Complex life cycles and eco-evolutionary dynamics in changing environments

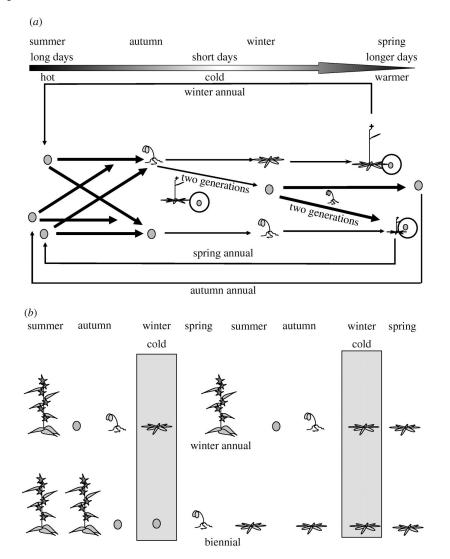
Module 5

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Winterschool "Forward-time individual-based simulations in Ecology and Evolutionary Biology" Mar 20-24 Tvärminne

Complex life cycle

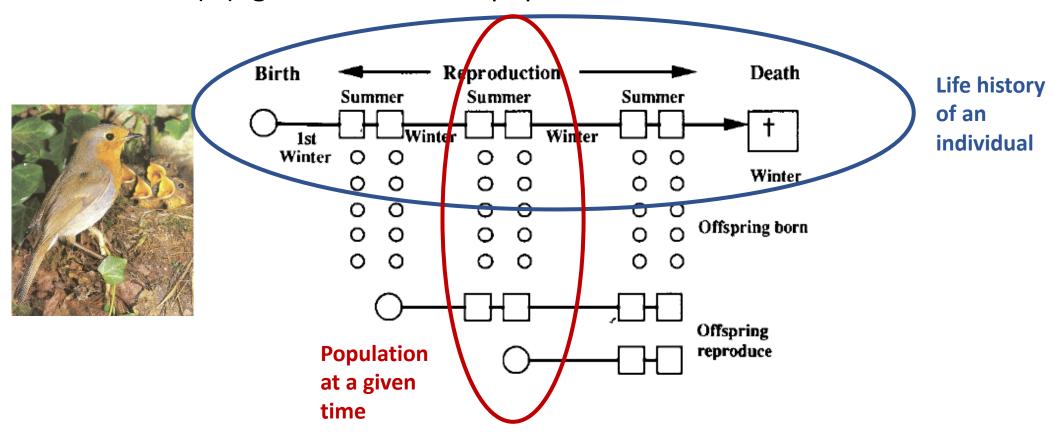
Life history





Structured populations

- What is it?
 - Population heterogeneous regarding the (st)age of individuals
 - Several (st)age co-occur in the population



Structured populations

« ...to understand the mechanistic basis of population fluctuations, we also generally need to allow for the demographic structure of the population. Animals of different age and sex often expend different amounts of energy at different times of the year, associated with behavior, the costs and timing of reproductive effort, growth, and maintenance. These differences can lead to variation in the strength and form of the association between vital rates, density, and extrinsic factors across age and sex classes. Fluctuations in the structure of the population, independent of the total population size, may then lead to interactions between density, weather, and the population structure to introduce heterogeneities that have an important impact on the population dynamics.» Coulson et al. 2001 Science

How to describe changes in population structure and size with time?

