Introduction to secondgeneration sequencing

Review: DNA sequencing

- Technologies to determine the nucleotide sequences of a DNA molecule.
- Motivation: decipher the genetic codes hidden in DNA sequences for different biological processes.
- Genome projects: determine DNA sequences for different species, e.g., human genome project.
- Genomic research (in a nutshell): study the functions of DNA sequences and related components.

Sequencing technologies

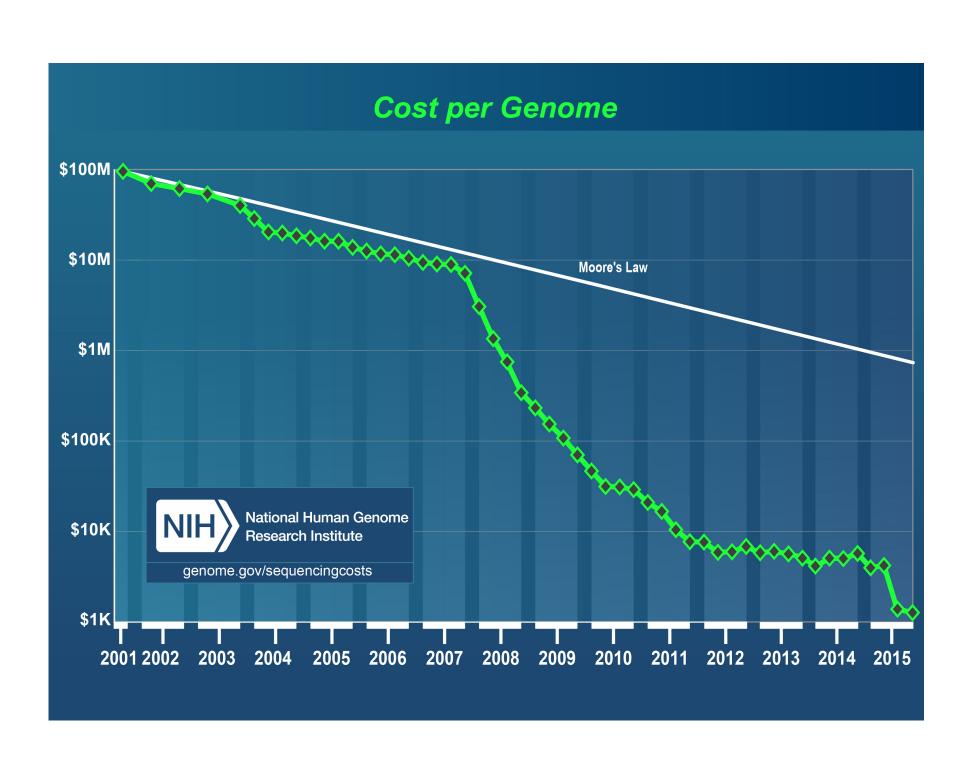
- Traditional technology: Sanger sequencing.
 - Slow (low throughput) and expensive: it took Human
 Genome Project (HGP) 13 years and \$3 billion to sequence
 the entire human genome.
 - Relatively accurate.
- New technology: different types of high-throughput sequencing.

Second generation sequencing

- Aka: high-throughput sequencing, next generation sequencing (NGS).
- Able to sequence large amount of short sequence segments in a short period:
 - high throughput: billions of sequences in a run.
 - Cheap: sequence entire human genome costs below one thousand dollars now.
 - short read length: up to several hundred bps.

