Estimating viability

Introduction

Being able to make predictions with known (or estimated) viabilities, doesn’t do us a heck of a lot of good unless we can figure out what those viabilities are. Fortunately, figuring them out isn’t too hard.\footnote{I almost said that it was easy, but that would be going a bit too far.} If we know the number of individuals of each genotype before selection, it’s really easy as a matter of fact. Consider that our data looks like this:

<table>
<thead>
<tr>
<th>Genotype</th>
<th>$A_1A_1$</th>
<th>$A_1A_2$</th>
<th>$A_2A_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number in zygotes</td>
<td>$n_{11}^{(z)}$</td>
<td>$n_{12}^{(z)}$</td>
<td>$n_{22}^{(z)}$</td>
</tr>
<tr>
<td>Viability</td>
<td>$w_{11}$</td>
<td>$w_{12}$</td>
<td>$w_{22}$</td>
</tr>
<tr>
<td>Number in adults</td>
<td>$n_{11}^{(a)} = w_{11}n_{11}^{(z)}$</td>
<td>$n_{12}^{(a)} = w_{12}n_{12}^{(z)}$</td>
<td>$n_{22}^{(a)} = w_{22}n_{22}^{(z)}$</td>
</tr>
</tbody>
</table>

In other words, estimating the absolute viability simply consists of estimating the probability that an individuals of each genotype that survive from zygote to adult. The maximum-likelihood estimate is, of course, just what you would probably guess:

$$w_{ij} = \frac{n_{ij}^{(a)}}{n_{ij}^{(z)}},$$

Since $w_{ij}$ is a probability and the outcome is binary (survive or die), you should be able to guess what kind of likelihood relates the observed data to the unseen parameter, namely, a binomial likelihood. In JAGS notation:\footnote{You knew you were going to see this again, didn’t you?}

$$n_{11}.adult \sim \text{dbin}(w_{11}, n_{11})$$
$$n_{12}.adult \sim \text{dbin}(w_{12}, n_{11})$$
$$n_{22}.adult \sim \text{dbin}(w_{22}, n_{11})$$

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Estimating relative viability

To estimate absolute viabilities, we have to be able to identify genotypes non-destructively, because we have to know what their genotype was both before the selection event and after the selection event. That’s fine if we happen to be dealing with an experimental situation where we can do controlled crosses to establish known genotypes or if we happen to be studying an organism and a trait where we can identify the genotype from the phenotype of a zygote (or at least a very young individual) and from surviving adults. What do we do when we can’t follow the survival of individuals with known genotype? Give up? Remember that to make inferences about how selection will act, we only need to know relative viabilities, not absolute viabilities. We still need to know something about the genotypic composition of the population before selection, but it turns out that if we’re only interested in relative viabilities, we don’t need to follow individuals. All we need to be able to do is to score genotypes and estimate genotype frequencies before and after selection. Our data looks like this:

<table>
<thead>
<tr>
<th>Genotype</th>
<th>$A_1A_1$</th>
<th>$A_1A_2$</th>
<th>$A_2A_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Frequency in zygotes</td>
<td>$x^{(z)}_{11}$</td>
<td>$x^{(z)}_{12}$</td>
<td>$x^{(z)}_{22}$</td>
</tr>
<tr>
<td>Frequency in adults</td>
<td>$x^{(a)}_{11}$</td>
<td>$x^{(a)}_{12}$</td>
<td>$x^{(a)}_{22}$</td>
</tr>
</tbody>
</table>

We also know that

$$x^{(a)}_{11} = \frac{w_{11}x^{(z)}_{11}}{\bar{w}}$$

$$x^{(a)}_{12} = \frac{w_{12}x^{(z)}_{12}}{\bar{w}}$$

$$x^{(a)}_{22} = \frac{w_{22}x^{(z)}_{22}}{\bar{w}}$$

Suppose we now divide all three equations by the middle one:

$$x^{(a)}_{11}/x^{(a)}_{12} = \frac{w_{11}x^{(z)}_{11}}{w_{12}x^{(z)}_{12}}$$

$$1 = 1$$

$$x^{(a)}_{22}/x^{(a)}_{12} = \frac{w_{22}x^{(z)}_{22}}{w_{12}x^{(z)}_{12}}$$

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3How many organisms and traits can you think of that satisfy this criterion? Any? There is one other possibility: If we can identify an individual’s genotype after it’s dead and if we can construct a random sample that includes both living and dead individuals and if we the probability of including an individual in the sample doesn’t depend on whether that individual is dead or alive, then we can sample a population after the selection event and score genotypes both before and after the event from one set of observations.

4Would I be asking the question if the answer were “Yes”?

5At least that’s true until we start worrying about how selection and drift interact.
or, rearranging a bit

\[
\frac{w_{11}}{w_{12}} = \left( \frac{x_{11}^{(z)}}{x_{12}^{(z)}} \right) \left( \frac{x_{12}^{(a)}}{x_{11}^{(a)}} \right) \\
\frac{w_{22}}{w_{12}} = \left( \frac{x_{22}^{(z)}}{x_{12}^{(z)}} \right) \left( \frac{x_{12}^{(a)}}{x_{22}^{(a)}} \right)
\]

This gives us a complete set of relative viabilities.

