# Bioinformatics for NGS analysis From Reads to Variants

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# Bioinformatics for NGS analysis

From Reads to Variants

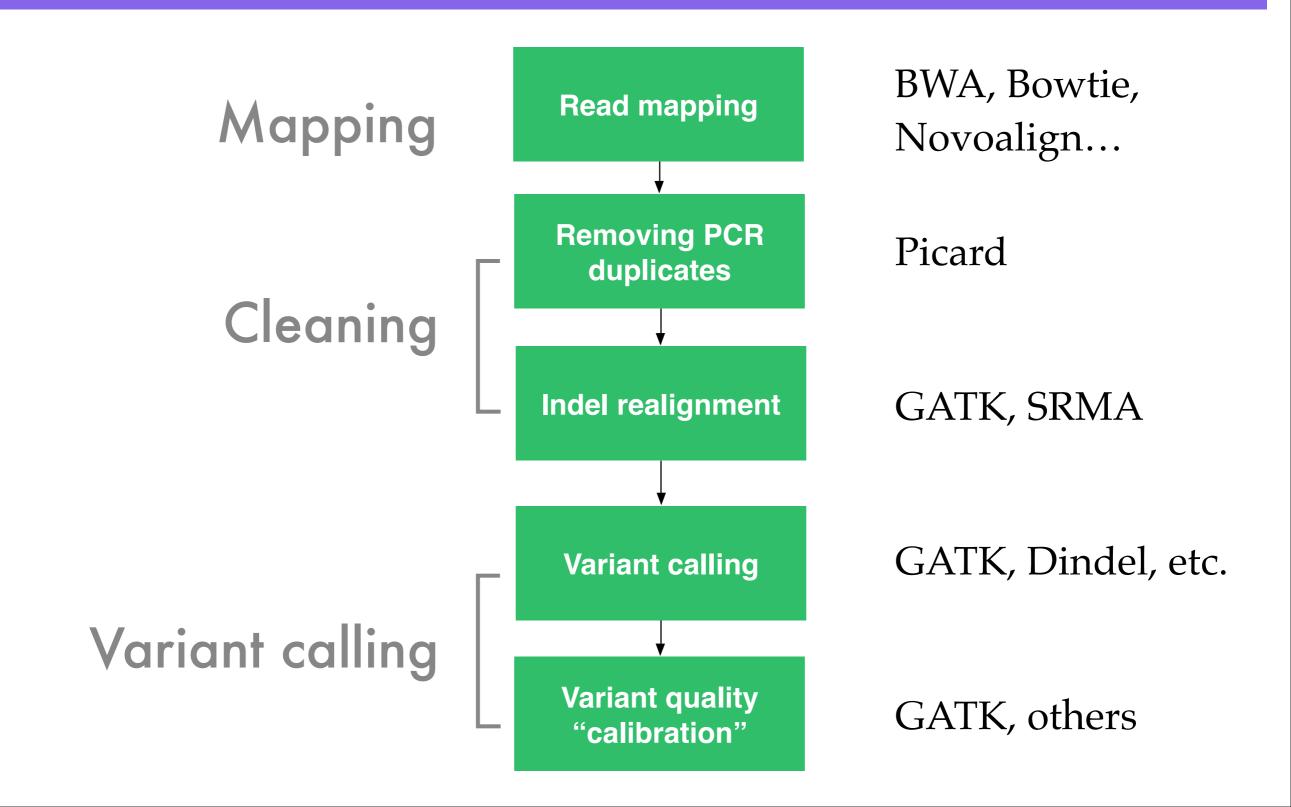
Tim Yu, MD, PhD

Disclosure:
Co-founder & Principal consultant,
Claritas Genomics



 Goal: arm you with concepts and vocabulary to understand how NGS data is analyzed, and to ask critical questions

