

# Notes on PPGtk

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# 1 Allele frequency estimation

## 1.1 The likelihood

The likelihood of an individuals' read data given the population allele frequency can be computed by summing over the possible genotypes:

$$\mathcal{L}_i(p) = P(r_i|p) = \sum_{a=0}^{m_i} P(r_i|a)P(a|p), \quad (\text{S1})$$

where  $P(r_i|a)$  is the genotype likelihood for genotype  $a = 0, \dots, m_i$  (e.g., calculated using GATK), and

$$P(a|p) = \binom{m_i}{a} p^a (1-p)^{m_i-a}.$$

For multiple samples, we take the product of the individual likelihoods:

$$\mathcal{L}(p) = \prod_i \mathcal{L}_i(p) = \prod_i \left( \sum_{a=0}^{m_i} P(r_i|a)P(a|p) \right). \quad (\text{S2})$$

Taking the natural log gives us the log likelihood of the population allele frequency at a single site:

$$\ell(p) = \log \mathcal{L}(p) = \sum_i \log \left( \sum_{a=0}^{m_i} P(r_i|a)P(a|p) \right). \quad (\text{S3})$$

## 1.2 Metropolis-Hastings algorithm

$$P(p) \sim \text{beta}(\alpha = 0.5, \beta = 0.5). \quad (\text{S4})$$

$$P(p|r) \propto P(r|p)P(p) = \left( \sum_a P(r|a)P(a|p) \right) P(p) \quad (\text{S5})$$

$$\alpha = \min \left\{ 1, \frac{P(r|p^*)P(p^*)}{P(r|p)P(p)} \right\} \quad (\text{S6})$$

## 1.3 Genotype estimation

$$P(g_i = a|r_i) = \frac{P(r_i|g_i = a)P(g_i = a|\hat{p})}{\sum_{j=0}^{m_i} P(r_i|g_i = j)P(g_i = j|\hat{p})} \quad (\text{S7})$$

## 2 Inbreeding

We introduce another parameter,  $\phi$ , that is related to the inbreeding coefficient ( $F$ ) through the following equation:

$$F = \frac{1}{1 + \phi} \quad (\text{S8})$$

### 2.1 The likelihood

$$\mathcal{L}_i(p, \phi) = P(r_i|p, \phi) = \sum_{a=0}^{m_i} P(r_i|a)P(a|p, \phi) \quad (\text{S9})$$

where  $P(r_i|a)$  is the genotype likelihood for genotype  $a = 0, \dots, m_i$  (e.g., calculated using GATK), and

$$P(a|p, \phi) = \binom{m_i}{a} \frac{\mathcal{B}(a + p\phi, m_i - a + (1 - p)\phi)}{\mathcal{B}(p\phi, (1 - p)\phi)},$$

which is the probability density function for the beta-binomial distribution with  $n = m_i$ ,  $k = a$ ,  $\alpha = p\phi$ , and  $\beta = (1 - p)\phi$ . Here,  $\mathcal{B}()$  is the beta function:

$$\mathcal{B}(\alpha, \beta) = \int_0^1 x^{\alpha-1} (1 - x)^{\beta-1} dx.$$

### 2.2 Metropolis-Hastings algorithm

## 3 Beta mixture model

$$\begin{aligned} P(g = a|p, \phi) &\sim \text{beta-binomial}(n = m_i, k = a, \alpha = p\phi, \beta = (1 - p)\phi) \\ P(p|\gamma, \pi, \theta_1, \theta_2) &\sim \gamma P(p|\pi\theta_1, (1 - \pi)\theta_1) + (1 - \gamma)P(p|\pi\theta_2, (1 - \pi)\theta_2), \\ P(\pi) &\sim \text{beta}(0.5, 0.5), \\ P(\gamma) &\sim \text{beta}(1, 1), \\ P(\phi) &\sim \text{gamma}(2, 0.1), \\ P(\theta_1) &\sim \text{gamma}(2, 0.1), \\ P(\theta_2) &\sim \text{gamma}(2, 0.1). \end{aligned}$$

**3.1 The likelihood**

**3.2 Metropolis-Hastings algorithm**

## **4 Population admixture model**

**4.1 The likelihood**

**4.2 Metropolis-Hastings algorithm**