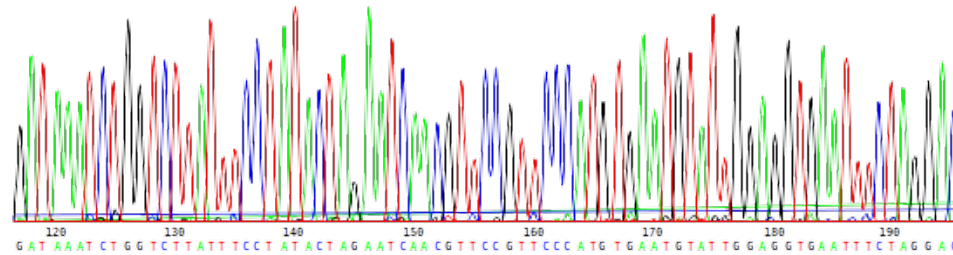
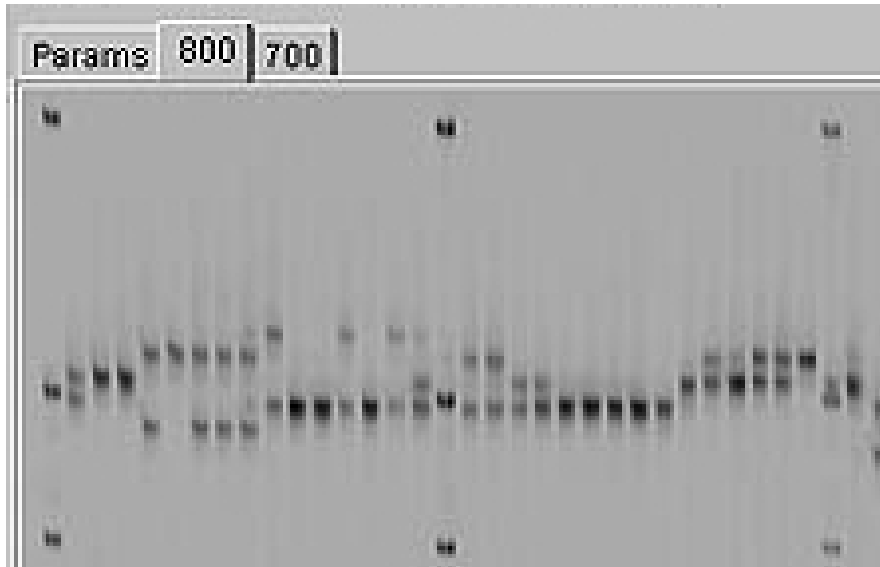


# Genotype likelihoods, allele frequencies, and SNP calling from NGS data

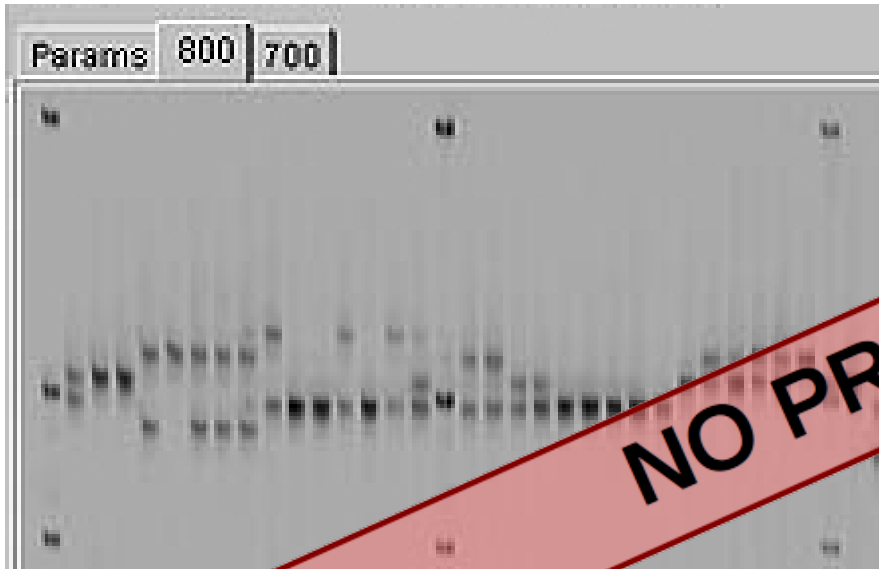
Tyler Linderoth  
Physalia lcWGS course 2023

