

Genome Informatics

Ryan E. Mills, Ph.D.

Department of Computational Medicine & Bioinformatics

Department of Human Genetics

University of Michigan Medical School

Ann Arbor, MI, USA

Genetics and Genomics



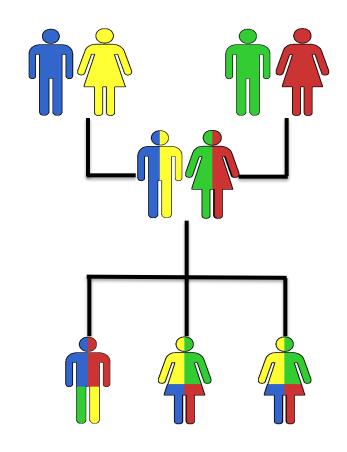
 Genetics is primarily the study of individual genes, mutations within those genes, and their inheritance patterns in order to understand specific traits.

 Genomics expands upon classical genetics and considers aspects of the entire genome, typically using computer aided approaches.



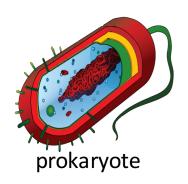


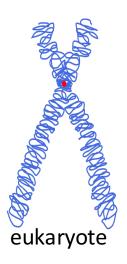
The total genetic material of an organism by which individual traits are encoded, controlled, and ultimately passed on to future generations

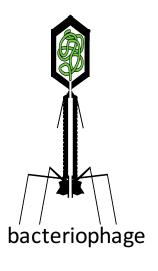




Genomes come in many shapes



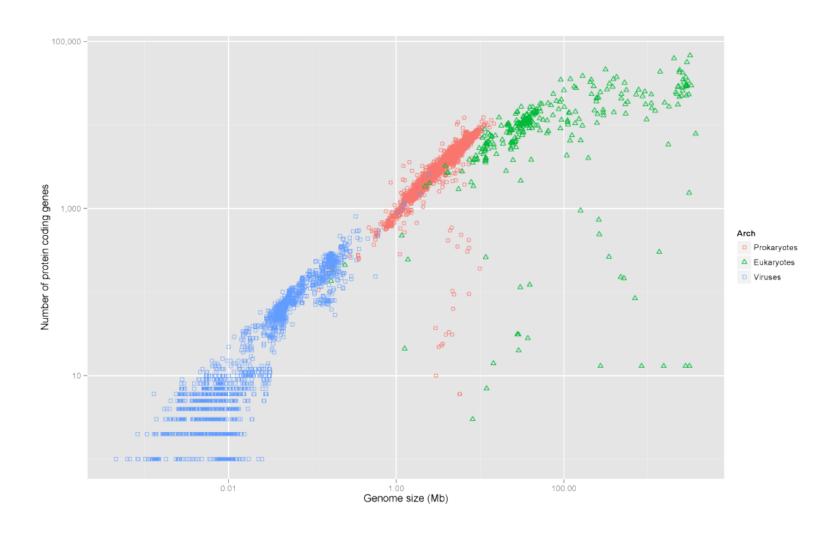




- Primarily DNA, but can be RNA in the case of some viruses
- Some genomes are circular, others linear
- Can be organized into discrete units (chromosomes) or freestanding molecules (plasmids)



Genomes come in many sizes

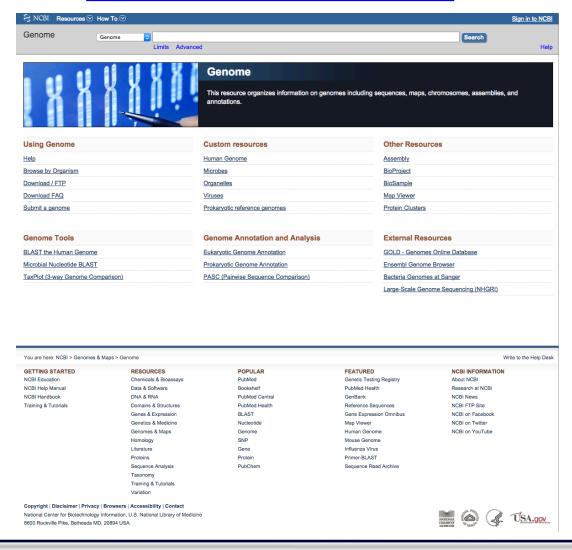




Genome Databases

NCBI Genome:

http://www.ncbi.nlm.nih.gov/genome

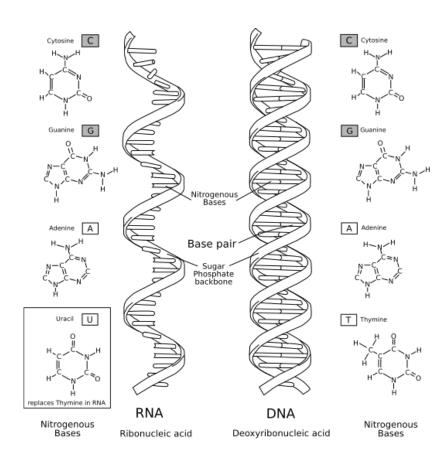


University of Michigan Medical School

Characteristics of Genomes

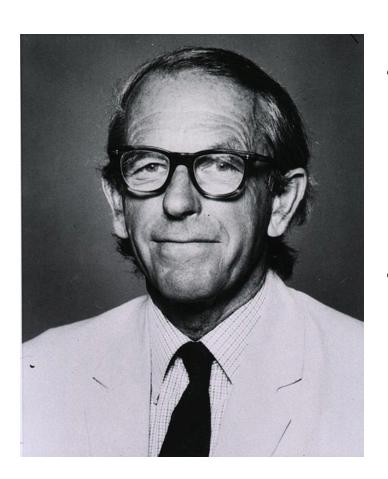
- All genomes are made up of nucleic acids
 - DNA and RNA: Adenine (A), Cytosine (C),
 Guanine (G)
 - DNA Only: Thymine (T)
 - RNA Only: Uracil (U)
- Typically (but not always), DNA genomes are double stranded (double helix) while RNA genomes are single stranded
- Genomes are described as long sequences of nucleic acids, for example:

GGACTTCAGGCAACTGCAACTACCTTAGGA





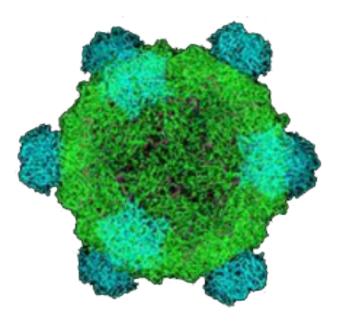
Early Genome Sequencing



- Chain-termination "Sanger" sequencing was developed in 1977 by Frederick Sanger, colloquially referred to as the "Father of Genomics"
- Sequence reads were typically 750-1000 base pairs in length with an error rate of ~1 / 10000 bases

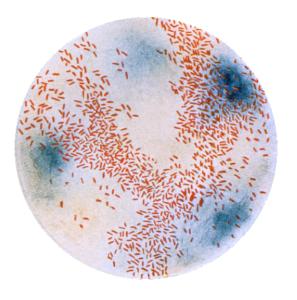


The First Sequenced Genomes



Bacteriophage φ-X174

- Completed in 1977
- 5,386 base pairs, ssDNA
- 11 genes



Haemophilus influenzae

- Completed in 1995
- 1,830,140 base pairs, dsDNA
- 1740 genes

http://en.wikipedia.org/wiki/Phi_X_174

University of Michigan Medical School

The Human Genome Project

- The Human Genome Project (HGP) was an international, public consortium that began in 1990
 - Initiated by James Watson
 - Primarily led by Francis Collins
 - Eventual Cost: \$2.7 Billion
- Celera Genomics was a private corporation that started in 1998
 - Headed by Craig Venter
 - Eventual Cost: \$300 Million
- Both initiatives released initial drafts of the human genome in 2001
 - ~3.2 Billion base pairs, dsDNA
 - 22 autosomes, 2 sex chromosomes
 - ~20,000 genes



University of Michigan Medical School

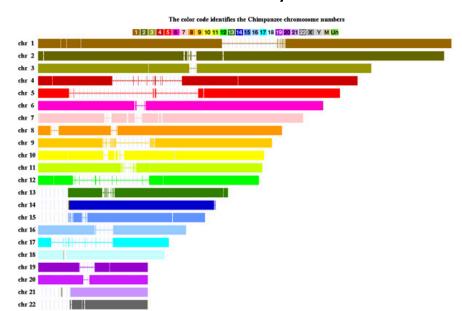
What can we do with a Genome?

