eco-evolutionary models

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library(deSolve)
library(reshape2)
library(ggplot2)
theme_set(theme_bw())

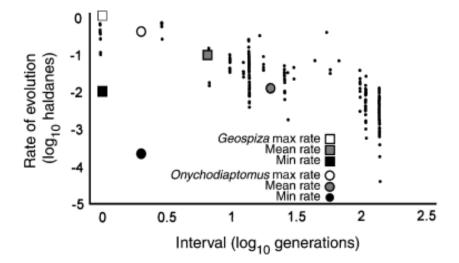
what are eco-evolutionary models?

- Slobodkin/Hutchinson: evolutionary theater, ecological play (i.e. ecological rates >> evolutionary rates)
- what if the rates are similar?
 - ecological dynamics (e.g. logistic equation)
 - evolutionary dynamics (changes in traits)
- alternatively, population genetics plus population dynamics

why aren't all models like this?

- eco/evo time scales are often different
- it's hard!
 - most population genetic models assume constant population size
 - most ecological models assume constant traits

(Ellner, Geber, and Hairston 2011)



(1 Haldane = change by a factor of 1 standard deviation/generation)

endpoints

- ecology only: e.g. Lotka-Volterra predator-prey or competition; Rosenzweig-MacArthur model ...
- evolution (population genetics); e.g. see Lande, Lande and Arnold ... (discrete-time matrix models)

how do we do it?

- range of realism/complexity
- individual or agent-based models
 - each individual has a genotype and a phenotype
 - rules for life history and interactions
- models for the distribution of a continuous trait
 - partial differential equations
 - reaction-diffusion equations
 - includes demography, mutation
- moment equations
 - simplify PDEs to equations for the means and variances of traits, plus population densities
- Price equations:
 - further simplify to equations for the means of traits (assume constant variance)

individual-based models

- maximum detail, realism
- slow
- in R: try to be clever
- also: Python/Cython, C, NetLogo

distribution models

- write equations to follow the fractions of the population in each bin
- i.e. write out $dN(\alpha)/dt$ for $\alpha = 0.01, 0.02, ...$
- move between bins via mutation kernel, e.g. $m(-1+C\sum_{i\neq j}\exp(-((\alpha_i-1))))$ $(\alpha_i)/\sigma^2$) where *C* is a normalization constant, i.e. make sure the sum is 1/ population is balanced
- boundary conditions!

PDEs

- advection (e.g. growth) $\partial N/\partial \alpha$
- diffusion (e.g. mutation): $\partial^2 N/\partial \alpha^2$

second partial derivatives by

moment equations

- figure out the dynamics of the mean and variance
- in general; mean follows
- Birch and Bolker (2015)

Price equations

- often used to decompose multi-level effects (Collins and Gardner 2009)
- simplest version reduces to $d\bar{T} = V_0 \partial (\text{fitness}) / \partial (T)$
- epidemic models: Troy Day and Proulx (2004); T. Day and Gandon (2006)

epidemic model

$$\frac{dS}{dt} = m(N - S) - \beta(\bar{\alpha})SI$$

$$\frac{dI}{dt} = \beta(\bar{\alpha})SI - (m + \alpha)I$$

$$\frac{d\bar{\alpha}}{dt} = h\left(S\frac{\partial\beta}{\partial\bar{\alpha}} - 1\right)$$

$$\beta(\bar{\alpha}) = c\bar{\alpha}^{1/8amma}$$

R implementation

```
## tradeoff curve
beta = function(alpha, c0, gamma = 2) {
    c0 * alpha^(1/gamma)
}
derivbeta = function(alpha, c0, gamma = 2) {
    c0/gamma * alpha^(1/gamma - 1)
}
derivfun1 = function(t, y, parms) {
    derivs = with(as.list(c(y, parms)), c(m *
        (N - S) - beta(alpha, c0, gamma) * S *
        I, (beta(alpha, c0, gamma) * S - (m +
        alpha)) * I, h * (S * derivbeta(alpha,
        c0, gamma) - 1)))
```

