

## Supplementary Material B Cortical threshold calculations

Let  $f(z) = E(e^{-\theta T} | Z_0 = z)$  be the probability distribution of the probability of survival given that the decision will be reached at time  $T$ , with the log of the initial likelihood ratio being  $Z_0 = z$ .

Taking a small time slice,  $\delta t$ , from  $T$ , we let  $T = \delta t + T'$ , which gives:  $e^{-\theta T} = e^{-\theta \delta t} e^{-\theta T'} \approx (1 - \theta \delta t) e^{-\theta T'}$ .

$$\begin{aligned} \text{Then, } f(z) &\approx E((1 - \theta \delta t) e^{-\theta T'} | Z_0 = z) \\ &= (1 - \theta \delta t) E(e^{-\theta T'} | Z_0 = z) \\ &= (1 - \theta \delta t) E(f(z + \delta Z)) \end{aligned}$$

where we take the movement  $\delta Z$  to have occurred during  $\delta t$ . Note that although the time slice is known, the amount of movement in that time is a random variable.

Using a Taylor expansion, we obtain

$$f(z) \approx (1 - \theta \delta t) \left( f(z) + E(\delta Z) f'(z) + \frac{E((\delta Z)^2)}{2} f''(z) \right).$$

Discarding terms smaller than  $\delta t$ :

$$0 \approx E(\delta Z) f'(z) + \frac{E((\delta Z)^2)}{2} f''(z) - \theta \delta t f(z). \quad (1)$$

In this case, the movement in position with each timestep,  $\delta t$ , is normally distributed according to  $N(\mu \delta t, \eta^2 \delta t)$ , where  $\mu$  and  $\eta$  are calculated as shown in Supplementary Material A (with  $\mu = \mu_+$  as there is a predator present). From the definition of the computational formula for variance,  $Var(\delta Z) = E((\delta Z)^2) - (E(\delta Z))^2$ , so  $\eta^2 \delta t = E((\delta Z)^2) - (\mu \delta t)^2$ . Substituting in equation (1), this gives:

$$0 \approx E(\delta Z) f'(z) + \frac{(\mu \delta t)^2 + \eta^2 \delta t}{2} f''(z) - \theta \delta t f(z)$$

The  $(\delta t)^2$  term can be discarded (in-keeping with the above). We also know that  $E(\delta Z) = \mu \delta t$ . Substituting, dividing through by  $\delta t$  and rearranging, we obtain:

$$0 \approx -\theta f(z) + \mu f'(z) + \frac{\eta^2}{2} f''(z)$$

As  $\delta t$  goes to zero and the information gain becomes continuous, the equation becomes exact.

Solving for  $f(z)$ , the characteristic equation  $0 = -\theta + \mu k + \frac{\eta^2}{2} k^2$  has solutions  $k_1 = \frac{-\mu + \sqrt{\mu^2 + 2\theta\eta^2}}{\eta^2}$ ,  $k_2 = \frac{-\mu - \sqrt{\mu^2 + 2\theta\eta^2}}{\eta^2}$ .

The general solution is then given by:

$$f(z) = K_1 e^{k_1 z} + K_2 e^{k_2 z}. \quad (2)$$

We define  $A$  to be the ratio of likelihoods at (or above) which the animal should act as though a predator is present. i.e., for a sequence of signals,  $(x_1, x_2, \dots, x_k)$ , the animal should take evasive action if and only if  $\frac{f_1(x_1)f_1(x_2)\dots f_1(x_k)}{f_0(x_1)f_0(x_2)\dots f_0(x_k)} \geq A$ . Then we know that  $f(\ln A) = 1$  and  $f(z) \rightarrow 0$  as  $z \rightarrow -\infty$ .

Substituting the second of these boundary conditions into equation (2) and noting that  $k_2$  is negative, we find that  $K_2 = 0$ . Substituting the first boundary condition, we obtain  $K_1 = e^{-k_1 \ln A} = A^{-k_1}$ .

Therefore, knowing that our process will start at  $Z_0 = 0$ , the probability of survival,  $R$ , is given by:

$$R = sf(0) = s(K_1 + K_2) = sA^{-k_1}. \quad (3)$$

As we are dealing with a continuous process, any decision to take anti-predator action will be reached at the point of hitting the boundary (as opposed to overshooting it), so rather than having to approximate  $A$ , the formula supplied by Wald (1945) is exact:  $A = \frac{1-\beta}{\alpha}$ , where  $\alpha$  is the probability of a false alarm (taking evasive action when no predator is present) and  $\beta$  is the probability of a miss (deciding that no predator is present and ignoring any further signals). In this case, as the probability of a miss is zero, we have  $A = \frac{1}{\alpha}$ .

Putting this into equation (3), we obtain:  $R = s\alpha^{k_1}$ , so  $\frac{dR}{d\alpha} = sk_1\alpha^{k_1-1}$ .

Our original ODE (in the main paper) was:

$$(1-p)c = p(v_f - c) \frac{dR}{d\alpha}$$

Substituting for  $\frac{dR}{d\alpha}$ , we obtain:

$$\alpha^{k_1-1} = \frac{(1-p)c}{p(v_f - c)k_1s}.$$