Introduction to NGS Analysis

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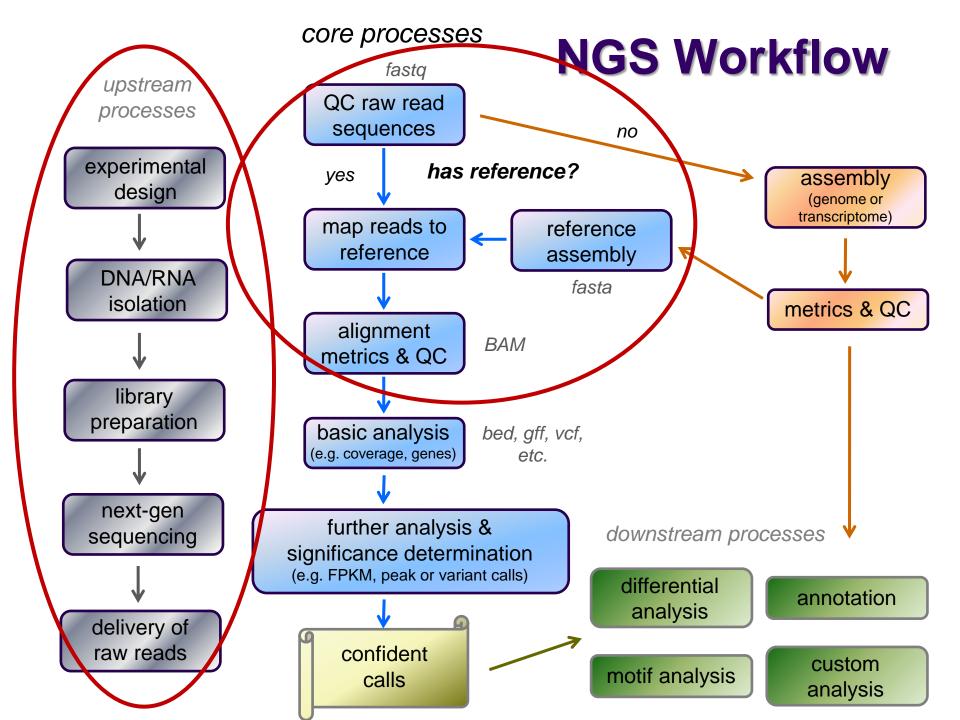
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Center for Biomedical Research Support (CBRS)

Bioinformatics Consulting Group (BCG) Biomedical Research Computing Facility (BRCF) Genome Sequencing & Analysis Facility (GSAF)

Center for Systems and Synthetic Biology (CSSB) Ed Marcotte & Vishwanath Iyer labs



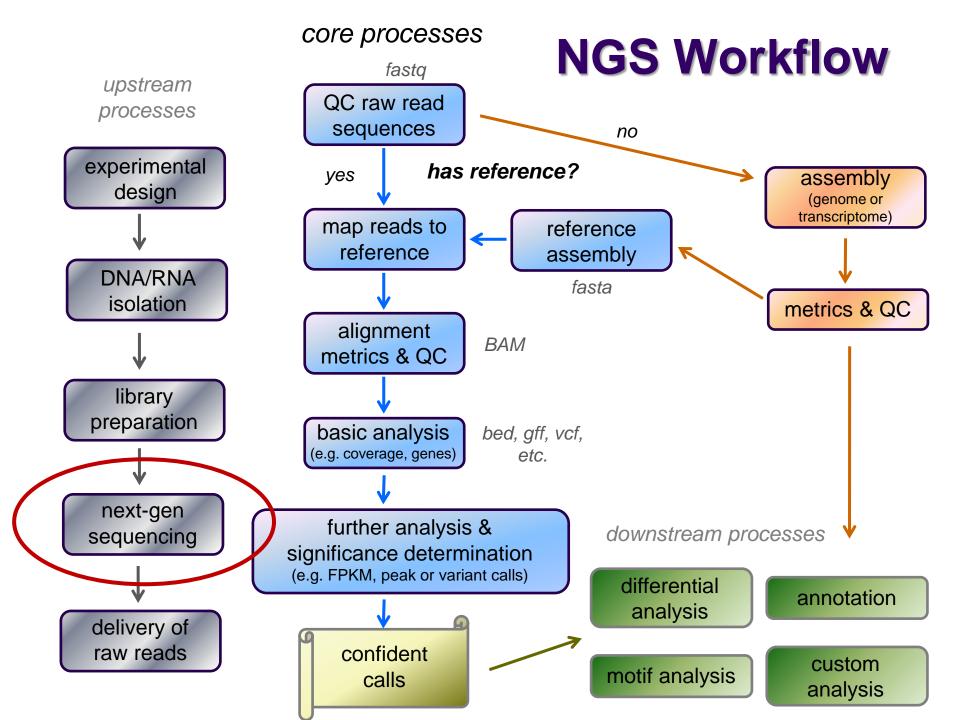
Outline



- 1. Overview of sequencing technologies
- 2. 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

а

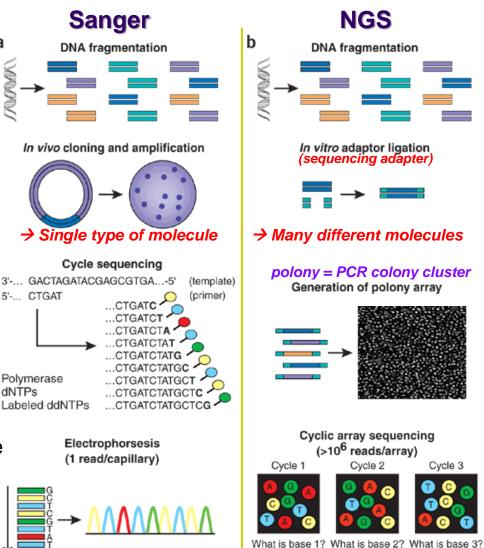
- Massively parallel
 - simultaneously sequence "library" of *millions* of different DNA fragments

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







Cycle 3

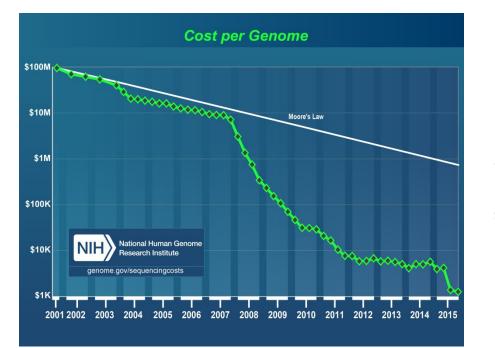
→ Single type of molecule

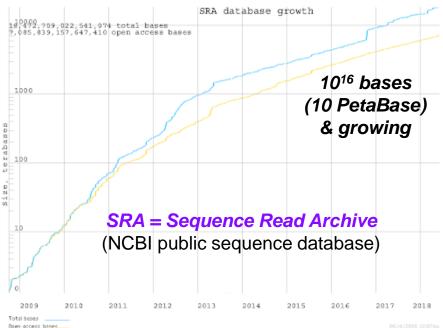
"Next Generation" sequencing (2nd generation)

Pro's:

- much faster!
- much lower cost!
- both deeper and wider coverage!

