

1 Going against the flow: The effects of non-Gaussian dispersal kernels
2 and reproduction over multiple generations

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5 RUNNING HEAD: Multiple Generations and non-Gaussian kernels

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Abstract

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Our previous studies discuss when populations and gradients in allele frequency can persist in the presence of a mean downstream dispersal of propagules from the parents' location. These studies assume that reproduction is uniform over the lifetime of the adult, and that the larval dispersal kernel is "nearly" Gaussian in shape. These results are extended below to include variable reproduction over the lifetime of an adult and non-Gaussian dispersal kernels. It is found that persistence is governed by the lifetime reproductive output of the adults. The impact of non-Gaussian dispersal kernels is quantified in terms of the excess kurtosis of the dispersal kernel.

KEYWORDS: Advection, Retention, Planktonic Larvae, Coastal, Physical-Biological coupling, Dispersal

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Introduction

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Byers & Pringle (2006) and Pringle & Wares (2007) (hereafter BWP) discuss when populations and gradients in allele frequency can be retained in the presence of a mean downstream dispersal of propagules from the parents' location, as occurs, for example, in benthic species in the coastal ocean. However, in their analysis they make a number of assumptions which they do not fully justify. In particular, they assert that it is the lifetime fecundity of the species in the absence of density dependence that governs the persistence of the species and alleles; however, they only test this result for the case in which reproductive output is constant with time over the lifetime of the organism. This is clearly not always a good assumption; many invertebrate marine species produce more larvae as the adults grow older and larger (Llodra, 2002). They also assume a Gaussian dispersal kernel, and assure the reader that any kernel that is "close to a Gaussian" will lead to similar results to those obtained with a Gaussian kernel. Unfortunately, they do not define what "close" means, thus leaving the reader uncertain as to how to apply the results to real world dispersal kernels. Thus the reader is unsure what to do when, to use an example from Byers & Pringle (2006), a species reproduces in multiple seasons in which the mean and variability of the currents are different, so that even if the

47 dispersal kernel for each reproductive event is Gaussian, the net larval dispersal
 48 kernel is a composite of the distribution for each spawning event and will not be
 49 Gaussian.

50 In the following section, we will provide two results that address these issues.
 51 First, we will show that the criteria for retention of a species is, for a Gaussian
 52 dispersal kernel,

$$\frac{L_{\text{adv}}^2}{2L_{\text{diff}}^2} < \ln(N), \quad (1)$$

53 where N is the total number of larvae which would recruit and reach reproductive
 54 competency in the absence of density dependent effects per adult per *lifetime*. L_{adv}
 55 and L_{diff} are the mean and standard deviation of the larval dispersal distance, as
 56 discussed in BWP. This is the same as equation (6) of Byers & Pringle (2006) when
 57 the reproductive output of the adults is constant with age, but is also correct when
 58 the reproductive output of the adult varies with age. Second, we show that if the
 59 kernel is non-Gaussian, the criteria for retention becomes approximately

$$\frac{L_{\text{adv}}^2}{2L_{\text{diff}}^2} < \ln(N) + \frac{\gamma_2}{6} \ln(N)^2 + \frac{\gamma_2^2}{144} \ln(N)^3, \quad (2)$$

60 where γ_2 is the “excess kurtosis”, and is equal to $\mu_4/L_{\text{diff}}^4 - 3$, where μ_4 is the fourth
 61 central moment of the kernel. The kurtosis is a measure of how many of the larvae
 62 of a given kernel are in the edges of the dispersal kernel relative to the center of
 63 the kernel for a given standard deviation of larval dispersal distance, L_{diff} . This
 64 can be seen in the two panels of figure 1, where the kernels with larger and more
 65 positive kurtosis have more larvae in their tails and, in order to keep the standard
 66 deviation constant, more larvae concentrated at the center of the distribution. Ex-
 67 cess kurtosis is defined with respect to a Gaussian distribution, so that a positive
 68 excess kurtosis indicates more larvae in the tails of the distribution than a Gaussian
 69 kernel would have, and a negative value indicates fewer. Thus a Gaussian kernel
 70 has an excess kurtosis γ_2 of zero, and the above criteria reduces to that of BWP for
 71 that kernel. The excess kurtosis for other kernels can be found in Lutscher (2007).
 72 This criteria for retention can be used to judge how important the deviations from
 73 a Gaussian kernel are, and, if the effect is large, numerical methods can then be

74 used to calculate the exact retention criterion (the code to do so is available in the
75 online supplement to this article). This result can be transferred to Pringle & Wares
76 (2007) by substituting the righthandside of (2) for the righthandside of equations
77 (3), (4) and (10) in that paper. It is also shown below that when there is no mean
78 downstream dispersal of larvae, the excess kurtosis does not affect the persistence
79 of a species, consistent with Lockwood et al. (2002).