HiSeq X Instrument Performance Parameters*

	Dual Flow Cell	Single Flow Cell
Output per Run	1.6-1.8 Tb	800-900 Gb
Reads Passing Filter	5.3-6 billion	2.6-3 billion
Supported Read Length	2 × 150 bp	
Run Time	< 3 days	
Quality Scores	≥ 75% of bases above Q30 at 2 × 150 bp	
Supported Library Preparation	TruSeq DNA PCR-Free Library Prep Kit, TruSeq Nano DNA Library Prep Kit	



Available platforms

- Major player:
 - Illumina: HiSeq, MiSeq.
 - LifeTech (now ThermoFisher): SOLiD, Ion Torrent.
- Others (third-generation sequencing):
 - Pacific Bioscience (SMRT)
 - Oxford Nanopore

High-throughput sequencers







Second-generation sequencing technologies

Second-generation sequencing technologies

- Complicated and involves a lot of biochemical reactions.
 - Sequencing by synthesis:
 https://www.youtube.com/watch?v=fCd6B5HRaZ8.
 - Sequencing by ligation.
 - Pyrosequencing.

• In a nutshell:

- Cut the long DNA into smaller segments (several hundreds to several thousand bases).
- Sequence each segment: start from one end and sequence along the chain, base by base.
- The process stops after a while because the noise level is too high.
- Results from sequencing are many sequence pieces. The lengths vary, usually a few thousands from Sanger, and several hundreds from NGS.
- The sequence pieces are called "reads" for NGS data.

Single-end vs. paired-end sequencing

- Sequence one or both ends of the DNA segments.
- Single-end sequencing: sequence one end of the DNA segment.
- Paired-end sequencing: sequence both ends of a DNA segments.
 - Result reads are "paired", separated by certain length (the length of the DNA segments, usually a few hundred bps).
 - Paired-end data can be used as single-end, but contain extra information which is useful in some cases, e.g., detecting structural variations in the genome.
 - Modeling technique is more complicated.

Applications of Second-generation sequencing

Applications

- NGS has a wide range of applications.
 - DNA-seq: sequence genomic DNA.
 - RNA-seq: sequence RNA products.
 - ChIP-seq: detect protein-DNA interaction sites.
 - Bisulfite sequencing (BS-seq): measure DNA methylation strengths.
 - A lot of others.
- Basically replaced microarrays with better data: greater dynamic range and higher signal-to-noise ratios.

Technology	Brief description
ChIP-seq	Locate protein-DNA interaction or histone modification sites.
CLIP-seq	Map protein-RNA binding sites
RNA-seq	Quantify expression
SAGE-seq	Quantify expression
RIP-seq	capture TF-bound transcripts
GRO-seq	evaluate promoter-proximal pausing
BS-seq	Profile DNA methylation patterns
MeDIP-seq	Profile DNA methylation patterns
TAB-seq	Profile DNA hydroxyl-methylation patterns
MIRA-seq	Profile DNA methylation patterns
ChiRP-seq	Map IncRNA occupancy
DNase-seq	Identify regulatory regions
FAIRE-seq	Identify regulatory regions
FRT-seq	Quantify expression
Repli-seq	Assess DNA replication timing
MNase-seq	Identify nucleosome position
Hi-C	Infer 3D genome organization
ChIA-PET	Detect long distance chromosome interactions
4C-seq	Detect long distance chromosome interaction
Sono-seq	Map open-chromatin sites
NET-seq	determine in vivo position of all active RNAP complexes.
NA-seq	Map Nuclease-Accessible Sites

DNA-seq

- Sequence the untreated genomic DNA.
 - Obtain DNA from cells, cut into small pieces then sequence the segments.

Goals:

- Compare with the reference genome and look for genetic variants:
 - Single nucleotide polymorphisms (SNPs)
 - Insertions/deletions (indels),
 - Copy number variations (CNVs)
 - Other structural variations (gene fusion, etc.).
- De novo assembly of a new genome.

Variations of DNA-seq

- Targeted sequencing, e.g., exome sequencing.
 - Sequence the genomic DNA at targeted genomic regions.
 - Cheaper than whole genome DNA-seq, so that money can be spent to get bigger sample size (more individuals).
 - The targeted genomic regions need to be "captured" first using technologies like microarrays.
- Metagenomic sequencing.
 - Sequence the DNA of a mixture of species, mostly microbes, in order to understand the microbial environments.
 - The goal is to determine number of species, their genome and proportions in the population.
 - De novo assembly is required. But the number and proportions of species are unknown, so it poses challenge to assembly.

