

Appendix S2. Deriving Fourier Coefficients from a Periodic Function

Let us assume we want to model periodic dispersal rates in species who breed once per year, and estimate that signal in a dynamic occupancy model. Therefore, let us consider a periodic pulse (Fig. S1) that illustrates such a biological pattern and define it over a single period:

$$x_P(t) = \begin{cases} A, & |t| \leq \frac{P_u}{2} \\ 0, & |t| > \frac{P_u}{2} \end{cases}, \quad -\frac{P}{2} < t < \frac{P}{2} \quad (\text{S1})$$

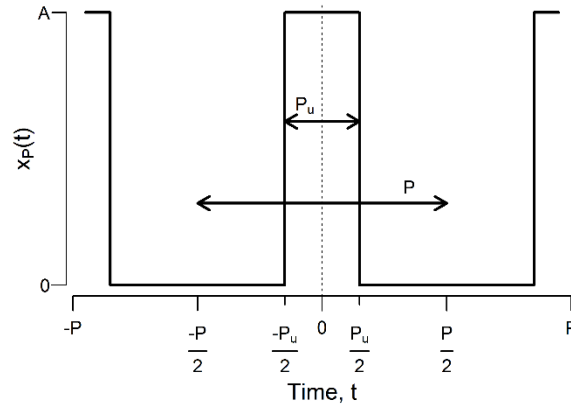


Figure S1. A graphical representation of the periodic pulse function in Eq. S1. Here, a pulse occurs with amplitude A once over period P .

Eq. S1 has a period of P , an amplitude of A , and a pulse width of P_u . Over a single period when the absolute value of t is less than or equal to $P_u/2$, Eq. S1 takes the value A . Otherwise, it is zero. This represents a short pulse that is separated by a longer trough over time (Fig. S1).

With our species and data, Eq. S1 could represent periodic pulses of increased colonization rates as juveniles disperse at a particular season over the period $P = 4$ (i.e., from sampling that occurs in the spring, summer, fall, and winter). As Eq. S1 is symmetric around its origin of zero, it is an even function (Kreyszig 2010). Therefore, $b_n = 0$ for all n and this function can be approximated with a Fourier cosine series

$$x_P(t) = a_0 + \sum_{n=1}^{\infty} (a_n \cos(nw_0 t)) \quad (\text{S2})$$

Which, following Kreyszig (2010), means that a_n can be written as

$$a_n = \frac{2}{P} \int_P x_P(t) \cos(nw_0 t) dt, \quad n \neq 0 \quad (\text{S3})$$

Given $x_P(t) = A$ at any time between $-P_u/2$ to $P_u/2$ and 0 at any other point (Eq. S1), a_n can be simplified and then solved. We illustrate this over the interval $-P_u/2$ to $P_u/2$

$$a_n = \frac{2}{P} \int_{-\frac{P_u}{2}}^{\frac{P_u}{2}} x_P(t) \cos(nw_0 t) dt \quad (\text{S4})$$

$$= \frac{2}{P} \frac{A}{nw_0} \sin(nw_0 t) \Big|_{-\frac{P_u}{2}}^{+\frac{P_u}{2}} \quad (\text{S5})$$

$$= \frac{2}{P} \frac{A}{nw_0} \left(\sin(nw_0 \frac{P_u}{2}) - \sin(-nw_0 \frac{P_u}{2}) \right) \quad (\text{S6})$$

Furthermore, as $\sin(a) - \sin(-a) = 2\sin(a)$ and $w_0 = 2\pi/P$, Eq. S6 can be reduced to

$$a_n = \frac{2A}{n\pi} \sin\left(\frac{n\pi P_u}{P}\right) \quad (\text{S7})$$

As $n \rightarrow \infty$, Eq. S2 more closely approximates the graphical representation of the pulse in Fig. S1 over continuous time. However, when included within a statistical model it is impossible to include infinite terms and computational efficiency is reduced as more n are included.

Fortunately, when modeling events over discrete time (e.g., $t = 1, 2, \dots, T$) fewer n are needed to reach a steady state. Our own trials suggest that n should at least range from $1, 2, \dots, P$ for model fitting. In general, dynamic occupancy models necessitate modeling with discrete time and the estimated parameters associated to a species colonization or persistence rates are related to the length of time between primary sampling periods (MacKenzie et al. 2006). If data are collected on a more continuous scale (e.g., year-round monitoring), an investigator should discretize these data in such a way that is relevant to a species biology and the question at hand. For example, if a species breeds once per year and juvenile dispersal occurs during the fall, setting primary sampling periods in the spring, summer, fall, and winter would be an adequate approach so that a periodic component may be included within a model.

The pulse width, P_u , is only a parameter of interest if one expects the pulse to occur over multiple time steps, which may be the case if P is large and the time between steps is small. If the pulse should only happen at one specific time step, as in our case, then P_u can be reduced to a value such that $t + P_u/2 < t + 1$, so the pulse does not reach into the surrounding time steps. The natural choice is to set P_u to 1 so that it can then be removed from the equation. Further, Eq. S2 allows us to estimate the amplitude of a pulse, A , but it does not let us estimate when the pulse occurs over P . We add a phase shift term, δ , to the equation:

$$x_P(t) = a_0 + \sum_{n=1}^P \frac{2A}{\pi n} \sin\left(\frac{\pi n}{P}\right) \cos\left(\frac{2\pi n}{P}(t - \delta)\right) \quad (\text{S8})$$

So that the model can then estimate the presence of a periodic pulse at any time step over period P with two parameters, A and δ (Fig. S2).

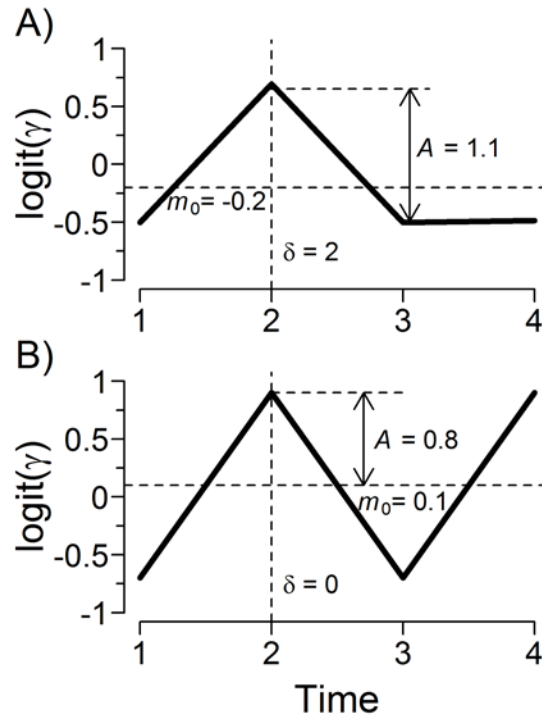


Figure S2. Examples of periodic curves when only discrete integers (i.e., $t = 1, 2, 3, 4$) are supplied for A) a single season pulse with period $P = 4$ as specified in Eq. S2 and B) a boom-bust sinusoidal curve with period $P = 2$ as specified in Eq. 10 of the main text. Plots A and B are on the logit scale. m_0 controls the average colonization rate, A controls the amplitude, and δ controls when a pulse occurs.

Literature Cited

Kreyszig, E., 2010. *Advanced engineering mathematics*. John Wiley & Sons.