

MTH notes on JKK's lecture 1

For discrete events, A_i in some sample space of \mathcal{S}_A of possible events, we can make statements about the probability of A_i occurring, denoted $\mathbb{P}(A_i)$.

$$0 \leq \mathbb{P}(A_i) \leq 1 \quad \forall A_i \in \mathcal{S}_A \tag{1}$$

$$1 = \sum_{A_i \in \mathcal{S}_A} \mathbb{P}(A_i) \tag{2}$$

\forall means “for all”.

\sum means a summation over values of a variable. So $\sum_{A_i \in \mathcal{S}_A}$ means we are going to do a summation over all values of A_i that possible in the sample space. We often number the outcomes from 1 to $|\mathcal{S}_A|$ in which case we can write the second eqn as:

$$1 = \sum_{i=1}^{|\mathcal{S}_A|} \mathbb{P}(A_i) \tag{3}$$

where i is indexing the possible events.

Compound event: combinations of simple events

E.g. Let x be an encoding of the result of a random trial into some specified sample space. So, if Trevor's student records the behavior of a lobster we could use the encoding:

Simple vent	code	$\mathbb{P}(x = \text{code})$
Motionless	0	0.5
Random	1	0.1
Walk toward	2	0.1
Walk away	3	0.1
Turn toward	4	0.1
Turn away	5	0.1

What is the probability of “some motion”?

$$\mathbb{P}(A \text{ OR } B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A, B) \tag{4}$$

$$\mathbb{P}(c \in [1, 2, 3, 4, 5]) = \mathbb{P}(c = 1) + \mathbb{P}(c = 2) + \mathbb{P}(c = 3) + \mathbb{P}(c = 4) + \mathbb{P}(c = 5) \tag{5}$$

$$= 0.1 + 0.1 + 0.1 + 0.1 + 0.1 = 0.5 \tag{6}$$

Note that each of the simple events are mutually exclusive with the others in our encoding. So the joint probability ($\mathbb{P}(A, B)$) of any of 2 of them occurring in the same random trial is 0. So we don't subtract anything. We could also solve this using eqn 2 above:

$$\mathbb{P}(A \text{ OR } \text{not}A) = \mathbb{P}(A) + \mathbb{P}(\text{not } A) = 1 \tag{7}$$

$$\mathbb{P}(\text{not } A) = 1 - \mathbb{P}(A) = 1 - .5 = 0.5 \tag{8}$$

If our dataset had observations for 2 second we could say: $X = [0, 2]$ or $X = [x_0, x_1]$ where $x_0 = 0$ is the encoding of “motionless in the first second” and $x_1 = 2$ meand “Walk toward in second two.”

The likelihood is the probability of data identical to what we have observed if the model were true:

$$\mathbb{P}(X) = \mathbb{P}(x_0, x_1) \tag{9}$$

$$= \mathbb{P}(x_0)\mathbb{P}(x_1 | x_0) \tag{10}$$

from the general multiplication rule of probabilities: $\mathbb{P}(A, B) = \mathbb{P}(A)\mathbb{P}(B | A) = \mathbb{P}(B)\mathbb{P}(A | B)$.

To make some progress we could assume that behavior in one second is independent of the behavior in every other seconds. So $\mathbb{P}(A | B) = \mathbb{P}(A)$. In this case: $\mathbb{P}(x_1 | x_0) = \mathbb{P}(x_1)$

$$\mathbb{P}(X) = \mathbb{P}(x_0)\mathbb{P}(x_1 | x_0) \tag{11}$$

$$= .5 \times 0.1 = 0.05 \tag{12}$$

What if we have genetic data. $X = [M_g = Aa, S_g = Aa]$ where M_g is the genotype of a mother at some locus. and S_g is the genotype of her son. What is the likelihood $\mathbb{P}(X)$?

We need a model. What if we assume that: mating is random, the frequency of A in the population is some unknown parameter q , and there is no selection on the locus or meiotic drive (or other weird stuff)?

$$\mathbb{P}(X | q) = \mathbb{P}(M_g = Aa | q)\mathbb{P}(S_g = Aa | M_g = Aa, q) \tag{13}$$

$$\mathbb{P}(M_g = Aa | q) = 2q(1 - q) \text{ from Hardy-Weinberg equil.} \tag{14}$$

How do we get $\mathbb{P}(S_g = Aa | M_g = Aa, q)$? We can use the law of total probability:

$$\mathbb{P}(A) = \sum_{B \in \mathcal{S}_B} [\mathbb{P}(A | B)\mathbb{P}(B)] \tag{15}$$

in this case use this to sum over the possible other event (B) that would be helpful to know: dad's genotype, D_g . Using Hardy-Weinberg again:

event, d	$\mathbb{P}(D_g = d)$	$\mathbb{P}(S_g = Aa M_g = Aa, D_g = d)$
AA	q^2	0.5 (from Mendel half AA, half Aa)
Aa	$2q(1 - q)$	0.5 (from Mendel one quarter AA, half Aa, one quarter aa)
aa	$(1 - q)^2$	0.5 (from Mendel half Aa, half aa)

Note that we could condition the third column on q , but the population allele frequencies does not matter when we know both parent's genotype. So we drop that from the notation for the sake of brevity.

$$\mathcal{G} = \{AA, Aa, aa\} \tag{16}$$

$$\begin{aligned} \mathbb{P}(S_g = Aa | M_g = Aa, q) &= \sum_{d \in \mathcal{G}} [\mathbb{P}(S_g = Aa | M_g = Aa, D_g = d)\mathbb{P}(D_g = d)] \\ &= 0.5 \sum_{d \in \mathcal{G}} [\mathbb{P}(D_g = d)] \\ &= 0.5 \end{aligned} \tag{17}$$

interestingly this componenet of the likelihood is not a function of q . But the full likelihood is:

$$\begin{aligned} \mathcal{L}(q) = \mathbb{P}(X | q) &= \mathbb{P}(M_g = Aa | q)\mathbb{P}(S_g = Aa | M_g = Aa, q) \\ &= 2q(1 - q) \times 0.5 \\ &= q(1 - q) \end{aligned} \tag{18}$$

So, the likelihood is a function of our unknown parameter q . If you play around with this, you will find that the highest likelihood is obtained when $q = 0.5$ and the likelihood is 0.25.