# Why Probabilistic Methods?

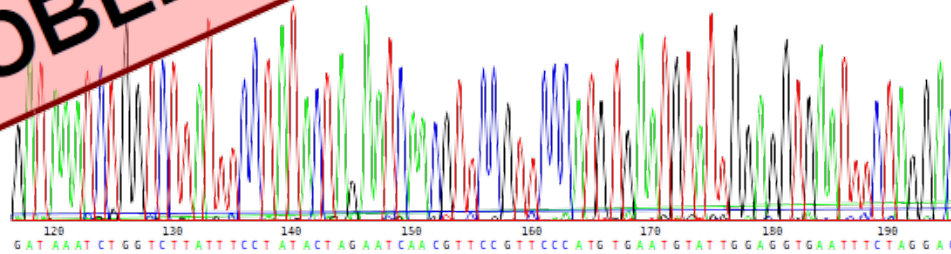


[http://www.licor.com/bio/products/software/saga\\_gt/details.html](http://www.licor.com/bio/products/software/saga_gt/details.html)

# Why Probabilistic Methods?

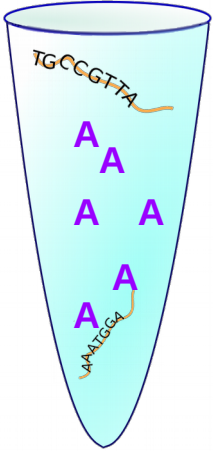


**NO PROBLEM!**

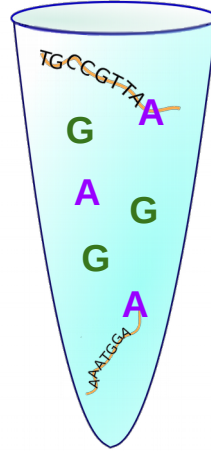


[http://www.ilcor.com/bio/products/software/saga\\_gt/details.html](http://www.ilcor.com/bio/products/software/saga_gt/details.html)

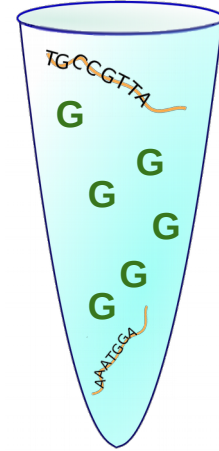
# Why Probabilistic Methods?



The library for an individual homozygous for the **A** allele will consist only of **As**.

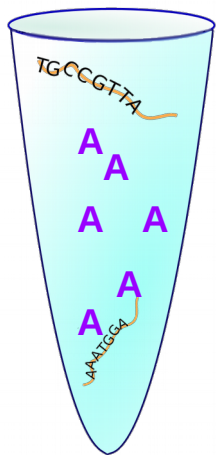


The library for a heterozygous individual at a site contains both **As** and **Gs**.

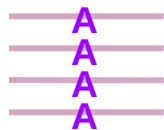


The library for an individual homozygous for the **G** allele will consist only of **Gs**.

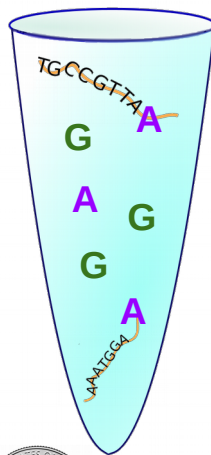
# Why Probabilistic Methods?



Sequence to  
average depth of  
4x.