#### A typical NGS processing pipeline



# NGS analysis is, in principle, a two step process

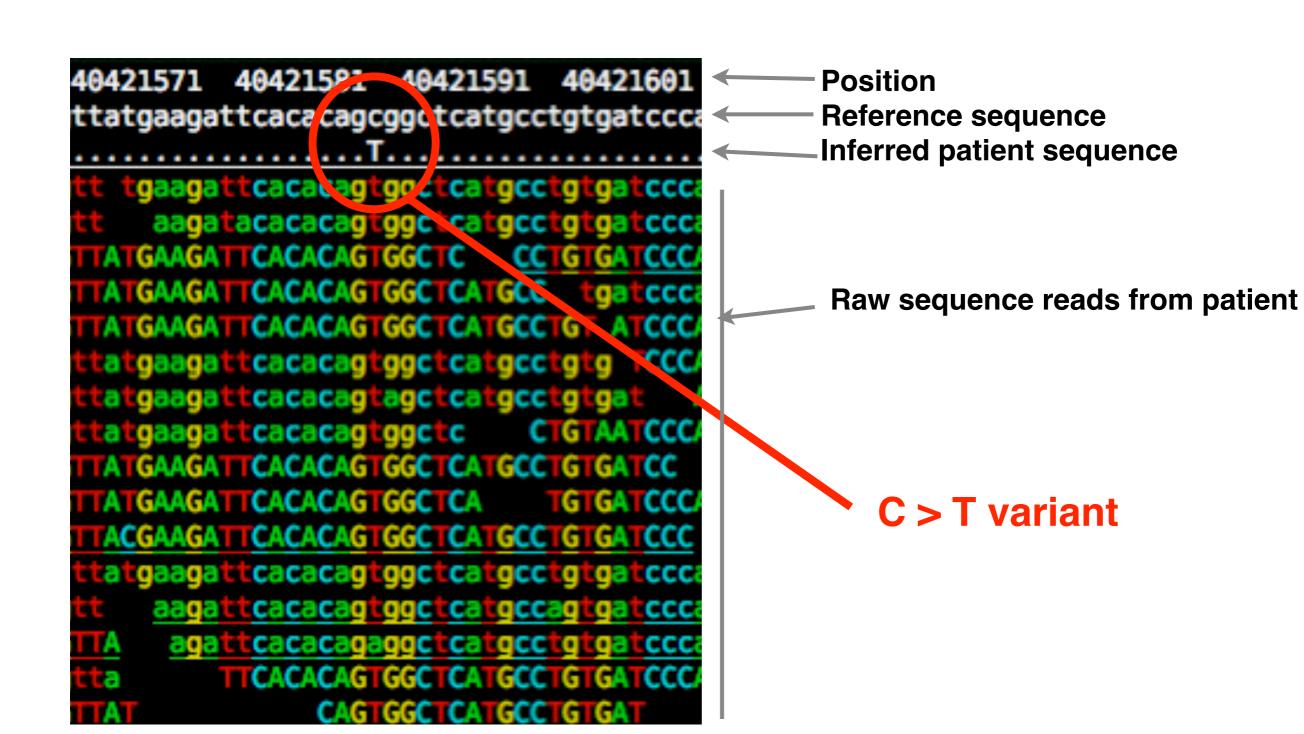
# 1. Millions/billions of reads are mapped en masse to a reference genome

```
48421551 48421561 48421571 48421581 48421591 48421681 48421611 48421621 48421631 48421641 48421651 48421661 48421671
721tttgagcagacctatataagatggttatgaagattcacacagcggctcatgcctgtgatcccagcactttgggaggctgaggcagtggagcacctgagatcatgagttcaagaccagcctggccaacatggtgaaaccccatctctactaaagatacaaaaattatccaggtgtgtg
 Α....Α....Τ
     tgaacagacctatataagatggtt tgaagattcacacagtggctcatgcctgtgatcccagcac tggggaggctgagtcaagtggagcacctgagatcatgagt
cagacctatataagatggtt aagatacacacagtggctcatgcctgtgatcccagcactt GGGAGGCTGAGGCAAGTGGAGCACCTGAGATCATGAGT
GACCTATATAAGATGGTTATGAAGATTCACACAGTGGCTC CCTGTGATCCCCAGCACTTTGGGAGGCAAGTGGAG ACCTGAGATCATGAGT
                                                                                               GGAGCACC GAGA CATGAGTIC cagcctggccaacatggtgaaaccccatctctactaaaga ACAAAAAT
GGAG ACCTGAGATCATGAGTTCAAGACCAGCCTGGACAACATGG AACCCCATCTCTACTAAAGATACAAAAAT
                                                                                                                                                AACCCCATCTCTACTAAAGATACAAAAATTATCCA
                ATATAAGATGGTTATGAAGATTCACACAGTGGCTCATGCC tgatcccagcact
                                                                                                  IGCACC TGAGA TCA TGAG
                                                                                                                      TCAAGACCA
                                                                                                                                    GCCAACATGGTGAAACCCCCATCTCTACTAAAGATACAAAA
                                       CACACAG GGC CA GCC G A CCCAGCAC
                                                                                                                      CAAGACCAGCC GGCCAACA GG GAAACCCCA TATACTAAAGA NCAAAA
                                                                                                           GATCACGAGTTCAAGACCAGCCTGCCCAACATGGTC AACCCCATCTCTACTAAAGATACAAAAAT
                                                                                                                               GCCTGGCCAACATGGTGAAA CCCATCTCTACTAAAGATACAAAAA
                                                                                                                                CCTGGCCAACATGGTGAAACCCCATCTCTACTAAAGATAC
      GAACAGACCTATATAAGA
                                                                                                  GCACC GAGATCA TGAGT CAAGACCAGCC TGGCCA
                                                                                 AGGCTGAGGCAAGTGGAGCAC GATCATGAGTTCAAGACCCGCCTGGCCAACATGGTGAAAC ccatctctactaaagatacaaa
     TGAACAGACCTATATAAGA
                               A GAAGAT CACACAG GGC CA GCC G GA CCCAGCA C GG
                                                                                                                                          CA IGG GAAACCCCA IC IC IAC IAAAGA TACAAAAA TA
                                                                                                                                          CATGG GAAACCCCATCTCT CTAAAGATACAAAAATTATCCAGG G
                                                                                                                                                GAAACCCCATC CTACTAAAGATACAAAAATGA
                                                                                                                                                    ACCCCATCTCTACTAAAGATACAAAAATTA CCAG
                                                                                                                                                  saaccccatctctactaaagatacaaaaattatccag
                                                                                                                                                 GAAA CCCA CTCTACTAAAGATACAAAAATTA TCCAGG
                                                                                          CAAG GGAGCACC GAGA CA GAG T CAAGACCAG
                                                                                                                                                   aaccccatctctgctgaagatgcaaaaat
                                                                                                                                                    ACCCCATC TCTACTAAAGATACAAAAATTATCCAGGTG
                                                                                                                                                   aaccccatctctactaaagatccaaaa
                                                                                                                                                  AACCCCATCTCTACTAAAGATACAAAAA
                                                                                                                                                   ACCCCGTT CTACTAAAGATACAAAAA
                                                                                                                                                        CATCTCTAATAAAGATACAAAAA
                                                                                                                                                       CATCTCTACTAAAGATACAAAAATTATCC
                                                                                                                                                       CG C C AC AAAGA ACAAAAA AA CCA
                                                                                                                                                       CA IC IC IAC IAAAGA TACAAAAA TIA ICCA
```

# 1. Millions/billions of reads are mapped en masse to a reference genome

```
40421571 40421581 40421591 40421601
                                                                        18421611 48421621 48421631 48421641 48421651 48421661 48421671 48421681 48421691 48421701
721tttgagcagacctatataagatg
                            ttatgaagattcacacagcggctcatgcctgtgatcccag<mark>:</mark>actttgggaggctgaggcaagtggagcacctgagatcatgagttcaagaccagcctggccaacatggtgaaaccccatctctactaaagatacaaaaattatccaggtgtggtg
                             .....T......T.....
                                                                                                                            ACCAGCCTGGCCAACATGGTGAAACCCCATCTCTACTAAA ATACAAAAA
                              t tgaagattcacacagtggctcatgcctgtgatcccag
                              t aagatacacacagtggctcatgcctgtgatcccac
TATGAAGATTCACACAGTGGCTC CCTGTGATCCCAC
                              TATGAAGATTCACACAGTGGCTCATGCC tgatccca
                              A TGAAGAT I CACACAG I GGC T CATGCC I GT A TCCCA
                                   AGA CACACAG GGC CA GCC G GA CC
                                                                                                                                gcclggccaacatggtgaaaccccatctctactaaagat
                                                                                                                                      CCAACATGGTGAAA CCCATCTCTACTAAAGATACAAA
                                                                                                                                 CCTGGCCAACATGGTGAAACCCCATCTCTACTAAAGATAC
 CATTIGAACAGACCTATATAAGA
                               ATGAAGATTCACACAGTGGCTCATGCCTGTGATCCCAGCA TC
                                                                                                                                                     ccccatctctactaaagatacaaa
                                                                                                                                                 GAAACCCCA C C C AC AAAGA TACAAAAA T A CCAGG
                                                                                                GGAGCACC TGAGATCA TGAGT TCAAGACCAG
                                                                                          gcaatttgagctcctgagatcatgagttcaagacc
GCAAGTGGAGCACCTGAGATCA
```