- We can compare genomes, both within and between species, to identify regions of variation and of conservation
- We can model genomes, to find interesting patterns reflecting functional characteristics
- We can edit genomes, to add, remove, or modify genes and other regions for adjusting individual traits

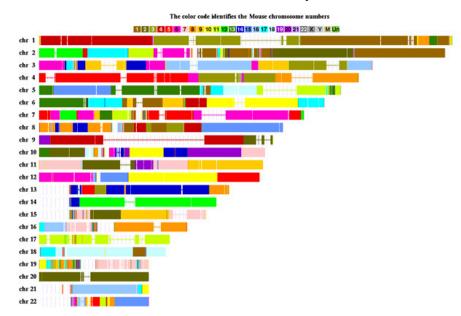


Comparative Genomics

~6-7 million years



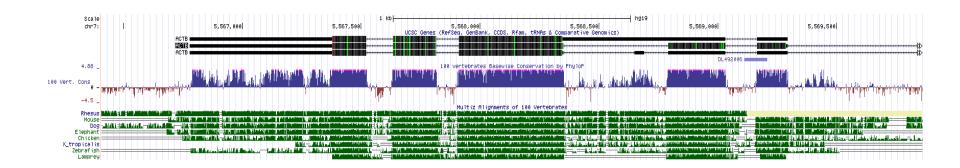
~60-70 million years





Conservation Suggests Function

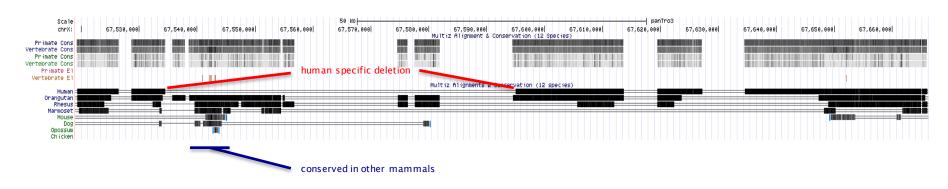
- Functional regions of the genome tend to mutate slower than nonfunctional regions due to selective pressures
- Comparing genomes can therefore indicate segments of high similarity that have remained conserved across species as candidate genes or regulatory regions



University of Michigan Medical School

Conservation Indicates Loss

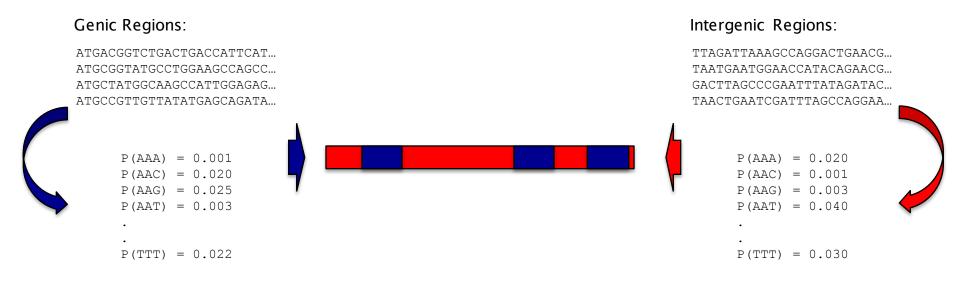
- Comparing genomes allows us to also see what we have lost over evolutionary time
- A model example of this is the loss of "penile spines" in the human lineage due to a human-specific deletion of an enhancer for the androgen receptor gene (McLean et al, Nature, 2011)





Gene Modeling and Prediction

Genomic features such as codon usage patterns can be modeled to identify novel genic regions





Gene Prediction Software

GeneMark:

http://exon.gatech.edu/GeneMark/

GeneMark

A family of gene prediction programs developed at Georgia Institute of Technology, Atlanta, Georgia, USA. What's New: A new algorithm, BRAKER1, an RNA-seq based eukaryotic genome annotation pipeline using GeneMark-ET and AUGUSTUS



Gene Prediction in Bacteria, Archaea, Metagenomes and Metatranscriptomes



Novel genomic sequences can be analyzed either by the self-training program <code>GeneMarkS</code> (sequences longer than 50 kb) or by <code>GeneMark.hmm</code> with <code>Heuristic</code> models. For many species pre-trained model parameters are ready and available through the <code>GeneMark.hmm</code> page. Metagenomic sequences can be analyzed by <code>MetaGeneMark</code>, the program optimized for speed.

Gene Prediction in Eukaryotes



Novel genomes can be analyzed by the program **GeneMark-ES** utilizing unsupervised training. Note that GeneMark-ES has a special mode for analyzing fungal genomes. Recently, we have developed a semi-supervised version of GeneMark-ES, called GeneMark-ET that uses RNA-Seq reads to improve training. For several species pre-trained model parameters are ready and available through the **GeneMark.hmm** page.

Gene Prediction in Transcripts



Sets of assembled eukaryotic transcripts can be analyzed by the modified GeneMarkS algorithm (the set should be large enough to permit self-training). A single transcript can be analyzed by a special version of GeneMark.hmm wi Heuristic models. A new advanced algorithm GeneMarkS-T was developed recently (manuscript sent to publisher); The GeneMarkS-T software (beta version) is available for download

Gene Prediction in Viruses, Phages and Plasmids



Sequences of viruses, phages or plasmids can be analyzed either by the GeneMark.hmm with Heuristic models (if the sequence is shorter than 50 kb) or by the self-training program GeneMarkS.

All the software programs mentioned here are available for download and local installation.

The software of GeneMark line is a part of genome annotation pipelines at NCBI, JGI, Broad Institute as well as the following software packages:

- QUAST: quality assessment tool for genome assemblies
- using GeneMarkS
 MetAMOS : a modular and open source metagenomic assembly and analysis -- using MetaGeneMark
- MAKER2: a eukaryotic genome annotation pipeline -- using GeneMark-ES (along with SNAP and AUGUSTUS)
- BRAKER1 : an RNA-seq based eukaryotic genome annotation pipeline
 -- using GeneMark-ET and AUGUSTUS

For more information see Background and Publications.

Gene Prediction Programs

- GeneMark
 GeneMark.hmm
 GeneMarkS
 Heuristic models
 MetaGeneMark
 Mirror site at NCBI
 GeneMarkS+
 BRAKER1

Information

- PublicationsSelected CitationsBackgroundFAQContact

Downloads

Other Programs

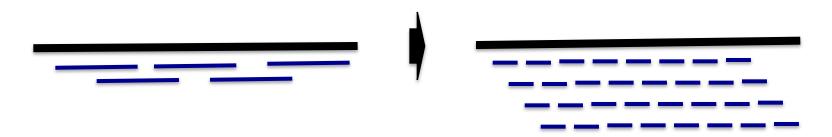
In silico Biology International Conferences

Bioinformatics Studies at Georgia Tech

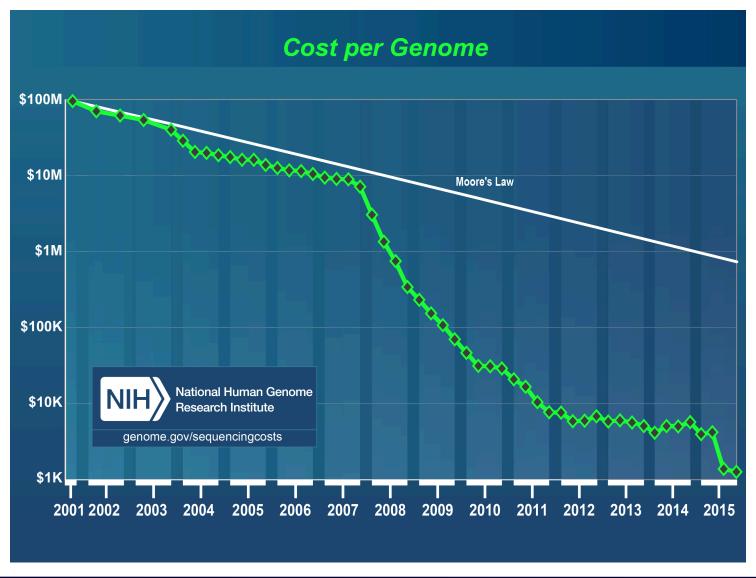
- PhD Program
 Center for Bioinformatics and

