

**DTU**



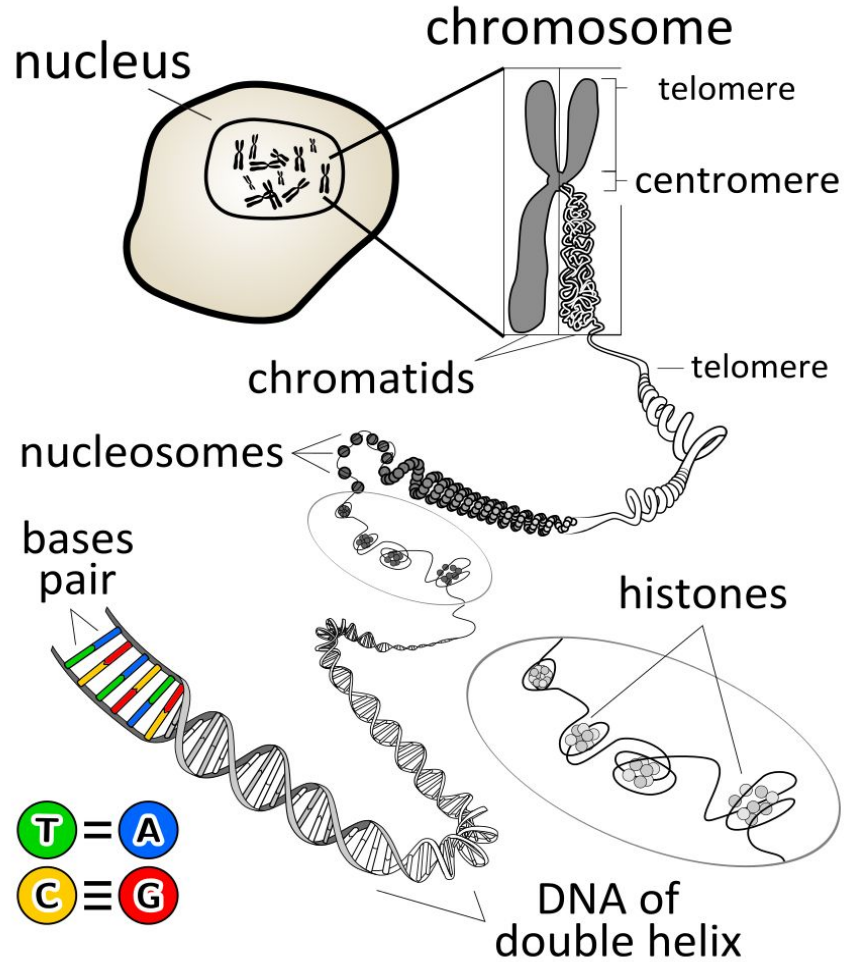


**DTU Health Technology  
Bioinformatics**

# **Alignment post-processing and variant calling part 1**

*Gabriel Renaud  
Associate Professor  
Section of Bioinformatics  
Technical University of Denmark  
gabriel.reno@gmail.com*

# A brief reminder

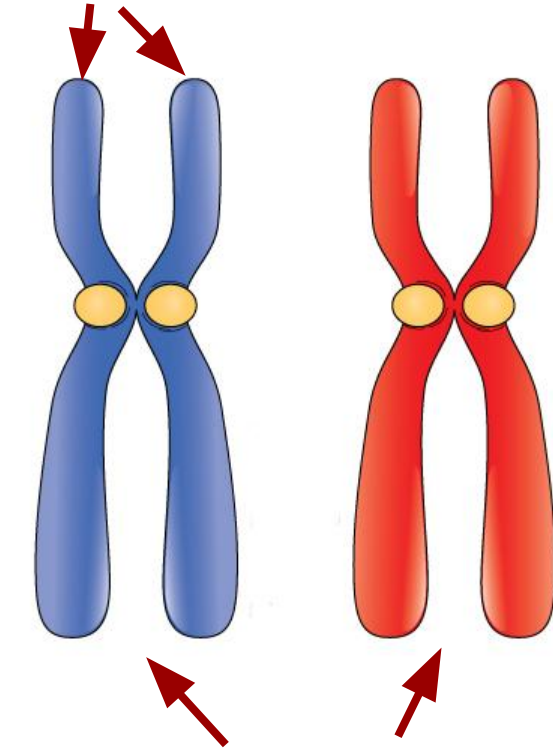


A brief reminder

sister chromatids

Paternal

Maternal



homologous chromosomes

# Heterozygosity

TACAAATAT  
TACAGATAT

M:



P:

Heterozygous sites

# Heterozygosity



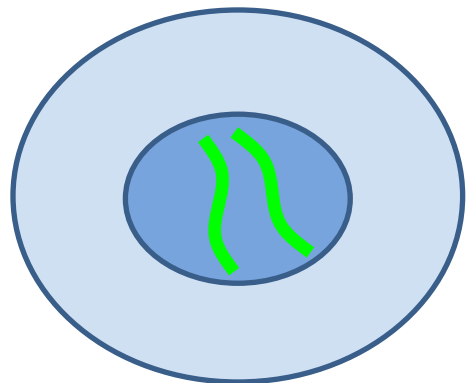
M:



P:



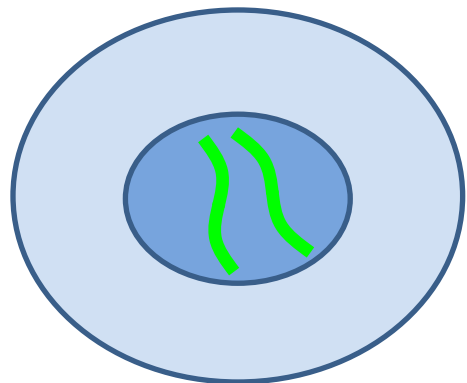
Homozygous sites



ind#A

M : **TACAAATAT**

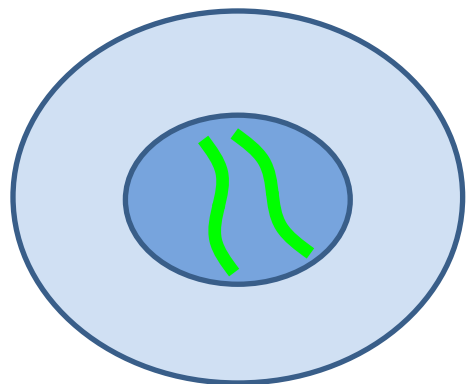
P : **TACAGATAT**



ind#B

M : **TACAGATCT**

P : **TACAGATCT**

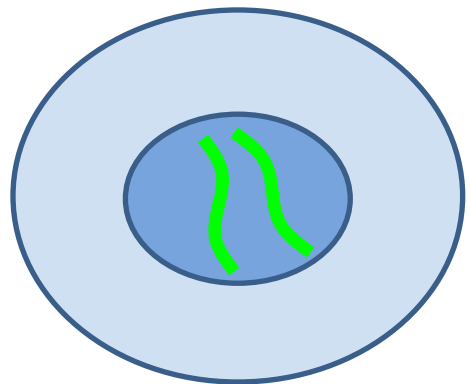
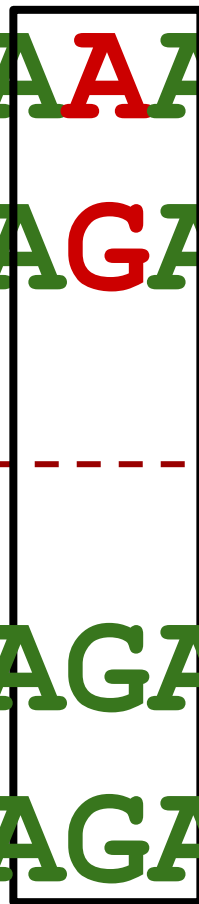


ind#A

M: TACAAATAT

P: TACAGATAT

Heterozygosity

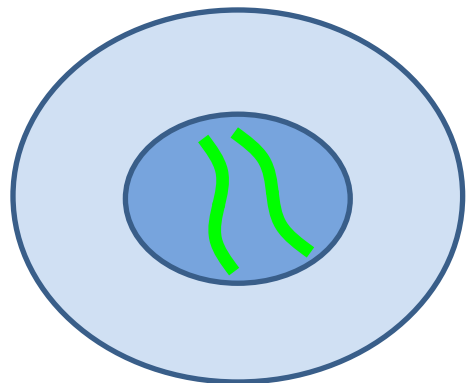


ind#B

M: TACAGATCT

P: TACAGATCT



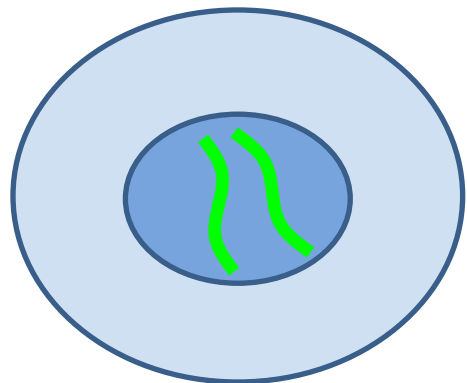
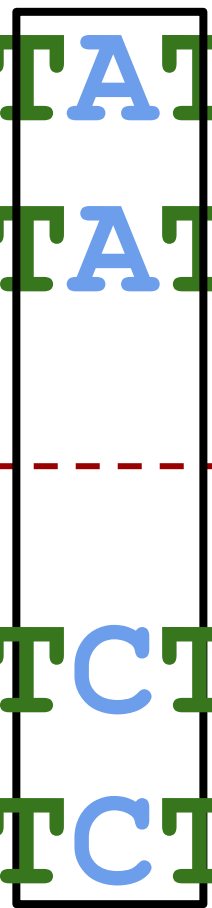


ind#A

Homozygous variant

M: TACAAATAT

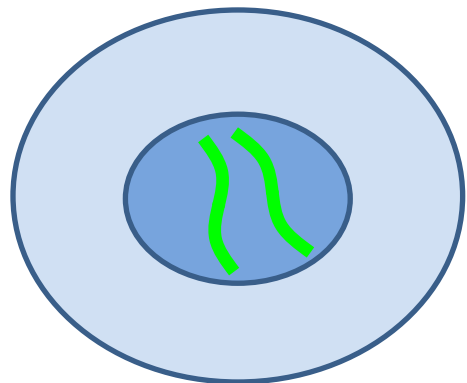
P: TACAGATAT



ind#B

M: TACAGATCT

P: TACAGATCT



ind#A

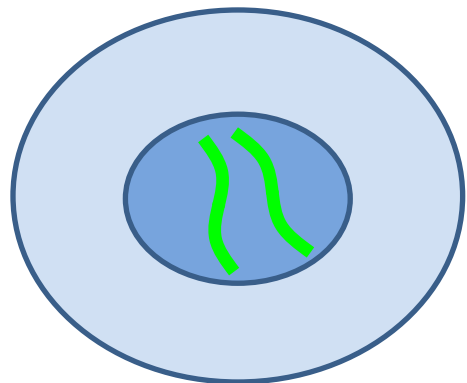
M :

TACAAATAT

P :