80 The impact of non-Gaussian kernels, such as those shown in figure 1, can be
81 very important when there is a mean downstream transport of larvae. In figure
82 2, the critical value of N needed to allow retention is shown as a function of the
83 mean larval dispersal distance L_{adv} for these three non-Gaussian dispersal kernels,
84 each representative of a certain kind of deviation from an idealized Gaussian kernel.
85 For each kernel, the standard deviation of the dispersal distance of successfully
86 recruiting larvae L_{diff} is 30km while the mean dispersal distance varies from 1 to
87 50km. The first kernel is the composite of two Gaussian kernels, such as might
88 occur if the species spawns in two different seasons with different mean currents.
89 This has a negative γ_2 of -1.89, indicating that the tails of the dispersal kernel are
90 relatively small for a given value of L_{diff} . Thus fewer larvae settle far from the center
91 of the larval recruitment distribution, increasing the N needed to allow retention
92 above the prediction of (1) for a given L_{adv} and L_{diff} . The second kernel is tent
93 shaped and there is no dispersal outside of a finite distance. Its excess kurtosis is
94 $\gamma_2=-0.6$. This models the truncation of the tails of the dispersal kernel which can
95 occur because there is a practical upper limit on the speed of the currents in the
96 ocean, and thus a limit to the dispersal distance of a larvae. This tends to increase
97 the N needed to allow retention relative to the Gaussian prediction by eliminating
98 the rare long-distance dispersal of larvae. The third set of results are for a Gaussian
99 kernel, for which $\gamma_2 = 0$ and the results of BWP and (2) agree. The last kernel is a
100 Laplacian kernel, with $\gamma_2 = 3.0$. With this kernel, the density of larval recruitment
101 is increased both near the center of the larval distribution and far from the center,
102 leading to a sharply peaked distribution with long tails of dispersal distance. These
103 tails tend to increase the ability of larvae to be retained for a given fecundity and
104 recruitment rate N relative to an equivalent Gaussian kernel.

105 For each non-Gaussian kernel, the approximation in (2) is more accurate than

106 the criteria in BWP calculated under the assumption of a Gaussian kernel. It
 107 successfully captures the increase in N needed for retention when $\gamma_2 < 0$, and the
 108 decrease when $\gamma_2 > 0$. Further discussion of the dynamics behind these results is
 109 given in the context of invasion speeds in Lutscher (2007). When the magnitude of
 110 the excess kurtosis of the dispersal kernel or L_{adv} becomes large, the approximate
 111 formula (2) becomes increasingly inaccurate. In practice, it would be prudent to use
 112 (2) to estimate the effect of the excess kurtosis on the critical value of N needed to
 113 allow retention and, if this effect is large, then to calculate the exact critical value
 114 of N numerically using the code in the online supplement.

115 The following sections lay out the derivations of these results, but present no
 116 further results.

117 Retention and Reproduction

118 Neubert & Caswell (2000) derive a method for determining the invasion speed of
 119 a population in a stage structured population given any dispersal kernel with ex-
 120 ponentially bounded tails. They argue that it is not necessary to include density
 121 dependent effects in this calculation, as long as there are no Allee or long-distance
 122 density dependence effects. We assume discrete generations of an organism that
 123 only disperses when spawned, and is afterwards sessile, as in BWP. Following Neu-
 124 bert & Caswell (2000), and assuming that the dispersal kernel K is independent of
 125 the parents location and includes the mean transport of the larvae, the population
 126 at location x and time $t + 1$ in generations, $\mathbf{p}(x, t + 1)$, is given by the convolution

$$\mathbf{p}(x, t + 1) = \int_{-\infty}^{\infty} dy (\mathbf{A}(x - y)\mathbf{p}(y, t)). \quad (3)$$

127 Here, $\mathbf{p}(x, t)$ is a vector whose elements are the population age structure, y is a
 128 dummy variable of integration, and \mathbf{A} is the matrix that describes how the larval
 129 settlement is distributed in space and how each generation ages (the combined $\mathbf{K} \circ \mathbf{B}$