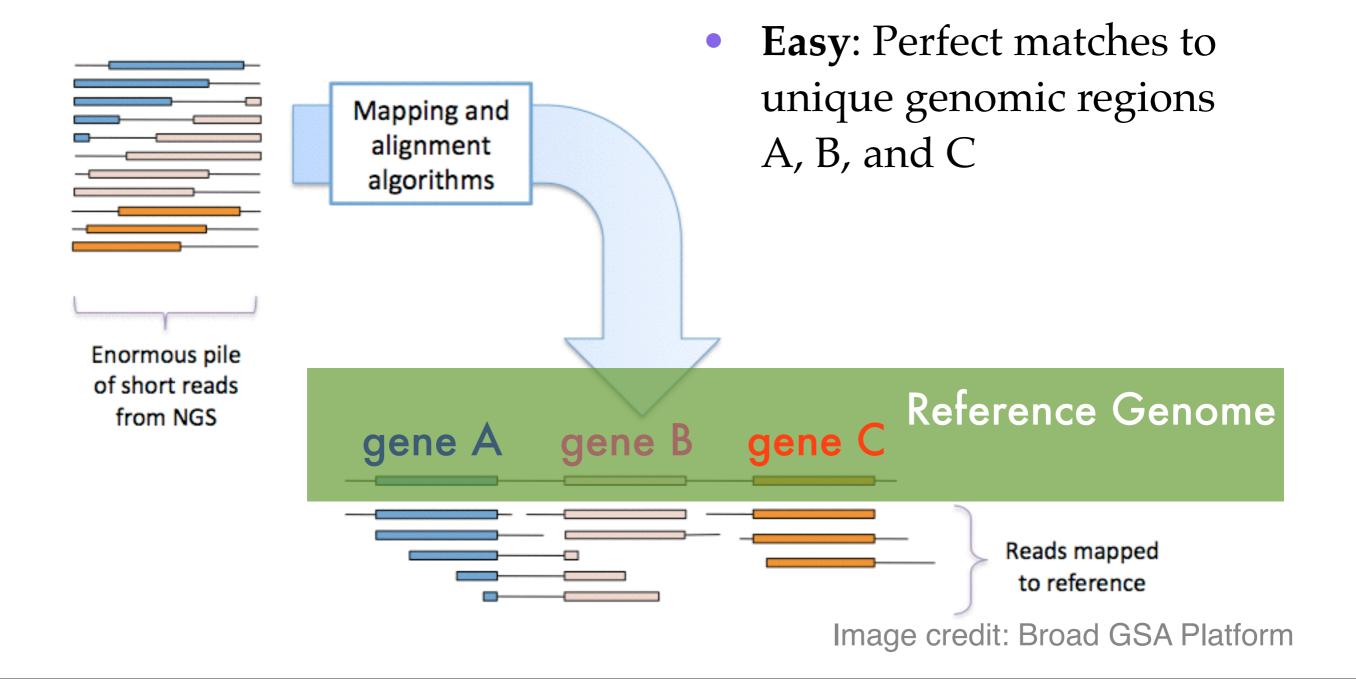
# 2. Variants are detected when enough reads disagree with reference