- Con's:
 - data deluge!
 - storage requirements!
 - analysis lags!

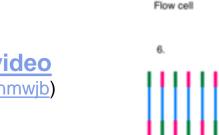


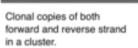


Illumina sequencing

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

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

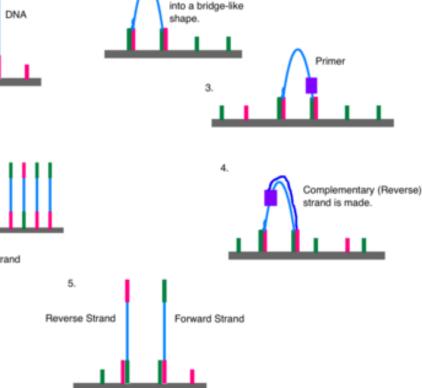






- 2 PCR amplifications performed
 - 1. during *library preparation*
 - 2. during *cluster generation*
- amplification always introduces bias!



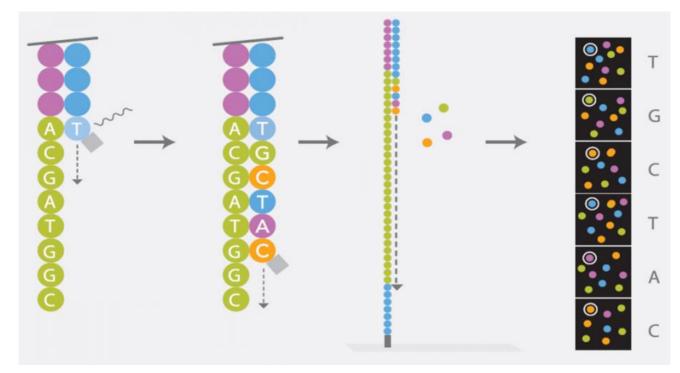


DNA folds over

2.

Illumina sequencing

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

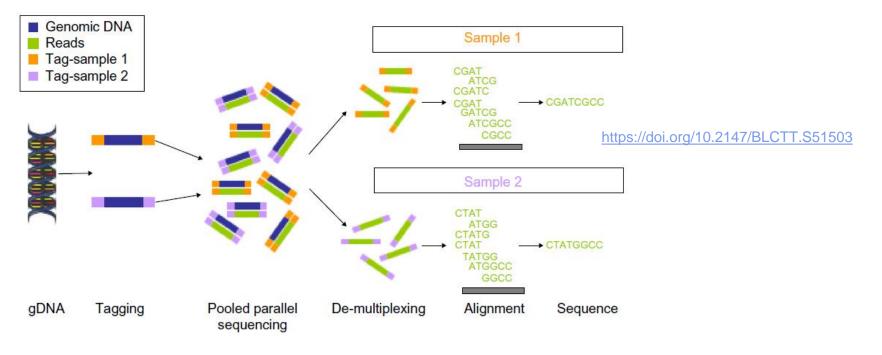


http://www.cegat.de/

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



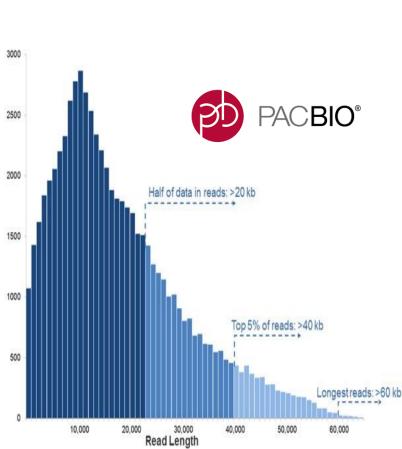
Long read sequencing

Read

- 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
 - difficult to detect large structural variations like inversions
- 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
 - and fewer reads are generated (~100 K)

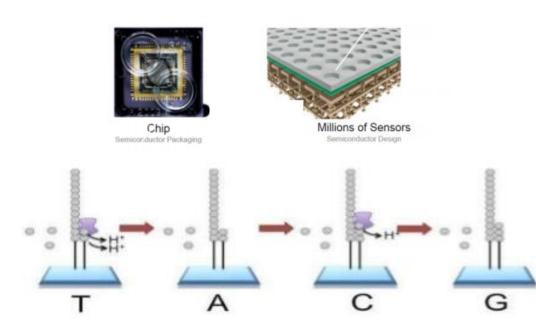


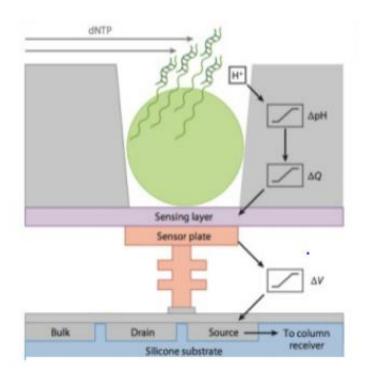


Long read sequencing



- 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%)