RNA-seq

 Sequence the "transcriptome": the set of RNA molecules.

Goals:

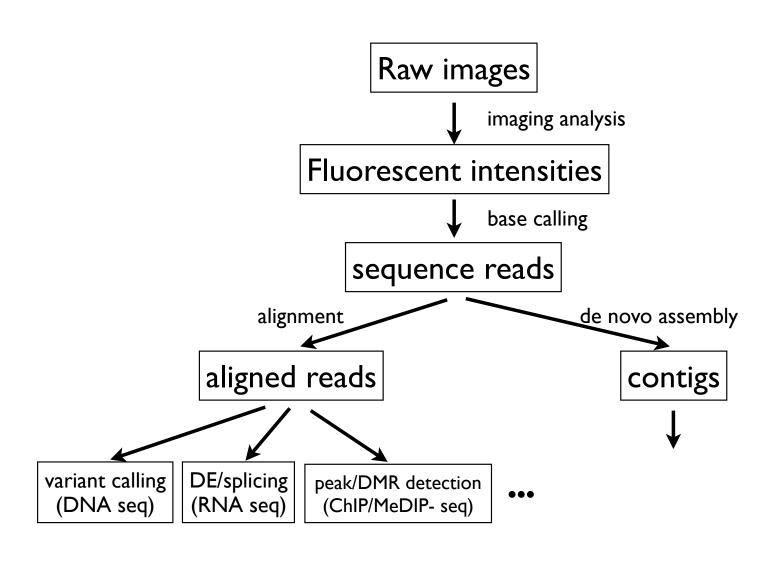
- Catalogue RNA products.
- Determine transcriptional structures: alternative splicing, gene fusion, etc.
- Quantify gene expression: the sequencing version of gene expression microarray.

ChIP-seq

- Chromatin-Immunoprecipitation (ChIP) followed by sequencing (seq): sequencing version of ChIP-chip.
- Used to detect locations of certain "events" on the genome:
 - Transcription factor binding.
 - DNA methylations and histone modifications.
- A type of "captured" sequencing. ChIP step is to capture genomic regions of interest.
- Similar technologies: MeDIP-seq, hMe-seal, etc.

Second-generation sequencing data analyses

Workflow of second generation sequencing data analysis



Imaging analysis

- Extract intensity values from images.
 - On Illumina and SOLiD systems, there are four images per cycle, one for a nucleotide/color.
- Similar to that in microarrays.
- Involves many statistical methods to extract signals from noisy data.
- Results of the imaging analysis: a 3-dimensional matrix: nreads x 4 x nbases.

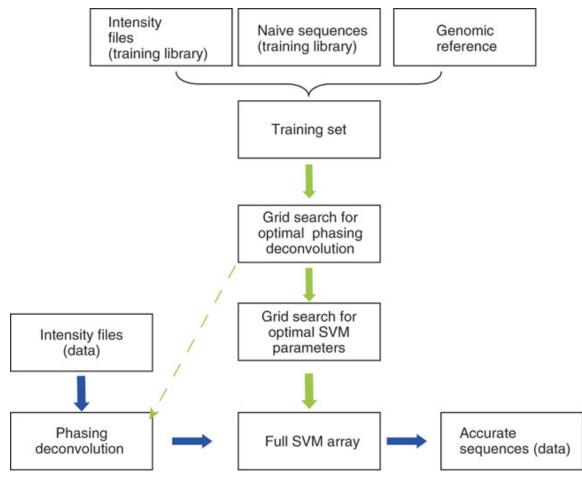
Base calling

- For each read, at each position, convert four fluorescent intensities (continuous) into a base or color (categorical).
- It's a classification problem.

```
Basel Base2 Base3 Base4 Base5
A 0.290 0.046 0.014 0.026 0.010
C 0.014 0.654 0.132 0.803 0.006
G 0.062 0.009 0.001 0.016 0.712
T 0.016 0.010 0.455 0.046 0.768
```