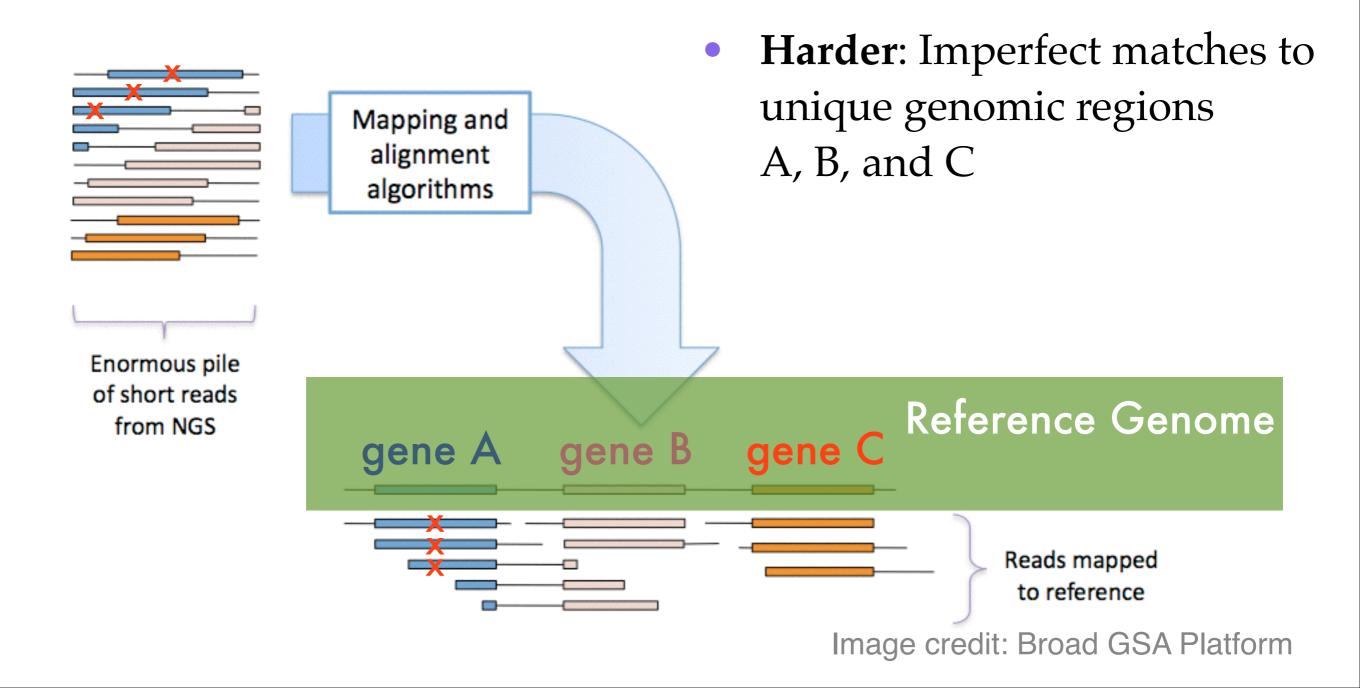
### Complicating factors

- Mapping can be tricky
- Sequencing coverage is biased
- Not all variant calls are created equal
- Beyond SNPs and small indels

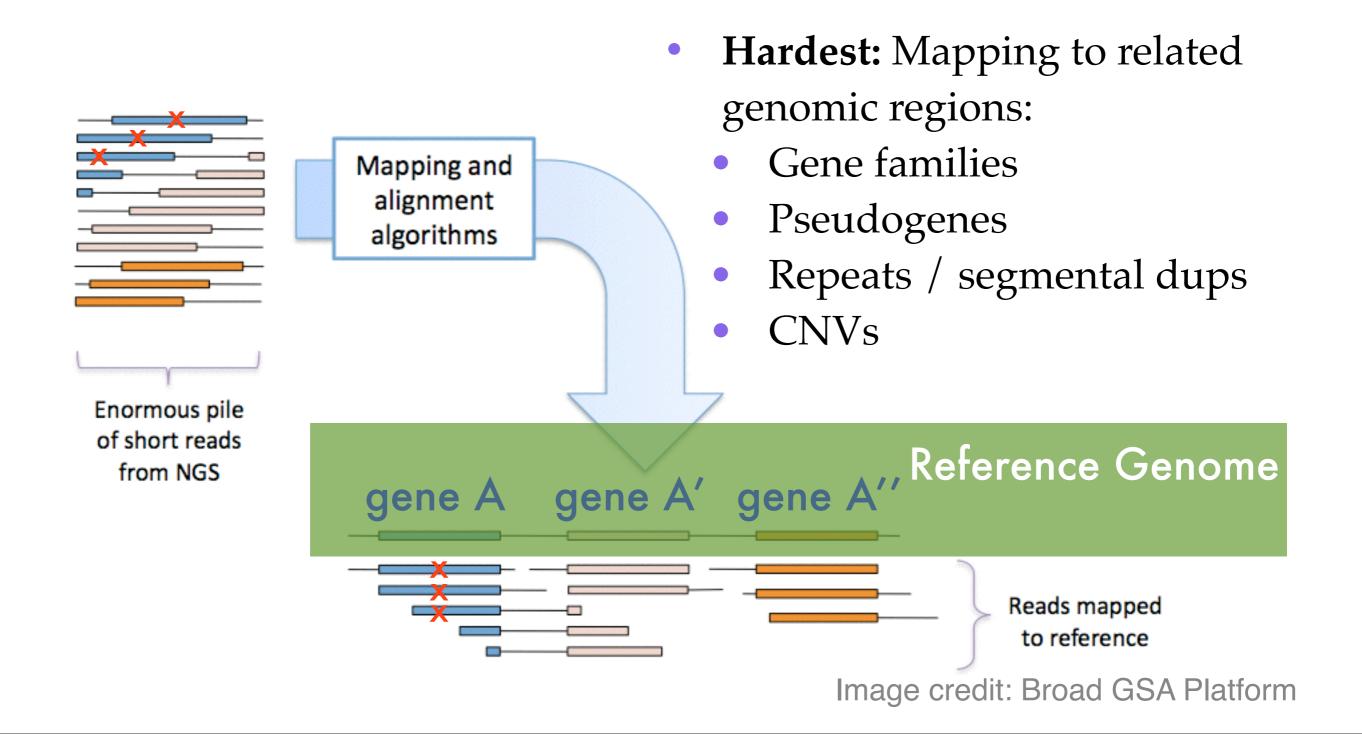
## 1. Mapping can be tricky



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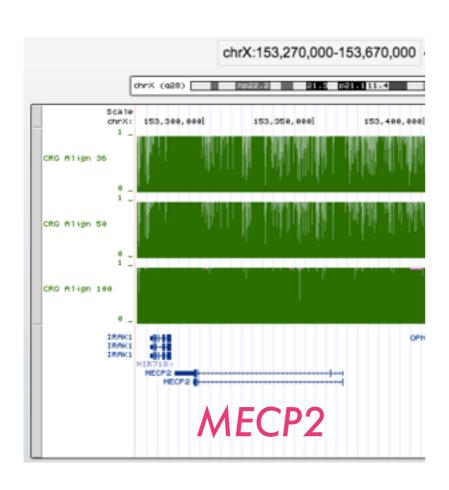


## 1. Mapping can be tricky



### Mapping confidence/mapability

- Mapping confidence is a prerequisite for good variant calls
- But mapability can vary quite a bit!