Life stages



- Embryos
- Yearlings
- Juveniles
- Mature females
- Post-reproductive females

Life history events

Census at t	Reproduction	Survival	Growth	Census at t+1
$N_0(t)$	$N_2f_2 + N_3f_3 + N_0$	$(N_2f_2 + N_3f_3 + N_0)s_0$	$(N_2f_2 + N_3f_3 + N_0)s_0(1 - g_0)$	$N_0(t+1)$
$N_1(t)$	N_1	N_1s_1	$(N_2f_2 + N_3f_3 + N_0)s_0(1 - g_0) + N_1s_1(1 - g_1)$	$N_1(t+1)$
$N_2(t)$	N_2	N_2s_2	$N_2 s_2 (1 - g_2) + N_1 s_1 g_1$	$N_2(t+1)$
$N_3(t)$	N_3	N_3s_3	$N_3 s_3 (1 - g_3) + N_2 s_2 g_2$	$N_3(t+1)$
$N_4(t)$	N_4	N_4s_4	$N_4 s_4 + N_3 s_3 g_3$	$N_4(t+1)$

Number of females

If fecundities account for both males and females: account for the sex ratio

How to describe changes in population structure and size with time?
 Newborn stage embedded in yearling stage; yearling stage for one year

$$g0 = 1$$

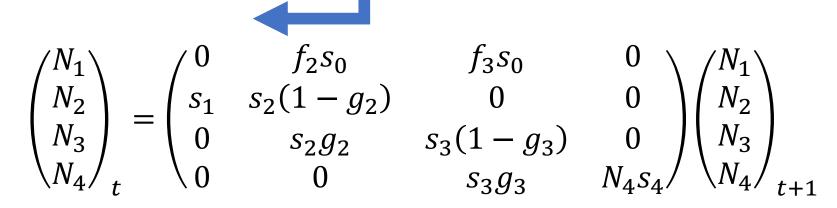
$$g1 = 1$$

Census at t	Reproduction	Survival	Growth	Census at t+1
0	$N_2f_2 + N_3f_3$	$(N_2f_2 + N_3f_3)s_0$		
$N_1(t)$	N_1	N_1s_1	$(N_2f_2 + N_3f_3)s_0$	$N_1(t+1)$
$N_2(t)$	N_2	N_2s_2	$N_2 s_2 (1 - g_2) + N_1 s_1$	$N_2(t+1)$
$N_3(t)$	N ₃	N_3s_3	$N_3 s_3 (1 - g_3) + N_2 s_2 g_2$	$N_3(t+1)$
$N_4(t)$	N_4	N_4s_4	$N_4 s_4 + N_3 s_3 g_3$	$N_4(t+1)$

Column

contributes

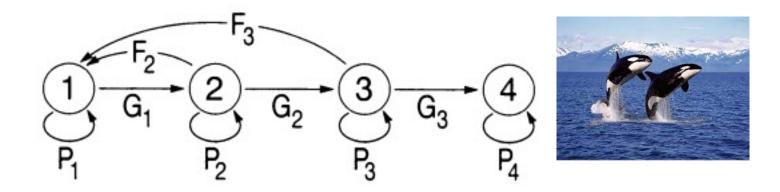
Matrix representation

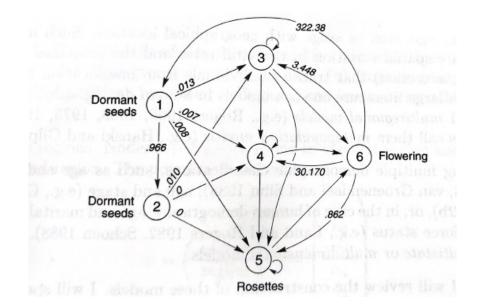


$$N(t+1) = A N(t)$$

The elements of **A** (i.e. the a_{ij}) are the **demographic transitions**

Age structure and stage structure







Continuous vs Discrete

Continuous time

$$\frac{d N_{x}(t)}{dt} = \sum_{y} N_{y}(t)b_{y\to x}(t) - N_{x}(t) \sum_{y} g_{x\to y}$$

Age/Other continuous trait

$$N_0(t+1) = \int_0^\infty N_x(t) f_x(t) dx$$

In practice,

- Wild populations are observed at discrete time intervals
- Continuous characters can structure the population (e.g. size) but can always be discretized into stages (see IPM framework)

Simulations with NEMOAGE

Simulate the demography of the killerwhale (Caswell and Brault 1993)