TACAGATAT

Homozygous invariant



ind#B

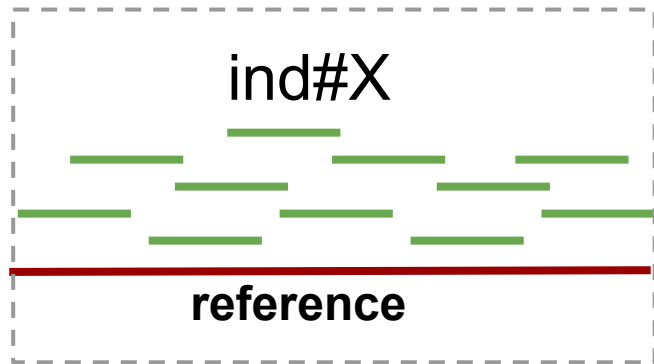
M :

TACAGATCT

P :

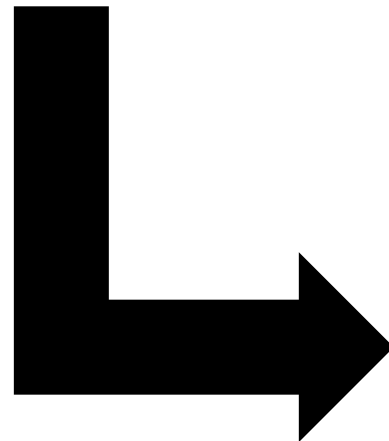
TACAGATCT

# Genotyping



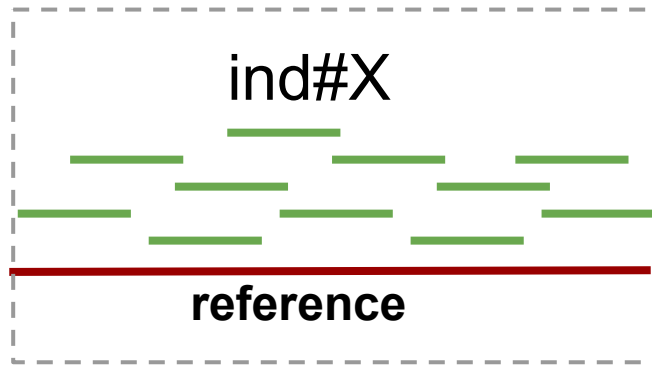
TACAAATAT  
TACAGATAT

Which of the 10 possible genotypes is the most likely?

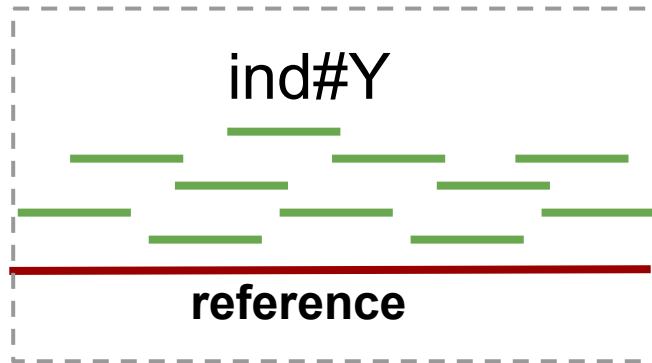


- AA
- AC
- AG
- AT
- CC
- CG
- CT
- GG
- GT
- TT

# Joint Genotyping



TACA**A**AT**A**T  
TACA**G**AT**A**T



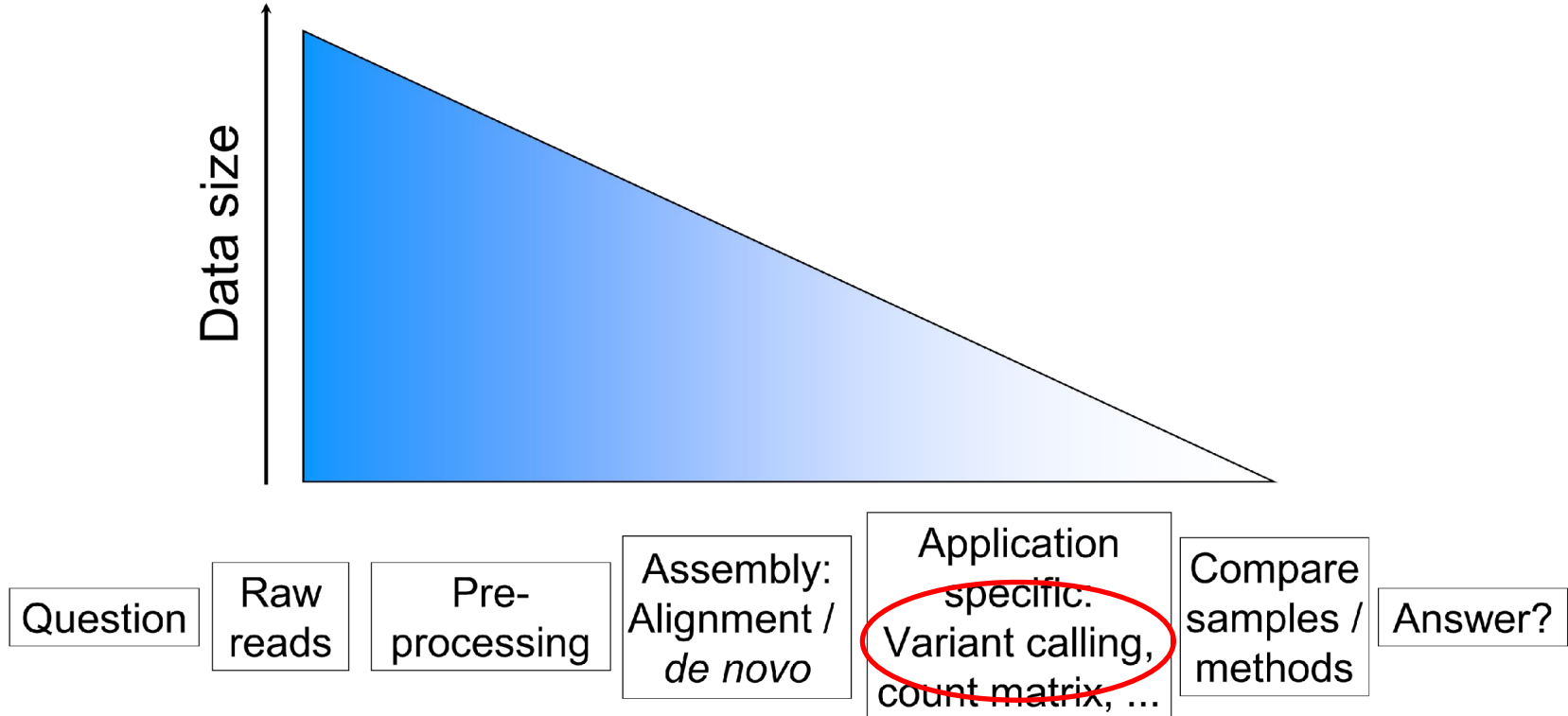
TACAGAT**C**T  
TACAGAT**C**T



# Menu

- Introduction
- From aligned reads to genomic variation
- Alignment post-processing
- Variant effect

# Generalized NGS analysis



# What is genotyping?

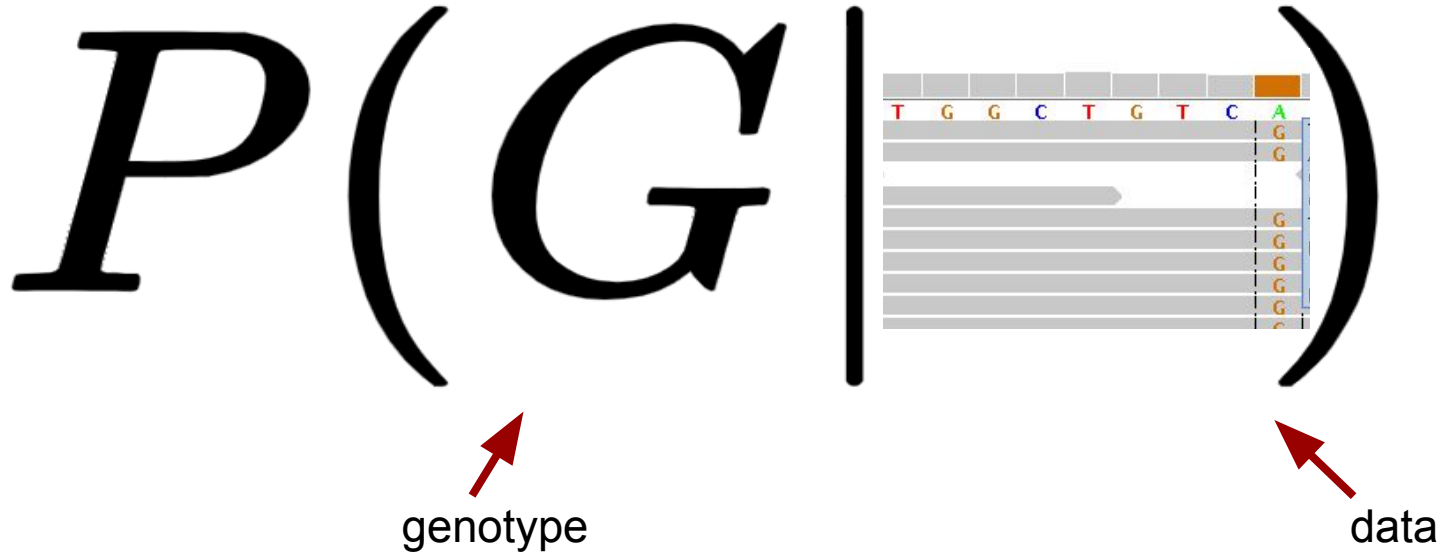
Genotyping is determining which genotype maximizes:

$$P(G|D)$$


The diagram shows the mathematical expression  $P(G|D)$  in a large, bold, serif font. Below the letter 'G', a red arrow points upwards to it, with the word 'genotype' written below the arrow. Similarly, below the letter 'D', a red arrow points upwards to it, with the word 'data' written below the arrow.