130 of Neubert & Caswell (2000)):

$$\mathbf{A} = \begin{bmatrix} N_1K & N_2K & \dots & N_nK \\ \delta r_1 & 0 & \dots & 0 \\ 0 & \delta r_2 & \dots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{bmatrix}. \quad (4)$$

131 Subscripts indicate the age in generations, r_i is the likelihood that an individual of
 132 age i will live to $i + 1$, N_i is the fecundity of an adult of age i , n is total number
 133 of generations an organism can live, δ is the Dirac delta function, indicating that
 134 adults are sessile, and both K and δ are functions of $(x - y)$. (In BWP, N_i is uniform
 135 for each generation, and is called $N_{\text{fec.}}$) Neubert & Caswell (2000) show that the
 136 invasion speed can be found from the matrix formed by computing the moment
 137 generating function of each element of \mathbf{A} , which is calculated by multiplying each
 138 element by e^{sx} , and integrating from $-\infty$ to ∞ . This results in

$$\mathbf{B} = \begin{bmatrix} N_1M(s) & N_2M(s) & \dots & N_nM(s) \\ r_1 & 0 & \dots & 0 \\ 0 & r_2 & \dots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{bmatrix} \quad (5)$$

139 where $M(s)$ is the moment generating function of the dispersal kernel K .

140 If the mean dispersal is towards smaller x , the upstream dispersal speed is given
 141 by the minimum

$$c = \min_{s>0} \left\{ \frac{1}{s} \ln(\rho(s)) \right\} \quad (6)$$

142 as a function of s , where $\rho(s)$ is the largest eigenvalue of \mathbf{B} (Neubert & Caswell,
 143 2000). Now, at the critical value of population growth that just allows a population
 144 to be retained, the upstream invasion speed must be zero. If the invasion speed is
 145 negative, the population is being washed downstream, and if it is greater than zero,
 146 the population has more than enough growth to persist (BWP). Thus, at the critical

147 growth rate for retention, $\rho(s)$ must be 1. The largest eigenvalue of the matrix \mathbf{B} is
 148 given by largest root of the characteristic polynomial (Strang, 1988)

$$\begin{aligned} \rho^n - N_1 M(s) \rho^{n-1} - r_1 N_2 M(s) \rho^{n-2} - r_1 r_2 N_3 M(s) \rho^{n-3} \\ \cdots - (r_1 r_2 \cdots r_{n-1}) N_n M(s) = 0. \end{aligned} \quad (7)$$

149 Setting $\rho = 1$, and assuming that K and thus $M(s)$ does not vary with the age of
 150 the parents, the critical condition of $c = 0$ becomes

$$\sum_{j=1}^n \left\{ \left(\prod_{l=1}^{j-1} r_l \right) N_j \right\} M(s) = 1. \quad (8)$$

151 The sum of the products above is just the likelihood that an adult reaches age
 152 j , $\prod_{l=1}^{j-1} r_l$, multiplied by the reproductive success for that age, N_j , i.e., it is just
 153 the expected total reproduction of an individual N , neglecting density dependence
 154 effects. Thus the minimum total lifetime reproduction needed for retention is given
 155 when condition (8) is satisfied for the smallest value of $M(s)$, i.e.,

$$N \min_{s>0} M(s) = 1. \quad (9)$$

156 This condition depends only on the dispersal kernel (which sets $M(s)$) and N ,
 157 regardless of how the fecundity of the organism varies with age. Thus retention
 158 is governed by the expected total number of larvae which would settle and reach
 159 reproductive competency in the absence of density dependent effects over the lifetime
 160 of an adult, N , and the dispersal kernel. For the Gaussian kernel with mean dispersal
 161 distance L_{adv} , the moment generating function is given by

$$M(s) = \exp(L_{\text{adv}} s + L_{\text{diff}}^2 s^2 / 2). \quad (10)$$

162 Solving (9) with (10), we obtain (1) for species with Gaussian dispersal kernels. We
 163 can recover the results of Byers & Pringle (2006) by noting that their N_{fec} is N_i
 164 above and is the same for each generation, their N_{gen} is n , and they assume $r_i = 1$.
 165 In these limits, $N_{\text{fec}} N_{\text{gen}} = N$, and so their criteria for retention in iteroparous
 166 species reduces to (1).