Leslie matrix for the female population

$$\mathbf{A} = \begin{pmatrix} 0 & 0.0043 & 0.1132 & 0 \\ 0.9775 & 0.9111 & 0 & 0 \\ 0 & 0.0736 & 0.9534 & 0 \\ 0 & 0 & 0.0452 & 0.9804 \end{pmatrix}$$



- Just demography (no trait or selection)
- ➤ No density regulation
- > 300 generations
- ➤ Vary initial stage structure
- Pay attention to the mating system

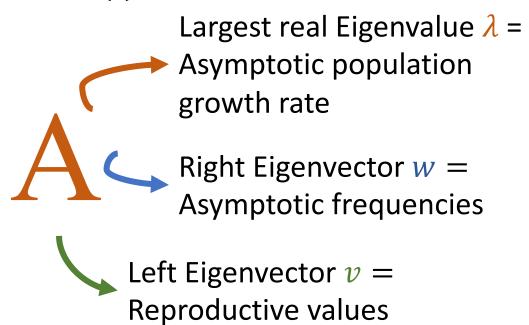
Simulations with NEMOAGE

Simulate the demography of the killerwhale (Caswell and Brault 1993)

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 Observe that the population reaches a constant (st)age structure corresponding to a constant exponential growth rate (log representation) after a transient period

- Exponential growth rate and equilibrium population size
 - Asymptotic behavior
 - Directly obtained from the Eigen properties of the transition matrix $\mathbf{N}(t+1) = \mathbf{A} \ \mathbf{N}(t)$



Simulations with NEMOAGE

- Compare simulation result with deterministic expectation
- > Use R to obtain deterministic expectations and compare with simulations

Simulations with NEMOAGE

- Compare simulation result with deterministic expectation
- > Use R to obtain deterministic expectations and compare with simulations

 Deterministic expectations are landmarks to check the simulation model

Sources of stochasticity in population dynamics

The fitness of individual i in a given year is

$$w_i = \mu_w + \delta_i$$
, with $E[\delta_i] = 0$

The rate of population increase in a given year is

$$\lambda = \overline{W} = \frac{1}{N} \sum_{i=1}^{N} w_i = \mu_w + \frac{1}{N} \sum_{i=1}^{N} \delta_i$$

Stochasticity in population dynamics results from yearly variation in population growth rate

$$\sigma_w^2 = \operatorname{Var}[\mu_w] + \frac{1}{N^2} \sum_{i=1}^N \operatorname{Var}[\delta_i] = \sigma_e^2 + \frac{1}{n} \sigma_d^2$$

Environmental stochasticity

Demographic stochasticity F(population size)

Sources of stochasticity in population dynamics

In stage-structured populations

$$\sigma_d^2 = \sum_{ij} \sum_{kl} \frac{\partial \ln \lambda}{\partial a_{ij}} \frac{\partial \ln \lambda}{\partial a_{kl}} \frac{\operatorname{cov}(\delta_{ij}, \delta_{kl})}{\partial a_{kl}}$$

Effect of the transitions on fitness

Stochastic (co)variation in transitions

Sources of stochasticity in population dynamics

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Effect of the transitions on fitness

Stochastic (co)variation in transitions

Depends on life history!

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Effect of the transitions on fitness

Stochastic (co)variation in transitions

Individual based simulations

Depends on life history!

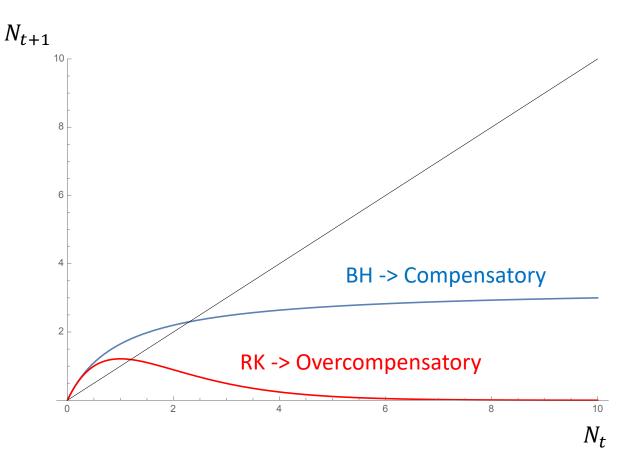
Density-dependence

• Beverton-Holt

Coming in stage i

$$N_i(t+1) \propto \frac{N_j(t)}{1 + b_k N_k(t)}$$
Stength of Competitor from competition stage k

