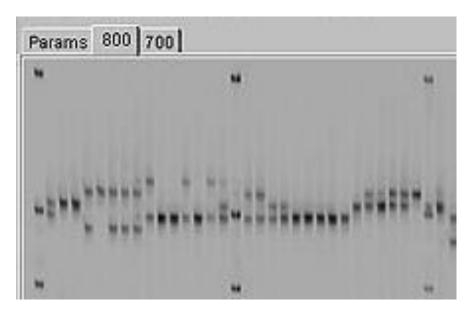
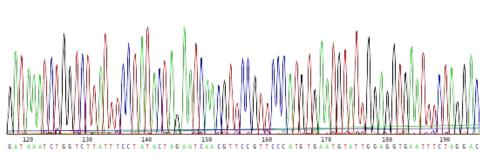
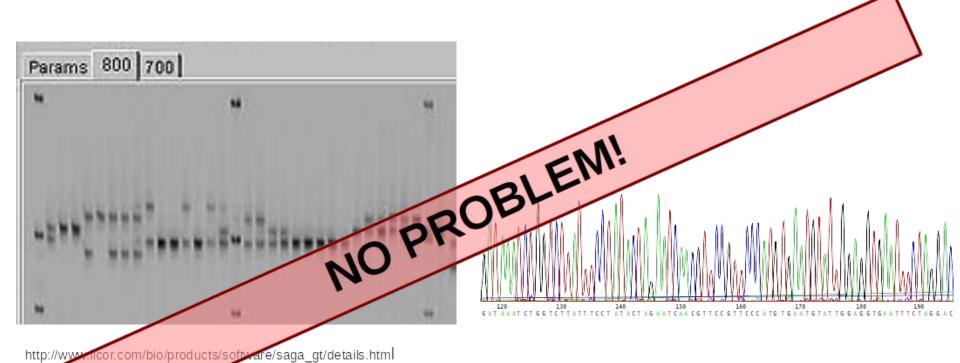
# Estimation of allele frequencies, SNP calling, and genotype calling from NGS data

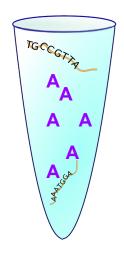
Tyler Linderoth Copenhagen Popgen Summer Course 2021



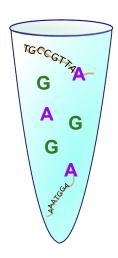


http://www.licor.com/bio/products/software/saga\_gt/details.html

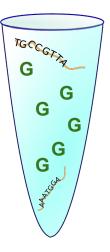




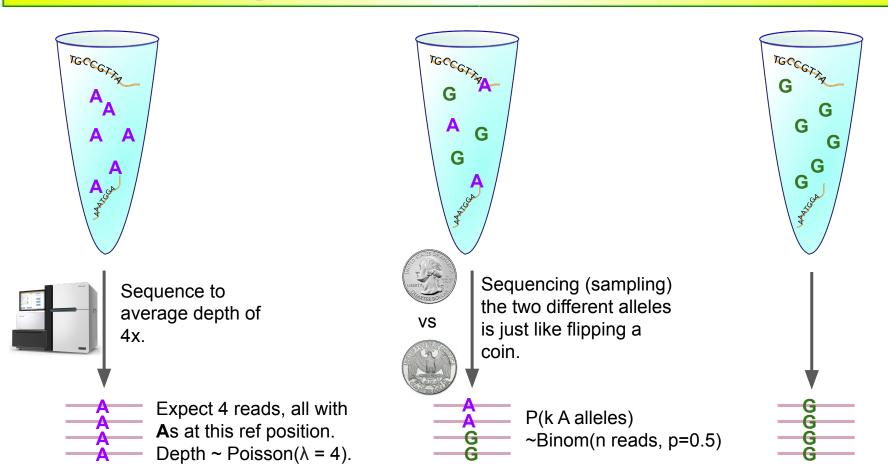
The library for an individual homozygous for the **A** allele will consist only of **A**s.