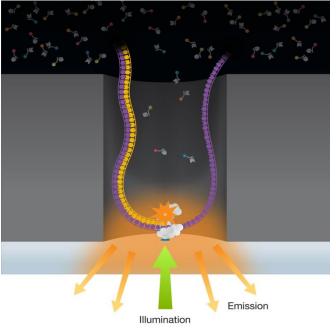
Long read sequencing

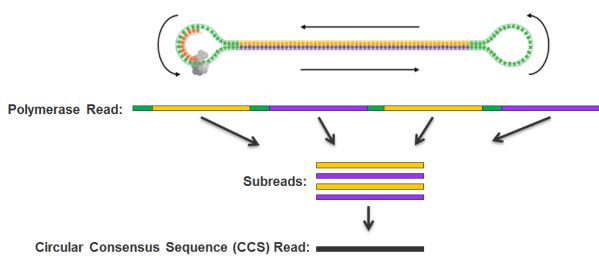
PacBio SMRT system

http://www.pacb.com/smrt-science/smrt-sequencing/



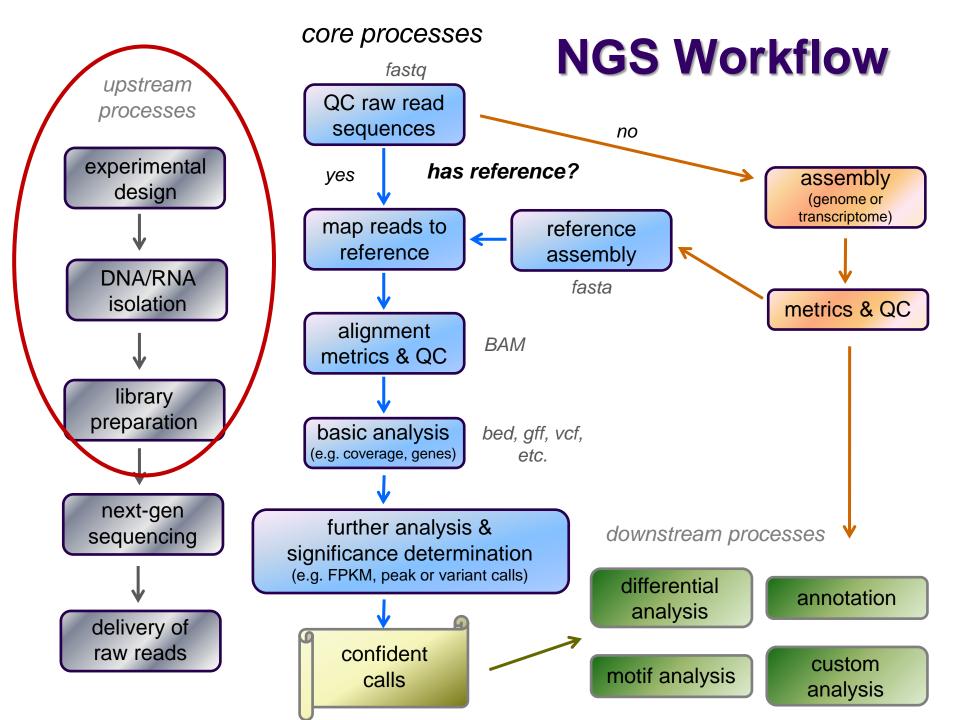
- Sequencing by synthessis in **Zero-Mode Waveguide** (ZMW) wells
- DNA is circularized then repeatedly sequenced to achieve "consensus"
 - reduces error rate (~1-2%), but equipment quite expensive





Part 2: NGS Concepts & Terminology

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





Library Complexity

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

Many different molecules \rightarrow high complexity Few different molecules \rightarrow low complexity

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

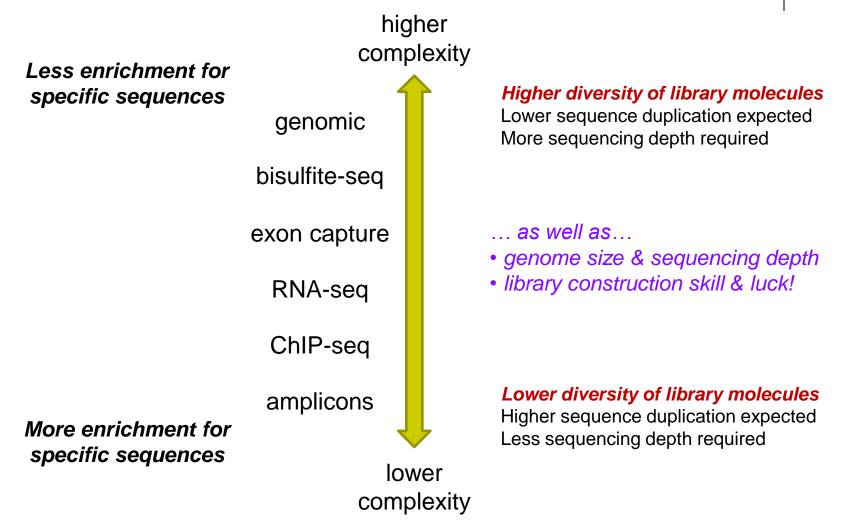
Popular Experiment Types

- Whole Genome sequencing (WGS)
 - *library*: all genomic DNA
 - *complexity*: high (fragments must cover the entire genome)
 - *applications*: genome assembly, variant analysis
- Exome sequencing (WXS)
 - *library*: DNA from eukaryotic exonic regions (uses special kits)
 - *complexity*: high/med (only ~5% of eukaryotic genome is in exons)
 - applications: polymorphism/SNP detection; genotyping
- RNA-seq
 - *library*: extracted RNA converted to cDNA
 - complexity: med/high (only a subset of genes are expressed in any given tissue)
 - *applications*: differential gene expression
- Amplicon panels (targeted sequencing)
 - *library*: DNA from a set of PCR-amplified regions using custom primers
 - **complexity**: very low (only 1 to a few thousand different library molecules)
 - *applications*: genetic screening panels; metagenomics; mutagenesis

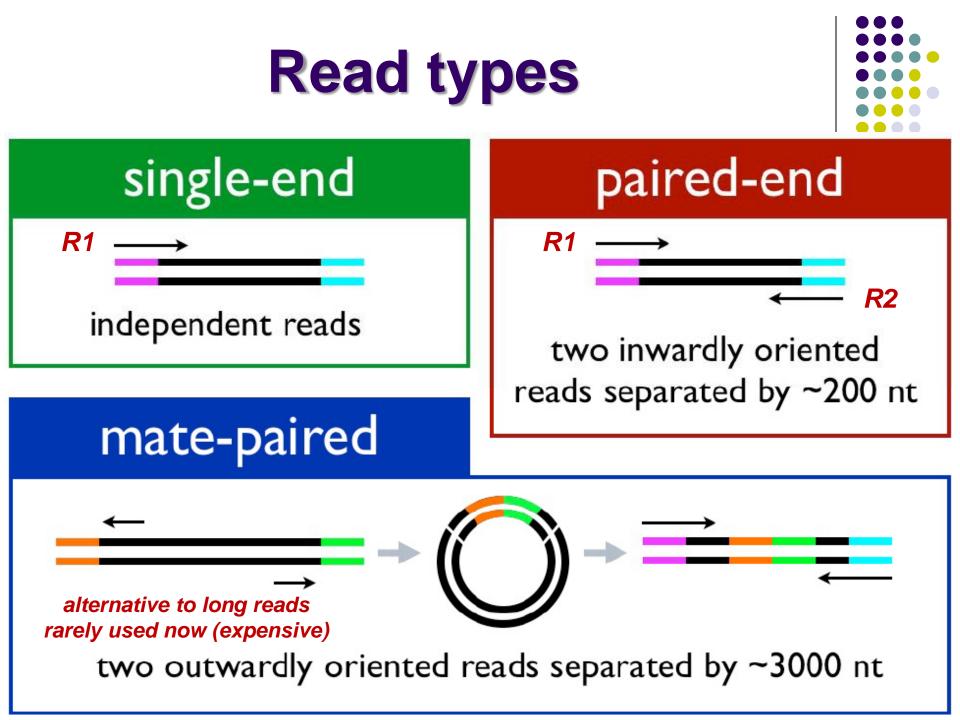


Туре	Library construction	Applications	Complexity
Whole genome (WGS)	 extract genomic DNA & fragment 	Genome assemblyVariant detection, genotyping	high
Bisulfite sequencing	 bisulfite treatment converts C → U but not 5meC 	 Methylation profiling (CpG) 	high
RAD-seq, ddRAD	 restriction-enzyme digest DNA & 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 expressionTranscriptome assembly	medium, medium-low for Tag-seq
Transposon seq (Tn-seq)	 create library of transposon- mutated genomic DNA amplify mutants via Tn-PCR 	Charcterize 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