Modern Genome Sequencing

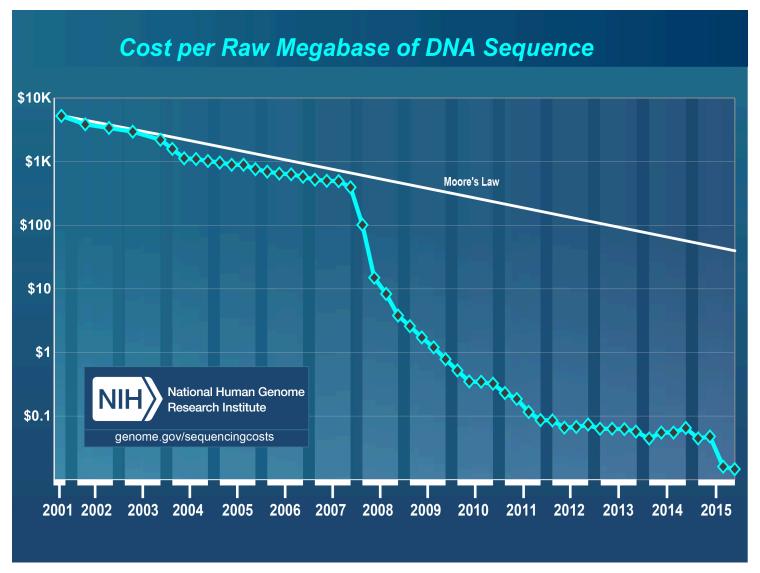
- Next Generation Sequencing (NGS) technologies have resulted in a paradigm shift from long reads at low coverage to short reads at high coverage
- This provides numerous opportunities for new and expanded genomic applications





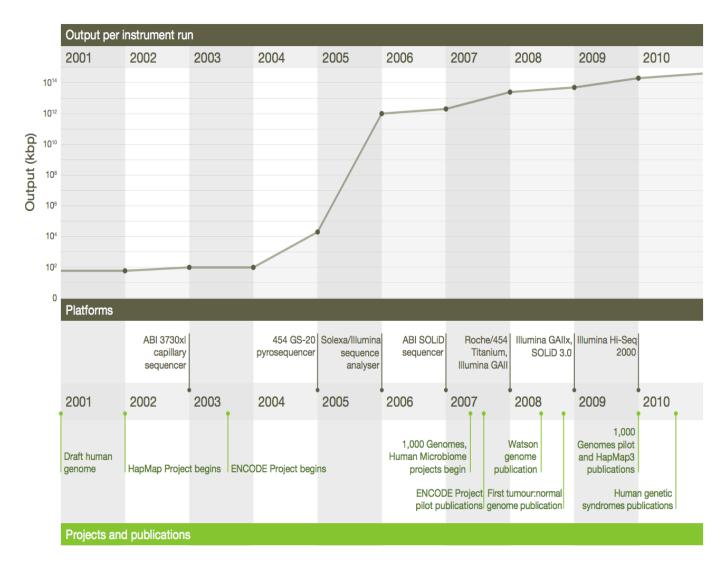








Timeline of Sequencing Capacity



DNA Sequencing Concepts

- Sequencing by Synthesis: Uses a polymerase to incorporate and assess nucleotides to a primer sequence
 - 1 nucleotide at a time
- Sequencing by Ligation: Uses a ligase to attach hybridized sequences to a primer sequence
 - 1 or more nucleotides at a time (e.g. dibase)

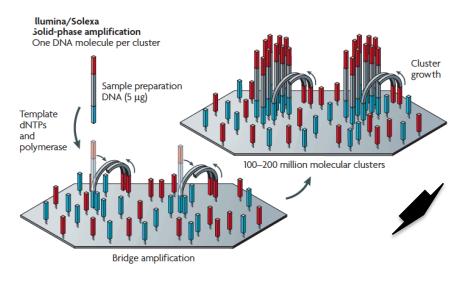


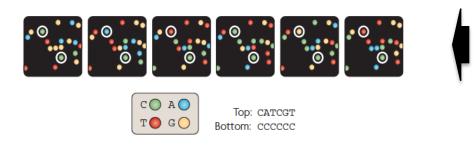
Modern NGS Sequencing Platforms

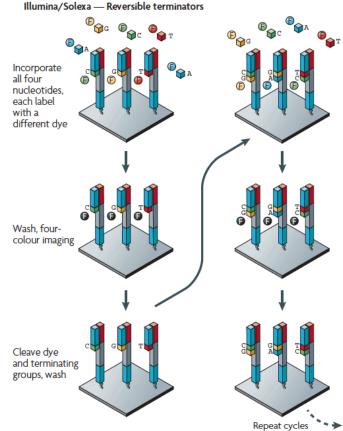
	Roche/454	Life Technologies SOLiD	Illumina Hi-Seq 2000
Library amplification method	emPCR* on bead surface	emPCR* on bead surface	Enzymatic amplification on glass surface
Sequencing method	Polymerase-mediated	Ligase-mediated addition of	Polymerase- mediated
	incorporation of unlabelled	2-base encoded fluorescent	incorporation of end-
	nucleotides	oligonucleotides	blocked fluorescent nucleotides
Detection method	Light emitted from secondary	Fluorescent emission from	Fluorescent emission
	reactions initiated by release of PPi	ligated dye-labelled oligonucleotides	from incorporated dye-labelled nucleotides
Post incorporation method	NA (unlabelled nucleotides are	Chemical cleavage removes	Chemical cleavage of
	added in base-specific fashion, followed by detection)	fluorescent dye and 3' end of oligonucleotide	fluorescent dye and 3' blocking group
Error model	Substitution errors rare, insertion/deletion errors at homopolymers	End of read substitution errors	
Read length	400 bp/variable length mate pairs	75 bp/50+25 bp	150 bp / 100 + 100 bp
(fragment/paired end)			



Illumina – Reversible terminators







(other sequencing platforms summarized at end of slide set)





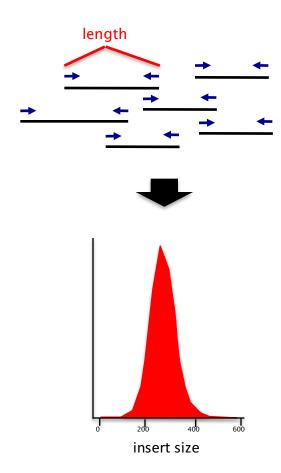




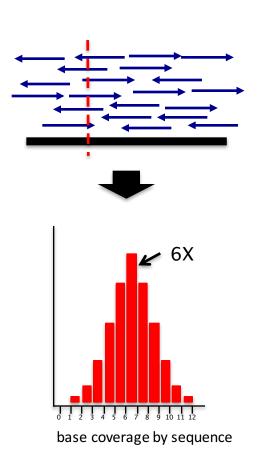


NGS Sequencing Terminology

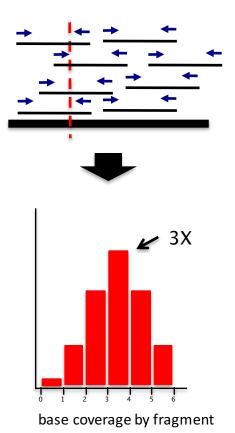
Insert Size



Sequence Coverage



Physical Coverage



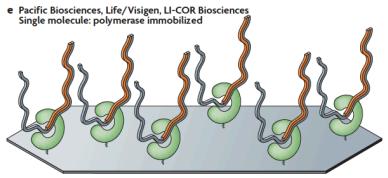
University of Michigan Medical School

Third Generation Sequencing

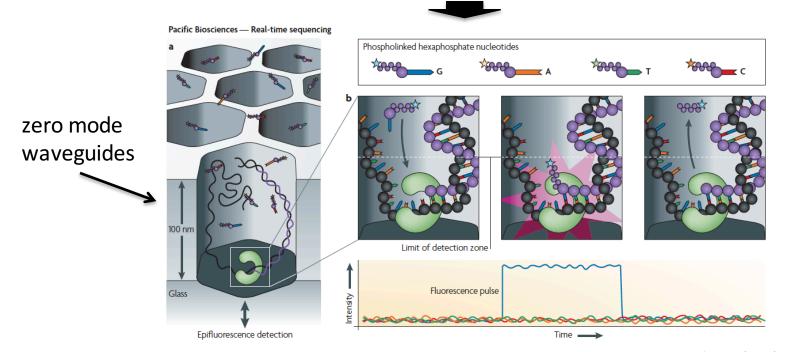
- Currently in transition / development
- Hard to define what "3rd" generation means
- Typical characteristics:
 - Long (1,000bp+) sequence reads
 - Single molecule (no amplification step)
 - Often associated with nanopore technology
 - But not necessarily!



