

# TODAYS MENU:

- What is a Genome?
  - Genome sequencing and the Human genome project
- What can we do with a Genome?
  - Comparative genomics
- Modern Genome Sequencing
  - Ist, 2nd and 3rd generation sequencing
- Workflow for NGS
  - RNA-Sequencing and discovering variation

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## **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.

## **Germline Variation**

- 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 simply 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

M (1983) Mol. Biol. Evol., 1(1), pp. 84-9

## Types of Genomic Variation

- Single Nucleotide Polymorphisms (SNPs) - mutations of one nucleotide to another
- AATCTGAGGCAT AATCTCAGGCAT

AATCTGAAGGCAT

AGGCAT

• Insertion/Deletion Polymorphisms (INDELs) - small mutations removing or adding one or more nucleotides at a particular locus

(SVs) - medium to large sized rearrangements of chromosomal

Structural Variation

DNA

AATCI

## **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



# 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

# <complex-block>

## **Discovering Variation: SVs**

- 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

## 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 (<u>http://http://grch37.ensembl.org/Homo\_sapiens/Tools/</u> VEP)
  - SeattleSeq (<u>http://snp.gs.washington.edu/</u> <u>SeattleSeqAnnotation134/</u>)
  - snpEff (<u>http://snpeff.sourceforge.net/</u>)

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GEMINI, http://gemini.readthedocs.org

## 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



# Variant Annotation Classes

#### Additional Reference Slides on FASTQ format, ASCII Encoded Base Qualities, FastQC, Alignment and SAM/BAM formats

More fu

# **ASCII Encoded Base Qualities**

@NS500177:196:HFTTTAFXX:1:11101:10916:1458 2:N:0:CGCGGCTG
ACACGACGATGAGGTGACAGTCACGGAGGATAAGATCAATGCCCTCATTAAAGCAGCCGGTGTAA

AAAAAEEEEEEEEE//AEEEAEEEEEEEEEEE/EE/<<<EE/AAEEAEE///EEEEAEEAEA<

- Each sequence base has a corresponding numeric quality score encoded by a single ASCII character typically on the 4th line (see ④ above)
- ASCII characters represent integers between 0 and 127
- Printable ASCII characters range from 33 to 126
- Unfortunately there are 3 quality score formats that you may come across...

# Raw data usually in **FASTQ format**

@NS500177:196:HFTTTAFXX:1:11101:10916:1458 2:N:0:CGCGGCTG
ACACGACGATGAGGTGACAGTCACGGAGGATAAGATCAATGCCCTCATTAAAGCAGCCGGTGTAA

#### Each sequencing "read" consists of 4 lines of data :

- 1 The first line (which always starts with '@') is a unique ID for the sequence that follows
- <sup>2</sup> The second line contains the bases called for the sequenced fragment
- The third line is always a "+" character

+

The forth line contains the quality scores for each base in the sequenced fragment (these are ASCII encoded...)

# Interpreting Base Qualities in R

	ASCII Range	Offset	Score Range		
Sanger, Illumina (Ver > 1.8)	33-126	33	0-93		
Solexa, Ilumina (Ver < 1.3)	59-126	64	5-62		
Illumina (Ver 1.3 -1.7)	64-126	64	0-62		

<pre>&gt; library(seqinr) &gt; library(gtools) &gt; phred &lt;- asc( s2c("DDDDCDEDCDDDDBBDDDCC@") ) - 33 &gt; phred ## D D D D C D E D C D D D D B B D D D C C @ ## 35 35 35 34 35 36 35 34 35 35 35 35 33 33 35 35 35 34 34 31</pre>
> prob <- 10**(-phred/10)

# FastQC Report



# FASTQC

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

<u>http://www.bioinformatics.babraham.ac.uk/projects/fastqc/</u>



# 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 Bowtie SOAP2 Novoalign mr/mrsFast Eland Blat	BarraCUDA CASHx GSNAP Mosiak Stampy SHRiMP SeqMap	RMAP SSAHA etc
Bfast	SLIDER	

# SAM Format

- <u>Sequence Alignment/Map</u> (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



# SAM Utilities

 <u>Samtools</u> is a common toolkit for analyzing and manipulating files in SAM/ BAM format

- http://samtools.sourceforge.net/

• <u>**Picard</u>** is a another set of utilities that can used to manipulate and modify SAM files</u>

- 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)

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/



Hands-on worksheet: <u>http://tinyurl.com/bggn213-L15</u>







## "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)	<mark>8</mark> h	\$31	4.4
(314 chip)	\$80,490 <sup>a,b</sup>	\$225°	10 Mb (100 bases)	3 h	\$22.5	3.3
(316 chip)		\$425	100 Mb <sup>d</sup> (100 bases)	3 h	\$4.25	33.3
(318 chip)		\$625	1,000 Mb (100 bases)	3 h	\$0.63	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		