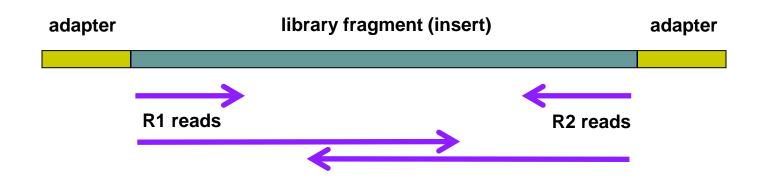






Reads and Fragments

- 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
 - also called tags
- An R1 and its associated R2 form a read pair
 - a readout of part (or all) of the fragment molecule





Single end vs Paired end

- single end (SE) reads are less expensive
- *paired end* (PE) reads can be mapped more reliably
 - especially against lower complexity genomic regions
 - an unmapped read can be "rescued" if its mate maps well
 - they provide more bases around a locus
 - e.g. for analysis of polymorphisms
 - actual fragment sizes can be easily determined
 - from the alignment records for each dual-mapping "proper pair"
 - also help distinguish the true complexity of a library
 - by clarifying which *fragments* are duplicates (vs *read* 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





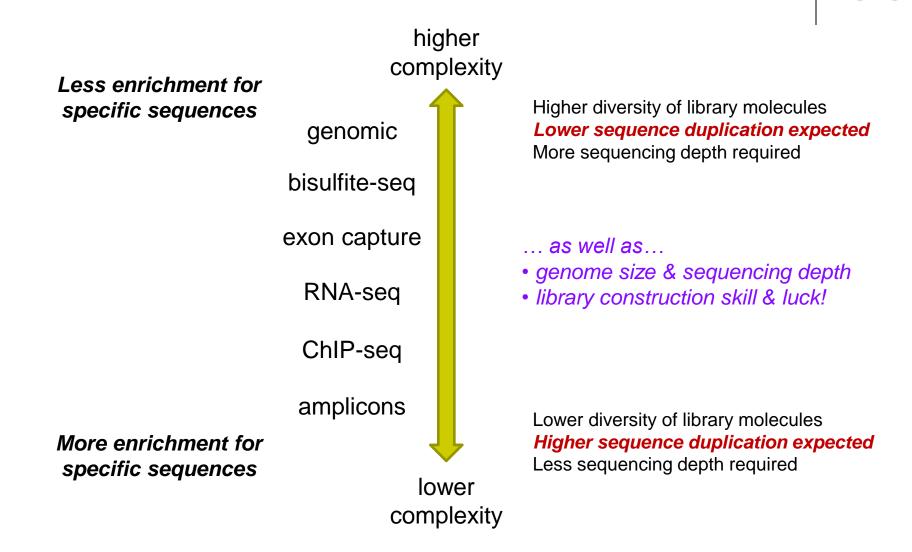
Sequence Duplication



- Duplication can arise from:
 - 1. sequencing of species enriched in your library (*biological good!*)
 - each read comes from a different DNA molecule
 - 2. 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!*
- Standard best practice is to "mark duplicates" during initial processing
 - then decide what to do with them later...
 - e.g. retain (use all), remove (use only non-duplicates), dose (use some)
- Different experiment types have different *expected* duplication

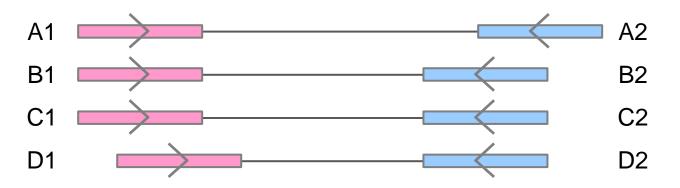


Expected sequence duplication is primarily a function of experiment type



Read vs Fragment duplication

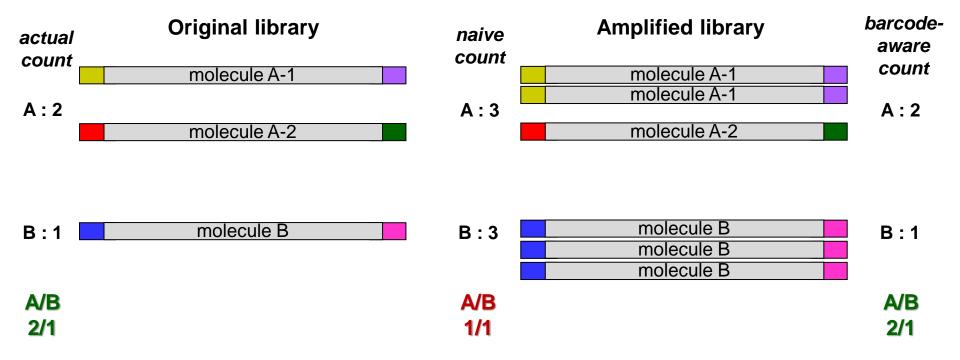
- Consider the 4 "aligned" fragments below
 - 4 R1 reads (pink), 4 R2 reads (blue)
- Duplication when only 1 end considered
 - A1, B1, C1 have identical sequences, D1 different
 - 2 unique + 2 duplicates = 50% duplication rate
 - B2, C2, D2 have identical sequences, A2 different
 - 2 unique + 2 duplicates = 50% duplication rate
- Duplication when both ends considered
 - fragments B and C are duplicates (same external sequences)
 - 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 UMIs (<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
 - e.g. UMI de-duplication for 3' Tag-seq, scRNA-seq





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
 - examples:
 - cells in different layers/regions of somatic tissue
 - cells in different areas of a tumor
 - essentially a very sophisticated library preparation technique
- Typical protocol (RNA-seq)
 - 1. isolate a few thousand cells (varying methods)
 - 2. the single-cell platform partitions each cell into an emulsion droplet
 - e.g. 10x Genomics (<u>https://www.10xgenomics.com/solutions/single-cell/</u>)
 - 3. 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