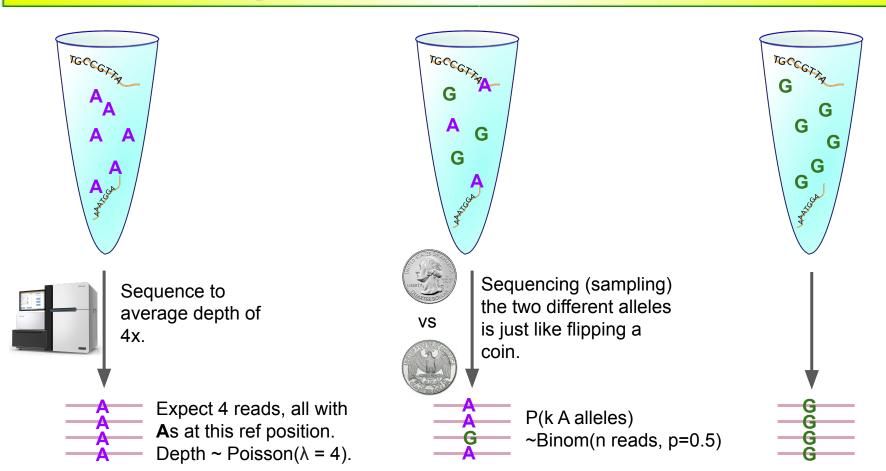


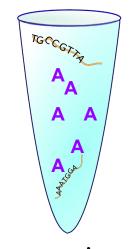
The library for a heterozygous individual at a site contains both **A**s and **G**s.



The library for an individual homozygous for the **G** allele vonsist only of **G**s.

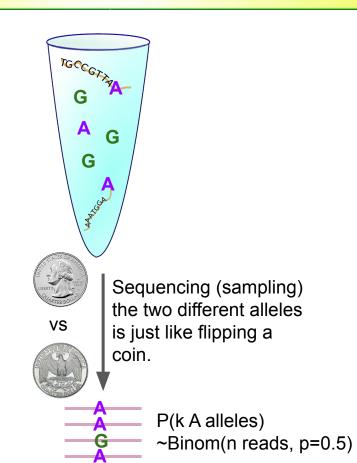






A sequencing error occurs. This can occur at rates of around 0.1% in Illumina data.

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

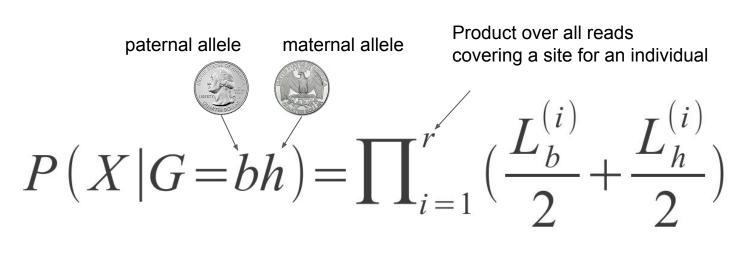


G

G

G

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

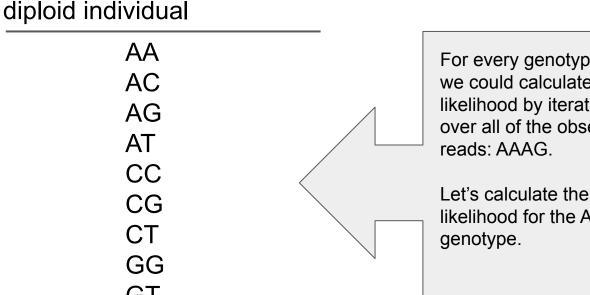


$$b, h \in \{A, C, G, T\}$$

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

Example for an individual with observed sequencing reads **AAAG** at a site. 
$$L_{h}^{(i)}$$

 $P(X|G=bh) = \prod_{i=1}^{r} \left(\frac{L_b^{(i)}}{2}\right)$ 10 potential genotypes for a



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

likelihood for the AC

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

Example for an individual with observed sequencing reads **AAAG** at a site. 
$$P(X|G=bh) = \prod_{i=1}^{r} \left(\frac{L_b^{(i)}}{2} + \frac{L_h^{(i)}}{2}\right)$$

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

$$A = C = \text{Probability of an error}$$

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

$$L_C^{(1)} = \frac{\epsilon}{3}$$

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

ε = Probability of an error

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

Genotype	Likelihood (log10)	
AA	-2.49	
AC	-3.38	
AG	-1.22	A
AT	-3.38	A
CC	-9.91	Α
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			
AG	-1.22			
AT	-3.38			
CC	-9.91 -7.74 -9.91 -7.44			
CG				
CT				
GG				
GT	-7.74			
TT	-9.91			
<u> </u>	-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 w you choose a uniform prior probability distribution for the genotypes.

### Major and minor alleles

#### Likelihood function

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

AAAG &  $\epsilon = 0.01$ 

Allele	Likelihood
Α	-2.49
C	-3.38
G	-1.22
Т	-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?

