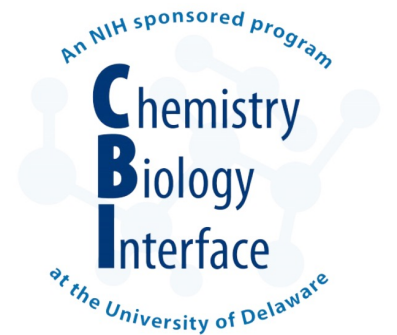


Variant Analysis

FAIR Data Practices for Omics Analysis Workshop

University of Delaware

April 21 (Day 4)



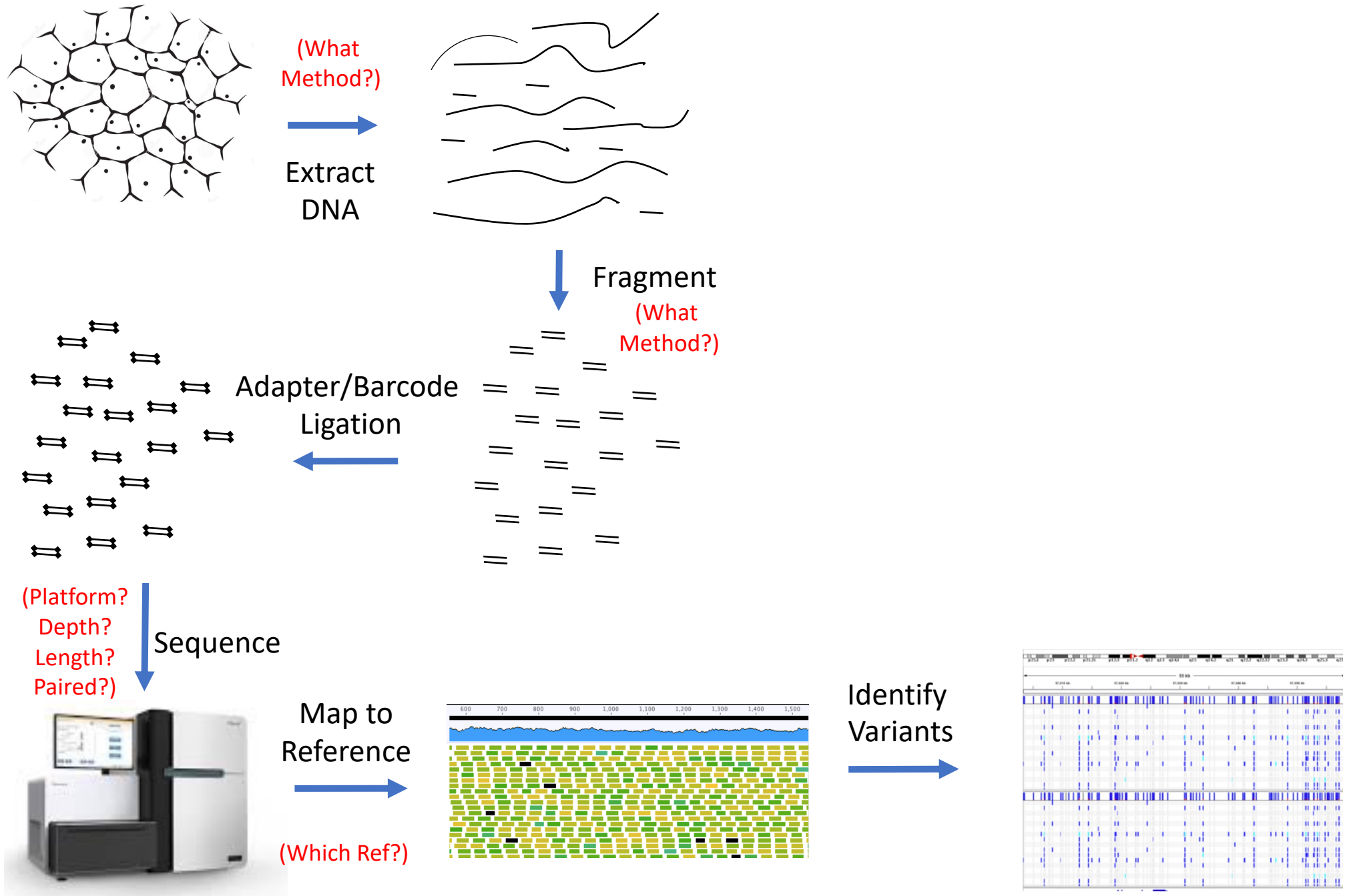
How do we detect variants?

Genome Resequencing

- Detection of small genomic variants like SNV's and InDels is usually done through Genome Resequencing (imprecisely referred to as DNA-Seq)
 - Start with a known reference genome
 - Sequence sample(s) of interest
 - Map Reads to reference
 - Identify differences
- To compare two samples to each other in this manner, you compare each to the reference genome and then look for differences

Typical Small Variant Analysis

- **Design Experiment**
- **Sample and Library Preparation**
- **Sequencing**
- Assess Sequence Quality
- Trim and Filter Reads
- Map Reads to Reference
- Identify Variants
- Explore and Verify Variants
 - Classify and Annotate Variants
 - Contextualize
 - Back to the lab: PCR and Sanger