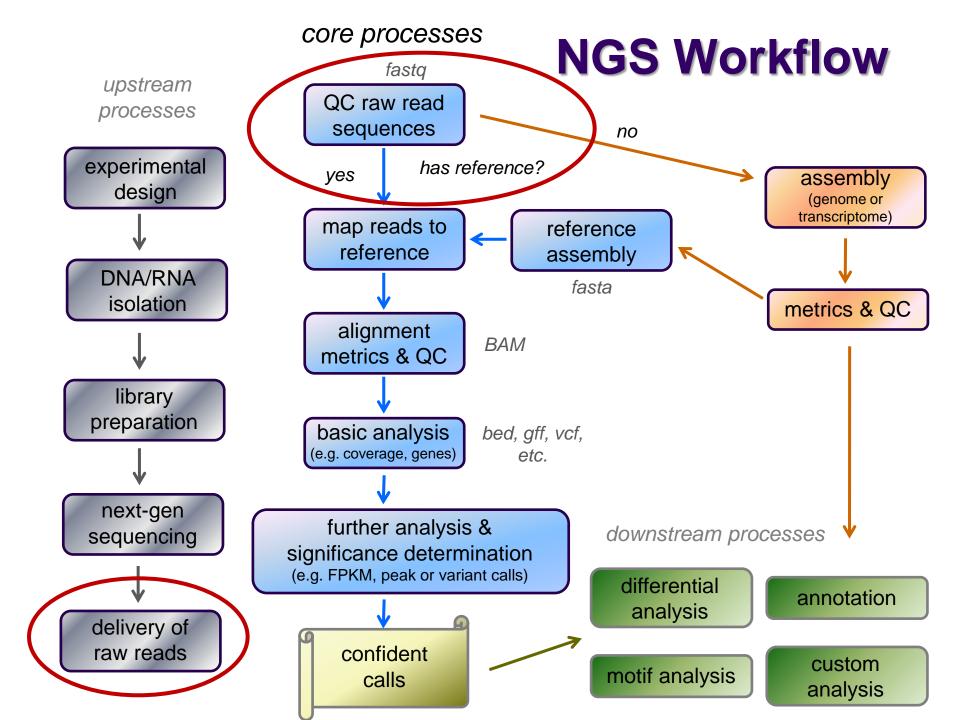
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Cells or nuclei

enzyme

Part 3: The FASTQ format, Data QC & preparation

- The FASTQ format
- 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:
 - 1. @read name (plus extra information after a space)
 - R1 and R2 reads have the same read name
 - 2. called base sequence (ACGTN) always 5' to 3'; usually excludes 5' adapter
 - 3. +optional read 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?



FASTQ quality scores

- Base qualities expressed as *Phred* scores
 - log scaled, *higher = better*
 - $20 = 1/10^2 = 1/100$ errors, $30 = 1/10^3 = 1/1000$ errors

Probability of Error = $10^{-Q/10}$

• Integer Phred score converted to Ascii character (add 33)

	http://www.asciitable.com/					
Quality character	!"#\$%&'()*+,/0123456789:;<=>?@ABCDEFGHIJ					
ASCII Value	33	43	53	63	73	
Base Quality (Q)	0	10	20	30	40	

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



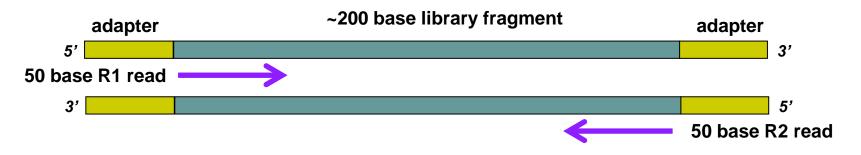
Raw sequence quality control

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

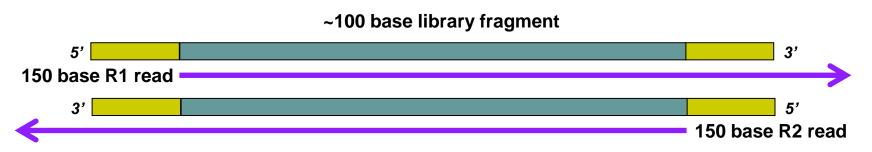


3' Adapter contamination

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.

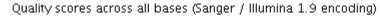


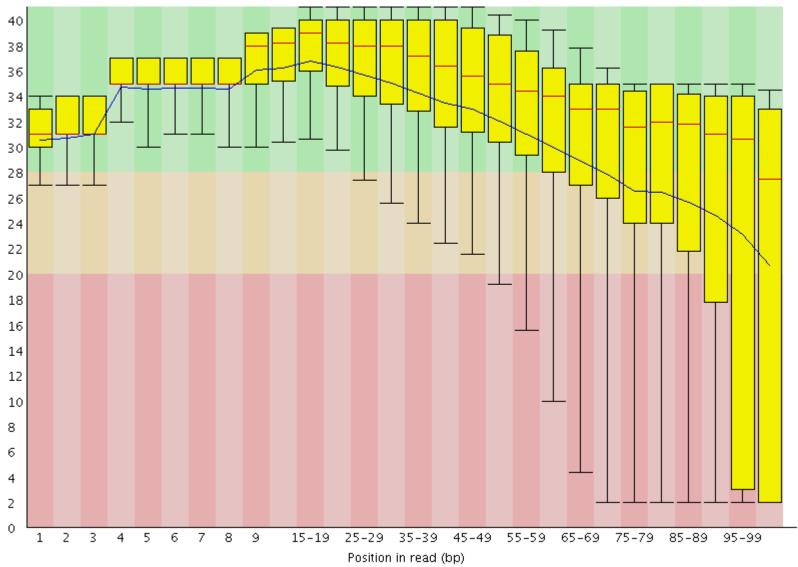


- Quality Assurance tool for FASTQ sequences http://www.bioinformatics.babraham.ac.uk
- Most useful reports:
 - 1. Per-base sequence quality Report
 - Should I trim low quality bases?
 - 2. Sequence duplication levels Report
 - How complex is my sequence library?
 - 3. Overrepresented sequences Report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

1. FastQC Per-base sequence quality report







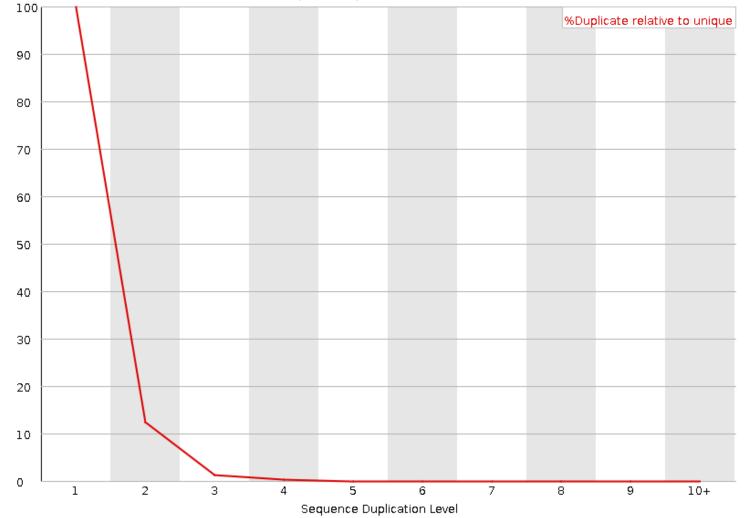
2. FastQC Sequence duplication report Yeast ChIP-seq