• Ricker $N_i(t+1) \propto N_i(t) \exp(-b_k N_k(t))$



Simulations with NEMOAGE

Compare Beverton-Holt and Ricker regulation

$$\mathbf{A} = \begin{pmatrix} 0 & 0 & f_3 \\ 0.45c(\mathbf{N}) & 0 & 0 \\ 0 & 0.05 & 0.1 \end{pmatrix}$$

 $c(\mathbf{N})$ measures the effect of density-dependent competition on survival

- \triangleright Vary the orders of magnitude of f_3
- > Start with a1 individuals only
- Vary the mode of regulation

From demography to fitness

Asexual case with two genotypes

$$\mathbf{A}_i = E[\mathbf{A}|i]$$

Projection matrix given that the individual has genotype i

λ

Largest Eigenvalue of A

-> Asymptotic growth rate of genotype i

$$\bar{\lambda} = p\lambda_1 + q\lambda_2$$

Mean asymptotic growth rate

Change in genotypic frequencies

$$\Delta q = \frac{pq(\lambda_2 - \lambda_1)}{\bar{\lambda}}$$

The asymptotic growth rate can be used as a measure of fitness



Sexual case for a quantitative trait

$$z=g+e$$
 Quantitative trait with value given by an additive genetic component and an environmental component

Distribution in each stage given by $p_i(g, z)$

Sexual case for a quantitative trait

$$z = g + e$$

$$a_{ij}(z) = t_{ij}(z) + f_{ij}(z)$$

- Transitions between stages depend on the trait (selection)
- We need to separate transitions of individuals (e.g. survival) from production of newly produced individuals (e.g. reproduction)

Sexual case for a quantitative trait

$$z = g + e$$

$$a_{ij}(z) = t_{ij}(z) + f_{ij}(z)$$

$$N_j(t+1) = \sum_i \bar{a}_{ij} N_i(t)$$

$$N(t+1) = \mathbf{A} N(t)$$

Demography depends on the mean transitions

Joint distribution of phenotype and genetic value at t + 1

$$p'_{i}(g,z) = \theta_{i}(g,z) \frac{\sum_{j} N_{j}(t)\bar{t}_{ij}}{N_{i}(t+1)} + \varphi_{i}(g,z) \frac{\sum_{j} N_{j}(t)f_{ij}}{N_{i}(t+1)}$$

• Joint distribution of phenotype and genetic value at t + 1

$$p_{i}'(g,z) = \theta_{i}(g,z) \frac{\sum_{j} N_{j}(t)\bar{t}_{ij}}{N_{i}(t+1)} + \varphi_{i}(g,z) \frac{\sum_{j} N_{j}(t)f_{ij}}{N_{i}(t+1)}$$

Bookkeeping...

How to be predictive?

- Further assumptions for analytical tractability
- Simulations

- Normal approximation: The joint distribution of genotypic values
- Weak selection: the ecological processes occur much faster than the evolutionary processes
- Lande's theorem

$$\bar{z}(t+1) - \bar{z}(t) = G \frac{\partial \ln \bar{\lambda}}{\partial \bar{z}}$$

Change in the mean trait value between two censuses (including all life history events)

Heritable genetic variance

Selection gradient = Effect of the trait on fitness

Classical breeder's equation

Decomposition by stage

$$\bar{z}(t+1) - \bar{z}(t) = \frac{1}{\bar{\lambda}} \sum_{ij} G_i \frac{\partial \bar{\lambda}}{\partial \bar{a}_{ij}} \frac{\partial \bar{a}_{ij}}{\partial \bar{z}}$$