167 Non-Gaussian kernels

168 For any arbitrary kernel, the critical value of N or L_{adv} needed to allow retention
 169 can be found from (9). However, while straightforward to do so computationally, it
 170 is difficult to make analytical headway in this manner. Instead, we take advantage
 171 of two results. First, Pachepsky et al. (2005) note that the criterion for retention of a
 172 species with a mean larval dispersal distance per generation ($L_{\text{adv}} \neq 0$) can be found
 173 by computing the upstream invasion distance per generation that would result from
 174 its dispersal kernel in the absence of a mean downstream dispersal ($L_{\text{adv}} = 0$). If
 175 this invasion speed exceeds the mean dispersal distance per generation, the species
 176 can be retained. Secondly, Lutscher (2007) computes an approximate invasion speed
 177 for an arbitrary dispersal kernel which has no mean downstream dispersal. When
 178 Lutscher (2007)'s invasion speed in distance per generation exceeds the L_{adv} , the
 179 population will persist.

180 Following Weinberger (1982), we note that the invasion speed for an semelparous
 181 organism with $L_{\text{adv}} = 0$ is given by

$$c = \min_{s>0} \frac{1}{s} \ln(NM(s)) \quad (11)$$

182 where $M(s)$ is the moment generation function of the kernel discussed above. (This
 183 is just (6) written for the semelparous case. As shown in the last section, the
 184 semelparous results will also be applicable to the iteroparous case.) $M(s)$ can be
 185 expanded in the raw moments of the dispersal kernel $M(s) = 1 + \sum_{i=1 \dots \infty} \frac{s^i}{i!} \mu'_i$. Since
 186 the mean downstream transport of the larvae in this calculation is zero, the first raw
 187 moment is zero, and we shall assume that the dispersal kernel is not skewed, so that
 188 the third raw moment is zero. Keeping the expansion to $O(s^4)$ in s of $M(s)$, making
 189 a Taylor series expansion of $\ln(NM(s))$ in (11) to the same order, and finding the
 190 minimum of this expansion, leads to an estimate of the upstream invasion speed in
 191 units of distance per generation of

$$c^* = \sqrt{2L_{\text{diff}}^2 \ln N} \left(1 + \frac{\gamma_2}{12} \ln N\right) \quad (12)$$

192 where γ_2 is the excess kurtosis as defined in the introduction. When c^* is greater
 193 than L_{adv} , the species can persist. Solving for this criterion leads to the expression
 194 given in (2).

195 Equation (2) provides an estimate of how the excess kurtosis of a dispersal kernel
196 can alter the estimates of the population growth rate needed to allow a population
197 to persist. However, if one is using dispersal kernels derived either from observation
198 or numerical model, one is likely to find the kernel has non-negligible skew (the third
199 central moment of the dispersal kernel), and may have higher moments that further
200 modify the persistence criterion away from (2). However, it is straightforward to
201 compute the persistence criteria numerically, and in the supplemental online mate-
202 rial for this article there is a Python program to compute the persistence criteria
203 for an arbitrary dispersal kernel. This code directly solves (11) for $c = 0$ for a user
204 defined dispersal kernel.

205 Excess kurtosis only affects population persistence when there is a mean down-
206 stream dispersal of larvae (e.g. $L_{adv}=0$), as has been pointed out by, among others,
207 Lockwood et al. (2002). These results are consistent with (2), for when $L_{adv}=0$, the
208 criterion for persistence becomes $\ln(N) > 0$ or, equivalently, $N > 1$. When $\ln(N)$
209 approaches 0, the terms involving the excess kurtosis in (2) approach zero, and thus
210 no longer affect the criterion.

211 **Acknowledgements**

212 John Wares and Jeb Byers provided some of the biological underpinnings and exam-
213 ples for this paper. This work was funded by NSF grant OCE-0453792 to J. Pringle
214 and the NASA-UNH Research and Discover Program to E. Glick.