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

Sequence Duplication Level >= 31.9%



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)
GAGATCGGAAGAGCACACGTCTGAACTCCAGTCACCTCAGAATCTCGTAT	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
TATTCTGGTGTCCTAGGCGTAGAGGAACAACACCAATCCATCC	494014	2.8091456822607364	No Hit
TCAAACGAGGAAAGGCTTACGGTGGATACCTAGGCACCCAGAGACGAGGA	446641	2.539765344040083	No Hit
TAAAACGACTCTCGGCAACGGATATCTCGGCTCTCGCATCGATGAAGAAC	179252	1.0192929387357474	No Hit
GAAGGTCACGGCGAGACGAGCCGTTTATCATTACGATAGGGGTCAAGTGG	171681	0.9762414422996221	No Hit
AACGACTCTCGGCAACGGATATCTCGGCTCTCGCATCGATGAAGAACGTA	143415	0.8155105483274229	No Hit
AGAACATGAAACCGTAAGCTCCCAAGCAGTGGGAGGAGCCCTGGGCTCTG	111584	0.6345077504066322	No Hit
AAAACGACTCTCGGCAACGGATATCTCGGCTCTCGCATCGATGAAGAACG	111255	0.6326369351474214	No Hit
ATTACGATAGGTGTCAAGTGGAAGTGCAGTGATGTATGCAGCTGAGGCAT	73682	0.41898300890326096	No Hit
GAAGGTCACGGCGAGACGAGCCGTTTATCATTACGATAGGTGTCAAGGGG	71661	0.4074908580252516	No Hit
GGATGCGATCATACCAGCACTAATGCACCGGATCCCATCAGAACTCCGCA	69548	0.3954755612388914	No Hit
ATATTCTGGTGTCCTAGGCGTAGAGGAACAACACCAATCCATCC	54017	0.30716057099328803	No Hit



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
 - 3. 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 <u>https://code.google.com/p/cutadapt/</u>
 - trimmomatic <u>http://www.usadellab.org/cms/?page=trimmomatic</u>
 - FASTX-Toolkit <u>http://hannonlab.cshl.edu/fastx_toolkit/</u>
- 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)
 - 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" adapter sequences will be "soft clipped" (ignored)

global (end-to-end) alignment of query

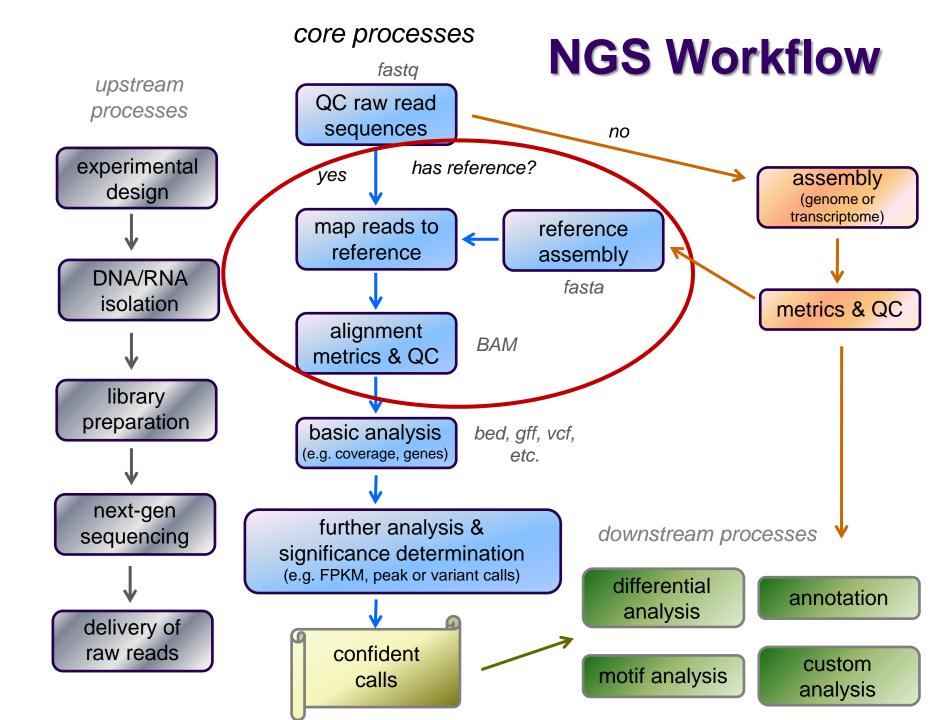
local (subsequence) alignment of query

CACAAGTACAATTATACACCTAGCTTATCGCCCTGAAGGACTTACATACACAAGTACAATTATACACAGACATTAGTTCTTATCGCCCTGAAAATTCTCC

reference sequence

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)
- 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 <u>http://bowtie-bio.sourceforge.net/bowtie2/manual.shtml</u>
 - 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
 - is a Phred score: P(incorrectly mapped) = 10^{-mappingQuality/10}
 - alignment score describes fit
 - reflects the correspondence between the read and the reference sequence

	Read 1	Rea	d 2	
 Maps to one location high mapping quality Has 2 mismatches low alignment score 	GCGTAGTCTGCC	ATCGGGAGATCC	r ATCGGGAGATCC TTATCGGGAGATCCGC	 Maps to 2 locations low mapping quality Matches perfectly high alignment score

reference sequence

a Spaced seeds

Seed index

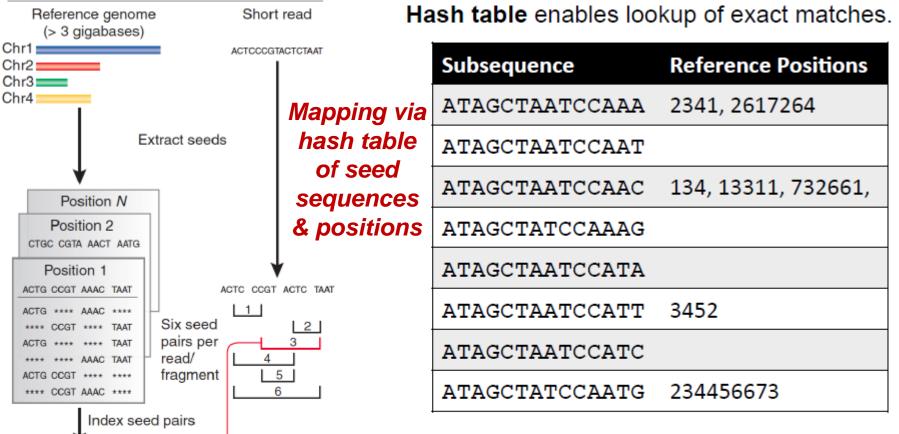
(tens of gigabytes)

ACTG **** AAAC ****

**** CCGT **** TAAT

ACTG **** **** TAAT

**** CCGT AAAC ****



Look up each pair

Hits identify positions in genome where spaced seed pair

of seeds in index

is found

Confirm hits

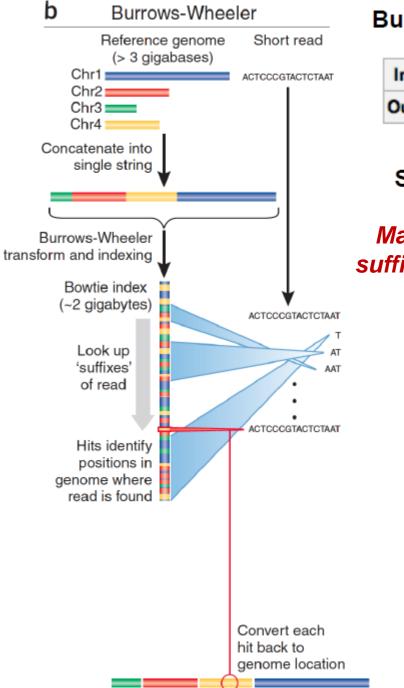
by checking

"****" positions

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.