Expect 4 reads, all with  
**A**s at this ref position.  
Depth  $\sim$  Poisson( $\lambda = 4$ ).

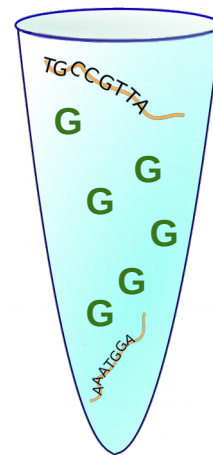


vs

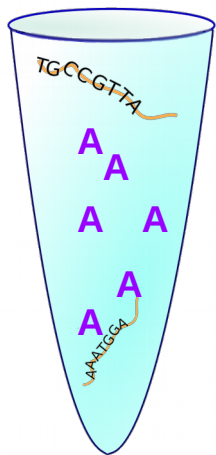
Sequencing (sampling)  
the two different alleles  
is just like flipping a  
coin.



$P(k \text{ A alleles})$   
 $\sim \text{Binom}(n \text{ reads}, p=0.5)$



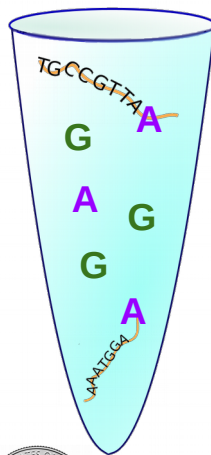
# Why Probabilistic Methods?



Sequence to  
average depth of  
4x.

A  
A  
A  
A

Expect 4 reads, all with  
**As** at this ref position.  
Depth  $\sim$  Poisson( $\lambda = 4$ ).



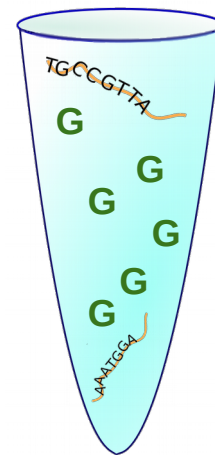
vs



Sequencing (sampling)  
the two different alleles  
is just like flipping a  
coin.

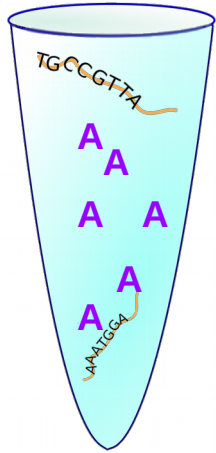
A  
A  
G  
A

$P(k \text{ A alleles})$   
 $\sim \text{Binom}(n \text{ reads}, p=0.5)$



G  
G  
G  
G

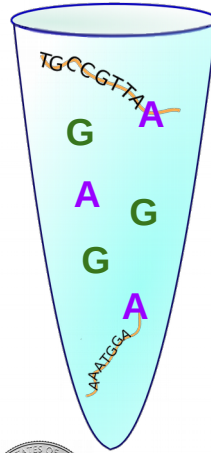
# Why Probabilistic Methods?



A sequencing error occurs. This can occur at rates of around 0.1% in Illumina data.



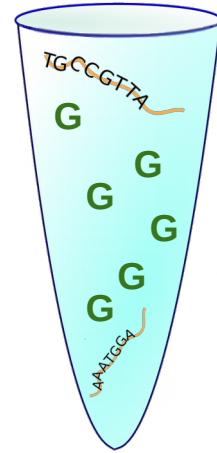
Expect 4 reads, all with  
**As** at this ref position.  
Depth  $\sim$  Poisson( $\lambda = 4$ ).



Sequencing (sampling)  
the two different alleles  
is just like flipping a coin.