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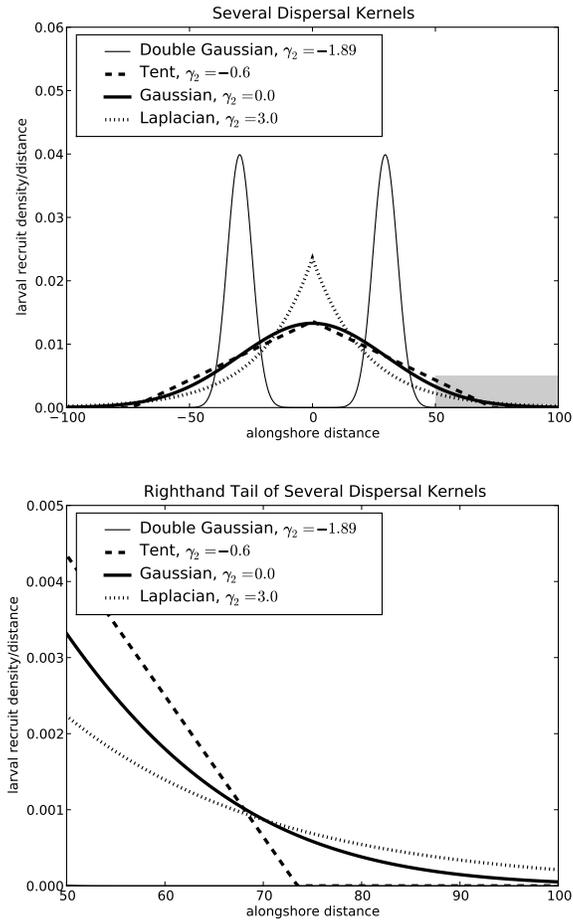


Figure 1: Top) The four larval dispersal kernels examined in figure 2. For each, the standard deviation of the distance the larvae disperse, L_{diff} , is 30km, while the mean distance L_{adv} is 0 in these plots (though not in figure 2). Bottom) The righthand tails of the same kernels, expanded from the shaded portion of the top plot. The double Gaussian kernel's tail is too close to zero to be seen on this scale.

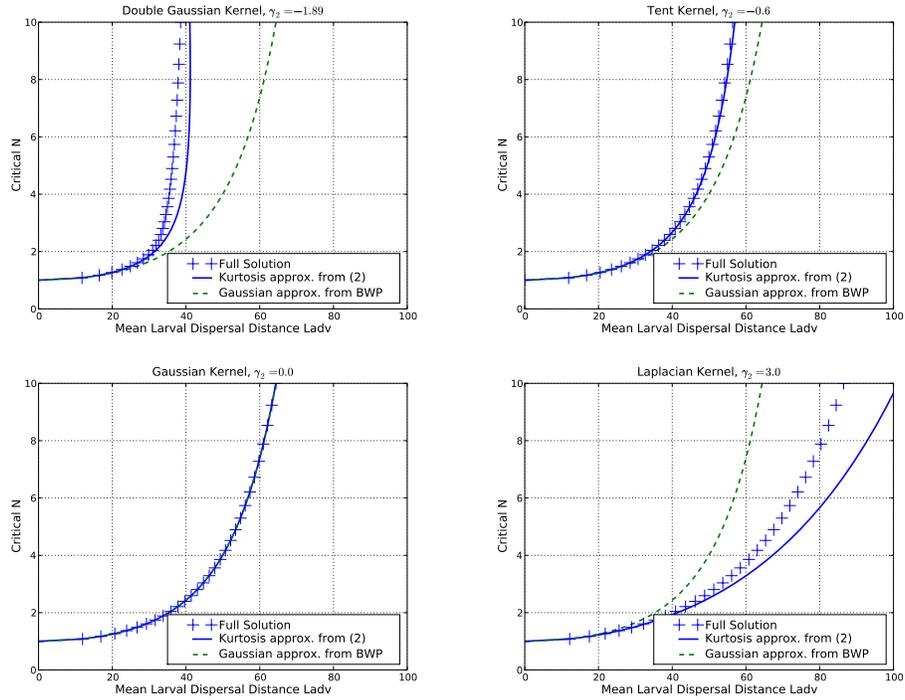


Figure 2: The critical value of N needed to allow the retention of a population as a function of the mean distance the larvae disperse, L_{adv} , for the various dispersal kernels shown in figure 1. Shown is the true value, computed numerically as described in the text, the estimate which includes the effect of the excess kurtosis of the dispersal kernel from equation (2), and the estimate assuming a Gaussian kernel from BWP. The plots are ordered from least excess kurtosis, the Double Gaussian kernel with $\gamma_2 = -1.89$, to the greatest excess kurtosis, the Laplacian kernel with $\gamma_2 = 3.0$.