Example of base calling method: Alta-Cyclic for Illumina data



The training process (green arrows) starts with creation of the training set, beginning with sequences generated by the standard Illumina pipeline, by linking intensity reads and a corresponding genome sequence (the 'correct' sequence). Then, two grid searches are used to optimize the parameters to call the bases. After optimization, a final SVM array is created, each of which corresponds to a cycle. In the base-calling stage (blue arrows), the intensity files of the desired library undergo deconvolution to correct for phasing noise using the optimized values and are sent for classification with the SVM array. The output is processed, and sequences and quality scores are reported.

Raw sequence reads from second generation sequencing after base calling

- Large text file (millions of lines) with simple format.
 - Most frequently used: fastq format

BCBB@AB@1ABBBBBBAAB?BBBBBBAB<A?AA>BB@?1ABBA@BBBA@;B>>:

Sequence alignment and assembly

- Sequence a known genome? --- Alignment
 - Use the known genome (called "reference genome") as a blue print.
 - Determine where each read is located in the reference genome.
- Sequence a whole new genome? --- Assembly
 - New genome: a species with unknown genome, or the genome is believed to be very different from reference (e.g., cancer).
 - Basically the short reads are "stitched" together to form long sequences called "contigs".
 - Overlaps among sequence reads are required, so it needs a lot of reads (deep coverage).
 - More computationally intensive.

Alignment

- Need: sequence reads file and a reference genome.
- It is basically a string search problem: where is the short (50-letter) string located within the reference string of 3 billion letters.
- Brute-force searching is okay for a single read, but computationally infeasible to alignment millions of reads.
- Clever algorithms are needed to preprocess the reference genome (indexing), which is beyond the scope of this class.

Popular general alignment software

- Bowtie: fast, but less accurate.
- BWA (Burrows-Wheeler Aligner): same algorithm as bowtie, but allow gaps in alignments.
 - about 5-10 times slower than bowtie, but provide better results especially for paired end data.
- Maq (Mapping and Assembly with Qualities): with SNP calling capabilities.
- ELAND: Illumina's commercial software.
- A lot of others. See
 <u>http://en.wikipedia.org/wiki/List_of_sequence_alignment_software</u> for more details.

Other technology-specific alignment software

- RNA-seq:
 - Tophat, STAR
 - Pseudoaligner: Salmon, kallisto. Very fast.
- Bisulfite sequencing:
 - Bismark
 - BSMAP
 - Merman

bowtie

Bowtie is an ultrafast, memory-efficient short read aligner. It aligns short DNA sequences (reads) to the human genome at a rate of over 25 million 35-bp reads per hour. Bowtie indexes the genome with a Burrows-Wheeler index to keep its memory footprint small: typically about 2.2 GB for the human genome (2.9 GB for paired-end).

Use bowtie: build alignment index

- Alignment index files are built based on reference genome (can be download as text files from UCSC).
- Note that pre-built indexes for many genomes are available from bowtie page, check that before building your own index.
- Command example for Human hg18 genome. Assume we have the hg18 sequence file ready called hg18.fa:

```
bowtie-build hg18.fa hg18
```

- Results: several ebwt files.
- Tips: the index files can be stored in a common place and shared among colleagues.

Use bowtie: alignment

Test whether it works:

bowtie -c hg18 GGTATATGCACAAAATGAGATGCTTGCTTA

Align a read files

bowtie -v 3 -f hg18 reads.fa reads.map

bowtie: commonly used parameters

Input file format:

- -q: query input files are FASTQ .fq/.fastq (default)
- -f: query input files are (multi-)FASTA .fa/.mfa
- -r: query input files are raw one-sequence-per-line

Aligment:

- -v: allowing v mismatches.
- -5: ignoring some based from 5' end.
- -3: ignoring some based from 3' end.