Pacific Biosystems – Real Time Sequencing



Thousands of primed, single-molecule templates

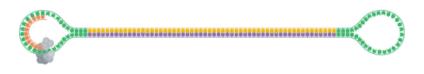


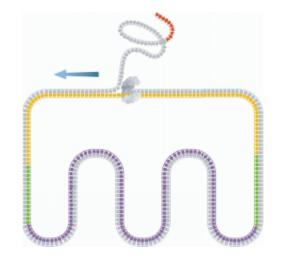


Pacific Biosystems – Circular Consensus

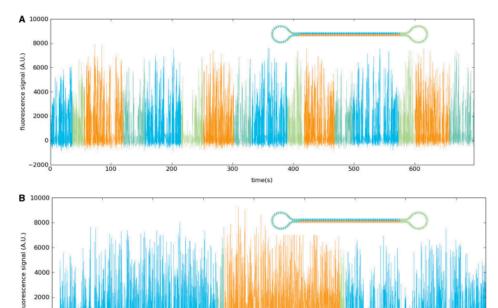
-2000

SMRTbell template





Subread Consensus Sequencing





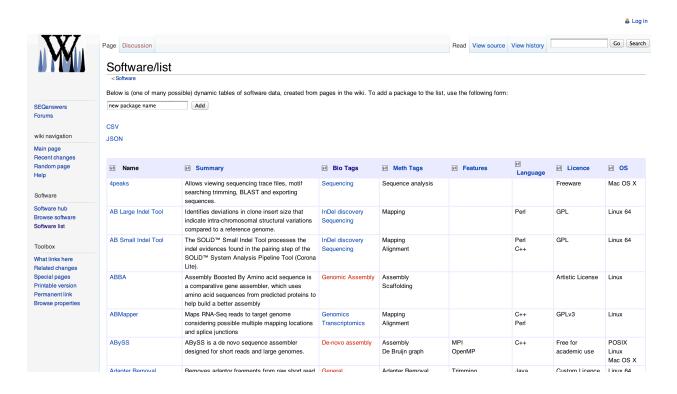
Summary: Generations of DNA Sequencing

	First generation	Second generation ^a	Third generation ^a
Fundamental technology	Size-separation of specifically end- labeled DNA fragments, produced by SBS or degradation	Wash-and-scan SBS	SBS, by degradation, or direct physical inspection of the DNA molecule
Resolution	Averaged across many copies of the DNA molecule being sequenced	Averaged across many copies of the DNA molecule being sequenced	Single-molecule resolution
Current raw read accuracy	High	High	Moderate
Current read length	Moderate (800–1000 bp)	Short, generally much shorter than Sanger sequencing	Long, 1000 bp and longer in commercial systems
Current throughput	Low	High	Moderate
Current cost	High cost per base	Low cost per base	Low-to-moderate cost per base
	Low cost per run	High cost per run	Low cost per run
RNA-sequencing method	cDNA sequencing	cDNA sequencing	Direct RNA sequencing and cDNA sequencing
Time from start of sequencing reaction to result	Hours	Days	Hours
Sample preparation	Moderately complex, PCR amplification not required	Complex, PCR amplification required	Ranges from complex to very simple depending on technology
Data analysis	Routine	Complex because of large data volumes and because short reads complicate assembly and alignment algorithms	Complex because of large data volumes and because technologies yield new types of information and new signal processing challenges
Primary results	Base calls with quality values	Base calls with quality values	Base calls with quality values, potentially other base information such as kinetics





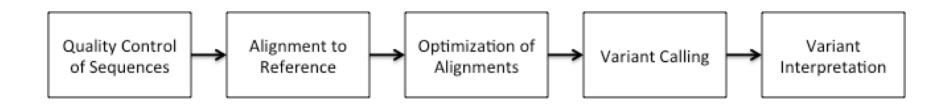
A good repository of analysis software can be found at http://seganswers.com/wiki/Software/list



Generic Workflow for NGS



- There are many different ways to analyze sequences generated from NGS, depending on the specific question you are investigating
- For the analysis of genomic sequence data, a typical (if generic) approach is as follows



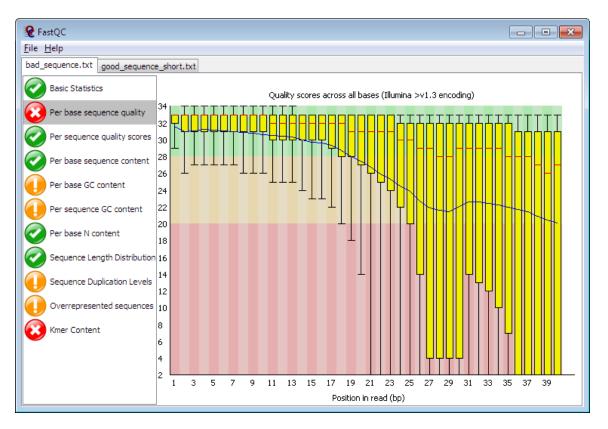


- Quality checks of raw sequence data are very important
- Common problems can include:
 - Sample mix-up
 - Sample contamination
 - Machine interruption
 - DNA quality
- It is crucial that investigators examine their sequences upon first receipt before any downstream analysis is conducted



FASTQC is one approach which provides a visual interpretation of the raw sequence reads

http://www.bioinformatics.babraham.ac.uk/projects/fastqc/





Sequence Alignment

- Once sequence quality has been assessed, the next step is to align the sequence to a reference genome
- There are many distinct tools for doing this; which one you choose is often a reflection of your specific experiment and personal preference

BWA BarraCUDA
Bowtie CASHx
SOAP2 GSNAP
Novoalign Mosiak
mr/mrsFast Stampy
Eland SHRiMP
Blat SeqMap

SLIDER

Bfast

RMAP SSAHA etc



- Sequence Alignment/Map (SAM) format is the almost-universal sequence alignment format for NGS
 - binary version is BAM
- It consists of a header section (lines start with '@') and an alignment section
- The official specification can be found here:
 - http://samtools.sourceforge.net/SAM1.pdf



Example SAM File

Header section

@HD	VN:1.0	SO:coordinate							
@SQ	SN:1	LN:249250621	AS:NCBI37	UR:file:/data	/local/ref/GATK/hur	man_g1k_v37.fasta	M5:1b22b98cd	eb 4a 9 30 4 cb 5d 4 80 2 6a 85 1 28	
@SQ	SN:2	LN:243199373	AS:NCBI37	UR:file:/data	/local/ref/GATK/hur	nan g1k v37.fasta	M5:a0d9851da	00 40 0 de c 10 98 a 92 5 5a c7 1 2e	
@ S Q	SN:3	LN:198022430	AS:NCBI37	UR:file:/data	/local/ref/GATK/hur	nan g1k v37.fasta	M5:fdfd81184	9c c2 f ad e bc 92 9 bb 9 25 90 2 e5	
@RG	ID:UM0098:1	PL:ILLUMINA	PU:HWUSI-EAS1707-	-615LHAAXX-L001	LB:80	DT:2010-05-05T20	:00:00-0400	SM:SD37743	CN:UMCORE
@RG	ID:UM0098:2	PL:ILLUMINA	PU:HWUSI-EAS1707-	-615LHAAXX-L002	LB:80	DT:2010-05-05T20	:00:00-0400	SM:SD37743	CN:UMCORE
a n.c	TD.brrn	77N7 • O E /							

Alignment section

1:497:R:-272+13M1	7D24M	113	1	497	37	37M	15	100338662	0
	CGGGTCTGACCTGAGGA	GAACT GT GCT C CG CC T TC A G	0;==-==9;>>>>=>>>	>>>>>>>>	XT:A:U	NM:i:0	SM:i:37	AM:i:0	X0:i:1
	X1:i:0	XM:i:0	XO:i:0	XG:i:0	MD: Z: 37				
19:20389:F:275+18	M2D19M	99	1	17644	0	37M	=	17919	314
TATGACTGCTAATAATACCTACACATGTTAGAACCAT		>>>>>>>>>>>		RG:Z:UM0098:1	XT:A:R	NM:i:0	SM:i:0	AM:i:0	
	X0:i:4	X1:i:0	XM:i:0	XO:i:0	XG:i:0	MD:Z:37			
19:20389:F:275+18M2D19M 147		1	17919	0	18M2D19M	=	17644	-314	
	GTAGTACCAACTGTAAG	TCCTTATCTTCATACTTTGT	;44999;499<8<8<<<8	<<><<><	XT:A:R	NM:i:2	SM:i:0	AM:i:0	X0:i:4
	X1:i:0	XM:i:0	XO:i:1	XG:i:2	MD:Z:18^CA19				
9:21597+10M2I25M:	R: -209	83	1	21678	0	8M2I27M	=	21469	-244
	CACCACATCACATATAC	CAAGCCTGGCTGTGTCTTCT	<;9<<5><<<>><	<><>>>9>>><>	XT:A:R	NM:i:2	SM:i:0	AM:i:0	X0:i:5
	X1:i:0	XM:i:0	XO:i:1	XG:i:2	MD:Z:35				