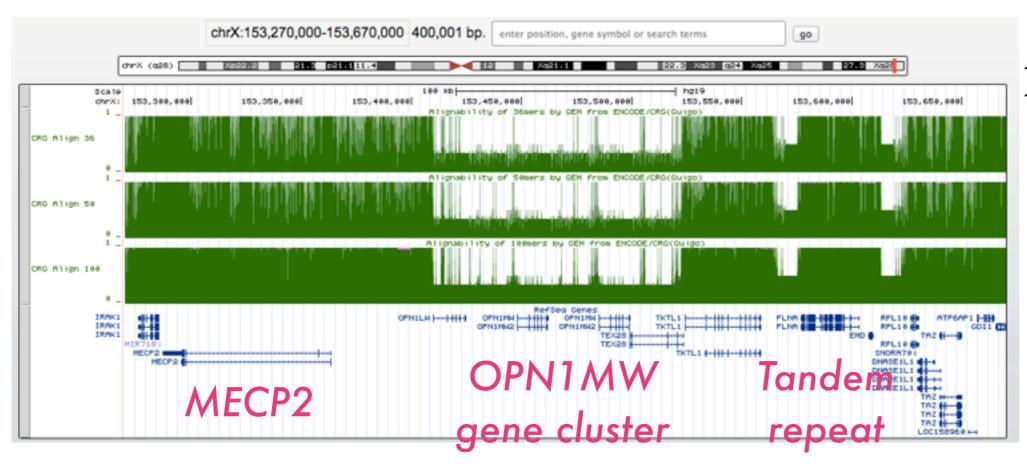


#### Read length

- 36 bp
- 75 bp
- 100 bp

### Mapping confidence/mapability

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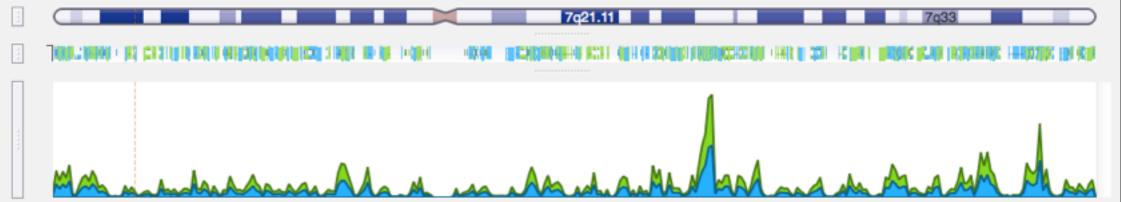
### Solutions to the mapability problem

- Longer reads
- Paired-end and mate-pair sequencing
- Better reference sequences (eg taking into account CN variable regions)

## 2. Sequencing coverage is biased

Read coverage on chr7 for a a typical WES (whole exome sequencing) experiment





#### **Contributing factors:**

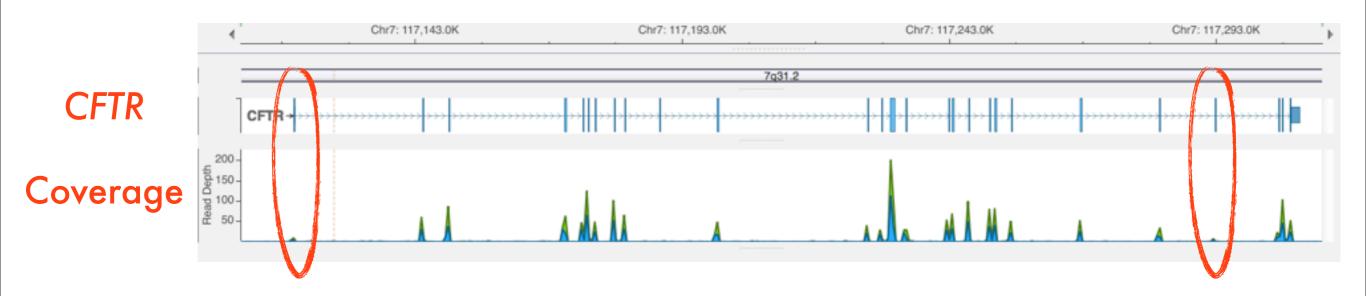
- intentional (eg, exome capture design)
- unintentional
  - PCR-related (eg, GC-rich regions)
  - mapability

# ...& may result in gaps in coverage (insufficient breadth)



- **Example:** absent read coverage over *CFTR* exon 10
- Consequence: variant dropout (false negatives)

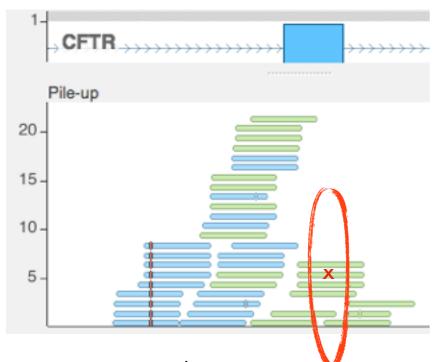
# ...or just inadequate coverage (insufficient depth)



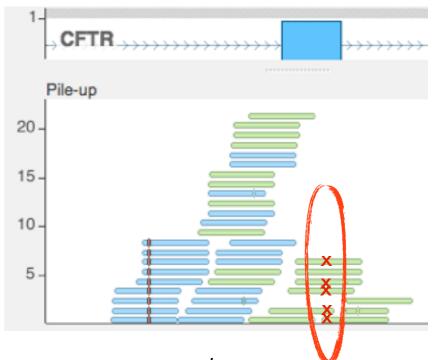
• Example: low read coverage over *CFTR* exons 1 & 24