Evolutionary change

Additive genetic variance in stage i

Sensitivity of fitness to a change in \bar{a}_{ij}

Effect of the trait on \bar{a}_{ij}

Lande 1982 Ecology Barfield et al. 2011 AmNat

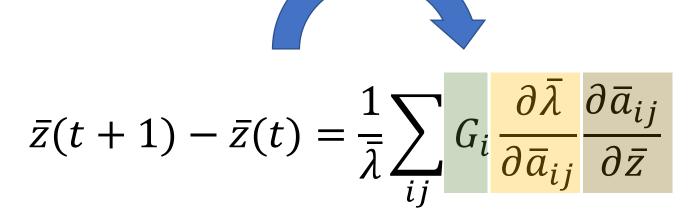
Demographic sensitivity

$$\bar{z}(t+1) - \bar{z}(t) = \frac{1}{\bar{\lambda}} \sum_{ij} G_i \frac{\partial \bar{\lambda}}{\partial \bar{a}_{ij}} \frac{\partial \bar{a}_{ij}}{\partial \bar{z}}$$

Evolutionary change

Genetics Demography Selection

- Demographic sensitivity
- Eco-evolutionary feedbacks



Evolutionary change

Genetics Demography Selection

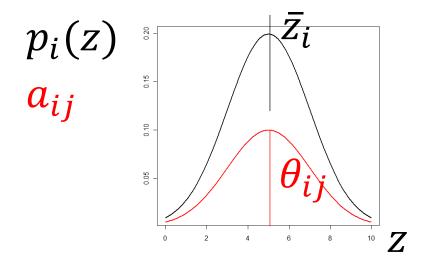
We need to specify how the transitions depend on the environment

Depends on knowledge on the case study



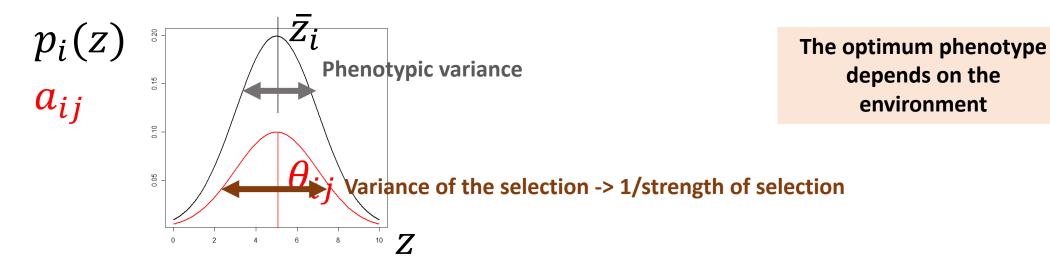
A general model: Moving optimum

- Gaussian selection
- Moving optimum model (in space or time)

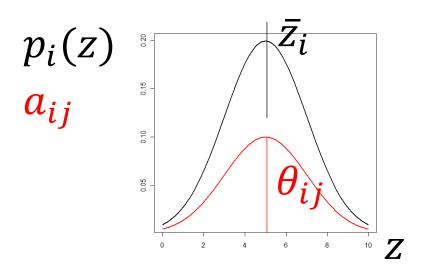


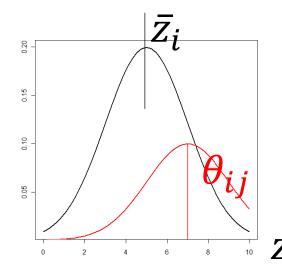
The optimum phenotype depends on the environment

- Gaussian selection
- Moving optimum model (in space or time)



- Gaussian selection
- Moving optimum model (in space or time)





The optimum phenotype depends on the environment

Simulations with NEMOAGE

How does a structured population adapted to a change in environment?