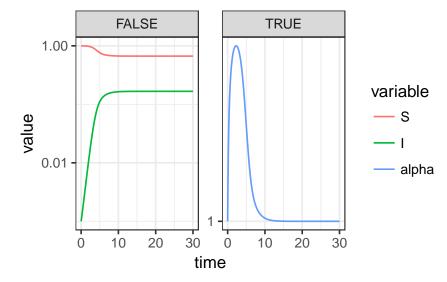
```
list(derivs)
}
Parameters
params3 = c(h = 5, c0 = 3, m = 1, N = 1, gamma = 2)
startvals = c(S = 0.999, I = 0.001, alpha = 1)
derivfun1(t = 0, y = startvals, parms = params3)
## [[1]]
## [1] -0.001997 0.000997 2.492500
L1 = ode(y = startvals, times = seq(0, 30, by = 0.1),
    parms = params3, func = derivfun1)
default plot (a bit ugly)
plot(L1)
                   s
                                                          ı
  1.00
  0.00
                                         0.10
  0.80
                                         0.05
  0.70
                                         0.00
          5
                                                     10
              10
                   15
                       20
                           25
                                                         15
                                                              20
                                                                  25
                                                                       30
                  time
                                                         time
                  alpha
  2.0
  6.
  1.6
  4.
  1.2
  0.
```

nicer plot

5 10 15 20 25 30

time

```
Lm1 <- melt(as.data.frame(L1), id.vars = "time") ## convert to long format</pre>
Lm1 <- transform(Lm1, vir = (variable == "alpha")) ## add var for faceting</pre>
```

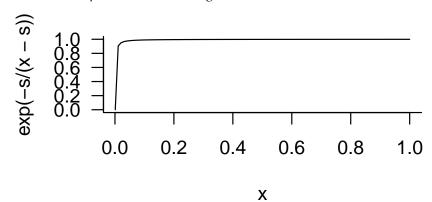


predator-prey system (Abrams and Matsuda 1997)

- MacArthur-Rosenzweig model: logistic growth + Holling type II predation
- trait *C* increases both growth rate and vulnerability of prey

$$\begin{aligned} \frac{dP}{dt} &= P\left(\frac{BCN}{1+hCN} - d\right) \\ \frac{dN}{dt} &= N\left(R + qC - kN - \frac{CP}{1+hCN}\right) \\ \frac{dC}{dt} &= V_0 \exp(-s/(C-s))\left(q - \frac{P}{1+hCN}\right) \end{aligned}$$

what is the exponential term doing there?



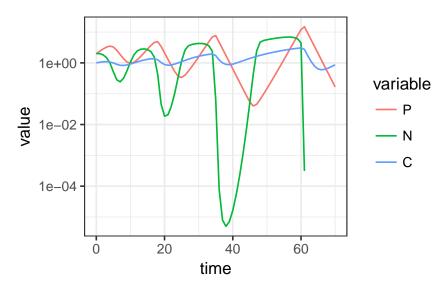
predator-prey system

Depending on parameters, this system can show

- expanding cycles
- stable cycles
- chaos
- stable equilibrium

R implementation

```
parms1 <- c(h = 1, B = 1, q = 1.2, d = 0.5, V0 = 0.1,
    R = 0.5, k = 0.5, s = 0.001)
init < c(P = 2, N = 2, C = 1)
AMgrad <- function(t, y, parms) {
    grad <- with(as.list(c(y, parms)), {</pre>
        c(P = P * (B * C * N/(1 + h * C * N)) -
            d), N = N * (R + q * C - k * N - C *
            P/(1 + h * C * N)), C = V0 * exp(-s/(C -
            s)) * (q - P/(1 + h * C * N)))
    })
    return(list(grad))
}
## check that gradient function works
AMgrad(t = 0, y = init, parms = parms1)
## [[1]]
##
                                   C
## 0.3333333 0.06666667 0.05327997
run the model
L2 = ode(t = 0.70, y = init, parms = parms1, func = AMgrad)
Lm2 <- melt(as.data.frame(L2), id.vars = "time")</pre>
plot
gg1 %+% Lm2
## Warning in self$trans$transform(x): NaNs
## produced
## Warning: Transformation introduced infinite
## values in continuous y-axis
## Warning: Removed 9 rows containing missing
## values (geom_path).
```