# What is genotyping?

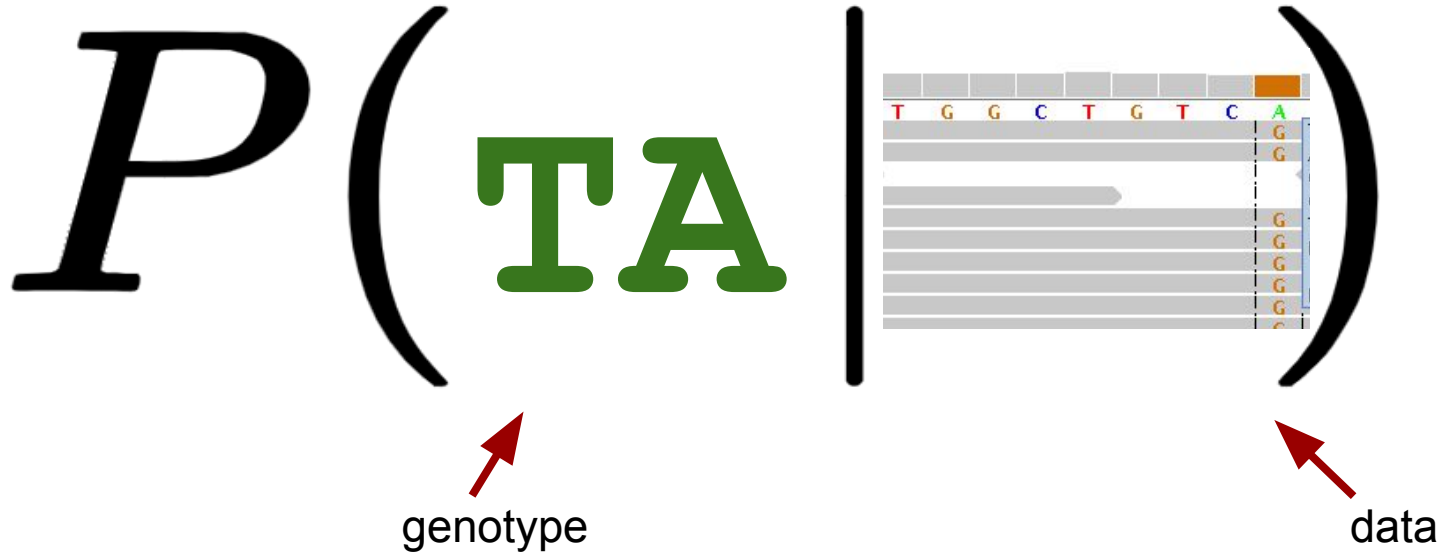
Genotyping is determining which genotype maximizes:





# What is genotyping?

Genotyping is determining which genotype maximizes:



**What is genotyping?**

$$P(G|D) = \frac{P(G)P(D|G)}{P(D)}$$

## What is genotyping?

prior: what is the probability of the genotype to begin with?

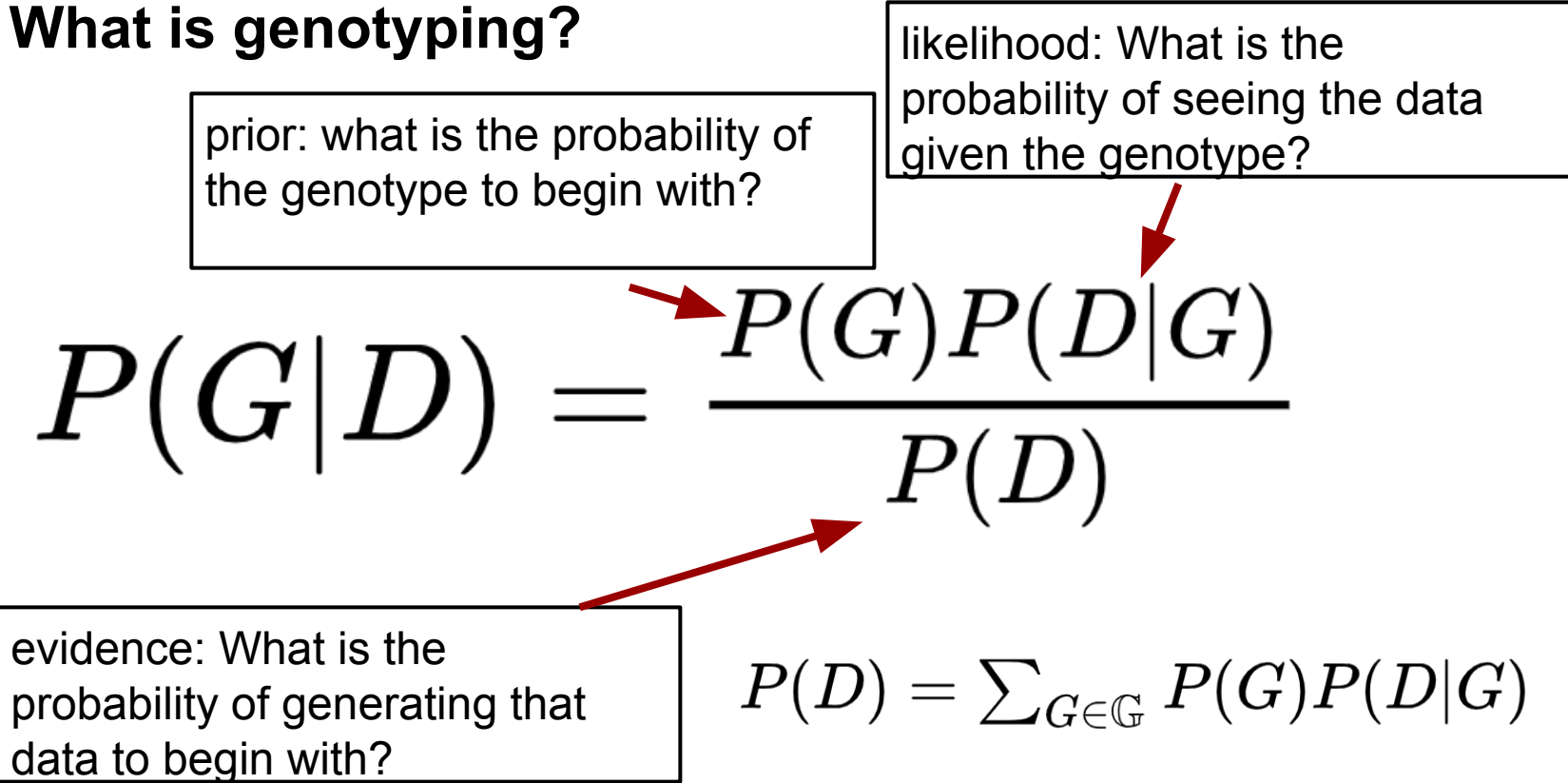
likelihood: What is the probability of seeing the data given the genotype?

$$P(G|D) = \frac{P(G)P(D|G)}{P(D)}$$

# What is genotyping?

prior: what is the probability of the genotype to begin with?

likelihood: What is the probability of seeing the data given the genotype?

$$P(G|D) = \frac{P(G)P(D|G)}{P(D)}$$


evidence: What is the probability of generating that data to begin with?

$$P(D) = \sum_{G \in \mathcal{G}} P(G)P(D|G)$$

## The likelihood

$$P(D|G) = \prod_{b \in \text{READS}} P(b|G)$$

i.e. each reads is an independent observation

# The likelihood $P(D|G)$

Toy example:

G

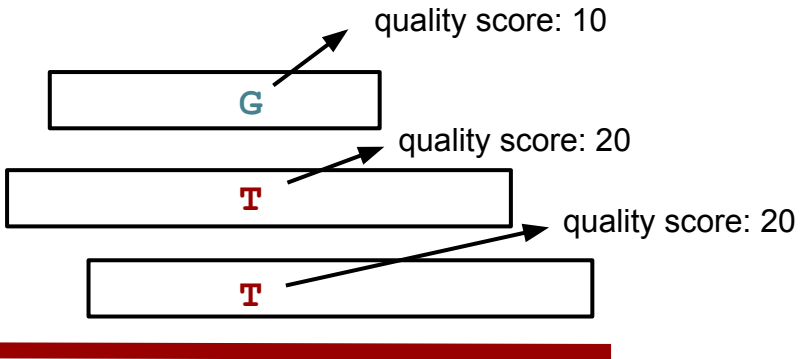
T

T



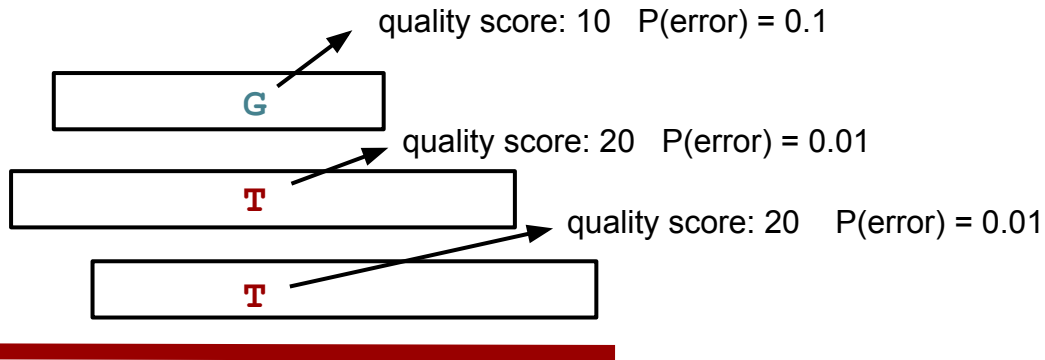
# The likelihood $P(D|G)$

Toy example:



# The likelihood $P(D|G)$

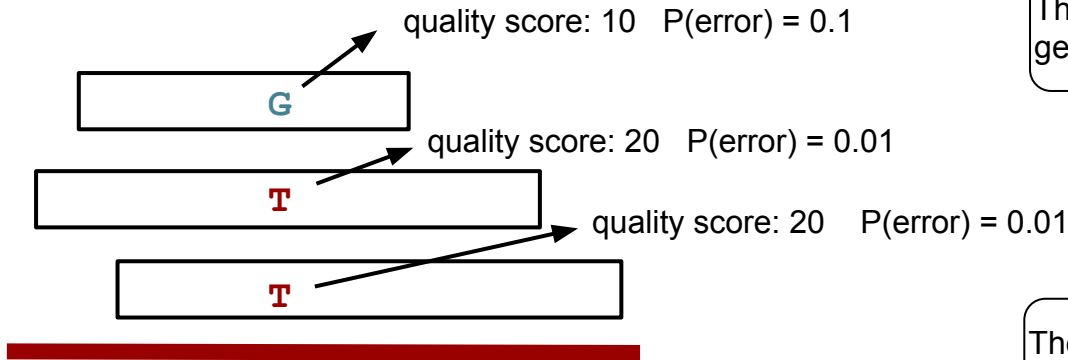
Toy example:





# The likelihood $P(D|G)$

Toy example:



The 2 Ts are sequencing errors!  
The genotype is GG



They are all correct and the  
genotype is GT

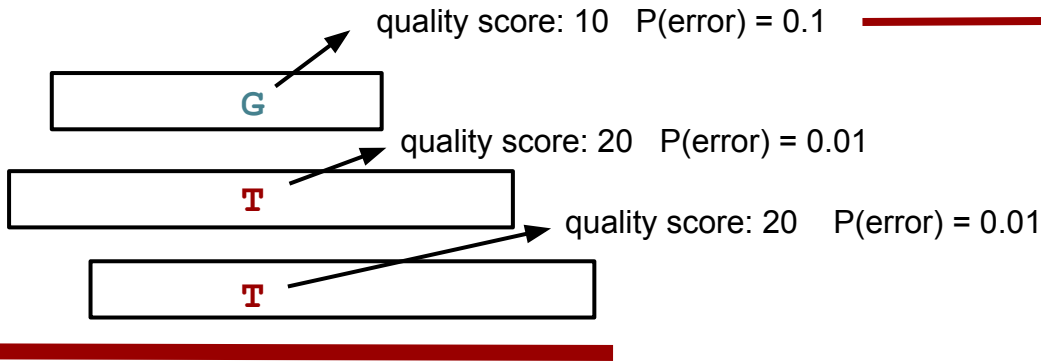


The G is a sequencing error! TT  
is the genotype



# The likelihood $P(D|G)$

Toy example:



# Error model

What I think is the base

probability of the data given the base

$$P(\text{G} | \text{A}) = 0.1 \frac{1}{3}$$

$$P(\text{G} | \text{C}) = 0.1 \frac{1}{3}$$

$$P(\text{G} | \text{G}) = 0.9$$

$$P(\text{G} | \text{T}) = 0.1 \frac{1}{3}$$

Let's evaluate 3 possible genotypes:

# The likelihood $P(D|G)$

Toy example:

GG

GT

TT

quality score: 10  $P(\text{error}) = 0.1$

G

quality score: 20  $P(\text{error}) = 0.01$

T

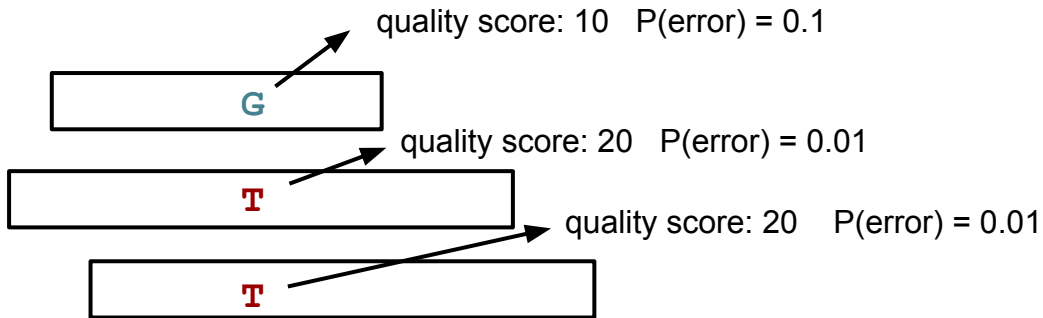
quality score: 20  $P(\text{error}) = 0.01$

T



$$P(D|GG)$$

The likelihood  $P(D|G)$

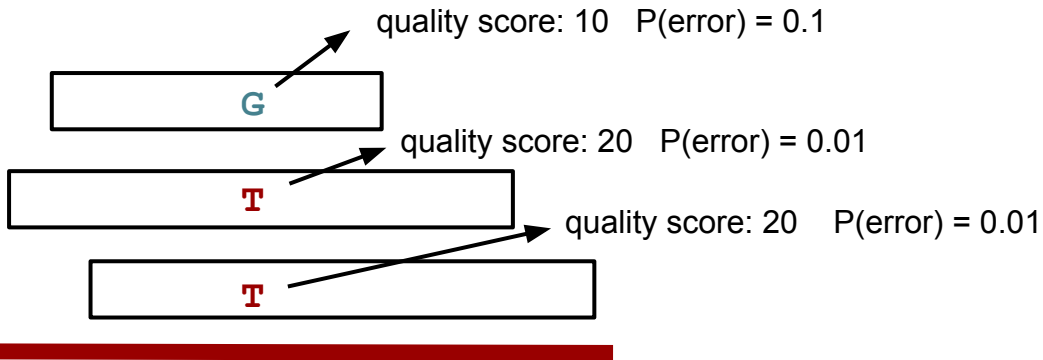


# The likelihood $P(D|G)$

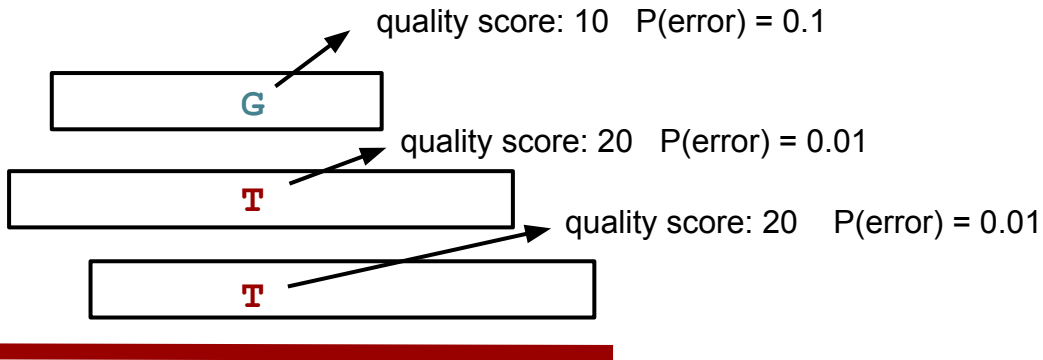
$$P(D|GG)$$

$$\frac{1}{2} G$$

$$\frac{1}{2} G$$



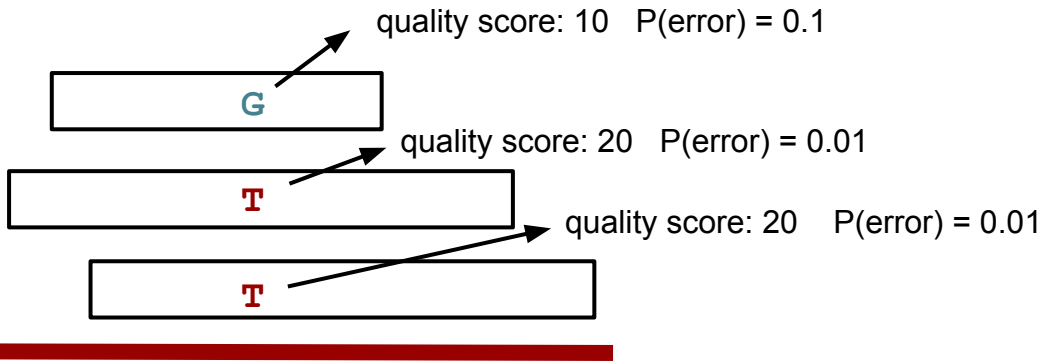
# The likelihood $P(D|G)$



$$P(D|GG)$$

$\frac{1}{2}$	G	$\frac{1}{2}$	G
✓	0.9	✓	0.9

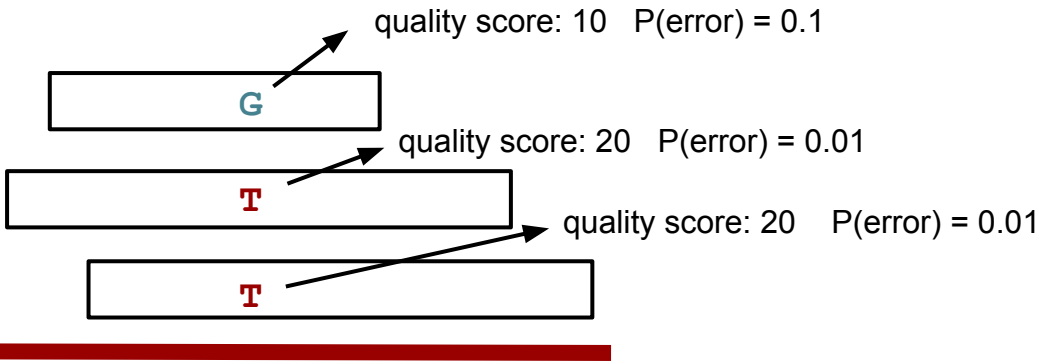
# The likelihood $P(D|G)$



$$P(D|GG)$$

$\frac{1}{2}$	G	$\frac{1}{2}$	G	
	✓	0.9	✓	0.9
	✗	$\frac{0.01}{3}$	✗	$\frac{0.01}{3}$

# The likelihood $P(D|G)$



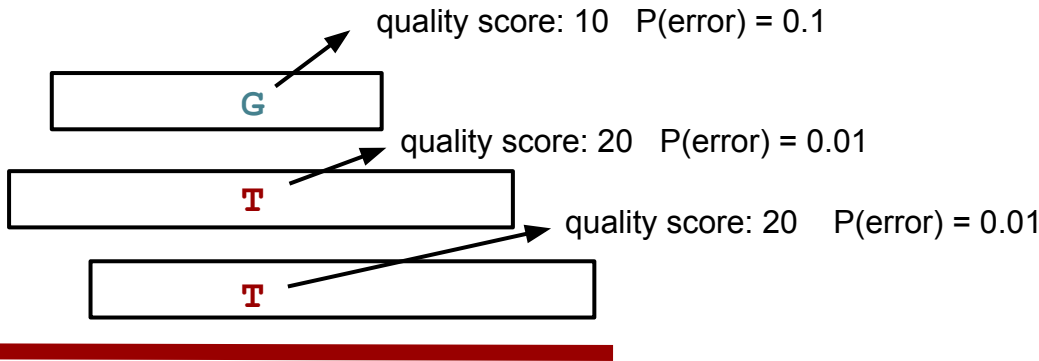
$$P(D|GG)$$

$\frac{1}{2}$ G		$\frac{1}{2}$ G	
✓	0.9	✓	0.9
✗	$\frac{0.01}{3}$	✗	$\frac{0.01}{3}$
✗	$\frac{0.01}{3}$	✗	$\frac{0.01}{3}$



# The likelihood $P(D|G)$

$$P(D|GG)$$

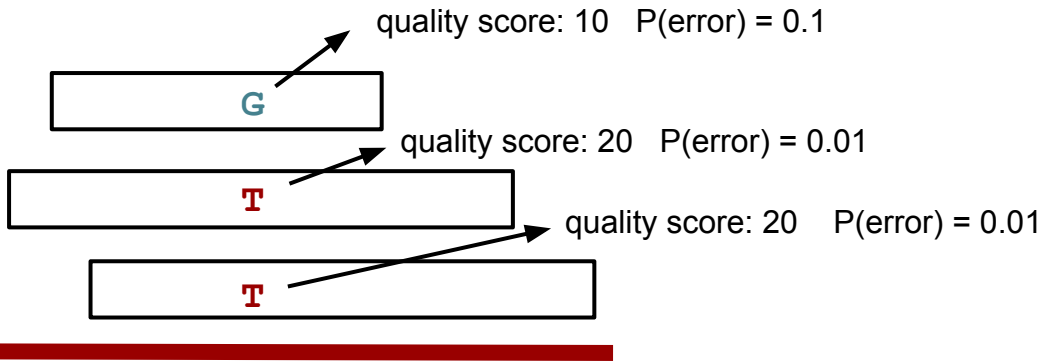


$\frac{1}{2}$ G		$\frac{1}{2}$ G	
✓	0.9	✓	0.9
✗	$\frac{0.01}{3}$	✗	$\frac{0.01}{3}$
✗	$\frac{0.01}{3}$	✗	$\frac{0.01}{3}$

$$\left(\frac{1}{2}0.9 + \frac{1}{2}0.9\right)\left(\frac{1}{2}\frac{0.01}{3} + \frac{1}{2}\frac{0.01}{3}\right)\left(\frac{1}{2}\frac{0.01}{3} + \frac{1}{2}\frac{0.01}{3}\right) = 0.00001$$