• Output format:

- S: output in SAM (sequence alignment map)format.
- Example: input is a single fa file, allowing 3 mismatches, ignore 5 bases from 3' end, output in SAM format:

```
bowtie -v 3 -3 5 -S hg18 reads.fa reads.sam
```

Output from bowtie

SAM format

```
:0HD
      VN:1.0 SO:unsorted
@SO
     SN:phage
                 LN:5386
@PG
     ID:Bowtie
                 VN:0.12.7
                            CL: "bowtie -v 3 -f -S phage reads.fa reads.sam"
5 143 428 832 4
                                             0
                                                   0
                                                         GATATTGTAGCATAACGCAACTTGGGAGGTGAGCTT
                                                                                      5 143 984 487
           0
                 phage
                       3948
                            255
                                  36M
                                              0
                                                         GTTTTCATGCCTCCAAATCTTGGAGGCTTTTTTATG
                                                                                      XA:i:0
                                                                                                                        MD:Z:36 NM:i:0
5 143 963 690
                 phage
                       3503
                            255
                                  36M
                                                         GGTATATGCACAAAATGAGATGCTTGCTTATCAACA
                                                                                      XA:i:0 MD:Z:36 NM:i:0
5 143 957 461
                       3903
                            255
                                  36M
                                                         TTGTCTAGGAAATAACCGTCAGGATTGACACCCTCC
                                                                                      MD:Z:36 NM:i:0
           16
                 phage
5 143 808 403
                       4122
                            255
                                  36M
                                                         GATAACCGCATCAAGCTCTTGGAAGAGATTCTGTCT
                                                                                      MD:Z:36 NM:i:0
                 phage
                                                                                                                   XA:i:0
5 143 986 385
                                                         GATGCTGAAGGAACTTGGTAAAATTTATCTGGAGAA
                                                                                      XM:i:0
5 143 981 626
                       1522
                            255
                                  36M
                                                         TCCTCCTGAGACTGAGCTTTCTCGCCAAATGACGAC
                                                                                      XA:i:0 MD:Z:36 NM:i:0
           16
                 phage
5 143 470 717
           16
                       2061
                            255
                                  36M
                                                         ATGCGCCTTCGTATGTTTCTCCTGCTTATCACCTTC
                                                                                      XA:i:0 MD:Z:36 NM:i:0
                 phage
5 143 992 626
                                                                                      GCCCAGAAGGGGCGGTTAAATGGTTTTTGGAGAAAG
                                                                                                                   XM:i:0
5 143 400 771
                 phage
                       3816
                                  36M
                                                         GATATTTTCATGGAATTGATAAAGCTGTTGCCGAT
                                                                                      XA:i:1 MD:Z:14T21
NM:i:1
5 143 962 110
                       5074
                                  36M
                                                         AATGGAACAACTCACTAAAAACCAAGCTGTCGCTAC
                                                                                      XA:i:0 MD:Z:36 NM:i:0
                 phage
5 143 774 100
                       3810
                                  36M
                                                         GTGGTTGATATTTTTCATGGTATTGATAAAGCTGTT
                                                                                      XA:i:0 MD:Z:36 NM:i:0
                 phage
```

Bowtie format

```
4397
                                                    0
5 143 961 681
               phage
                          GCTGCTGAACGCCCTCTTAAGGATATTCGCGATGAG
5 143 996 500
               phage
                     2537
                          GGTTAATGCTGGTAATGGTGGTTTTCTTCATTTCAT
                                                    0
                                                                                    32:G>T
5 143 468 916
                     339
               phage
                          GGATTACTATCTGAGTCCGATGCTGTTCAACCACTA
                                                     5 143 972 467
                     3021
                                                                               0
               phage
                          GTGGCATTCAAGGTGATGTGCTTGCTACCGATAACA
                                                    5 143 953 471
                     5017
                          ATACGTTAACAAAAGTCAGATATGGACCTTGCTGC
                                                    0
               phage
5 143 687 97
               phage
                     1287
                          GACTGTTAACACTACTGGTTATATTGACCATGCCAC
                                                    0
                                                                                     34:G>A
                                                                               0
5 143 620 93
               phage
                     4463
                          GATGAGTGTTCAAGATTGCTGGAGGCCTCCACTATG
                                                     5 143 766 307
                                                                                     6:A>T,10:T>C,27:A>T
               phage
                          GCATTCTAGGCGATGTGCTTGCTACCGTTAACAATA
```