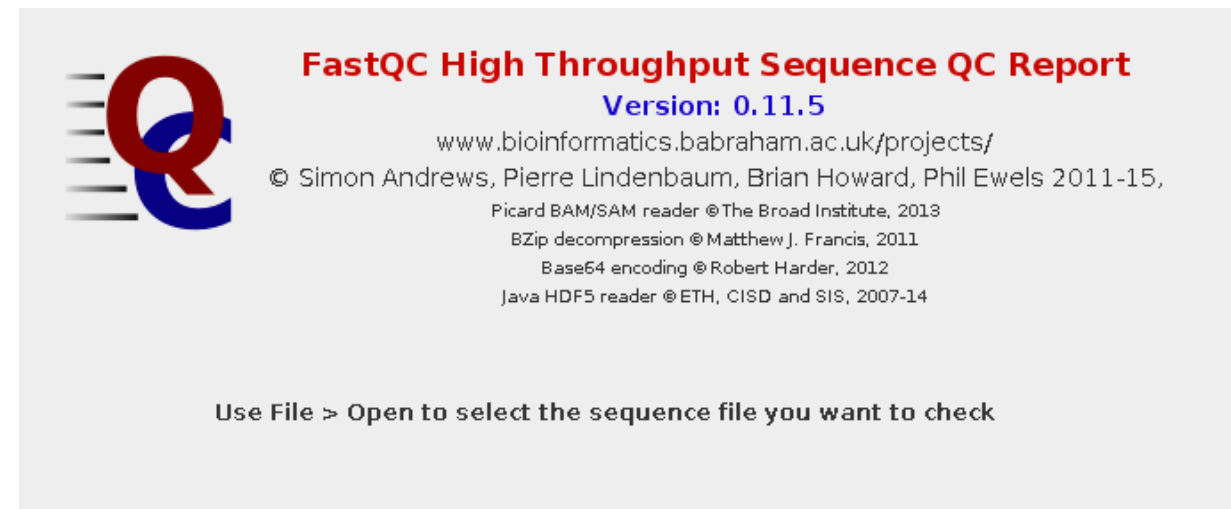


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Quality Control

- FASTQ – similar to RNA-Seq
- GC should be a normal curve
- High number of duplicate reads is generally bad
- Not likely to see base composition skew at beginning



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Trimming and Filtering

- Even if your data is high quality as a whole, it may have reads of regions of reads which are lower quality
- Cutadapt/TrimGalore! and Trimmomatic are popular choices for trimming and filtering Illumina data
 - Remove low quality read ends
 - Remove adapter sequences
 - Filter out short sequences, sequences with overall low quality, or excessive ambiguous bases (e.g. N's)

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Reference Genome

- The completeness, quality, and annotations available for the reference genome can affect the approach (sequencing and bioinformatics)
- A perfect reference genome would represent the Major Alleles for the entire population . . . not always true (or for some genomes even usually)
- If no reference is available you can attempt a *de novo* assembly to make a reference, very challenging . . .

Reference Mapping

- Once reads are trimmed the next step is to map them onto the reference
- Similar software to RNA-Seq reference mapping . . . But doesn't need to be splice aware
- Bowtie2 is the genome mapping equivalent of TopHat2
- bwa (mem algorithm) is another popular mapper for DNA-Seq and has become the recommended standard for many workflows
- Structural variants: would need to allow discordant mapping

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Variant Detection

- Turns out this is a lot more complex than just comparing the read to the reference
- SAMtools and other packages can do that naïve comparison . . . But they will detect many, many, many false variants
 - Sequencing error
 - Bad mapping
 - InDels cause alignment issues
 - Incomplete penetrance of mutations
 - Mixed cell types
 - Ploidy and multi-copy genes

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| - | OR | - |
| ACT-AT | | ACTA-T |

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| ACTAGT | | ACTAGT |
| - | OR | - X |
| ACTA-T | | ACT-AT |

Variant Detection

- Variant calling software needs to employ sophisticated strategies to overcome such issues and identify likely variants
 - Read depth masking (too much or too little is bad)
 - Allelic/multi-copy balance vs sequencing error
 - Insufficient coverage/masking vs no variant (comparing samples)
- Also such software usually also does double-duty as haplotype caller: 0/0, 0/1, 1/1, 1/2, 2/2, 0/2, 1/3, etc (can be separated like this 0|0)
- GATK (**G**enome **A**nalysis **T**ool**k**it) has become the dominant variant detection software, especially for Eukaryotic organisms

GATK Best Practices

- GATK posts extensive information on best practices for performing variant identification:
<https://software.broadinstitute.org/gatk/best-practices/>
- Best practices shift over time, so this is always a good resource to review before doing an analysis

Microbial Variant Detection

- For “Prokaryotes” the process is a little more straight forward? Why?
- We still have issues like sequencing error that need to be dealt with...but lack complexities of dealing with heterozygous alleles

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- So we can use faster and more streamlined tools like Snippy...which we will use for our project

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| ACTA-T | | ACT-AT |

- Snippy wraps the steps of reference mapping, variant calling, and variant filtering into a single step!

VCF File Format

- <https://samtools.github.io/hts-specs/VCFv4.2.pdf>

```
##fileformat=VCFv4.2
##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 ID REF ALT QUAL FILTER INFO FORMAT NA00001 NA00002 NA00003
20 14370 rs6054257 G A 29 PASS NS=3;DP=14;AF=0.5;DB;H2 GT:GQ:DP:HQ 0|0:48:1:51,51 1|0:48:8:51,51 1/1:43:5:.,.
20 17330 . T A 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
20 1110696 rs6040355 A G,T 67 PASS 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 1230237 . T . 47 PASS NS=3;DP=13;AA=T GT:GQ:DP:HQ 0|0:54:7:56,60 0|0:48:4:51,51 0/0:61:2
20 1234567 microsat1 GTC G,GTCT 50 PASS NS=3;DP=9;AA=G GT:GQ:DP 0/1:35:4 0/2:17:2 1/1:40:3
```