1 . 1 1 . 1 1 . 1

1 . 1 1 . 1 1 . 1

1.11.11.1

9

10

10

10

DABGIIIII

DABGIIIII

DABGIIIII

DARGITITIE

D3BGIIIIII

**D3BGIIIII** 

DEGEGG

DEGEGG

DEGEGG

DEGEGG

DEGEGGGBG

**DEGEGGGBG** 

. . . . . ,

. . . . . ,

. . . . . ,

.G...gG,,

. . . . . , . , ,

FGC ["]

3GGDDD

3GGBGB

3GGBGB 3GGBGB

DGIBGB

/GIBGB

.,,...

.,,...

>AB/A 6

>ABDA 6

>ABDA

>AR/A

>BBAA

>BBAA

6

6

,,.,.

,,.,.

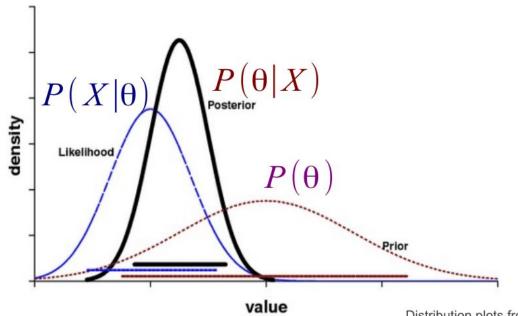
9	,.,^],	DEGEGGGBE	10	1.11.11.1.	DABGIIIII	5	,,.,.	>BBAA	6	.,,	DGIBGB
8	, . ,	DEGEGGGB	10	, . , , . , , . , .	DABGIIIII	5	,,.,.	>/BAA	6	.,,	DGIBGB
8	, . ,	DEGEGGGB	10		DABGIIIII	5	,,.,.	>ABAA	6	.,,	DGIBGB
8	, . ,	DEGEGGGB	10	, . , , . , , . , .	DABGIIIII	5	,,.,.	>AB/A	6	.,,	DGIBGB
8	,.^],	DEGEGGGB	10	, . , , . , , . , .	DABGIIIII	5	,,.,.	>ABDA	6	.,,	5GIBGB
7	, .	DEGEGGG	10	1.11.11.1.	DABGIIIII	5	,,.,.	>ABDA	6	.,,	5GIBGB
7	,^].	DEGEGGE	10	, . , , . , , . , .	DABGIIIII	5	,,.,.	>ABDA	6	.,,	5GIBGB
6	,	DEGEGG	10	1.11.11.1.	DABGIIIII	5	,,.,.	>ABDA	6	C,,	5G/BGB
6	,	DEGEGG	10	1.11.11.1.	DABGIIIII	5	,,.,.	>ABDA	6	.,,	BGGBGB
6	,	DEGEGG	10	1.11.11.1.	DABGIIIII	5	,,.,.	>AB/A	6	.,,	3GGBGB
U	,	DEGEGG	10	1.11.11.1.	DADOITITE	J	11.1.	- 7077	U	.,,	SOUDOD

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

1 . 1 / . / / . / .

#### Bayesian Inference

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



Distribution plots from Bink 2008

#### From genotype likelihoods to posterior probabilities

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

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

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

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

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

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

These are the genotype likelihoods (D<sub>i</sub> is the sequencing data for the ith individual) that we now know how to calculate.

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 under the ML minor

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

$$\lambda = -2*log(\mathcal{L}(f=0|D) - \mathcal{L}(f=ML f|D))$$
  
 $\lambda \sim Chi-square(1 d.f.)$ 

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

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

Geno	type	Likelihood	
AA	4	-5.73	
ΑC	G	-2.80	
G	Ĝ	-17.12	
Apply	Bayes	s' Theorem	Prior probability using $f(A) = 0.7$ and HWE
P(G	D)	$=$ $\frac{1}{\sum_{i=1}^{n}}$	$rac{(D G)\pi(G)}{P(D G)\pi(G)}$
		$G \in \{0,1,2\}$	2.}

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).