# The likelihood $P(D|G)$

$$P(D|GT)$$

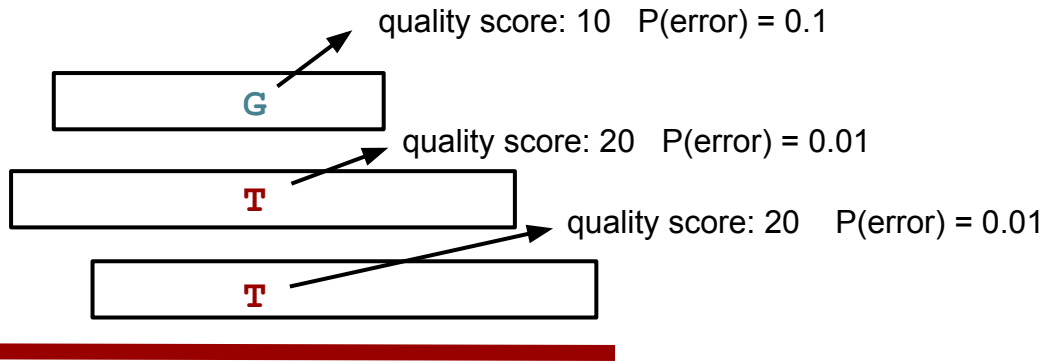


$\frac{1}{2}$ G	$\frac{1}{2}$ T		
✓	0.9	✗	$\frac{0.1}{3}$
✗	$\frac{0.01}{3}$	✓	0.99
✗	$\frac{0.01}{3}$	✓	0.99

$$\left(\frac{1}{2}0.9 + \frac{1}{2}\frac{0.1}{3}\right)\left(\frac{1}{2}\frac{0.01}{3} + \frac{1}{2}0.99\right)\left(\frac{1}{2}\frac{0.01}{3} + \frac{1}{2}0.99\right) = 0.1151163$$

# The likelihood $P(D|G)$

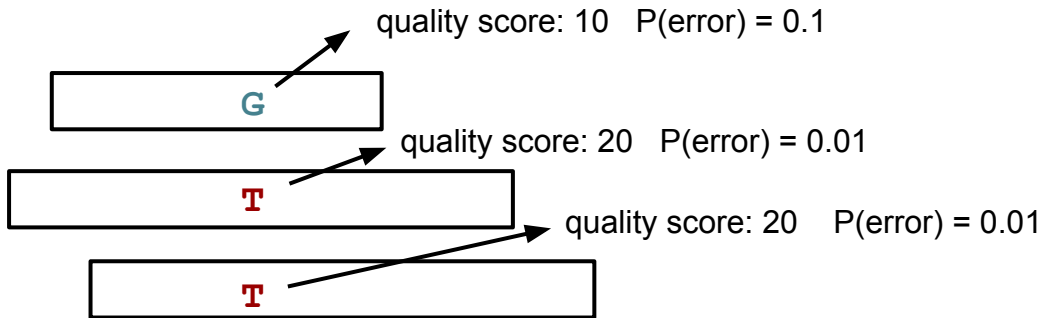
$$P(D|\mathbf{T}\mathbf{T})$$



$\frac{1}{2}$	<b>T</b>	$\frac{1}{2}$	<b>T</b>
<b>X</b>	$\frac{0.1}{3}$	<b>X</b>	$\frac{0.1}{3}$
<b>✓</b>	0.99	<b>✓</b>	0.99
<b>✓</b>	0.99	<b>✓</b>	0.99

$$\left(\frac{1}{2} \frac{0.1}{3} + \frac{1}{2} \frac{0.1}{3}\right) \left(\frac{1}{2} 0.99 + \frac{1}{2} 0.99\right) \left(\frac{1}{2} 0.99 + \frac{1}{2} 0.99\right) = 0.03267$$

## The likelihood $P(D|G)$



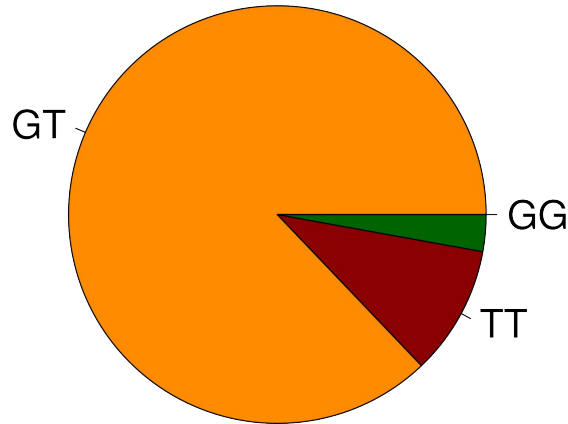
$$P(D|\mathbf{GG}) = 0.00001$$

$$P(D|\mathbf{GT}) = 0.11511$$

$$P(D|\mathbf{TT}) = 0.0327$$

The likelihood  $P(D|G)$ .

A likelihood in itself  
is not meaningful,  
you need to  
compare it to other  
models



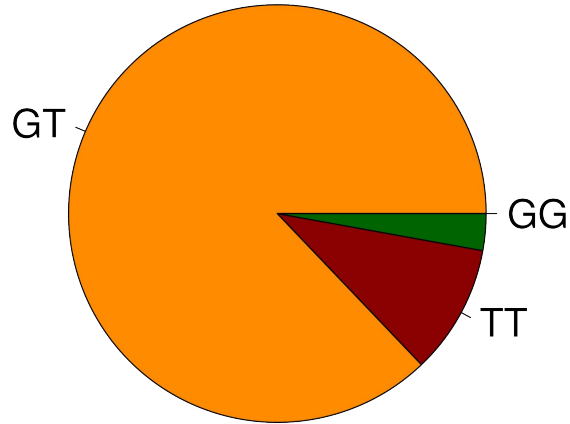
$$P(D|\mathbf{GG}) = 0.00001$$

$$P(D|\mathbf{GT}) = 0.11511$$

$$P(D|\mathbf{TT}) = 0.0327$$

$$P(D) = P(\mathbf{GG})P(D|\mathbf{GG}) + P(\mathbf{GT})P(D|\mathbf{GT}) + P(\mathbf{TT})P(D|\mathbf{TT})$$

## The likelihood $P(D|G)$



We will neglect  
the genotype  
prior this time

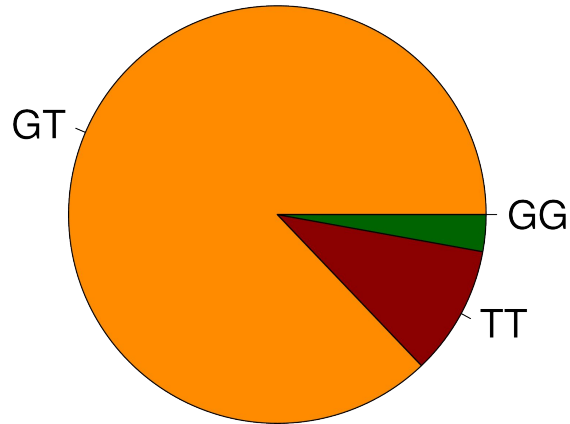
$$P(D|GG) = 0.00001$$

$$P(D|GT) = 0.11511$$

$$P(D|TT) = 0.0327$$

$$P(G|D) = \frac{P(G)P(D|G)}{P(D)}$$

## The likelihood $P(D|G)$

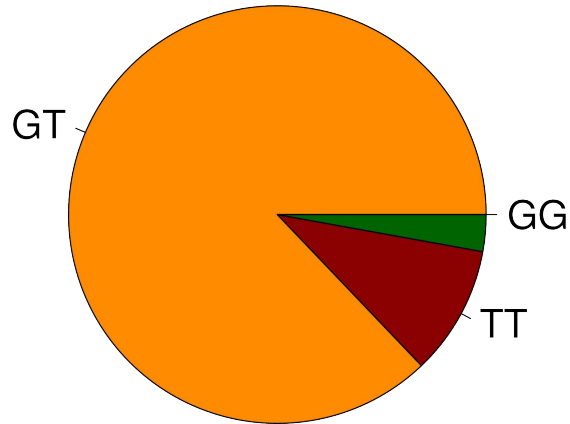


$$P(\mathbf{GG}|D) = 6.7\mathbf{e}-05$$

$$P(\mathbf{GT}|D) = 0.77888$$

$$P(\mathbf{TT}|D) = 0.22104$$

## The likelihood $P(D|G)$



$$P(\text{GG}|D) = 6.7 \times 10^{-5}$$

$$P(\text{GT}|D) = 0.77888$$

$$P(\text{TT}|D) = 0.22104$$

**Important point:** More coverage  $\rightarrow$  More multiplications  $\rightarrow$  The relative difference between models become larger



## The likelihood $P(D|G)$

$$P(\mathbf{GG}|D) = 6.7e-05$$

$$P(\mathbf{GT}|D) = 0.77888$$

$$P(\mathbf{TT}|D) = 0.22104$$

PHRED

41.70

1.09


6.56

PHRED-scaled

40.60

0.00

5.47



What is the difference between the **most likely** and **second most likely**

## Details I did not cover

- Error model
  - Most genotypers do not simply use raw quality scores

# Most common genotypers

- GATK
- SAMtools/BCFtools
- GraphTyper
- FreeBayes

# Deep Learning and genotyping?

Published: 24 September 2018

## A universal SNP and small-indel variant caller using deep neural networks

[Ryan Poplin](#), [Pi-Chuan Chang](#), [David Alexander](#), [Scott Schwartz](#), [Thomas Colthurst](#), [Alexander Ku](#), [Dan Newburger](#), [Jojo Dijamco](#), [Nam Nguyen](#), [Pegah T Afshar](#), [Sam S Gross](#), [Lizzie Dorfman](#), [Cory Y McLean](#) & [Mark A DePristo](#) 

*Nature Biotechnology* **36**, 983–987 (2018) | [Cite this article](#)

**26k** Accesses | **196** Citations | **319** Altmetric | [Metrics](#)

### Abstract


Despite rapid advances in sequencing technologies, accurately calling genetic variants present in an individual genome from billions of short, errorful sequence reads remains challenging. Here we show that a deep convolutional neural network can call genetic variation in aligned next-generation sequencing read data by learning statistical

## Accurate, scalable cohort variant calls using DeepVariant and GLnexus

[Taedong Yun](#), [Helen Li](#), [Pi-Chuan Chang](#), [Michael F Lin](#), [Andrew Carroll](#), [Cory Y McLean](#) 

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*Bioinformatics*, Volume 36, Issue 24, 15 December 2020, Pages 5582–5589,  
<https://doi.org/10.1093/bioinformatics/btaa1081>