$P(k \text{ A alleles})$   
 $\sim \text{Binom}(n \text{ reads}, p=0.5)$



## A basic model for a diploid individual's genotype

paternal allele



maternal allele



Product over all reads  
covering a site for an  
individual

$$P(X|G=bh) = \prod_{i=1}^r \left( \frac{L_b^{(i)}}{2} + \frac{L_h^{(i)}}{2} \right)$$

$$b, h \in \{A, C, G, T\}$$



Example for an individual with observed sequencing reads **AAAG** at a site.

$$P(X|G=bh) = \prod_{i=1}^r \left( \frac{L_b^{(i)}}{2} + \frac{L_h^{(i)}}{2} \right)$$

10 potential genotypes for a  
diploid individual

---

AA

AC

AG

AT

CC

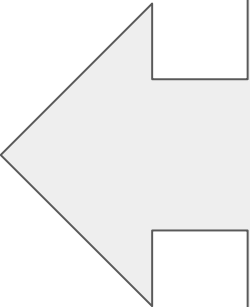
CG

CT

GG

GT

TT



For every genotype,  
we could calculate it's  
likelihood by iterating  
over all of the  
observed reads:  
AAAG.

Let's calculate the  
likelihood for the AC  
genotype.

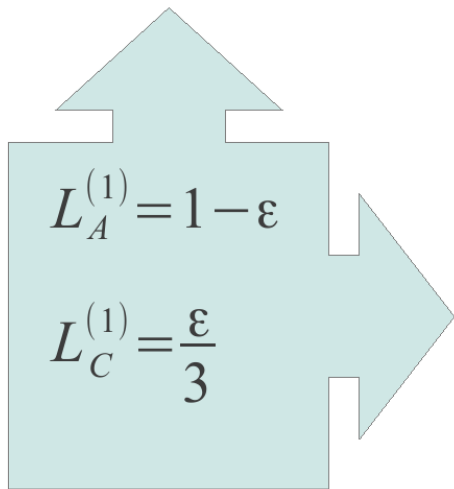
Example for an individual with observed sequencing reads **AAAG** at a site.

$$P(X|G=bh) = \prod_{i=1}^r \left( \frac{L_b^{(i)}}{2} + \frac{L_h^{(i)}}{2} \right)$$

A,A,A,G

$$P(X|G=AC) = \left( \frac{L_A^{(1)}}{2} + \frac{L_C^{(1)}}{2} \right) * \left( \frac{L_A^{(2)}}{2} + \frac{L_C^{(2)}}{2} \right) * \left( \frac{L_A^{(3)}}{2} + \frac{L_C^{(3)}}{2} \right) * \left( \frac{L_A^{(4)}}{2} + \frac{L_C^{(4)}}{2} \right)$$

||  
A



$\epsilon$  = Probability of an error

There are 3 potential erroneous reads for a error to turn into, hence the  $1/3 * \epsilon$

$$P(X=A|G=AC) = \frac{1-\epsilon}{2} + \frac{\epsilon}{6}$$

Example for an individual with observed sequencing reads **AAAG** at a site.

Genotype	Likelihood (log10)	
AA	-2.49	
<b>AC</b>	<b>-3.38</b>	
AG	-1.22	A
AT	-3.38	A
CC	-9.91	A
CG	-7.74	G
CT	-9.91	$\epsilon = 0.01$
GG	-7.44	
GT	-7.74	
TT	-9.91	

# Genotype calling

Genotype	Likelihood (log10)
AA	-2.49
AC	-3.38
<b>AG</b>	<b>-1.22</b>
AT	-3.38
CC	-9.91
CG	-7.74
CT	-9.91
GG	-7.44
GT	-7.74
TT	-9.91

AAAG &  $\epsilon = 0.01$

What is the genotype?  
AG.

## Maximum Likelihood

The simplest genotype caller: choose the genotype with the highest likelihood.

This is essentially what you are doing in ANGSD when you choose a uniform prior probability distribution for the genotypes.

# Major and minor alleles

## Likelihood function


$$\log P(D|G = A) = \sum_{i=1}^R \log L_{A_j,i}$$

AAAG &  $\epsilon = 0.01$

Allele	Likelihood
<b>A</b>	<b>-2.49</b>
C	-3.38
<b>G</b>	<b>-1.22</b>
T	-3.38

We can reduce the genotype space to 3 entries (from 10, for diploids).