# ...or just inadequate coverage (insufficient depth)

• **Example:** low read coverage over *CFTR* exons 1 & 24

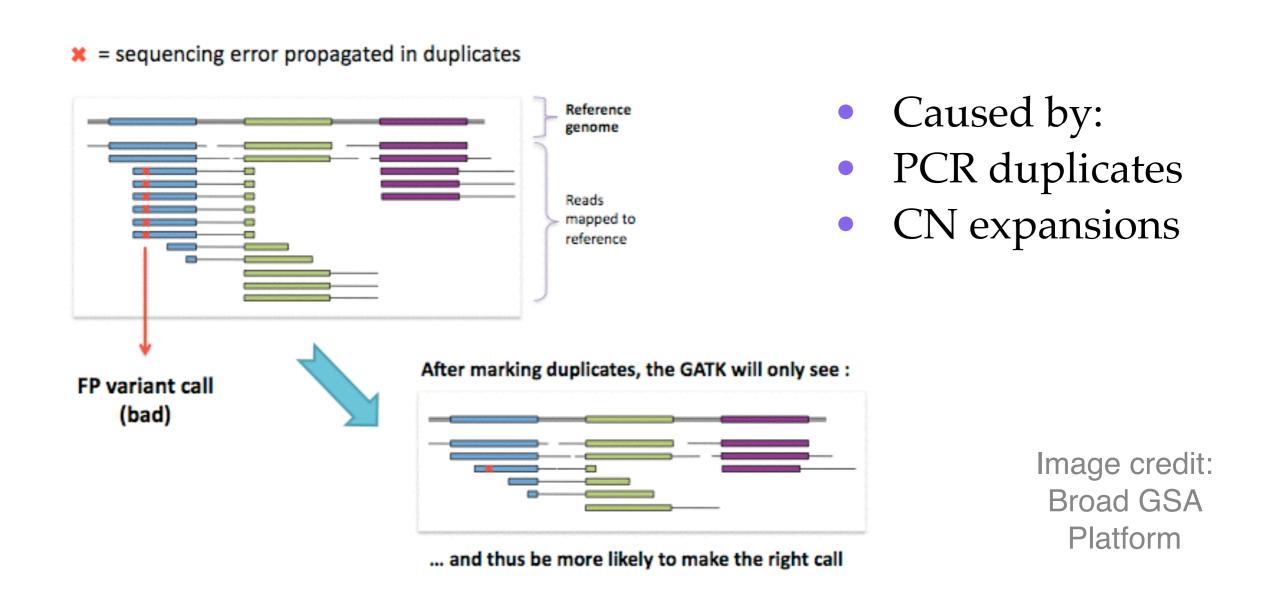


- SNP on 1/6 reads
- Is this a heterozygous variant?



- $\bullet$  SNP on 5/6 reads
- Is this a heterozygous or homozygous variant?
- Consequence: Inaccurate genotyping in areas of low read depth

#### Excess coverage is sometimes a red flag



 "Cleaning" alignments by finding and removing PCR duplicates evens out coverage, and reduces false positives

#### Depth and breadth are usually a tradeoff

Given fixed \$\$\$: Depth or Breadth, choose one!

#### **Shallow & wide**

more variants

less accurate genotypes

e.g., "exomes" at 50-150X

#### Narrow & deep

fewer variants

more accurate genotypes

e.g., "panels" at 500-1500X

 Costs are gradually dropping so hopefully this tradeoff will become moot!

#### Solutions to the coverage bias problem

- [Optimize mapability (longer reads, paired end sequencing, etc.)]
- Optimize library prep
  - Minimize PCR, or use PCR-free library prep methods
  - Normalize baits
- Informatically, find and remove PCR duplicates

### 3. Not all variant calls are created equal

• We do quite well with SNPs (i.e., single base substitutions)

• Calls are reliable: >99% concordance with chip-based SNP genotyping or other "truth sets"

#### 3. Not all variant calls are created equal

- But indels (i.e., small insertions or deletions) are significantly harder
- It is computationally hard to map a 100bp read to the genome if you allow for gaps
- Sensitivity estimates vary hugely (50-90%), & 2-10X more false positives (compared to SNPs)

#### Example: calling around homopolymers

Small insertions / deletions (especially near the ends)
 can trick mappers into misaligning with mismatches

10bp "T" homopolymer run

ref: TGACTCGTAACCAGGCTTTTTTTTTTTGCGGGCCGAA

#### Example: calling around homopolymers

Small insertions / deletions (especially near the ends)
 can trick mappers into misaligning with mismatches

10bp "T" homopolymer run

ref: TGACTCGTAACCAGGCTTTTTTTTTTTGCGGGCCGAA

reads: TCGTAACGAGGCTTTTTTTTTGCGGGC

AGGCTTTTTTTTTGCGGCCCGAA

GACTCGTAACGAGGCTTTTTTTTTGC

**CGAGGCTTTTTTTTTTGCGGGCCG** 

TGACTCGTAACGAGGCTTTTTTTTG

many single-bp mismatches?

#### Example: calling around homopolymers

 Small insertions/deletions (especially near the ends) can trick mappers into misaligning with mismatches

10bp "T" homopolymer run

ref: TGACTCGTAACCAGGCTTTTTTTTTTGCGGGCCGAA

reads: TCGTAACGAGGCTTTTTTTT^GCGGGC

AGGCTTTTTTTTT GCGGGCCGAA

GACTCGTAACGAGGCTTTTTTTTT^GC

CGAGGCTTTTTTTTT GCGGGCCGAA

TGACTCGTAACGAGGCTTTTTTTTT^G

Local realignment reveals a hidden 1bp delT

### Red flags that a variant may be suspicious

- In fact, raw indel calls are infested with false positives
- Statistics can be calculated that predict problematic variants:
  - Low read depth
  - Strand bias
  - Low mapping quality
  - Clusters of nearby variants
  - Nearby homopolymer run/other repeats

#### Variant Quality Scores

• "Variant quality score": These statistics can be combined to derive a score that expresses the confidence in a particular call

#### Solutions for calling difficult variants

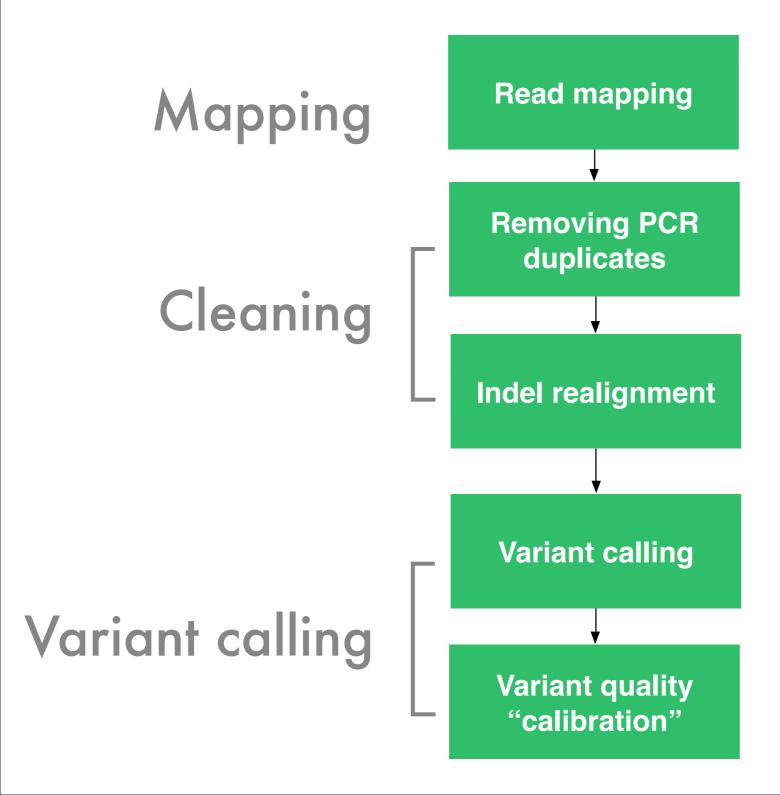
- Increase coverage
- Main advice: Be aware that variant calling is imperfect
  - SNPs pretty good
  - indels less so
- Investigational approaches:
  - Joint calling in large batches
  - Building custom references for specific difficult-to-catch variants
- Trust, but verify!