$$\mathbf{A} = \begin{pmatrix} 0.4 & 0 & 2.89 \\ 0.4 & 0 & 0 \\ 0 & 0.8c(\mathbf{N}) & 0.5 \end{pmatrix}$$

- Quanti trait
- > Choose a number of loci, the effect of mutations and the mutation rate
- \triangleright Gaussian selection with $\theta(t=0)=0$
- Compare expected genetic variance with the strength of selection
- \blacktriangleright Perform burn-in simulations for different life histories (e.g. vary s_3) and a constant optimum
- Verify that genetic variance has reached equilibrium

Simulations with NEMOAGE

How does a structured population adapted to a change in environment?

$$\mathbf{A} = \begin{pmatrix} 0.4 & 0 & 2.89 \\ 0.4 & 0 & 0 \\ 0 & 0.8c(\mathbf{N}) & 0.5 \end{pmatrix}$$

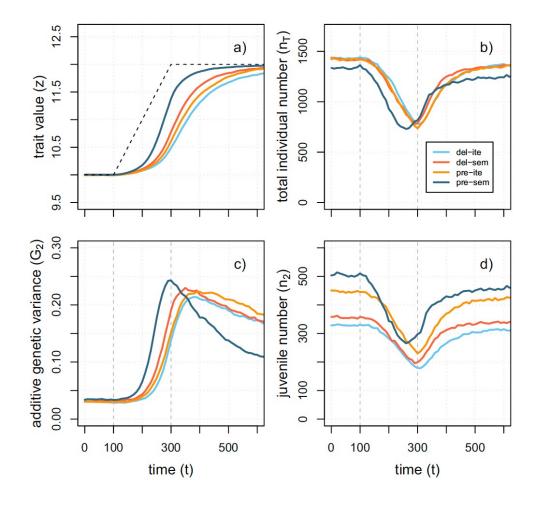
- > Implement environmental change
- selection_rate_environmental_change (@g0 0, @g250 0.05, @g500 0)
- ➤ Characterize the evolutionary and demographic dynamics
- ➤ How does the life history trait that you varied affect the evolutionary and demography dynamics?
- ➤ Add some cloning How does it affect the dynamics

Evolutionary Lag and effect on life history

Two life-history gradient

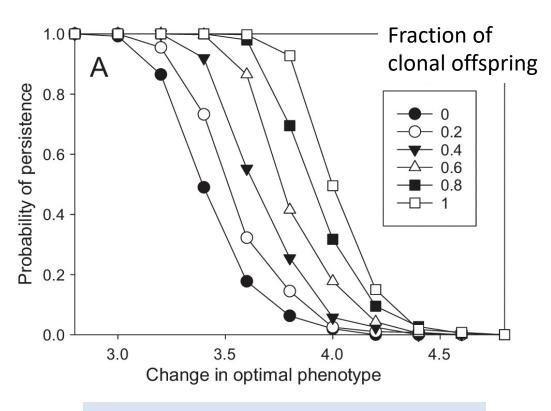
- Precocious delayed maturity
- Semelparous Iteroparous

