - creates a small, binary alignment index file (.bai)
 - quite fast

Handling Duplicates



- Optional step, but very important for many protocols
- Definition of alignment duplicates:
 - single-end reads or singleton/discordant PE alignment reads
 - alignments have the same start positions
 - paired-end reads
 - pairs have same external coordinates (5' + 3' coordinates of the insert)
- Two choices for handling:
 - samtools rmdup removes duplicates entirely
 - faster, but data is lost
 - Picard MarkDuplicates flags duplicates only
 - slower, but all alignments are retained
 - both tools are quirky in their own ways

obtain reference genome

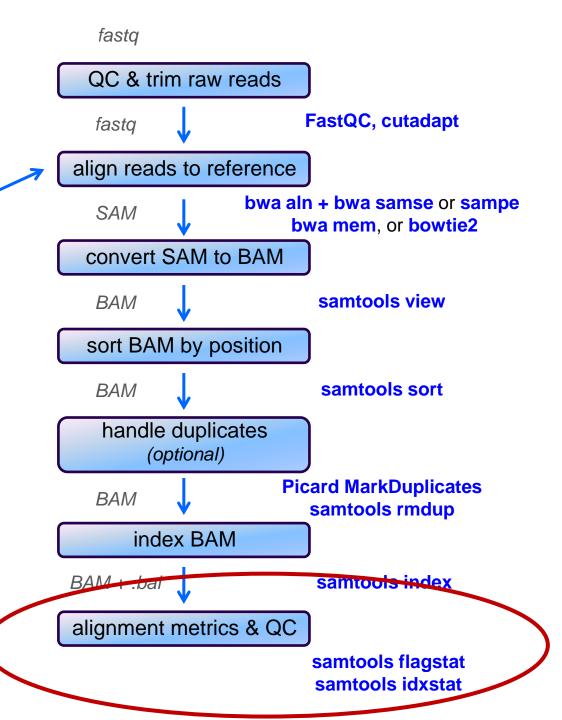
bwa index bowtie2-build

fasta

build aligner-specific reference index

custom binary index

Alignment Workflow



Alignment metrics

samtools flagstat

- simple statistics based on alignment record flag values
 - total sequences (R1+R2), total mapped
 - number properly paired
 - number of duplicates (0 if duplicates were not marked)

```
161490318 + 0 in total (QC-passed reads + QC-failed reads)
0 + 0 secondary
0 + 0 supplementary
31602827 + 0 duplicates
158093331 + 0 mapped (97.90% : N/A)
161490318 + 0 paired in sequencing
80745159 + 0 read1
80745159 + 0 read2
153721151 + 0 properly paired (95.19% : N/A)
156184878 + 0 with itself and mate mapped
1908453 + 0 singletons (1.18% : N/A)
1061095 + 0 with mate mapped to a different chr
606632 + 0 with mate mapped to a different chr (mapQ>=5)
```

Alignment wrap up



- Many tools involved
 - choose one or two and learn their options well
- Many steps are involved in the full alignment workflow
 - important to go through manually a few times for learning
 - but gets tedious quickly!
 - best practice
 - automate series of complex steps by wrapping into a pipeline script
 - e.g. bash or python script
 - the Bioinformatics team has a set of pipeline scripts available at TACC and BRCF "pods"
 - align_bowtie2_illumina.sh, align_bwa_illumina.sh, trim_adapters.sh, etc.
 - TACC: shared project directory /work/projects/BiolTeam/common/script/
 - BRCF pods: read-only mount /mnt/bioi/

Part 4 summary



- Short read aligners determine placement of query sequences (your reads) against a known reference (e.g. a genome)
 - Mapping determines one or more positions (a.k.a. seeds or hits)
 where a read shares a short sequence with the reference
 - Alignment starts with the seed and determines how read bases are best matched, base-by-base, around the seed
- The alignment workflow:
 - Obtain a suitable reference (FASTA) + annotations (gtf)
 - Build an aligner-specific index (one-time)
 - Align QC'd reads to the reference → SAM file; convert SAM → BAM & sort by position
 - Handle duplicates (optional, but informative)
 - Index the BAM for fast random access
 - Gather and interpret alignment statistics

Other NGS Resources at UT



- CBRS training courses
 - Intro to NGS, RNAseq, many others
 https://research.utexas.edu/cbrs/cores/cbb/educational-resources/
 - "Summer School" (4 or 5 half-day sessions in June)
 - lots of hands-on, including w/TACC
 - Short courses (3-4 hour workshops) starting next week!
 https://site.research.utexas.edu/cbrs/classes/short-courses/spring-2025-semester/
- Genome Sequencing & Analysis Facility (GSAF)
 - Jessica Podnar, Director, gsaf@utgsaf.org
- Bioinformatics consultants
 - Dennis Wylie, Dhivya Arasappan, Anna
 - BiolTeam wiki https://wikis.utexas.edu/display/bioiteam/
- Biomedical Research Support Facility (BRCF)
 - provides local compute and managed storage resources
 https://wikis.utexas.edu/display/RCTFUsers

Final thoughts

- Good judgement comes from experience unfortunately...
- Experience comes from bad judgement!
- So go get started making your 1st 1,000 mistakes....