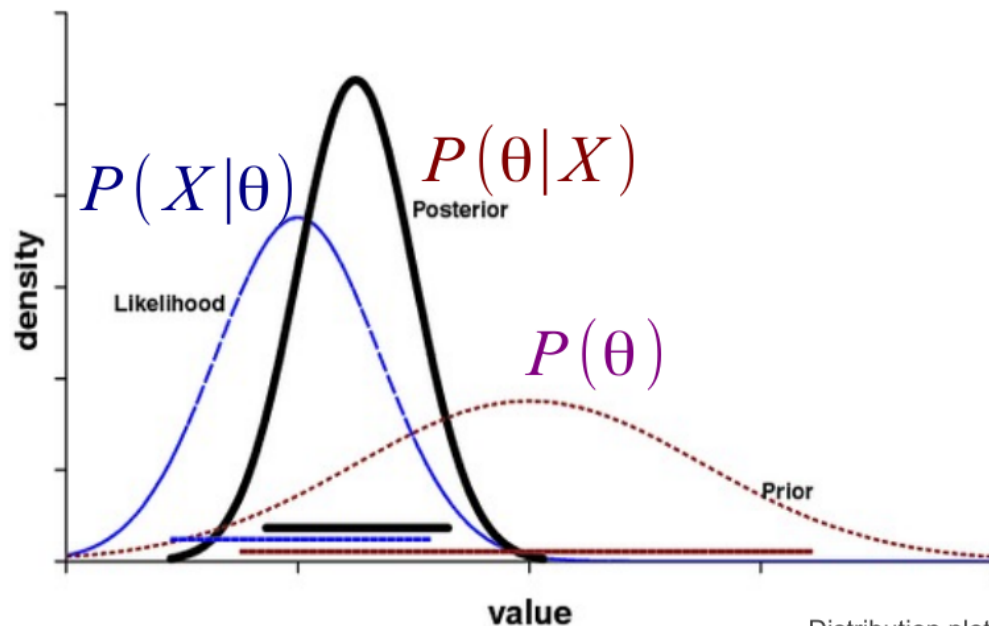
Can we somehow use other information present in our data to further increase our genotype calling accuracy?

6	.....,	DEGEGG	9	.....,	DABGIIIIII	5	.....	>AB/A	6	.....	FGG	
6	.....,	DEGEGG	9	.....,	DABGIIIIII	5	.....	>ABDA	6	.....	3GGGGG	
6	.....,	DEGEGG	9	.....,	DABGIIIIII	5	.....	>ABDA	6	.....	3GGGBG	
6	.....,	DEGEGG	10	.....,^].	DABGIIIIIE	5	.....	>AB/A	6	.....	3GGGBG	
6	.....,	DEGEGG	10	.....,	DABGIIIIII	5	.....	>AB/A	6	.....	3GGGBG	
6	.....,	DEGEGG	10	.....,	DABGIIIIII	5	.....	>ABDA	6	.....	BGGGBG	
6	.....,	DEGEGG	10	.....,	DABGIIIIII	5	.....	>ABDA	6	C,...	5G/BGB	
7	.....,^].	DEGEGGE	10	.....,	DABGIIIIII	5	.....	>ABDA	6	.....	5GIBGB	
7	.....,	DEGEGGG	10	.....,	DABGIIIIII	5	.....	>ABDA	6	.....	5GIBGB	
8	.....,^],	DEGEGGGB	10	.....,	DABGIIIIII	5	.....	>ABDA	6	.....	5GIBGB	
8	.....,	DEGEGGGB	10	.....,	DABGIIIIII	5	.....	>AB/A	6	.....	DGIBGB	
8	.....,	DEGEGGGB	10	.....,	DABGIIIIII	5	.....	>ABAA	6	.....	DGIBGB	
8	.....,	DEGEGGGB	10	.....,	DABGIIIIII	5	.....	>/BAA	6	.....	DGIBGB	
9	.....,^],	DEGEGGGBE	10	.....,	DABGIIIIII	5	.....	>BBAA	6	.....	DGIBGB	
9	.G...gG,,	DEGEGGGBG	10	.....,	D3BGIIIIII	5	.....,C	>BBAA	6	.....	DGIBGB	
9	.....,	DEGEGGGBG	10	.....,	D3BGIIIIII	5	.....	>BBAA	6	.....	/GIBGB	

Wouldn't it be awesome if you knew what the frequency of C was in the rest of the sample or population.