#### 4. Beyond SNPs and small indels

- Algorithms for other variant classes are coming, but still largely investigational:
  - CNVs\* and structural variants
  - Larger insertions (>20bp) or deletions (>50bp)
  - Repeat expansions / contractions
  - Transposable elements

# Take home points

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 A proper analytic pipeline mitigates many of the complications of NGS analysis

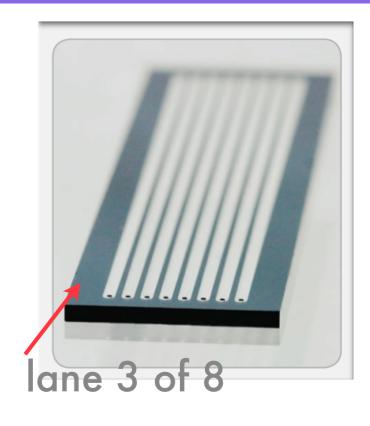
# Take home points

- Ask not just about mean coverage, but coverage <u>breadth and</u> <u>depth</u> ("95% coverage at 30X")
- Ask for a list of <u>coverage dropouts</u>. There is no such thing as a "whole" genome!
- Weigh the pros and cons of maximizing breadth (exome) vs. depth (panels)
- SNPs are generally high quality, but it is still important to weigh <u>variant quality</u> and other red flags. Especially for indels, trust but <u>verify</u>
- Recognize that CNV, SV, larger indels, repeat expansion/ contractions, and mobile elements are out of the scope of most clinical NGS pipelines

#### Questions

# Sequence reads

```
6 6 6
                       twy1@orchestra: /home/twy1/080904.3.chr7PCHchr4EP — ssh — #1
01:24:52 twyl@orchestra~/080904.3.chr7PCHchr4EP$ more s_3_sequence.txt
@HWI-EAS214_1:3:3:1375:979
CCCAACCAACCCNNCACATCCCAAACAACCCCAACC
+HWI-EAS214_1:3:3:1375:979
25 25 25 25 25 25 25 25 25 25 25 25 13 -2 -2 25 11 14 25 -2 25 25 4 25 25 25 25 25 25 25 21 19 19 2 17 19 15
@HWI-EAS214_1:3:3:1257:1681
GGCGACTTACCCNTCNTCATCTCATATATTTAAGTT
+HWI-EAS214_1:3:3:1257:1681
14 8 8 21 9 9 8 25 13 13 8 5 -2 3 4 -2 3 8 4 3 2 3 8 3 3 22 22 18 25 14 2 9 2 5 2 10
@HWI-EAS214_1:3:3:1365:902
ACCCACCAACCCNNATCCACGAAACCCACAAAAAAC
+HWI-EAS214_1:3:3:1365:902
25 25 25 25 25 18 22 13 22 25 25 25 -2 -2 5 5 8 13 14 25 2 25 14 21 13 22 21 13 11 13 7 3 11 13 11 13
@HWI-EAS214_1:3:3:773:1646
GCCGCTATTGCCNACATCTATGTTNTCGCGACCTAT
+HWI-EAS214_1:3:3:773:1646
25 8 2 25 11 25 14 8 9 5 8 9 -2 5 5 11 3 11 2 11 13 25 5 22 -2 3 9 3 13 -2 11 11 5 19 8 1
@HWI-EAS214_1:3:3:1330:1567
GGCAGAGTCTGCNAGCGGGATCCTGATACGTTTGCA
+HWI-EAS214_1:3:3:1330:1567
25 9 25 14 3 25 11 25 11 8 4 13 -2 5 5 4 5 18 5 5 9 3 2 2 4 9 22 25 11 4 4 9 2 3 -2 2
@HWI-EAS214_1:3:3:1730:897
GAGAGGGGGAGNNTNCGGGCCAGCAGTCACAGGTA
+HWI-EAS214_1:3:3:1730:897
25 8 25 25 5 25 13 25 25 14 9 25 -2 -2 8 -2 3 13 -2 13 9 2 13 25 22 4 21 25 11 -2 3 15 4 11 5 4
@HWI-EAS214_1:3:3:531:1102
GATCGGCAGATGNTGCCAAGCACTCTTATTGTTGTG
+HWI-EAS214_1:3:3:531:1102
25 9 25 25 25 25 8 25 3 13 8 25 -2 13 9 5 8 3 4 16 4 5 14 18 3 14 25 3 11 10 8 13 19 8 14 8
@HWI-EAS214_1:3:3:388:1074
GCAGGACGATGCNCGCTGTGTAGGGCTATGTACGTT
+HWI-EAS214_1:3:3:388:1074
25 25 22 22 2 9 11 22 4 8 5 5 -2 5 3 25 5 14 2 9 10 5 3 10 8 3 13 5 25 22 11 5 9 17 2 15
@HWI-EAS214_1:3:3:1726:412
GGGAATGGATGANNANTACCATATGCATATCAGCTG
+HWI-EAS214_1:3:3:1726:412
25 25 11 25 8 25 8 25 5 8 22 14 -2 -2 3 -2 8 25 11 5 13 10 11 3 2 2 14 25 5 8 5 5 3 3 5 5
@HWI-EAS214_1:3:3:1731:958
GACTTCAGCCAGNNCNTCCCTTAGGTCGCATGACCG
+HWI-EAS214_1:3:3:1731:958
25 11 25 21 11 25 25 22 25 12 11 25 -2 -2 11 -2 5 25 4 25 2 11 8 22 25 8 25 25 11 3 9 19 4 3 3 2
@HWI-EAS214_1:3:3:1718:1225
GGATTCGATTCTNNTNCTCGCTGACATTGCCGACAA
+HWI-EAS214_1:3:3:1718:1225
```