SAM Utilities



- Samtools is a common toolkit for analyzing and manipulating files in SAM/BAM format
 - http://samtools.sourceforge.net/
- Picard is a another set of utilities that can used to manipulate and modify SAM files
 - http://picard.sourceforge.net/
- These can be used for viewing, parsing, sorting, and filtering SAM files as well as adding new information (e.g. Read Groups)

Post-alignment Optimization



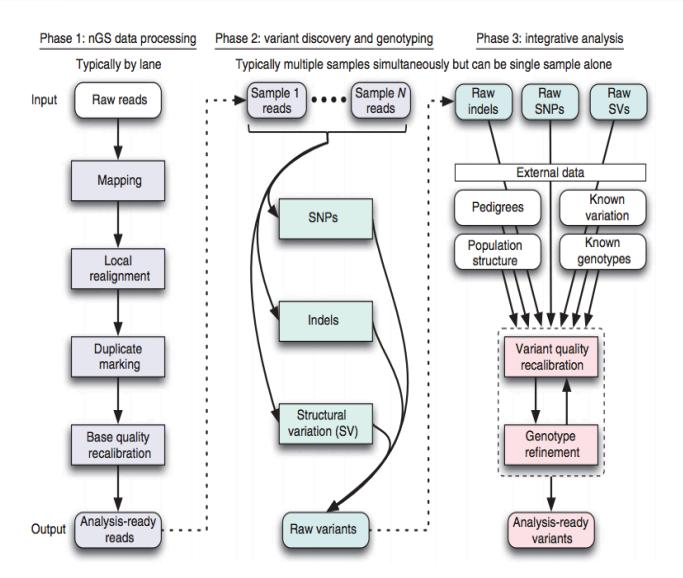
- A lot of research has been conducted to improve and optimize sequence alignments
- However, genomic sequences are very complex and by their very nature can preclude the ability to accurately determine where a sequence read originated
- New tools and approaches have been developed to help address these shortcomings and improve our overall ability to interpret the alignments

Genome Analysis Toolkit (GATK)

- Developed in part to aid in the analysis of 1000 Genomes Project data
- Includes many tools for manipulating, filtering, and utilizing next generation sequence data
- http://www.broadinstitute.org/gatk/



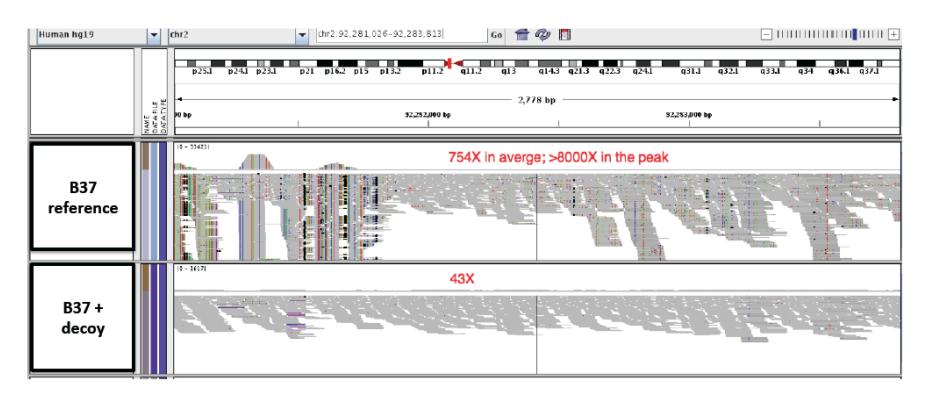
GATK Workflow



DePristo MA et al. (2011) Nat. Genet. 43, pp. 491-498



One approach is to allow reads to map to a "decoy" alignment of extra-chromosomal or unassembled sequences





Realignment around INDELs

- Insertions and deletions in samples can cause misalignments, resulting in false variant detection
- By identifying regions with known INDELs or reads which may have INDEL characteristics and performing multiple sequence alignments, these alignments can be rescued



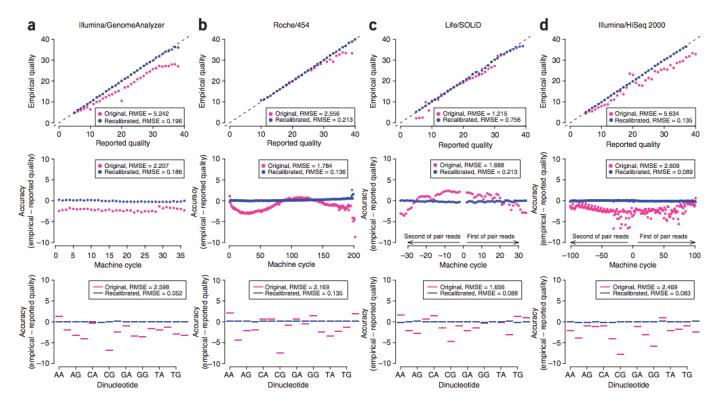


Marking/Removing Duplicate Sequences

- Sequence biases can arise from PCR amplification effects during the construction of the library
- There can also be optical duplicates which occur when sequences from one cluster are accidentally identified as arising as well from adjacent clusters
- Both Picard (MarkDuplicates) and Samtools (rmdup) have utilities for addressing one or both of these issues

Base Recalibration

- Provides empirically accurate base quality scores for each base in every read
- Also corrects for error covariates like machine cycle and dinucleotide content



Population Scale Analysis

We can now begin to assess genetic differences on a very large scale, both as naturally occurring variation in human and non-human populations as well somatically within tumors









"Variety's the very spice of life"

-William Cowper, 1785

"Variation is the spice of life"

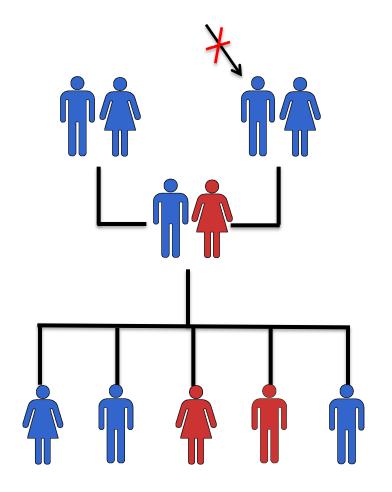
-Kruglyak & Nickerson, 2001

- While the sequencing of the human genome was a great milestone, the DNA from a single person is not representative of the millions of potential differences that can occur between individuals
- These unknown genetic variants could be the cause of many phenotypes such as differing morphology, susceptibility to disease, or be completely benign.



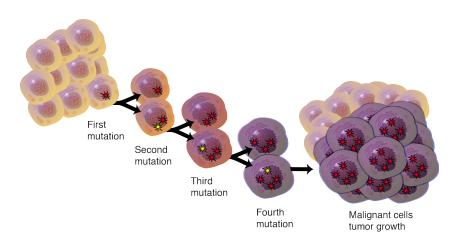


- Mutations in the germline are passed along to offspring and are present in the DNA over every cell
- In animals, these typically occur in meiosis during gamete differentiation





Somatic Variation



- Mutations in non-germline cells that are not passed along to offspring
- Can occur during mitosis or from the environment itself
- Are an integral part in tumor progression and evolution

Mutation vs Polymorphism

- A mutation must persist to some extent within a population to be considered polymorphic
 - >1% frequency is often used
- Germline mutations that are not polymorphic are considered rare variants

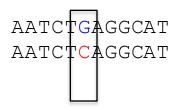
"From the standpoint of the neutral theory, the rare variant alleles are simple those alleles whose frequencies within a species happen to be in a low-frequency range (0,q), whereas polymorphic alleles are those whose frequencies happen to be in the higher-frequency range (q, 1-q), where I arbitrarily take q = 0.01. Both represent a phase of molecular evolution."