# Bayesian Inference

$$P(\theta|X) = \frac{P(X|\theta)P(\theta)}{P(X)} = \frac{P(X|\theta)P(\theta)}{\sum_{\theta} P(X|\theta)P(\theta)}$$



## From genotype likelihoods to posterior probabilities

Having an estimate of the allele frequency in the population would enable us to have prior knowledge about the probabilities of observing a particular genotype from principles like Hardy-Weinberg Equilibrium (HWE):

$$P(\text{Genotype} = 0 \text{ minor alleles}) = (1 - f_{\text{minor}})^2$$

$$P(\text{Genotype} = 1 \text{ minor allele}) = 2 * f_{\text{minor}} * (1 - f_{\text{minor}})$$

$$P(\text{Genotype} = 2 \text{ minor alleles}) = (f_{\text{minor}})^2$$

Things like inbreeding can easily be incorporated into these genotype probabilities. So, now we have to know how to estimate allele frequencies.



A simple model to estimate the population minor allele frequency,  $f$ , is given by

$$p(D_i | f) = \sum_{\mathbf{g} \in \{0,1,2\}} \underbrace{p(D_i | G_i = \mathbf{g})}_{\text{These are the genotype likelihoods (D}_i \text{ is the sequencing data for the } i\text{th individual) that we now know how to calculate.}} \underbrace{p(G_i = \mathbf{g} | f)}_{\text{And here is just the probability of the genotype given the minor allele frequency, which we can get through HWE.}}$$

$$\hat{f} = \arg \max_f \prod_i p(D_i | f) \longleftarrow$$

Figure out what minor allele frequency maximizes the above likelihood across all individuals in the sample, and you have a ML estimate of the minor allele frequency.

A simple model to estimate the population minor allele frequency,  $f$ , is given by

$$p(D_i|f) = \sum_{\mathbf{g} \in \{0,1,2\}} p(D_i|G_i = \mathbf{g})p(G_i = \mathbf{g}|f)$$

$$\hat{f} = \arg \max_f \prod_i p(D_i|f)$$

One thing to note here is that you can compare the likelihood for the ML minor allele frequency to the likelihood calculated from above with  $f$  set to zero:

$$\lambda = -2 \cdot \log(L(f=0) - L(f=f_{\text{ML}}))$$

$$\lambda \sim \text{Chi-square}(1 \text{ d.f.})$$

Now you have a way to test whether the ML MAF is statistically nonzero, i.e. whether the site is a SNP.

Now, getting back to this individual with sequencing data AAAG at a site. If we estimate  $f(A) = 0.7$  and we consider only the two most likely alleles (A and G) and  $\epsilon$  is always 0.01 (Phred quality of 20, remember that?), then the genotype likelihoods are

Genotype	Likelihood
AA	-5.73
AG	-2.80
GG	-17.12

Apply Bayes' Theorem

Prior probability using  $f(A) = 0.7$   
and HWE

$$P(G|D) = \frac{P(D|G) \pi(G)}{\sum_{G \in \{0,1,2\}} P(D|G) \pi(G)}$$

And you get genotype posterior probabilities!

Genotype	Likelihood (log)	Prior	Posterior
AA	-5.73	0.49	0.06
AG	-2.80	0.42	0.94
GG	-17.12	0.09	0

Now, we can call the genotype as AG based on the max posterior probability (and we also have an estimate of how reliable this call is).