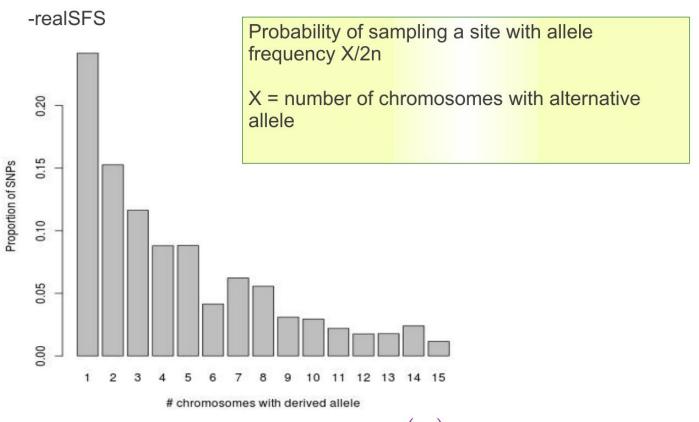
### Allele frequency likelihoods

#### -doSaf 1:

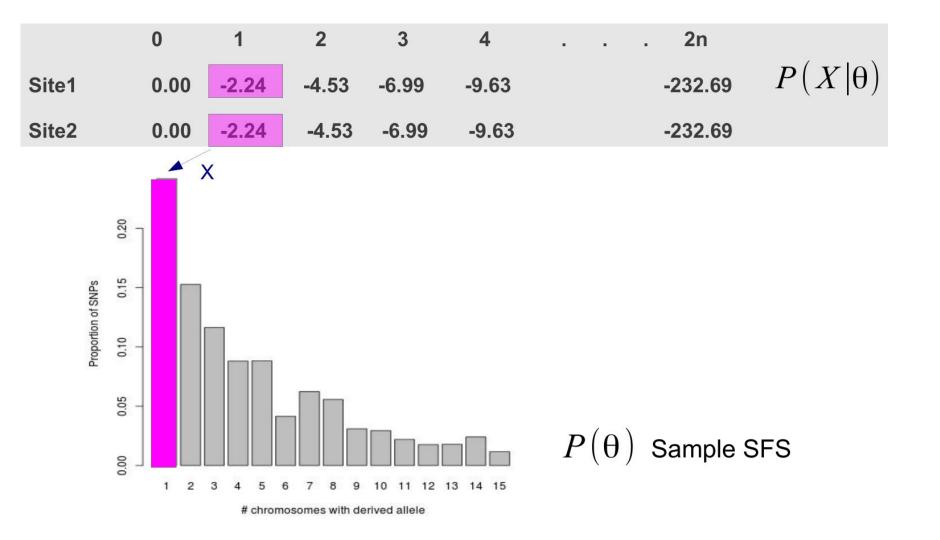
	0	1	2	3	4	2n
Site1	0.00	-2.24	-4.53	-6.99	-9.63	-232.69
Site2	0.00	-2.24	-4.53	-6.99	-9.63	-232.69
Site3	-76.63	-37.87	-10.42	0.00	-9.59	-467.13
Site4	0.00	-2.24	-5.53	-6.99	-9.63	-237.55
•						
Sitek	0.00	-8.62	-19.22	-30.67	-43.27	-626.78

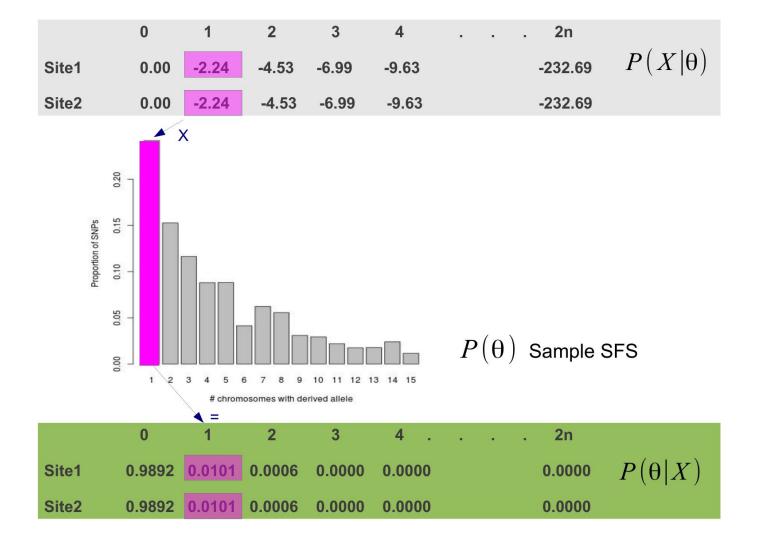
$$P(X|\theta)$$

### Site Frequency Spectrum



 $P(\theta)$ 





### Call SNPs

		0	1	2	3	4 .	2n
	Site1	0.9892	0.0101	0.0006	0.0000	0.0000	0.0000
	Site2	0.9892	0.0101	0.0006	0.0000	0.0000	0.0000
	Site3	0.0000	0.0000	0.0000	0.9999	0.0000	0.0000
Ī	Site4	0.9892	0.0101	0.0006	0.0000	0.0000	0.0000
		99.99 allele					
	Sitek	0.9999	0.0000	0.0000	0.0000	0.0000	0.0000

#### What about allele frequency posterior probabilities?