-Motoo Kimura

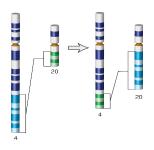


Types of Genomic Variation

- Single Nucleotide Polymorphisms (SNPs) – mutations of one nucleotide to another
- Insertion/Deletion Polymorphisms (INDELs) – small mutations removing or adding one or more nucleotides at a particular locus
- Structural Variation (SVs) medium to large sized rearrangements of chromosomal DNA



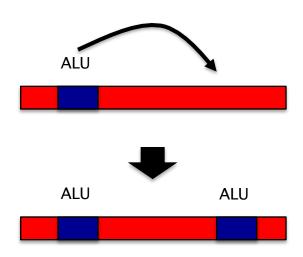






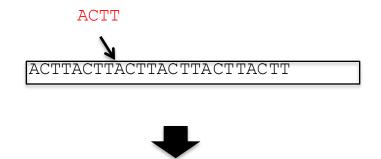
Variant Subtypes: Repetitive Elements

Mobile Elements / Retrotransposons



(in humans, primarily ALU, LINE, and SVA)

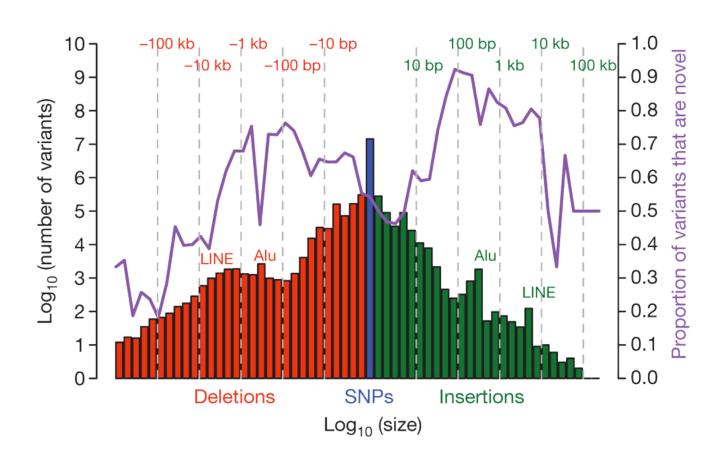
Repeat Expansions



ACTTACTTACTTACTTACTTACTT



Variant Length Distribution



Differences Between Individuals



The average number of genetic differences in the germline between two random humans can be broken down as follows:

- 3,600,000 single nucleotide differences
- 344,000 small insertion and deletions
- 1,000 larger deletion and duplications

Numbers change depending on ancestry!



Discovering Variation: SNPs and INDELs

- Small variants require the use of sequence data to initially be discovered
- Most approaches align sequences to a reference genome to identify differing positions
- The amount of DNA sequenced is proportional to the number of times a region is covered by a sequence read
 - More sequence coverage equates to more support for a candidate variant site



Discovering Variation: SNPs and INDELs

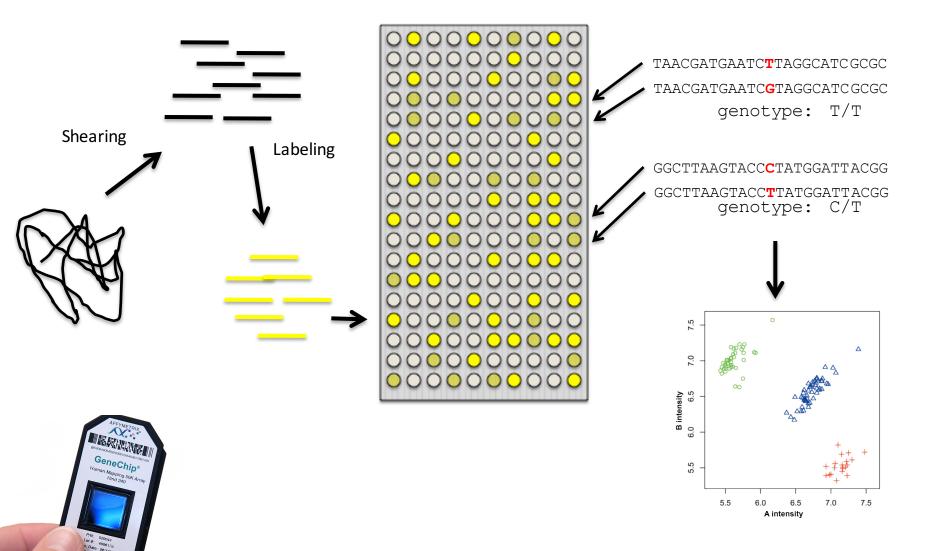
SNP

ATCCTGATTCGGTGAACGTTATCGACGATCCGATCGA sequencing error ATCCTGATTCGGTGAACGTTATCGACGATCGA or genetic variant? CGGTGAACGTTATCGAC GATCCGATCGAACTGTCAGC GGTGAACGTTATCGAC GTTCCGATCGAACTGTCAGCG TGAACGTTATCGACGTTCCGATCGAACTGTCATCGGC TGAACGTTATCGACGTTCCGATCGAACTGTCACCGGC TGAACGTTATCGACGTT CCGATCGAACTGTCAGCGGC GTTAT CGAC GAT C GAT CGAA CTGT CAGC GGCA AGCT TTATCGACGATCCGATCGAACTGTCAGCGGCAAGCT reference genome TTATC GACGATCC GATC GAAC TGTC AGCG GCAA GCT TCGACGATCCGATCGAACTGTCAGCGGCAAGCTGAT ATCCGATCGAACTGTCAGCGGCAAGCTGATCG CGA T TCCGAGCGAACTGTCAGCGGCAAGCTGATCG CGATC TCCGATCGA ACTGTCAGCGGCAAGCTGATCGATCGA GATCGAACTGTCAGCGGCAAGCTGATCG **CGATCGA** sequencing error AACTGTCAGCGGCAAGCTGATCG CGAT CGAT GCTA TGTCA GCGG CAAG CTGA TCGA TCGA TCGA TGCT AG or genetic variant? TCAGCGGCAAGCTGATCGATCGATCGATGCTAGTG

Genotyping Small Variants

- Once discovered, oligonucleotide probes can be generated with each individual allele of a variant of interest
- A large number can then be assessed simultaneously on microarrays to detect which combination of alleles is present in a sample

SNP Microarrays



Maggie Bartlett, Courtesy: National Human Genome Research Institute

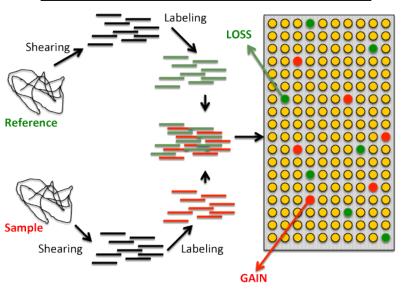


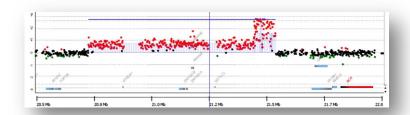
- Structural variants can be discovered by both sequence and microarray approaches
- Microarrays can only detect genomic imbalances, specifically copy number variants (CNVs)
- Sequence based approaches can, in principle, identify all types of structural rearrangements



Microarray-based CNV Discovery

Comparative Genomic Hybridization (CGH)

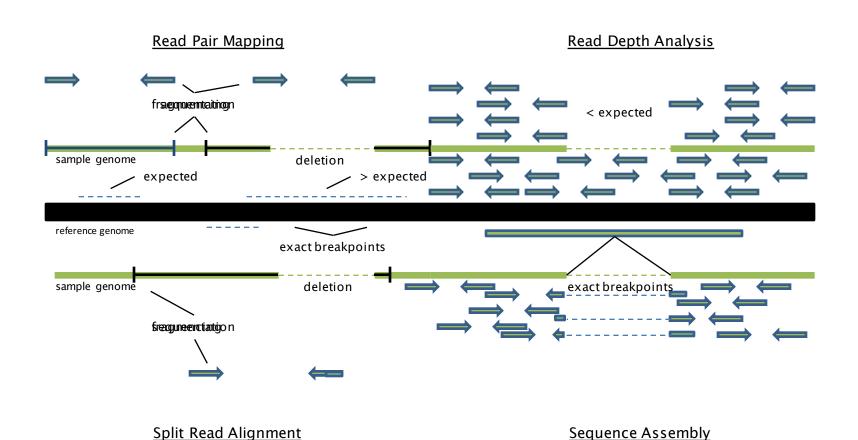








Sequenced-based SV Discovery



Variant Databases and Formats

- dbSNP repository for SNP and small INDELs
 - http://www.ncbi.nlm.nih.gov/SNP/
- VCF variant call format for reporting variation
 - https://github.com/samtools/hts-specs