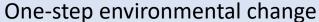
Evolvability-persistence trade-off

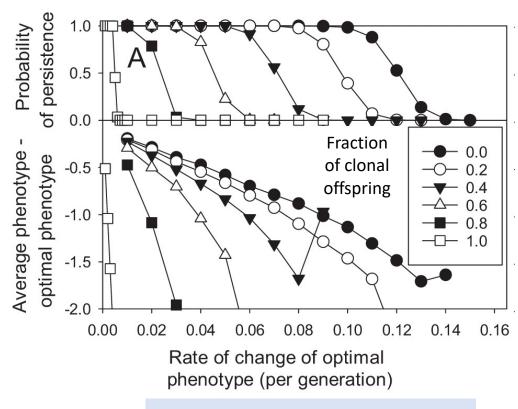


Orive et al. 2017 AmNat

Cloning



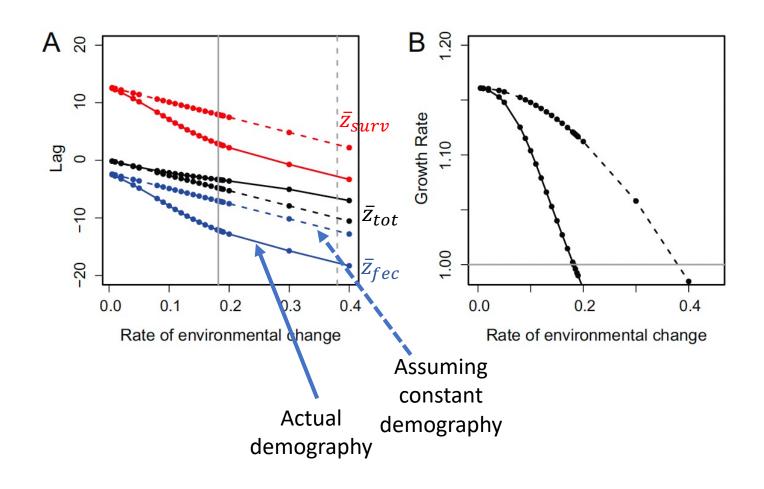




Linear environmental change

Effect of cloning depends on the type of environmental change

Eco-evolutionary feedback



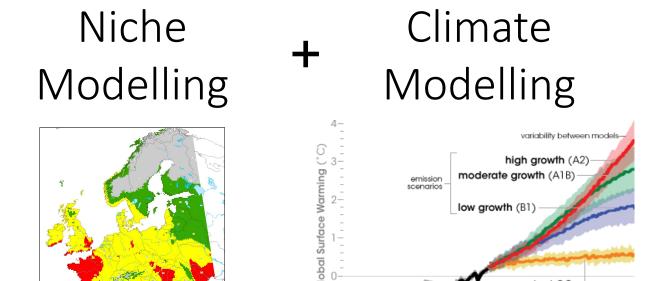
Individual based simulations in structured populations

 Modeling evolution in structured populations increases a lot the degree of freedom!

Individual based simulations

- Account for stochastic processes (demographic stochasticity and drift)
- Account for demography-genetics-evolution feedback loops
- Allow to make predictions in complex scenarios (e.g. spatiotemporal variation of the environment)

Predicting species range



"Today's scientists are facing the enormous challenge of predicting how climate change will affect species distributions and species assemblages. To do so, ecologists are widely using phenomenological models of species distributions that mainly rely on the concept of species niche and **generally ignore species**' **demography, species**' **adaptive potential, and biotic interactions**."

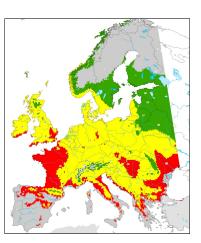
2000

2100

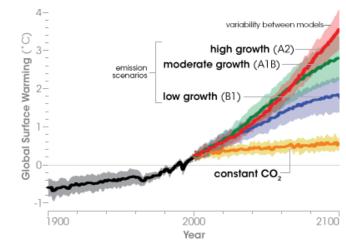
Lavergne et al. 2010 Annu. Rev. Ecol. Evol. Syst.

Predicting species range

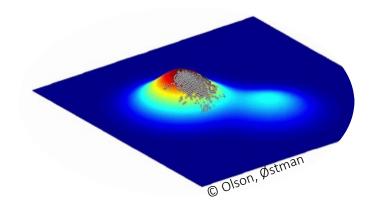
Niche Modelling



+ Climate Modelling



Eco-Evolutionary Modelling



- Complex life history
- Demography
- Dispersal capacity
- Evolution

Simulations with NEMOAGE

- Evolution of a species range
- Simulate a species distribution model (R-script)
- > Define a life history of your choice (see Compadre or Comadre databases if needed)
- Predict future species distribution (build-on previous simulations)
- ➤ How are the predictions sensitive to unknown parameters (e.g. genetic parameters choose one)?
- How does evolution play out? And demography?