Bioconductor alignment packages

- Rsubread, Rbowtie/Rbowtie2, QuasR, etc.
- Example code for Rsubread:

Once the reads are aligned

- Downstream analyses depend on purpose.
 - We will cover the analyses for RNA-seq, ChIP-seq, and BS-seq in next several lectures.
- Often one wants to manipulating and visualizing the alignment results. There are several useful tools:
 - file manipulating (format conversion, counting, etc.):
 samtools/Rsamtools, BEDTools, bamtools, IGV tools.
 - Visualizing: samtools (text version), IGV (Java GUI).

samtools

- samtools provide various utilities for manipulating alignments in the SAM format, including sorting, merging, indexing and generating alignments in a per-position format.
- Command line driven, meaning one needs to type command in a terminal window.
 - Installation could be tricky. Needs to install extra tools on Windows or Mac, such as Cygwin and perl on Windows and Xcode on Mac.

Main functionalities:

- view: SAM<->BAM conversion
- sort: sort alignment file
- mpileup: multi-way pileup
- depth: compute the coverage depth
- tview: text alignment viewer
- index: index alignment

samtools: generate sorted, indexed bam files

- BAM file: binary SAM. Smaller file sizes and faster operations.
- To convert from sam to bam:

```
samtools view -bS reads.sam > reads.bam
```

 Sort and index bam file. This sorts the reads by chromosome and position and makes subsequence analysis easier.

```
samtools sort reads.bam reads.sorted samtools index reads.sorted.bam
```

Another useful software: BEDTools

- A set of commands to manipulate BED/GFF/VCF files.
- Conversion tools: pairToBed(BAM), bamToBed, bedToBam, etc.
- Counting tools: coverageBed(BAM), windowBed (BAM)
- Others: sortBed, overlap, etc.

Bioconductor package: Rsamtools

- Provide functions to import BAM files to R.
- There are many tools (samtools, BEDTools, bamtools) available to convert different formats (BED, SAM, fasta, fastq, etc.) to BAM.
- Read alignment results should always be saved in BAM format because they are smaller and faster.

Read in a BAM file

```
> bamFile="reads.sorted.bam"
> bam <- scanBam(bamFile)
> names(bam[[1]])
  [1] "qname" "flag" "rname" "strand" "pos" "qwidth" "mapq" "cigar"
  [9] "mrnm" "mpos" "isize" "seq" "qual"
```

This gives the available information in the BAM file. One can specify what to read in (to save time and memory):

```
> what <-c("rname", "strand", "pos", "qwidth") ## fields to read in
> param <- ScanBamParam(what = what)
> bam <- scanBam(bamFile, param=param)[[1]]
> names(bam)
[1] "rname" "strand" "pos" "qwidth"
> bam$pos[1:10]
[1] 1 2 3 3 4 4 4 4 4 5
> bam$strand[1:10]
[1] + + + - - - - +
Levels: + - *
```

Summarize the read counts

- Remember each aligned read can be treated as a genomic interval. So the results from scanBam can be used to construct an GRanges object (of millions of intervals):
 - > GRange.reads=GRanges(seqnames=Rle(bam\$rname),
 ranges=IRanges(bam\$pos, width=bam\$qwidth))
- Then it becomes very handy, for example, we can:
 - compute genome coverage:
 - > cc=coverage(IRange.reads)
 - count number of reads in intervals (such as genes):
 - > countOverlaps(genes, GRange.reads)

