Summary

Introduction

Materials and Methods

Basic explanation of the models. We modeled a stage-structured population in two stages: immatures and matures. The demography is given by a transition matrix, with...

Under constant environment

Using Lande (2009), under weak selection we have:

$$\Delta \overline{z} = \frac{d \ln \overline{\lambda}(\overline{z})}{d\overline{z}} = \frac{1}{\overline{\lambda}(\overline{z})} \frac{d\overline{\lambda}(\overline{z})}{d\overline{z}} \tag{1}$$

And we have:

$$\overline{\lambda}(\overline{z}) = \sum_{i,j} v_i u_j \overline{a_{ij}}$$

$$= v_I u_I \overline{a_{II}} + v_I u_M \overline{a_{IM}} + v_M u_I \overline{a_{MI}} + v_M u_M \overline{a_{MM}}$$

With $\overline{a_{ij}}$ the expected values of the coefficient of the transition matrix. Thus,

$$\overline{\lambda}(\overline{z}) = v_I u_I \left[\overline{f_1}(\overline{z}) m s_0 + (1 - m) \overline{s_I}(\overline{z}) \right] + v_I u_M s_0 \overline{f_2}(\overline{z})
+ v_M u_I m s_M + v_M u_M s_M$$
(2)

$$\frac{d\overline{\lambda}(\overline{z})}{d\overline{z}} = v_I u_I \left[\frac{d\overline{f_1}(\overline{z})}{d\overline{z}} m s_0 + (1 - m) \frac{d\overline{s_I}(\overline{z})}{d\overline{z}} \right] + v_I u_M s_0 \frac{d\overline{f_2}(\overline{z})}{d\overline{z}}$$
(3)

Because f_i and s_I are gaussians we can write the population means $\overline{f_i}$ and $\overline{s_I}$ easily.

$$\overline{f_1}(\overline{z}) = f_1(\theta_f) \sqrt{\frac{\omega_f}{\omega_f + P_I}} \exp\left(-\frac{(\overline{z} - \theta_f)^2}{2(\omega_f + P_I)}\right)$$
(4a)

$$\overline{f_2}(\overline{z}) = f_2(\theta_f) \sqrt{\frac{\omega_f}{\omega_f + P_M}} \exp\left(-\frac{(\overline{z} - \theta_f)^2}{2(\omega_f + P_M)}\right)$$
(4b)

$$\overline{s_I}(\overline{z}) = s_I(\theta_s) \sqrt{\frac{\omega_s}{\omega_s + P_I}} \exp\left(-\frac{(\overline{z} - \theta_s)^2}{2(\omega_s + P_I)}\right)$$
(4c)

Thus we can derive these expression with respect to \overline{z} :

$$\frac{\partial \overline{f_1}(\overline{z})}{\partial \overline{z}} = f_1(\theta_f) \sqrt{\frac{\omega_f}{\omega_f + P_I}} \frac{\partial \exp\left(-\frac{(\overline{z} - \theta_f)^2}{2(\omega_f + P_I)}\right)}{\partial \overline{z}}$$

$$= f_1(\theta_f) \sqrt{\frac{\omega_f}{\omega_f + P_I}} \exp\left(-\frac{(\overline{z} - \theta_f)^2}{2(\omega_f + P_I)}\right) \frac{\theta_f - \overline{z}}{\omega_f + P_I}$$

$$= \overline{f_1}(\overline{z}) \frac{\theta_f - \overline{z}}{\omega_f + P_I}$$
(5)

We obtain similar formulas for $\overline{f_2}$ and $\overline{s_I}$. Plugging (5) into (3) we have:

$$\frac{d\overline{\lambda}(\overline{z})}{d\overline{z}} = v_I u_I \left[\frac{\theta_f - \overline{z}}{\omega_f + P_I} m s_0 + (1 - m) \frac{\theta_s - \overline{z}}{\omega_s + P_I} \right] + v_I u_M s_0 \frac{\theta_f - \overline{z}}{\omega_f + P_M}$$
 (6)

Using (6) into (1) gives us after rearranging: We have for variations of phenotype, under weak selection:

$$\Delta \overline{z} = (\theta_f - \overline{z}) \left[\frac{v_I u_I G_I s_0 m \overline{f_1}}{\lambda (P_I + \omega_f)} + \frac{v_I u_M G_M s_0 \overline{f_2}}{\lambda (P_M + \omega_f)} \right] + (\theta_s - \overline{z}) \left[\frac{v_I u_I G_I \overline{s_I} (1 - m)}{\lambda (P_I + \omega_s)} \right]$$
(7)

Within the square brackets, we see weighting average of fecundity and survival. Thus, we define them as γ_f and γ_s such as:

$$\gamma_f = \frac{v_I u_I s_0 m \overline{f_1}}{\lambda (P_I + \omega_f)} + \frac{v_I u_M \frac{G_M}{G_I} s_0 \overline{f_2}}{\lambda (P_M + \omega_f)}$$
(8a)

and

$$\gamma_s = \frac{v_I u_I \overline{s_I} (1 - m)}{\lambda (P_I + \omega_s)} \tag{8b}$$

We end up having a simpler expression for $\Delta \overline{z}$ under constant environment:

$$\Delta \overline{z} = -G_I \left[\gamma_f (\overline{z} - \theta_f) + \gamma_s (\overline{z} - \theta_s) \right] \tag{9}$$

Under varying environment

From Engen et al. (2011), we derived equations for mean variation of phenotype on our model.

We supposed an auto-correlated fluctuating environment influencing optimum such as $\theta_f = \overline{\theta}_f + \alpha_f \epsilon_t$. With $\epsilon_{t+1} = (1 - \rho)\overline{\epsilon} + \rho \epsilon_t + \xi$ with ξ a gaussian noise vector with variance σ_{ξ}^2 and mean 0.

Results

Subheading1

Subheading2

Discussion

Authors Contributions and Acknowledgments

References

(Barfield et al., 2011)

References

Barfield, M., Holt, R. D. and Gomulkiewicz, R. (2011). Evolution in Stage-Structured Populations (2 versions). The American Naturalist *177*, 397--409.

Engen, S., Lande, R. and Sæther, B.-E. (2011). Evolution of a Plastic Quantitative Trait in an Age-Structured Population in a Fluctuating Environment. Evolution *65*, 2893--2906.

Lande, R. (2009). Adaptation to an extraordinary environment by evolution of phenotypic plasticity and genetic assimilation. Journal of Evolutionary Biology *22*, 1435--1446.