**Published:** 05 January 2021 **Article history** 

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### Abstract

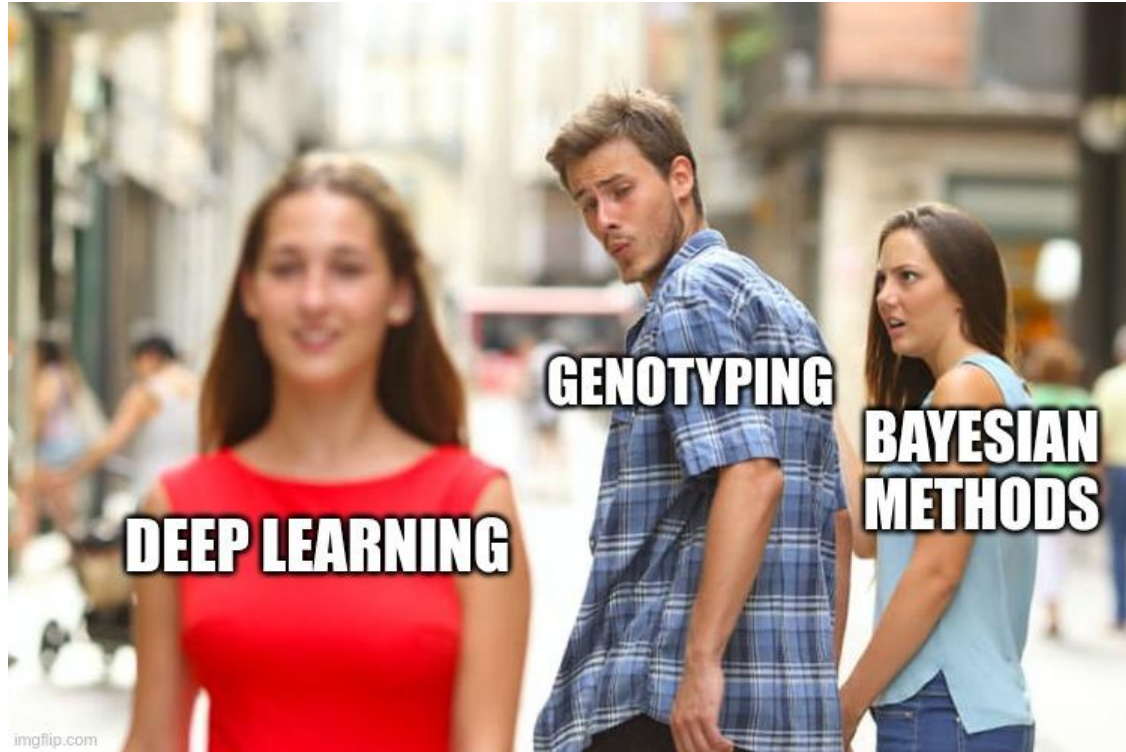
#### Motivation

Population-scale sequenced cohorts are foundational resources for genetic analyses, but processing raw reads into analysis-ready cohort-level variants remains challenging.

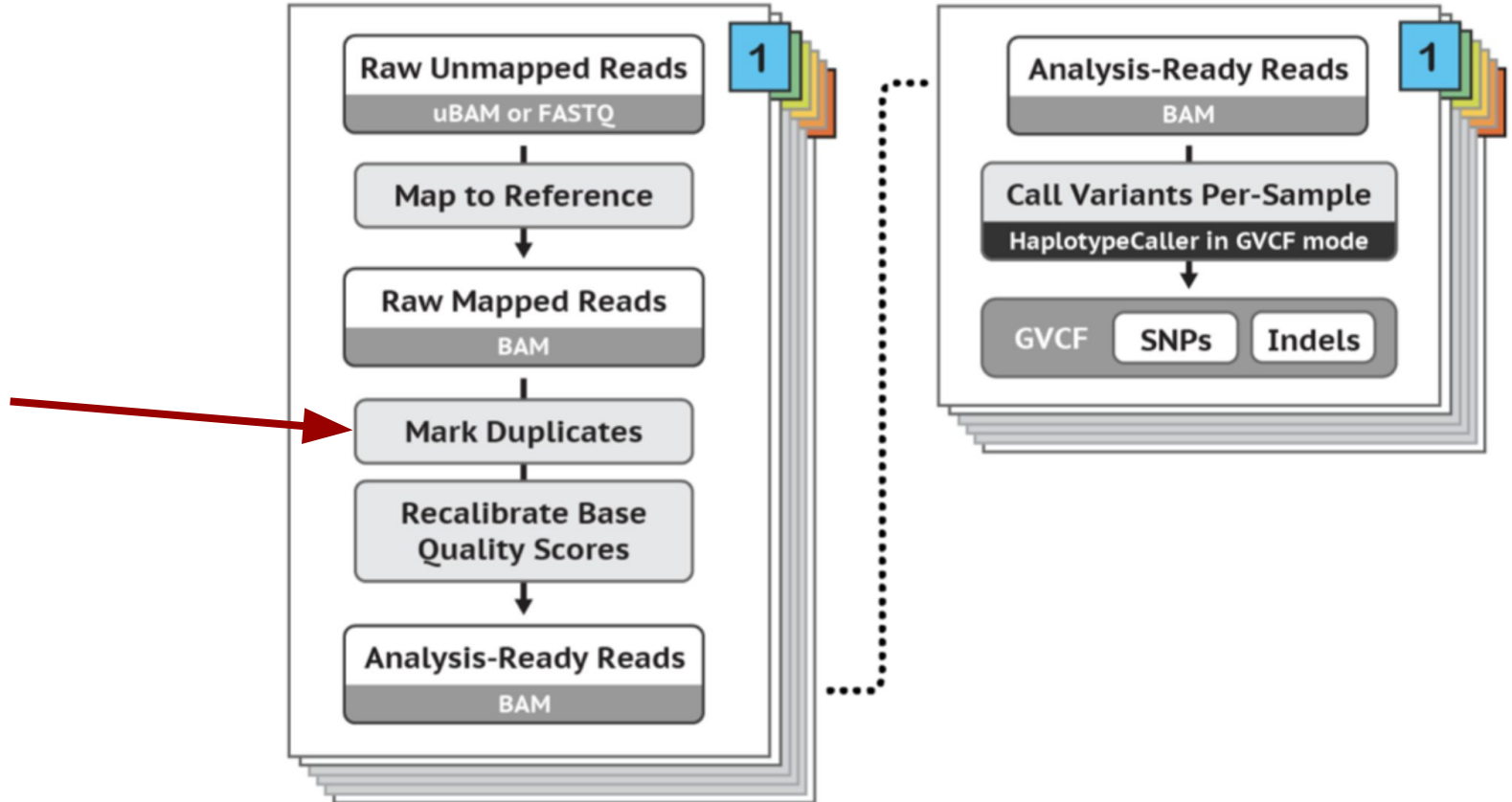
#### Results

We introduce an open-source cohort-calling method that uses the highly accurate caller DeepVariant and scalable merging tool GLnexus. Using callset

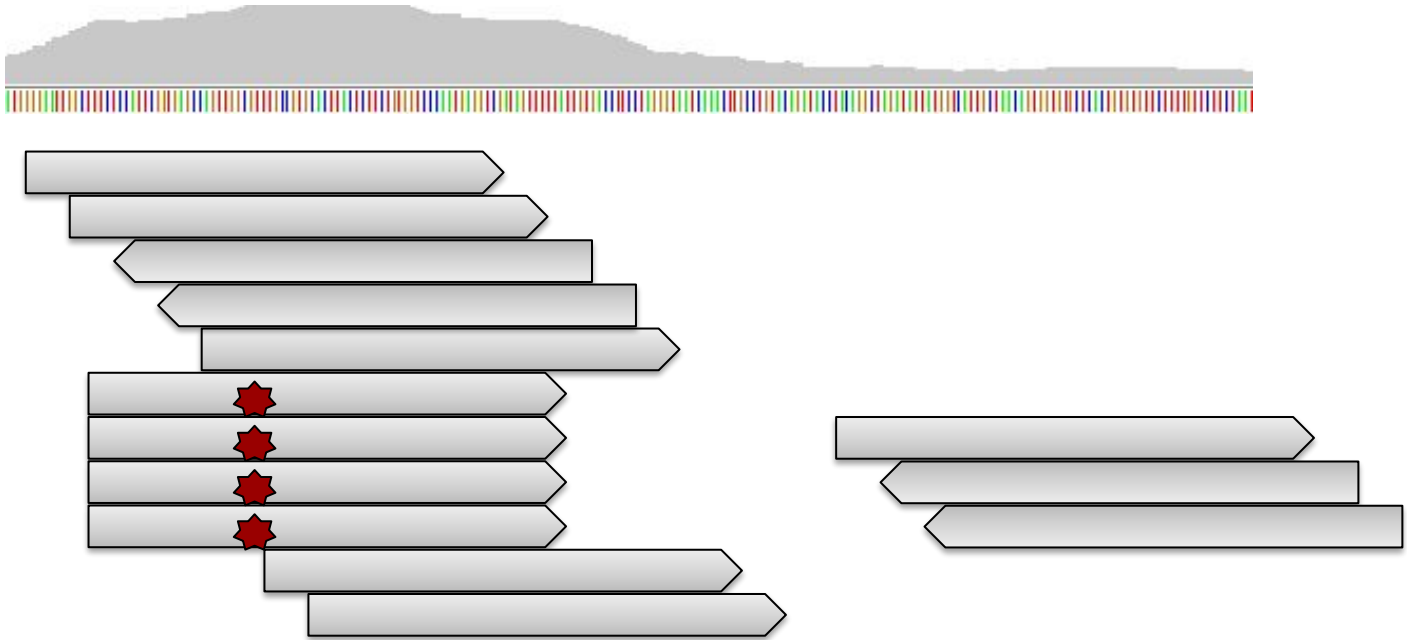
# Deep Learning and genotyping?



# GATK's recommended workflow



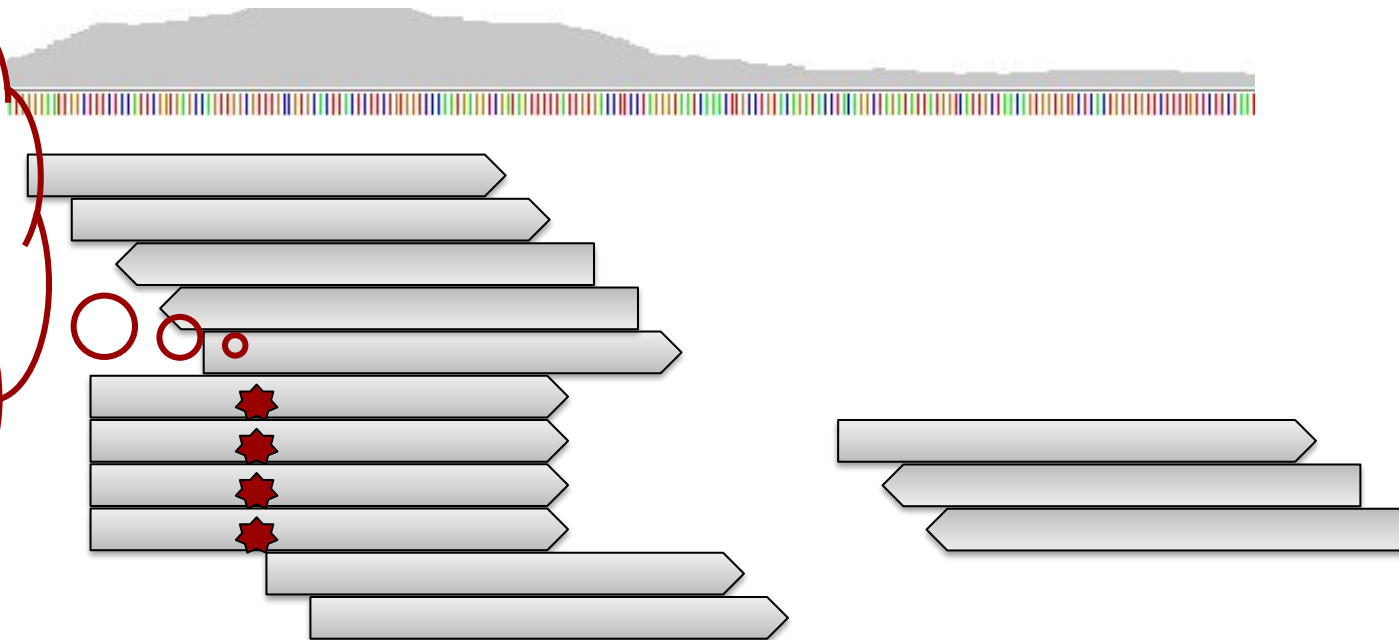
The PCR amplification step included in the majority of NGS library construction techniques can introduce duplicates in the data.