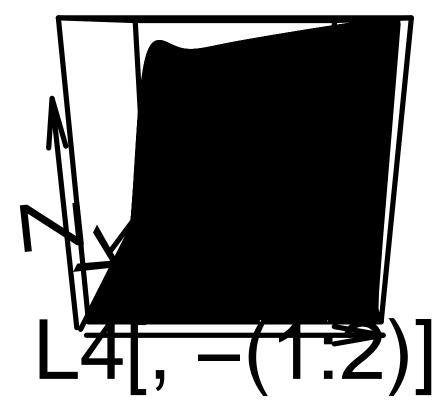
```
## limit cycles
parms2 <- c(h = 1, B = 1, q = 0.8, d = 0.5, V0 = 0.05,
    R = 0.5, k = 1, s = 0.001
## same but higher V0: chaotic/long-period
## limit cycles?
parms3 <- c(h = 1, B = 1, q = 0.8, d = 0.5, V0 = 0.075,
    R = 0.5, k = 1, s = 0.001
## expanding cycles for V0 larger (0.5) stable
## for V0 < 0.03148 ?
```

Full-distribution models

Now let's build a full model for the distribution.

```
## define vector of virulence values
alphavec <- seq(0.01, 10, by = 0.1)
## correspond vector of transmission rates
betavec <- beta(alphavec, c0 = 3, gamma = 2)
## mutation model
mut.sd <- 0.02 ## mutational std dev
## compute (alpha_i-alpha_j)^2 for all {i,j}
sqdist <- outer(alphavec, alphavec, "-")^2</pre>
## mutation distribution is Gaussian
M <- exp(-sqdist/(mut.sd)^2)</pre>
## make sure rows sum to 1 (conservation)
M <- sweep(M, MARGIN = 1, FUN = "/", STATS = rowSums(M))
all(abs(rowSums(M) - 1) < 1e-04) ## check
## [1] TRUE
## gradient function for distribution model
derivfun3 = function(t, y, parms) {
```

```
S < -y[1]
    I < -y[-1]
    derivs = with(as.list(parms), {
        inf <- M %*% (betavec * S * I) ## infection + mutation</pre>
        c(m * (N - S) - sum(inf), inf - (m + alphavec) *
            I)
    })
    list(derivs)
}
## initial values
init.I <- dnorm(alphavec, mean = 1, sd = 0.2)
init.I <- init.I/sum(init.I) * 0.001
startvals.d <- c(S = 0.999, init.I)
g1 \leftarrow derivfun3(t = 0, y = init, parms = params3)[[1]]
tvec = seq(0, 30, by = 0.1)
L4 = ode(y = startvals.d, times = tvec, parms = params3,
    func = derivfun3)
persp(L4[, -(1:2)])
```



```
Itot <- rowSums(L4[, -(1:2)])</pre>
## library(rgl)
## persp3d(tvec,alphavec,L4[,-(1:2)])
```

references

Abrams, Peter A., and Hiroyuki Matsuda. 1997. "Prey Adaptation as a Cause of Predator-Prey Cycles." Evolution 51 (6): 1742-50. doi:10.1111/j.1558-5646.1997.tb05098.x.

Birch, Michael, and Benjamin M. Bolker. 2015. "Evolutionary Stability of Minimal Mutation Rates in an Evo-Epidemiological Model." Bulletin of Mathematical Biology 77: 1985-2003. doi:10.1007/s11538-015-0112-6.

Collins, Sinéad, and Andy Gardner. 2009. "Integrating Physiological, Ecological and Evolutionary Change: A Price Equation Approach." Ecology Letters 12 (8): 744-57. doi:10.1111/j.1461-0248.2009.01340.x.

Day, T., and S. Gandon. 2006. "Insights from Price's Equation into Evolutionary Epidemiology." In Disease Evolution: Models, Concepts, and Data Analyses, edited by Zhilan Feng, Ulf Dieckmann, and Simon Levin.

Day, Troy, and Stephen R. Proulx. 2004. "A General Theory for the Evolutionary Dynamics of Virulence." The American Naturalist 163 (4): E40-E63. http://www.jstor.org/stable/10.1086/382548.

Ellner, Stephen P., Monica A. Geber, and Nelson G. Hairston. 2011. "Does Rapid Evolution Matter? Measuring the Rate of Contemporary Evolution and Its Impacts on Ecological Dynamics." Ecology Letters 14 (6): 603-14. doi:10.1111/j.1461-0248.2011.01616.x.