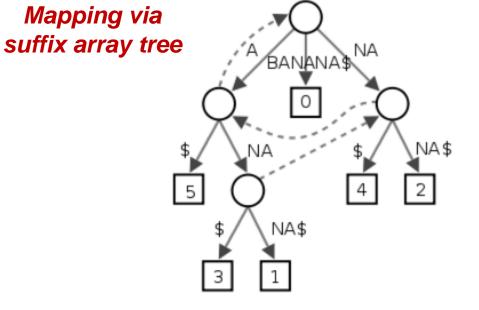
Trapnell, C. & Salzberg, S. L. How to map billions of short reads onto genomes. *Nature Biotech.* 27, 455–457 (2009).



Burrows-Wheeler transform compresses sequence.

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

Suffix tree enables fast lookup of subsequences.



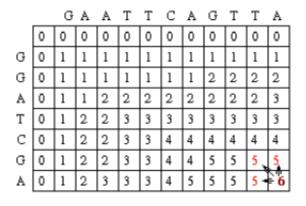
http://en.wikipedia.org/wiki/Suffix_tree

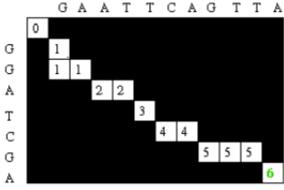
Exact matches at all positions below a node.

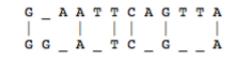
Trapnell, C. & Salzberg, S. L. How to map billions of short reads onto genomes. *Nature Biotech.* 27, 455–457 (2009).

Alignment via dynamic programming

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







- Alignment score = Σ
 - match reward
 - base mismatch penalty
 - gap open penalty
 - gap extension penalty

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

ATTTGCGATCGGATGAAGACG..AA

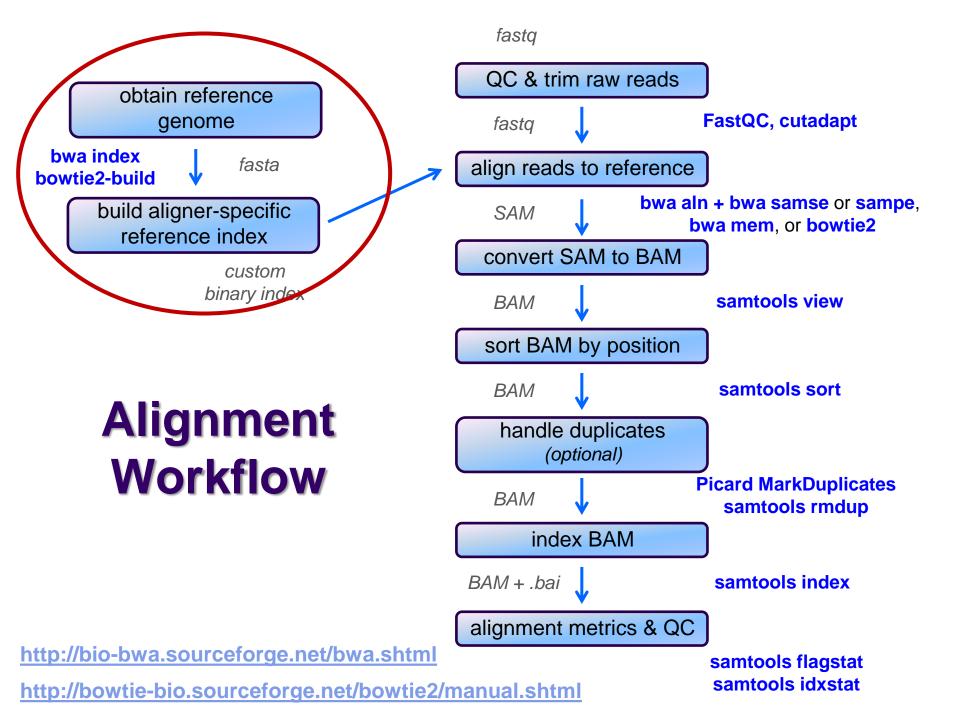
 rewards and penalties may be adjusted for quality scores of bases involved



Paired End mapping

- Having paired-end reads improves mapping
 - mapping one read with high confidence anchors the pair
 - even when its mate read by itself maps several places equally
- Three possible outcomes of mapping an R1/R2 pair
 - 1. only one of a pair might map (singleton/orphan)
 - 2. both reads can map within the most likely distance range and with correct orientation *(proper pair)*
 - 3. both reads can map but with an unexpected insert size or orientation, or to different contigs *(discordant pair)*
- Insert fragment size is reported in the alignment record
 - for both proper and discordant pairs





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 can be chromosomes, genes, etc. (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, .gff) may also be used to build the index, but will definitely be needed for downstream analysis

>chrM Mitochondiral Chromosome GATCACAGGTCTATCACCCTATTAACCACTCACGGGAGCTCTCCATGCAT TTGGTATTTTCGTCTGGGGGGGGTGTGCACGCGATAGCATTGCGAGACGCTG GAGCCGGAGCACCCTATGTCGCAGTATCTGTCTTTGATTCCTGCCTCATT

sequence name line

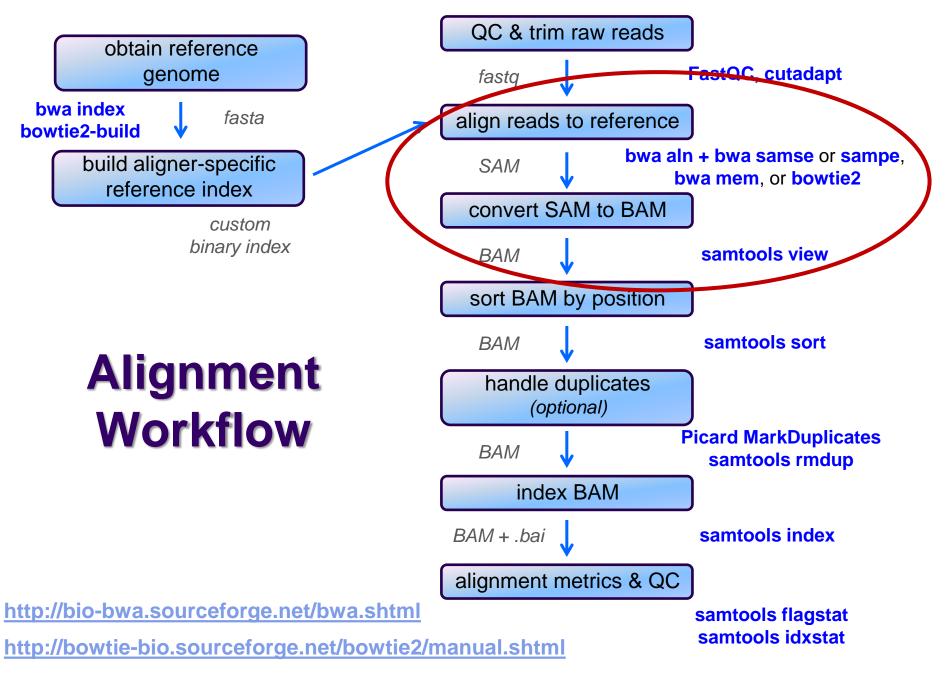
- always starts with >
- followed by a *name* and other (optional) descriptive information