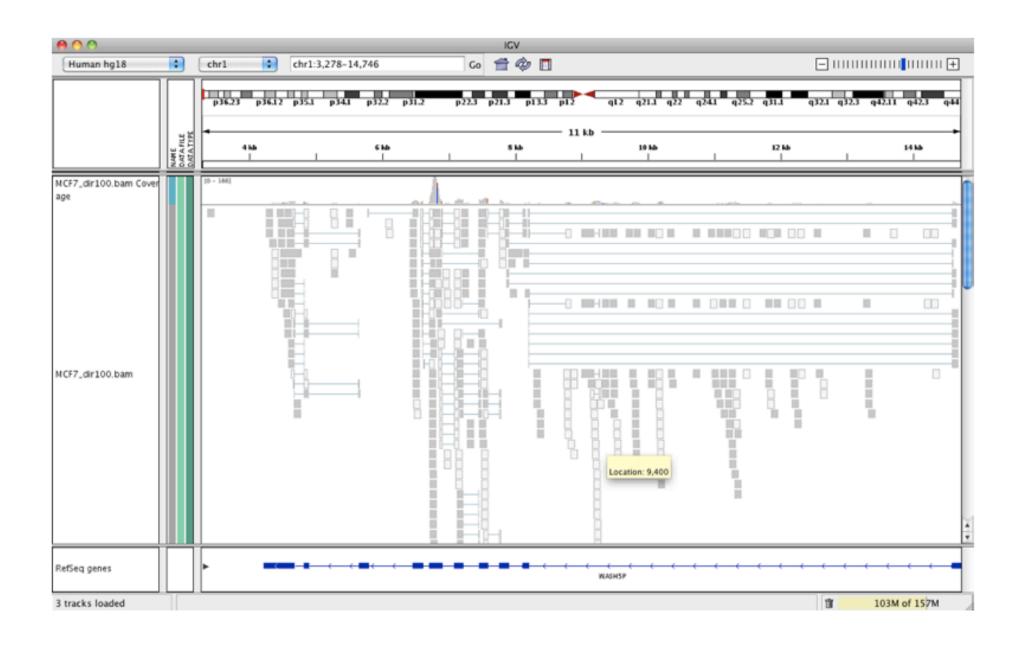
An example: obtaining RNA-seq reads mapped to exons and introns

```
library(GenomicRanges)
library(GenomicFeatures)
library(Rsamtools)
## get gene annotation, and extract exons/introns
refGene.hq18=makeTranscriptDbFromUCSC(genom="hq18",tablename="refGene")
ex=exonsBy(refGene.hq18, "tx")
intr=intronsByTranscript(refGene.hq18)
## read in RNA-seq BAM file
what=c("rname", "strand", "pos", "qwidth")
TSS.counts=NULL
param=ScanBamParam(what = what)
bam=scanBam("RNA-seq.bam", param=param)[[1]]
IRange.reads=GRanges(segnames=Rle(bam$rname),
    ranges=IRanges(bam$pos, width=bam$qwidth))
## obtain counts
counts.exon=countOverlaps(ex, IRange.reads)
counts.intron=countOverlaps(intr, IRange.reads)
```