##fileformat=VCFv4.2

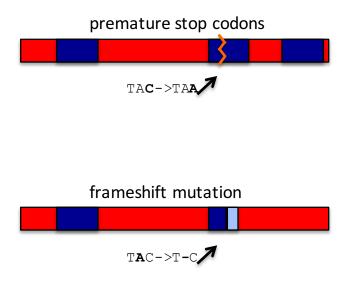
VCF Format Example

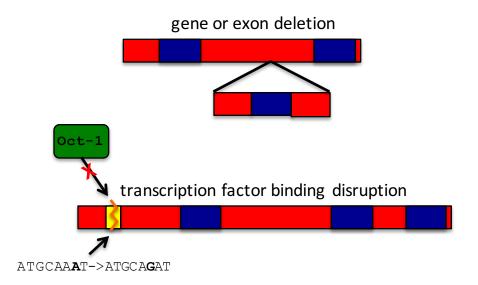
```
##fileDate=20090805
##source=myImputationProgramV3.1
##reference=file:///seq/references/1000GenomesPilot-NCBI36.fasta
##contig=<ID=20,length=62435964,assembly=B36,md5=f126cdf8a6e0c7f379d618ff66beb2da,species="Homo sapiens",taxonomy=x>
##phasing=partial
##INFO=<ID=NS, Number=1, Type=Integer, Description="Number of Samples With Data">
##INFO=<ID=DP, Number=1, Type=Integer, Description="Total Depth">
##INFO=<ID=AF, Number=A, Type=Float, Description="Allele Frequency">
##INFO=<ID=AA, Number=1, Type=String, Description="Ancestral Allele">
##INFO=<ID=DB, Number=0, Type=Flag, Description="dbSNP membership, build 129">
##INFO=<ID=H2, Number=0, Type=Flag, Description="HapMap2 membership">
##FILTER=<ID=q10, Description="Quality below 10">
##FILTER=<ID=s50, Description="Less than 50% of samples have data">
##FORMAT=<ID=GT, Number=1, Type=String, Description="Genotype">
##FORMAT=<ID=GQ, Number=1, Type=Integer, Description="Genotype Quality">
##FORMAT=<ID=DP, Number=1, Type=Integer, Description="Read Depth">
##FORMAT=<ID=HQ, Number=2, Type=Integer, Description="Haplotype Quality">
#CHROM POS
                                ALT
                                        OUAL
                                               FILTER
                                                                                             FORMAT
                                                                                                          NA00001
                                                                                                                          NA00002
                                                                                                                                           NA00003
20
       14370 rs6054257 G
                                        29
                                                PASS
                                                         NS=3; DP=14; AF=0.5; DB; H2
                                                                                            GT:GO:DP:HO 0|0:48:1:51,51 1|0:48:8:51,51 1/1:43:5:.,.
20
       17330
                                Α
                                         3
                                                q10
                                                         NS=3; DP=11; AF=0.017
                                                                                            GT:GQ:DP:HQ 0|0:49:3:58,50 0|1:3:5:65,3
                                                                                                                                           0/0:41:3
      1110696 rs6040355 A
                                                         NS=2; DP=10; AF=0.333, 0.667; AA=T; DB GT: GQ: DP: HQ 1 | 2:21:6:23, 27 2 | 1:2:0:18, 2
                                                                                                                                           2/2:35:4
20
                                G, T
                                         67
                                                PASS
      1230237 .
                                                PASS
                                                         NS=3; DP=13; AA=T
                                                                                             GT:GO:DP:HO 0|0:54:7:56,60 0|0:48:4:51,51 0/0:61:2
20
       1234567 microsat1 GTC
                                                PASS
                                                         NS=3:DP=9:AA=G
                                                                                            GT:GO:DP
                                                                                                          0/1:35:4
                                                                                                                          0/2:17:2
                                                                                                                                           1/1:40:3
```



Impact of Genetic Variation

There are numerous ways genetic variation can exhibit functional effects





Variant Annotation



- Variants are annotated based on their potential functional impact
- For variants falling inside genes, there are a number of software packages that can be used to quickly determine which may have a functional role (missense/nonsense mutations, splice site disruption, etc)
- A few examples are:
 - ANNOVAR (http://www.openbioinformatics.org/annovar/)
 - VAAST (http://www.yandell-lab.org/software/vaast.html)
 - VEP (http://http://grch37.ensembl.org/Homo_sapiens/Tools/VEP)
 - SeattleSeq (http://snp.gs.washington.edu/SeattleSeqAnnotation134/)
 - snpEff (<u>http://snpeff.sourceforge.net/</u>)

Variant Annotation Classes

High Impact

- exon_deleted
- frame shift
- splice_acceptor
- splice donor
- start loss
- stop_gain
- stop_loss
- non_synonymous_start
- transcript_codon_change

Medium Impact

- non_syn_coding
- inframe_codon_gain
- inframe_codon_loss
- inframe codon change
- codon_change_del
- codon_change_ins
- UTR_5_del
- UTR_3_del
- other_splice_variant
- mature miRNA
- regulatory_region
- TF_binding_site
- regulatory_region_ablation
- regulatory_region_amplification
- TFBS_ablation
- TFBS amplification

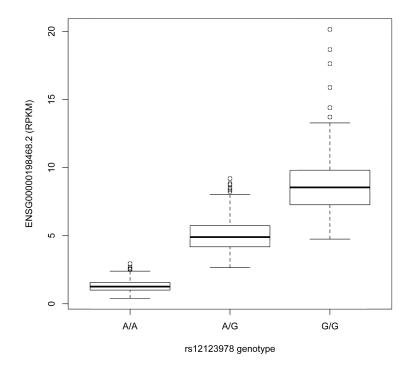
Low Impact

- synonymous stop
- synonymous coding
- UTR 5 prime
- UTR_3_prime
- intron
- CDS
- upstream
- downstream
- intergenic
- intragenic
- gene
- transcript
- exon
- start gain
- synonymous_start
- intron_conserved
- nc transcript
- NMD_transcript
- transcript_codon_change
- incomplete_terminal_codon
- nc_exon
- transcript_ablation
- transcript amplification
- feature elongation
- feature truncation



Variation and Gene Expression

- Expression quantitative trait loci (eQTLs) are regions of the genome that are associated with expression levels of genes
- These regions can be nearby (cis) or far away (trans) from the genes that they affect
- Genetic variants in eQTL regions are typically responsible through changes to regulatory elements





Geuvadis Consortium

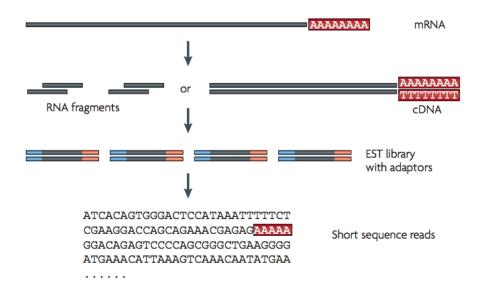
http://www.geuvadis.org/web/geuvadis







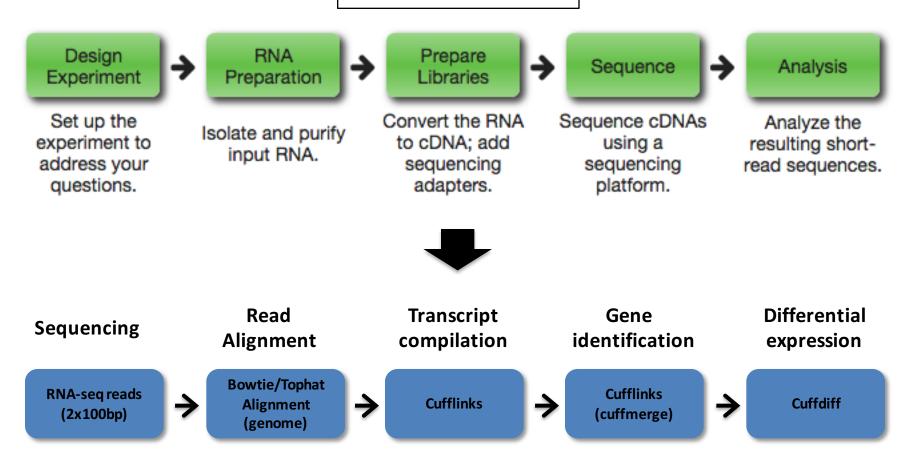
- Uses same technologies as DNA sequencing
- Primary difference is in library preparation
 - RNA converted to cDNA through reverse transcription





RNA-Seq Overview

Sample Preparation



Analysis Pipeline (Tophat)

Types of RNA-Seq Libraries

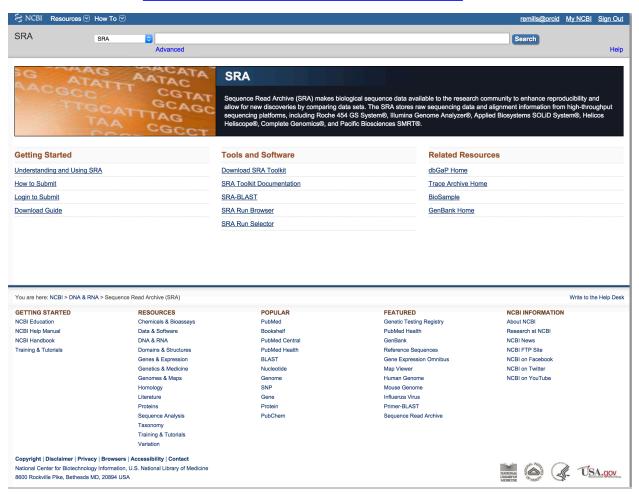
- poly(A) capture utilizes oligo(dT) to prime off of mature mRNA
 - Won't amplify non-coding mRNA (e.g. lincRNAs, miRNAs, etc)
 - Has enrichment biases between 3' and 5' ends due to RT drop-offs
 - Won't work with fragmented RNA
- random hexamer priming primes at random positions along the transcript
 - Will work with fragmented or degraded RNA (e.g. FFPE samples)
 - Removes positional biases of poly(A) capture
 - Requires some type of rRNA removal (e.g. Ribo-Zero) to address its overabundance