We want: remove or mark them to avoid false calls

genotyper:

the site below  
is probably  
**heterozygous**  
(i.e. the ★ is  
the second  
allele)

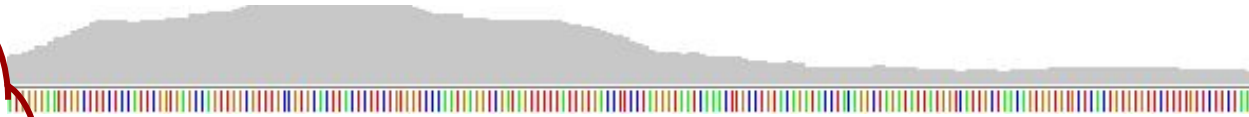





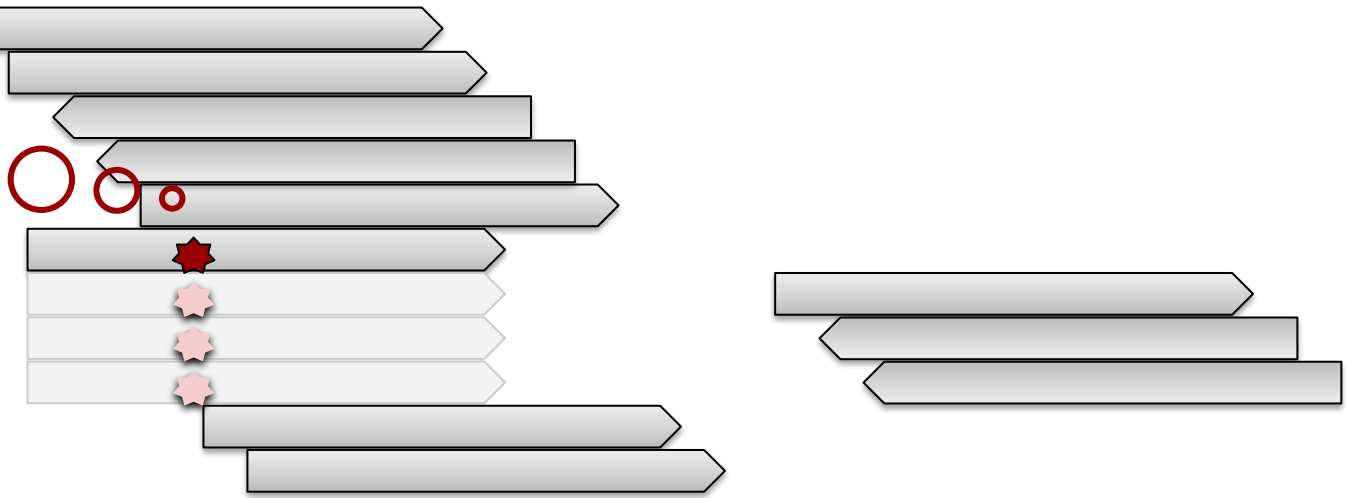
Genotypers will ignore reads marked as duplicates



genotyper:



the site below  
is probably  
**homozygous**  
(i.e. the  is a  
seq. error



# Duplicate/marking removal

Basic concepts of duplicate marking algorithm:

- Identify genomic position and strand for 5'-most bases.
- Mark reads that are duplicates of each other.
- Within a group of duplicate reads, the read with the highest sum of base quality scores is retained.

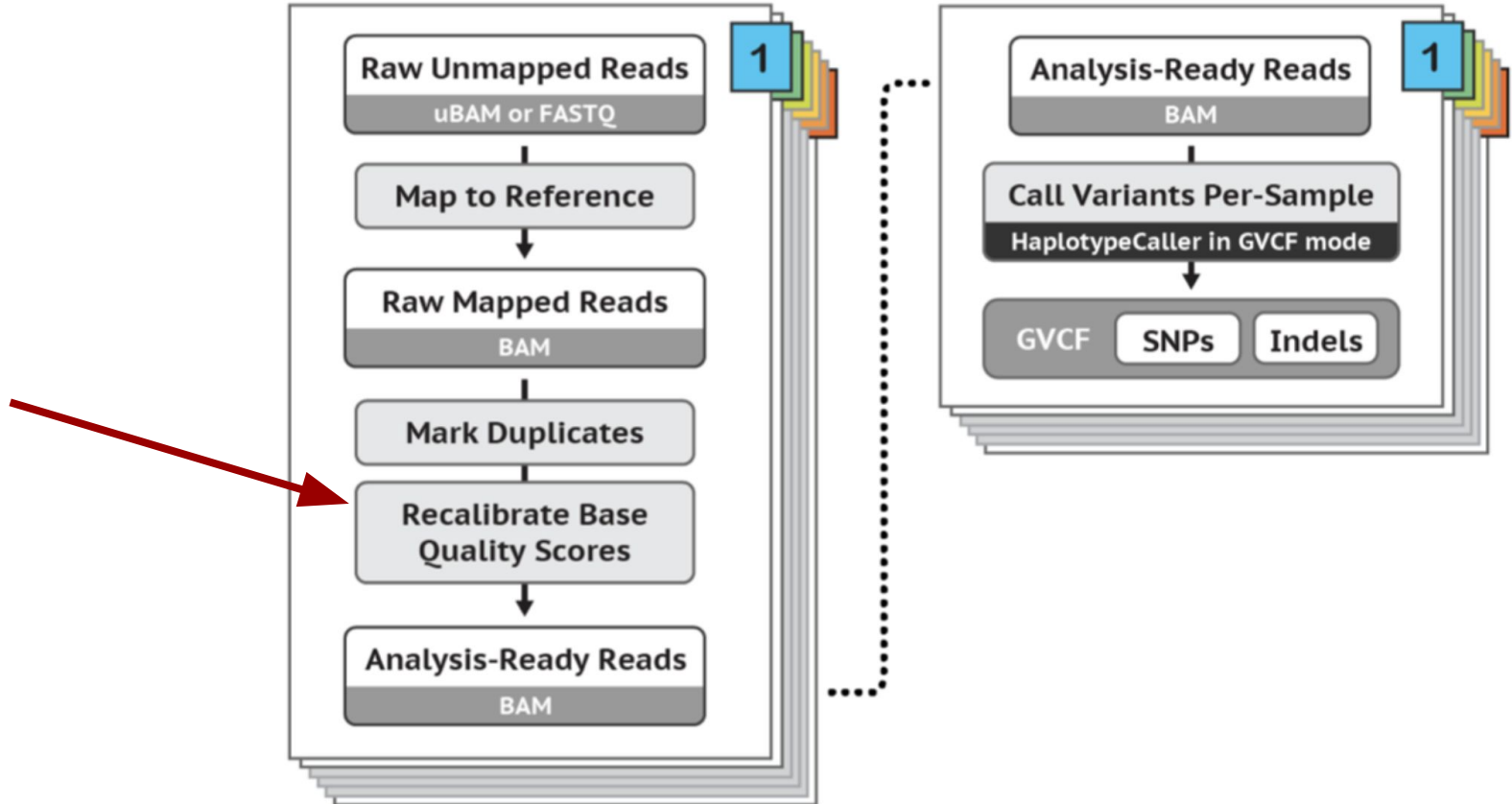
<http://picard.sourceforge.net/>

# Duplicate/marking removal

Problems:

- Does not account for sequencing errors.
- Does not account for natural duplicates.
- Does not account for duplicate reads with different mapping locations.

# GATK's recommended workflow



# Base quality score recalibration?

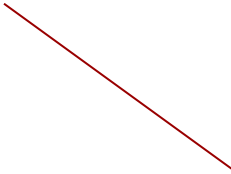
- remember those?

@ILLUMINA-C90280\_0030\_FC:5:1:2675:1090#NNNNNN/1

ATTCCCGGCCTTTTTCCAGGCCTGCCTGCTCGAGC

+

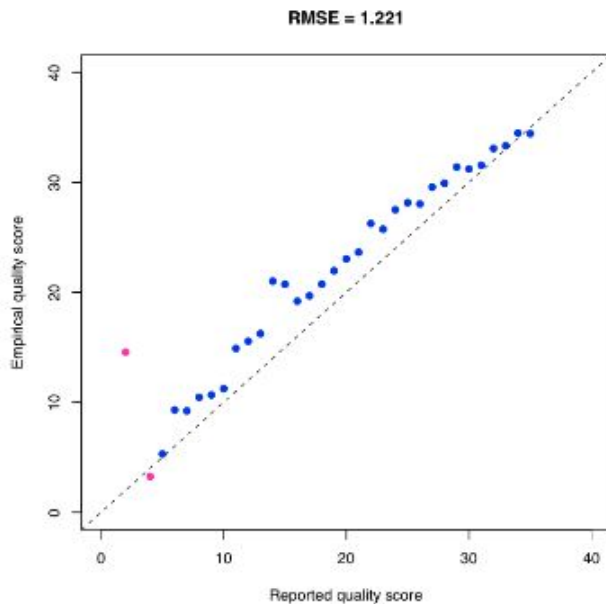
BAAAGECEE<EEDFEDF3DBDBB=A+==>9>>88?



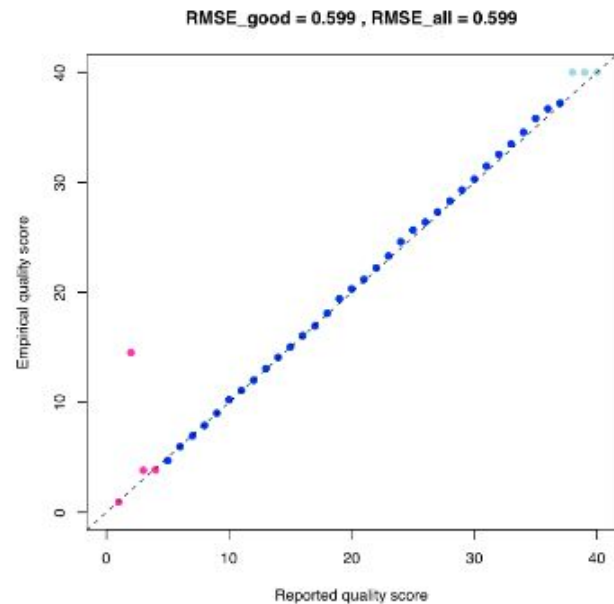
- There are supposed to reflect  $P(\text{error})$
- They are not always accurate: problem for genotyping

# Reported Quality vs. Empirical Quality

Idea: use documented variants in the genome



Original Data



After GATK Recalibration

# The Missing Diversity in Human Genetic Studies

[Giorgio Sirugo](#)  <sup>6</sup>  • [Scott M. Williams](#)  <sup>6</sup>  • [Sarah A. Tishkoff](#)  <sup>6</sup>  • [Show footnotes](#)

DOI: <https://doi.org/10.1016/j.cell.2019.02.048> •



The majority of studies of genetic association with disease have been performed in Europeans. This European bias has important implications for risk prediction of diseases across global populations. In this commentary, we justify the need to study more diverse populations using both empirical examples and theoretical reasoning.