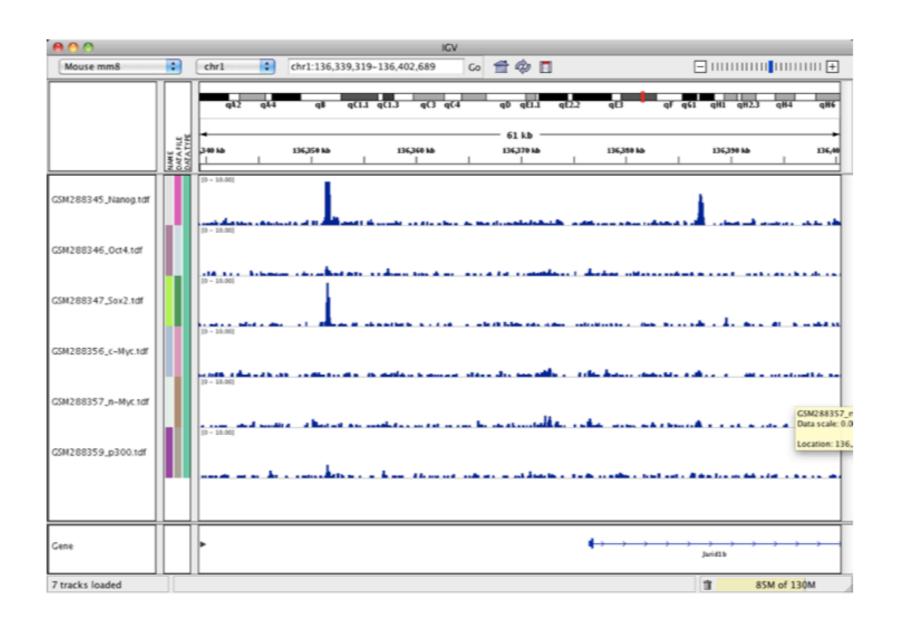
Visualization of sequencing data – Integrated Genome Viewer (IGV)

- "The Integrative Genomics Viewer (IGV) is a highperformance visualization tool for interactive exploration of large, integrated datasets. It supports a wide variety of data types including sequence alignments, microarrays, and genomic annotations."
- Written in Java and runs on all OS.
- Very versatile and fast.
- Ability to connect to data server and display some public data (from ENCODE, broad, etc.)

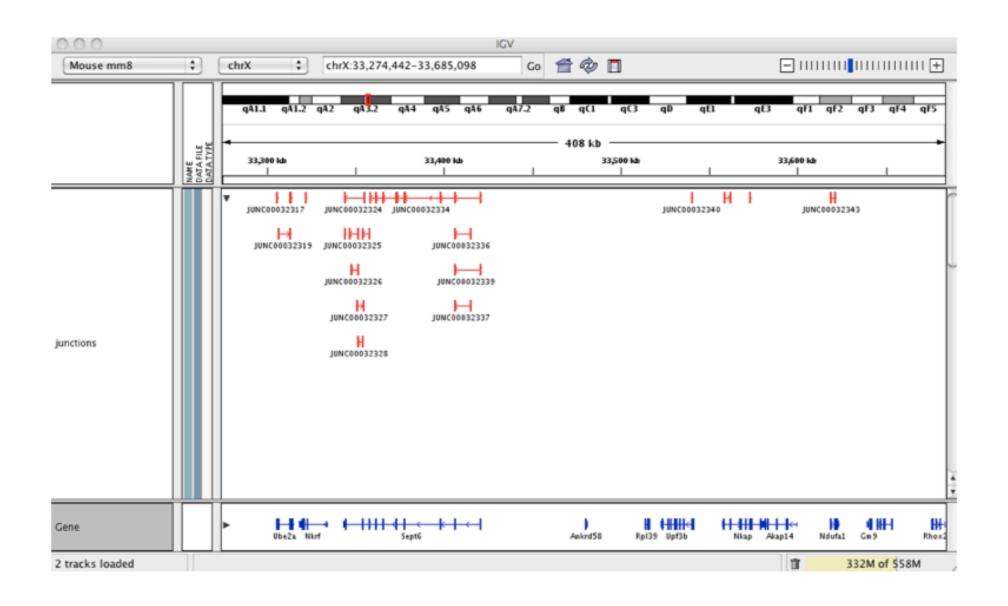
Aligned reads on IGV



ChIP-seq data on IGV



RNA-seq junction reads on IGV



Review

- We've covered
 - Basics of second generation sequencing: technology, data, application
 - Alignment using bowtie
 - Manipulation of alignment results using samtools
 - Import alignment into R using Rsamtools.
 - Visualization using IGV.