DNA- and RNA-Seq Databases

NCBI Short Read Archive (SRA):

http://www.ncbi.nlm.nih.gov/sra

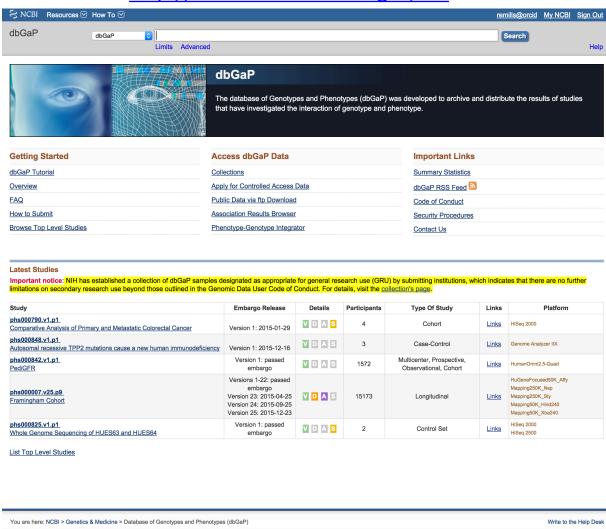




Protected Data - dbGaP

NCBI Database of Genotypes and Phenotypes (dbGaP):

http://www.ncbi.nlm.nih.gov/sra

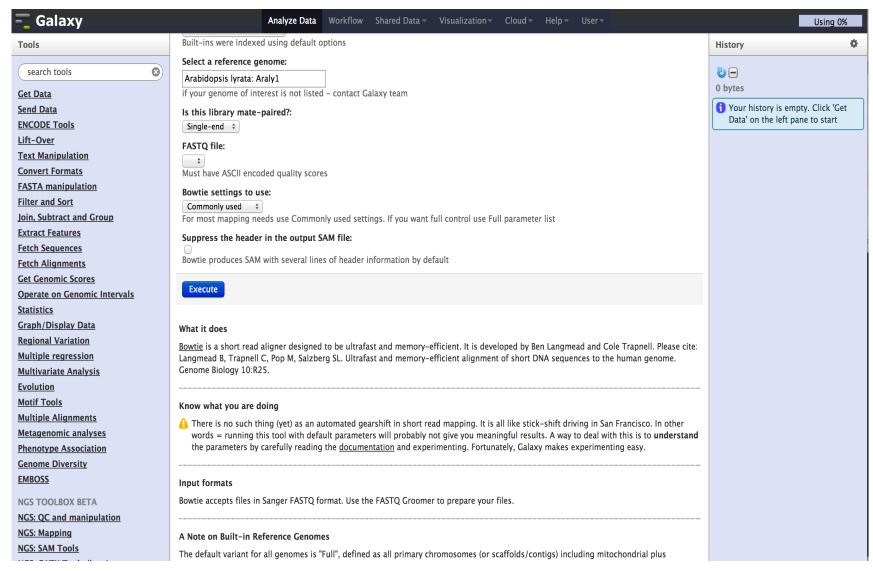




- Galaxy is a useful web-based application for the manipulation of NGS and non-NGS data sets
 - https://main.g2.bx.psu.edu/
- It contains many of the same utilities discussed today, and provides a more standardized approach to analyzing NGS
- However, it requires the uploading of data to their server, which typically precludes its application to protected data sets (e.g. human samples)
- You are also limited to only those tools which have been incorporated into their system



Galaxy Website





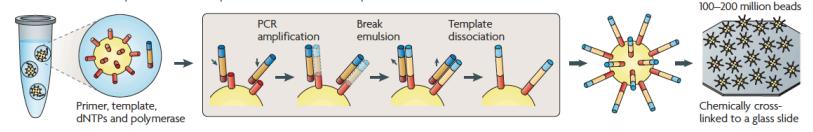
Additional Slides for Reference



Roche 454 - Pyrosequencing

a Roche/454, Life/APG, Polonator Emulsion PCR

One DNA molecule per bead. Clonal amplification to thousands of copies occurs in microreactors in an emulsion





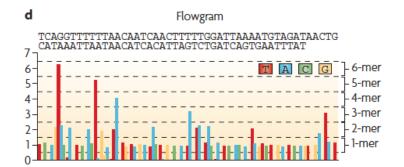


The polymerase PPP and provided in the property of the polymerase PPP.

Sulphurylase APS PPP.

Luciferin

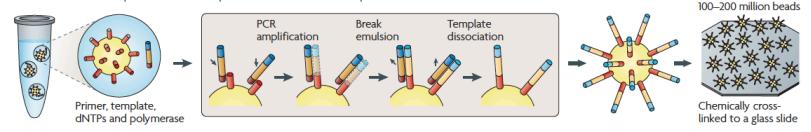
Light and oxyluciferin

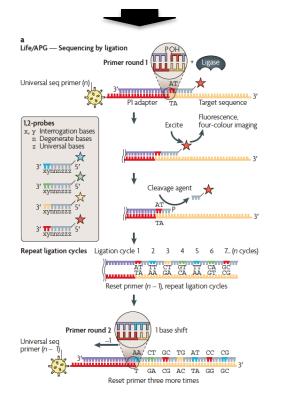


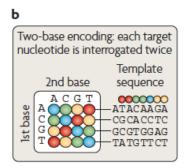
Life Technologies SOLiD – Sequence by Ligation

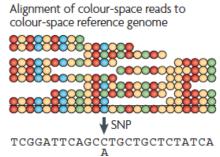
a Roche/454, Life/APG, Polonator Emulsion PCR

One DNA molecule per bead. Clonal amplification to thousands of copies occurs in microreactors in an emulsion





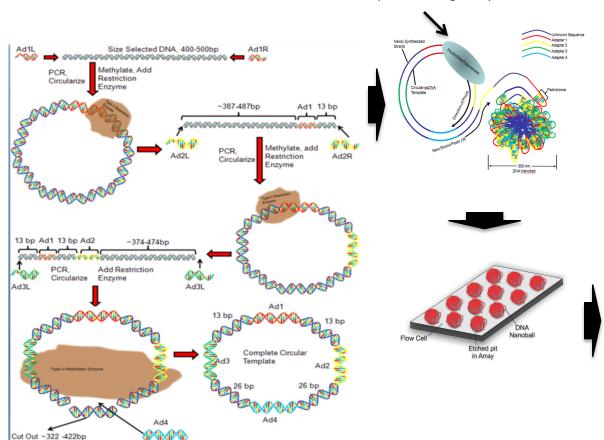


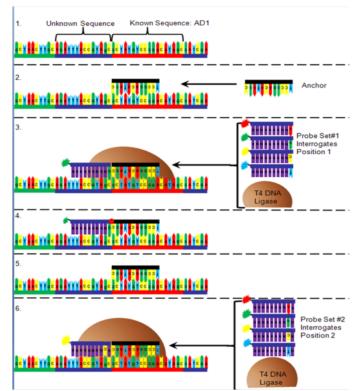




Complete Genomics – Nanoball Sequencing

Has proofreading ability!





"Benchtop" Sequencers

- Lower cost, lower throughput alternative for smaller scale projects
- Currently three significant platforms
 - Roche 454 GS Junior
 - Life Technology Ion Torrent
 - Personal Genome Machine (PGM)
 - Proton
 - Illumina MiSeq

Platform	List price	Approximate cost per run	Minimum throughput (read length)	Run time	Cost/Mb	Mb/h
454 GS Junior Ion Torrent PGM	\$108,000	\$1,100	35 Mb (400 bases)	8 h	\$31	4.4
(314 chip)	\$80,490 ^{a,b}	\$225°	10 Mb (100 bases)	3 h	\$22.5	3.3
(316 chip) (318 chip)		\$425 \$625	100 Mb ^d (100 bases) 1,000 Mb (100 bases)	3 h 3 h	\$4.25 \$0.63	33.3 333.3
MiSeq	\$125,000	\$750	1,500 Mb (2 × 150 bases)	27 h	\$0.5	55.5

Loman, NJ (2012), Nat. Biotech., 5, pp. 434-439



PGM - Ion Semiconductor Sequencing