# Base quality score recalibration

To work we need:

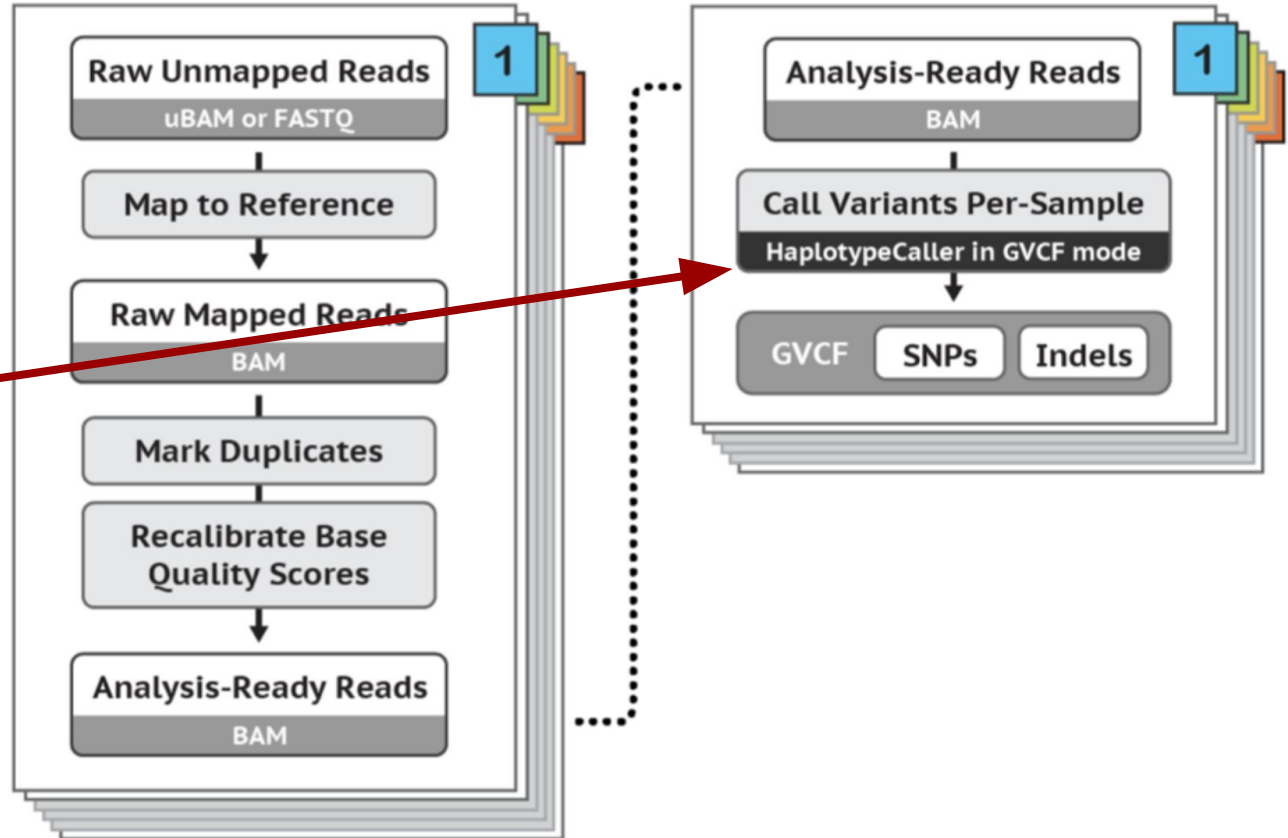
- East Asian or European (as in mostly West European) samples
- WGS
- Sufficient coverage

My biased opinion:

- Just don't bother



# GATK's recommended workflow



We covered this before

# Variant call format (VCF)

- Details which variants have been called
- Can be bgzip (block gzip) and indexed using tabix
- Using tabix, queries can be made like:
  - return all variants in the region chr22:323,340-361,152

# Variant call format (VCF)

```
20 51391523 . A G 173.96. AC=2;DP=5;MQ=52.03 GT:AD:DP:GQ:PL 1/1:0,5:5:15:188,15,0
20 51392469 . C T 146.14. AC=2;DP=4;MQ=60.00 GT:AD:DP:GQ:PL 1/1:0,4:4:12:160,12,0
20 51394015 . T C 97.64 . AC=1;DP=6;MQ=60.00 GT:AD:DP:GQ:PL 0/1:3,3:6:66:105,0,66
20 51395647 . A C 89.64 . AC=1;DP=7;MQ=57.28 GT:AD:DP:GQ:PL 0/1:4,3:7:97:97,0,100
20 51397399 . C T 93.64 . AC=1;DP=7;MQ=60.00 GT:AD:DP:GQ:PL 0/1:4,3:7:99:101,0,120
20 51402308 . C T 161.64. AC=1;DP=9;MQ=60.00 GT:AD:DP:GQ:PL 0/1:3,6:9:63:169,0,63
```

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

name of chromosome (ex: chr1, chr2 ...)

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

coordinate on chromosome

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

ID (ex: rs23534)

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

reference base

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

alternative base



# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

quality field

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Filter (ex: 'LowQual')

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Info field ex:

AC= allele count

DP = depth

MQ = root mean square of the mapping quality

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Format field, what do the next fields mean?

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Most likely genotype

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Allele distribution

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Depth

# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

Genotype quality



# Variant call format (VCF)

20	51391523	.	A	G	173.96.	AC=2;DP=5;MQ=52.03	GT:AD:DP:GQ:PL	1/1:0,5:5:15:188,15,0
20	51392469	.	C	T	146.14.	AC=2;DP=4;MQ=60.00	GT:AD:DP:GQ:PL	1/1:0,4:4:12:160,12,0
20	51394015	.	T	C	97.64	AC=1;DP=6;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,3:6:66:105,0,66
20	51395647	.	A	C	89.64	AC=1;DP=7;MQ=57.28	GT:AD:DP:GQ:PL	0/1:4,3:7:97:97,0,100
20	51397399	.	C	T	93.64	AC=1;DP=7;MQ=60.00	GT:AD:DP:GQ:PL	0/1:4,3:7:99:101,0,120
20	51402308	.	C	T	161.64.	AC=1;DP=9;MQ=60.00	GT:AD:DP:GQ:PL	0/1:3,6:9:63:169,0,63

PHRED-scaled likelihood

# The likelihood $P(D|G)$

PHRED

PHRED-scaled

$$P(\mathbf{GG} | D) = 6.7e-05$$

41.70

40.60

$$P(\mathbf{GT} | D) = 0.77888$$

1.09

0.00

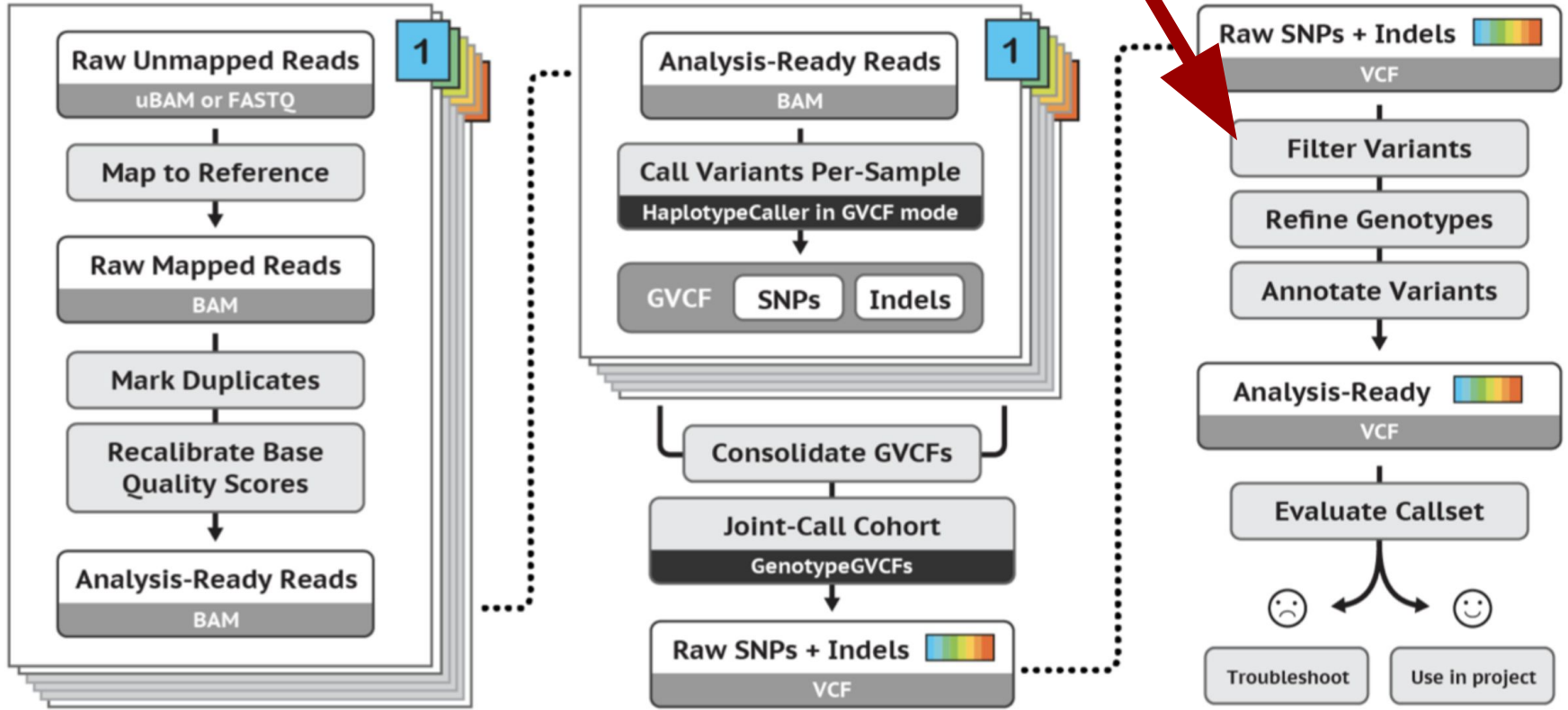
$$P(\mathbf{TT} | D) = 0.22104$$

6.56

5.47



# GATK's recommended workflow



# Later, we will...

- Filter variants
- Annotate the variants
- Other types of variants
- Final considerations about genomic variants

... but for now we have:

- removed duplicates to get independent observations
- Used them to call the most likely genotype
- Saw the VCF format

# Exercise time!

[http://teaching.healthtech.dtu.dk/22126/index.php/Postprocess\\_exercise](http://teaching.healthtech.dtu.dk/22126/index.php/Postprocess_exercise)