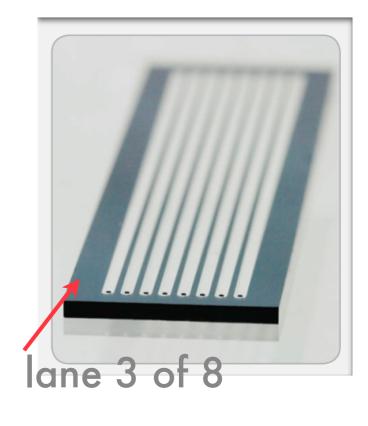


identifier



identifier

machine | lane | tile | X:Y



sequence

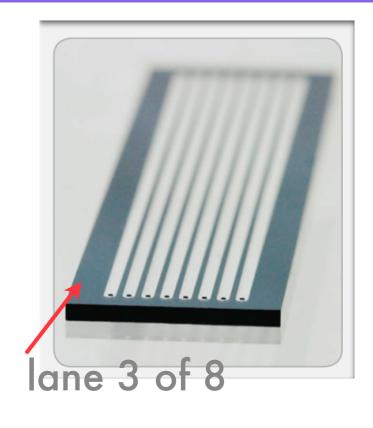
@HWI-EAS214\_1:3:3:1375:979 CCCAACCAACCCNNCACATCCCAAACAACCCCCAACC +HWI-EAS214\_1:3:3:1375:979

25 25 25 25 25 25 25 25 25 25 25 25 13 -2 -2 25 11 14 25 -2 25 25 4

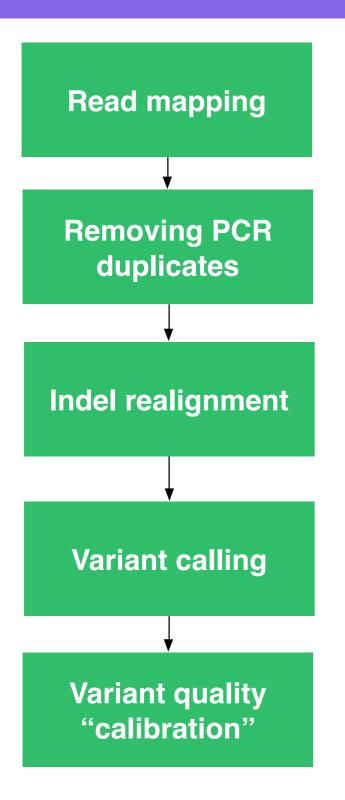


```
identifier (again!)
```

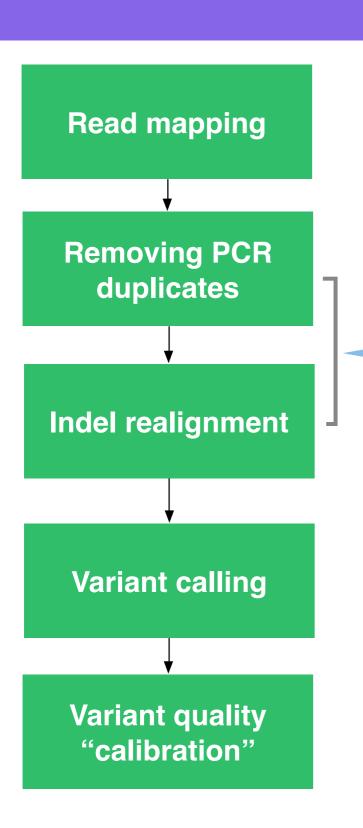
```
@HWI-EAS214_1:3:3:1375:979
CCCAACCAACCCNNCACATCCCAAACAACCCCCAACC
+HWI-EAS214_1:3:3:1375:979
25 25 25 25 25 25 25 25 25 25 25 25 25 13 -2 -2 25 11 14 25 -2 25 25 4
```



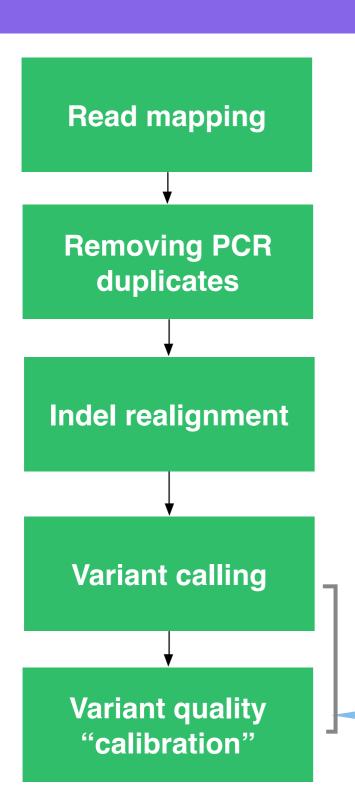
```
base qualities (higher=better)
```



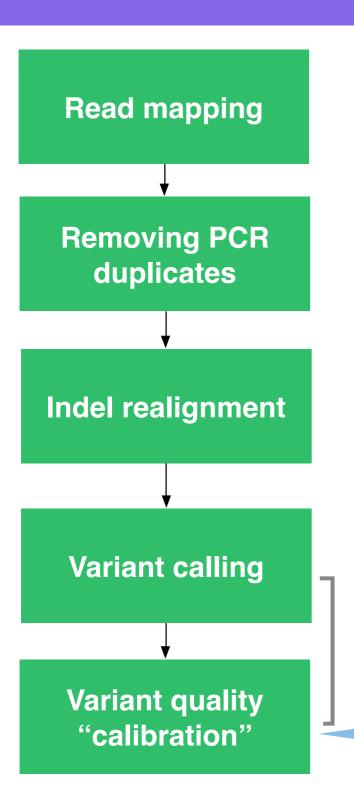
- Was sufficient breadth & depth of coverage achieved?
  - "85-95% coverage at >30X"
- What regions were missed?



 Was appropriate cleaning performed?



- Are the #s of variants called reasonable?
- Especially indels
- Is the percentage of "known SNPs" reasonable (98% in dbSNP)?



- Is the variant quality score sufficiently high to be believed?
- What was used to confirm these variants?