one or more line(s) of sequence characters

never starts with >







SAM / BAM file format

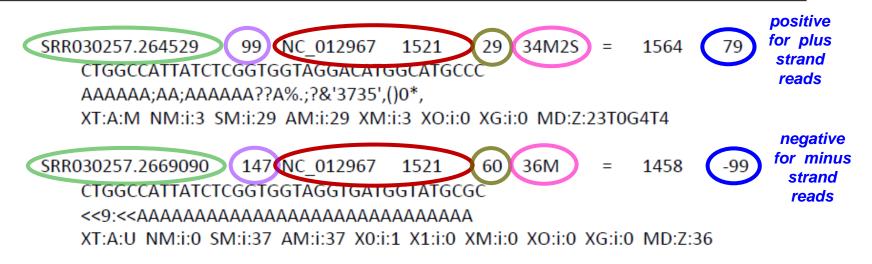
- Aligners take FASTQ as input, output alignments in Sequence Alignment Map (SAM) format
 - plain-text file format that describes how reads align to a reference
 - http://samtools.github.io/hts-specs/SAMv1.pdf (the Bible)
 - and now <u>https://github.com/samtools/hts-specs/blob/master/SAMtags.pdf</u>
- 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



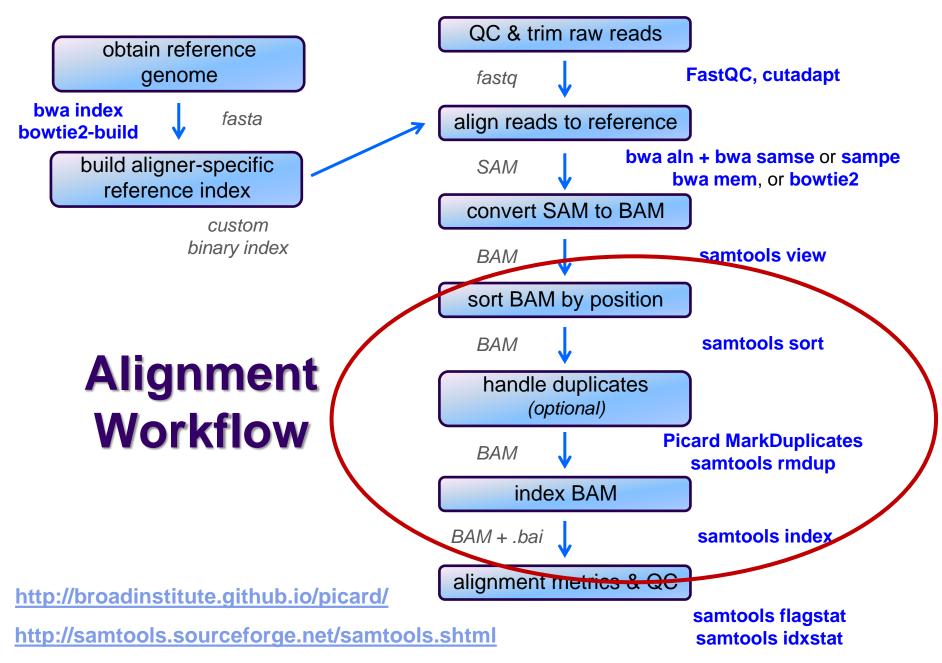
SAM file format Fixed fields (tab-separated)



Col	Field	Туре	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



fastq



Sorting / indexing BAM files

- 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*
 - by contig name then start position (leftmost)
 - 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

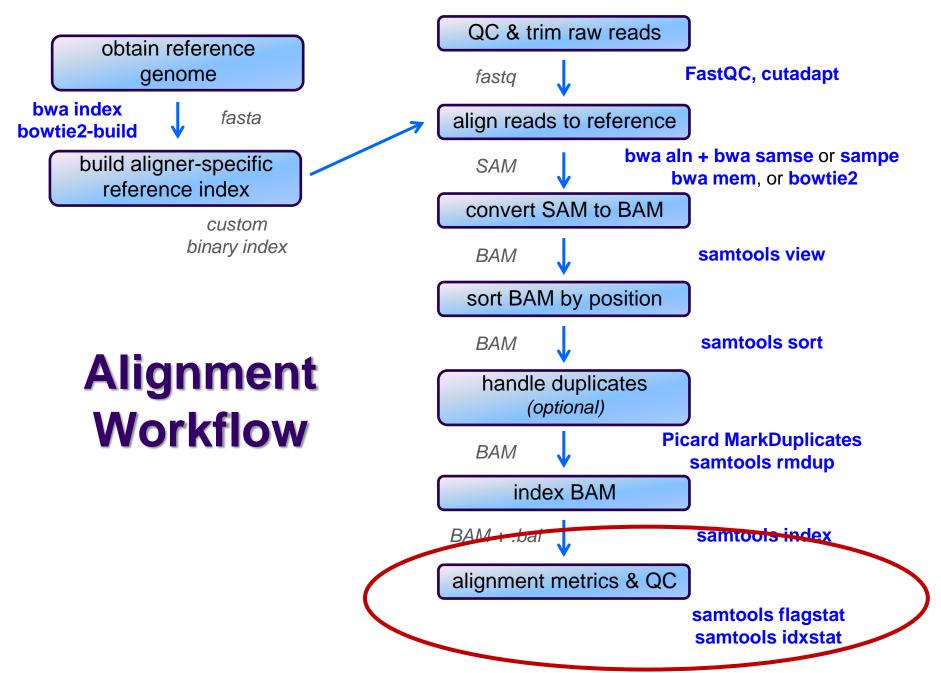


Handling Duplicates

- 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
 - properly paired 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 (0x400 bam flag)
 - slower, but all alignments are retained
 - both tools are quirky in their own ways



fastq



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/

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)
- Genome Sequencing & Analysis Facility (GSAF)
 - Jessica Podnar, Director, <u>gsaf@utgsaf.org</u>
- Bioinformatics consultants
 - Dennis Wylie, Dhivya Arasappan, Benni Goetz, Anna
 - BiolTeam wiki <u>https://wikis.utexas.edu/display/bioiteam/</u>
- Biomedical Research Support Facility (BRCF)
 - provides local compute and managed storage resources
 <u>https://wikis.utexas.edu/display/RCTFUsers</u>