<table>
<thead>
<tr>
<th>Genotype</th>
<th>(A_1A_1)</th>
<th>(A_1A_2)</th>
<th>(A_2A_2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Relative viability</td>
<td>(\frac{w_{11}}{w_{12}})</td>
<td>1</td>
<td>(\frac{w_{22}}{w_{12}})</td>
</tr>
</tbody>
</table>

If we use the maximum-likelihood estimates for genotype frequencies before and after selection, we obtain maximum likelihood estimates for the relative viabilities.\(^6\) If we use Bayesian methods to estimate genotype frequencies (including the uncertainty around those estimates), we can use these formulas to get Bayesian estimates of the relative viabilities (and the uncertainty around them).

**An example**

Let’s see how this works with some real data from Dobzhansky’s work on chromosome inversion polymorphisms in *Drosophila pseudoobscura*.\(^7\)

<table>
<thead>
<tr>
<th>Genotype</th>
<th>(ST/ST)</th>
<th>(ST/CH)</th>
<th>(CH/CH)</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number in larvae</td>
<td>41</td>
<td>82</td>
<td>27</td>
<td>150</td>
</tr>
<tr>
<td>Number in adults</td>
<td>57</td>
<td>169</td>
<td>29</td>
<td>255</td>
</tr>
</tbody>
</table>

You may be wondering how the sample of adults can be larger than the sample of larvae. That’s because to score an individual’s inversion type, Dobzhansky had to kill it. The numbers in larvae are based on a sample of the population, and the adults that survived

\(^6\)If anyone cares, it’s because of the invariance property of maximum-likelihood estimates. If you don’t understand what that is, don’t worry about it, just trust me.

\(^7\)Taken from [1].
were not genotyped as larvae. As a result, all we can do is to estimate the relative viabilities.

\[
\frac{w_{11}}{w_{12}} = \left( \frac{x_{11}^{(a)}}{x_{12}^{(a)}} \right) \left( \frac{x_{12}^{(z)}}{x_{11}^{(z)}} \right) = \left( \frac{57/255}{169/255} \right) \left( \frac{82/150}{41/150} \right) = 0.67
\]

\[
\frac{w_{22}}{w_{12}} = \left( \frac{x_{22}^{(a)}}{x_{12}^{(a)}} \right) \left( \frac{x_{12}^{(z)}}{x_{22}^{(z)}} \right) = \left( \frac{29/255}{169/255} \right) \left( \frac{82/150}{27/150} \right) = 0.52
\]

So it looks as if we have balancing selection, i.e., the fitness of the heterozygote exceeds that of either homozygote.

We can check to see whether this conclusion is statistically justified by comparing the observed number of individuals in each genotype category in adults with what we’d expect if all genotypes were equally likely to survive.

<table>
<thead>
<tr>
<th>Genotype</th>
<th>ST/ST</th>
<th>ST/CH</th>
<th>CH/CH</th>
</tr>
</thead>
<tbody>
<tr>
<td>Expected</td>
<td>(\frac{41}{150})</td>
<td>(\frac{82}{150})</td>
<td>(\frac{27}{150})</td>
</tr>
<tr>
<td>Observed</td>
<td>69.7</td>
<td>139.4</td>
<td>45.9</td>
</tr>
<tr>
<td>(\chi^2)</td>
<td>14.82</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

So we have strong evidence that genotypes differ in their probability of survival.

We can also use our knowledge of how selection works to predict the genotype frequencies at equilibrium:

\[
\frac{w_{11}}{w_{12}} = 1 - s_1
\]

\[
\frac{w_{22}}{w_{12}} = 1 - s_2
\]

So \(s_1 = 0.33\), \(s_2 = 0.48\), and the predicted equilibrium frequency of the \(ST\) chromosome is \(s_2/(s_1 + s_2) = 0.59\).

Now all of those estimates are maximum-likelihood estimates. Doing these estimates in a Bayesian context is relatively straightforward and the details will be left as an excerise.\(^8\) In outline we simply

1. Estimate the genotype frequencies before and after selection as samples from a multinomial.

\(^8\)In past years Problem #3 has consisted of making Bayesian estimates of viabilities from data like these and predicting the outcome of viability selection. This year Nora will illustrate the approach (unless you’d rather have her spend more time helping you with Problem #2).
2. Apply the formulas above to calculate relative viabilities and selection coefficients.

3. Determine whether the 95% credible intervals for $s_1$ or $s_2$ overlap 0.9

4. Calculate the equilibrium frequency from $s_2/(s_1 + s_2)$, if $s_1 > 0$ and $s_2 > 0$. Otherwise, determine which fixation state will be approached.

In the end you then have not only viability estimates and their associated uncertainties, but a prediction about the ultimate composition of the population, associated with an accompanying level of uncertainty.

References


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9Meaning that we don’t have good evidence for selection either for or against the associated homozygotes, relative to the heterozygote.