

Variation in thermal pressures and resource availability drives disease dynamics

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Table of contents

0.1	Model	1
0.1.1	Equilibrium Solutions	2
0.1.2	Brute-force Computation	2

0.1 Model

Vinton and Vasseur (2022) provide a basic model for temperature-dependent consumer-resource dynamics with a chemostat model of resource, R , supply, given by

$$\frac{dR}{dt} = D(S - R) - f(R, T)C, \quad (1)$$

where there is an inflow density S and an outflow rate D and $f(R, T)$ is the functional response of R with respect to temperature T . The biomass change of the consumer C is given by

$$\frac{dC}{dt} = (1 - \delta)f(R, T) - m(T)C, \quad (2)$$

where $(1 - \delta)$ is the consumption efficiency (denoted as a fraction), and m is the rate of respiration. We assume a Boltzmann-Arrhenius relationship and approximate that function $m(T)$ with

$$m(T) = m_a e^{m_b T} + m_c, \quad (3)$$

with m_a , m_b and m_c all > 0 . The functional response is a standard type II, with attack rate a , handling rate $(1/h)$, but re-state according to the Michaelis-Menten form of

$$f(R, T) = I_{max}(T) \times \frac{R}{R + R_{half}}, \quad (4)$$

and $I_{max}T$ is the maximum uptake rate which we state as equivalent to the handling rate $I_{max}T \equiv 1/h$, and then the half-saturation density is made equivalent via $R_{half} \equiv \frac{1}{a \times h}$. Note that the resource saturation is reached by $R/(R_{half} + R)$, which is independent of I_{max} . Last,

$$I_{max}(T) = e^{-(T-T_I)^2\gamma}, \quad (5)$$

is the equation governing the relationship with temperature, where T_I is the optimum temperature for consumption, and γ is the breadth of response.

0.1.1 Equilibrium Solutions

Vinton and Vasseur (2022) showed that there is a coexistence equilibrium solution where

$$R_e = \frac{mR_{half}}{(1 - \delta)I_{max}(T) - m}, \quad (6)$$

and

$$C_e = \frac{D(S - R_e)(R_e - R_{half})}{I_{max}(T)R_e}, \quad (7)$$

0.1.2 Brute-force Computation

We use the standard set of model parameters for the chemostat dynamics, $R_{half} = 2, T_I = 25, \gamma = 150, m_a = 0.01, m_b = 0.1, m_c = 0.05, D = 1, S = 1$.

With these values, we set up a brute-force computation that draws values of R

Here's a single run for R:

```
# Load the deSolve package
library(deSolve)
library(ggplot2)
library(magrittr)
library(dplyr)
```

```

# Define the differential equations
consumer_resource_model <- function(time, state, parameters) {
  R <- state[1]
  C <- state[2]
  T <- parameters["T"]
  D <- parameters["D"]
  S <- parameters["S"]
  delta <- parameters["delta"]
  m_a <- parameters["m_a"]
  m_b <- parameters["m_b"]
  m_c <- parameters["m_c"]
  a <- parameters["a"]
  h <- parameters["h"]
  e <- parameters["e"]
  gamma <- parameters["gamma"]
  T_I <- parameters["T_I"]
  R_half <- parameters["R_half"]

  # Functional response f(R, T)
  I_max_T <- exp(-(T - T_I)^2 * gamma)
  f_R_T <- I_max_T * R / (R + R_half)

  # Respiration rate m(T)
  m_T <- m_a * exp(m_b * T) + m_c

  # Differential equations
  dR_dt <- D * (S - R) - f_R_T * C
  dC_dt <- (1 - delta) * f_R_T - m_T * C

  return(list(c(dR_dt, dC_dt)))
}

# Define the parameters
parameters <- c(
  T = rnorm(1, mean = 25, sd = 1), # Random normal temperature
  D = 1, # Outflow rate
  S = 2, # Inflow resource density
  delta = 0.2, # Consumption efficiency
  e = 0.5,
  m_a = 0.01,
  m_b = 0.1,
  m_c = 0.05,

```

```

    a = 0.5, # Attack rate
    h = 0.1, # Handling rate
    gamma = 150, # Breadth of response
    T_I = 25, # Optimum temperature for consumption
    R_half = 0.5 # Half-saturation density
  )

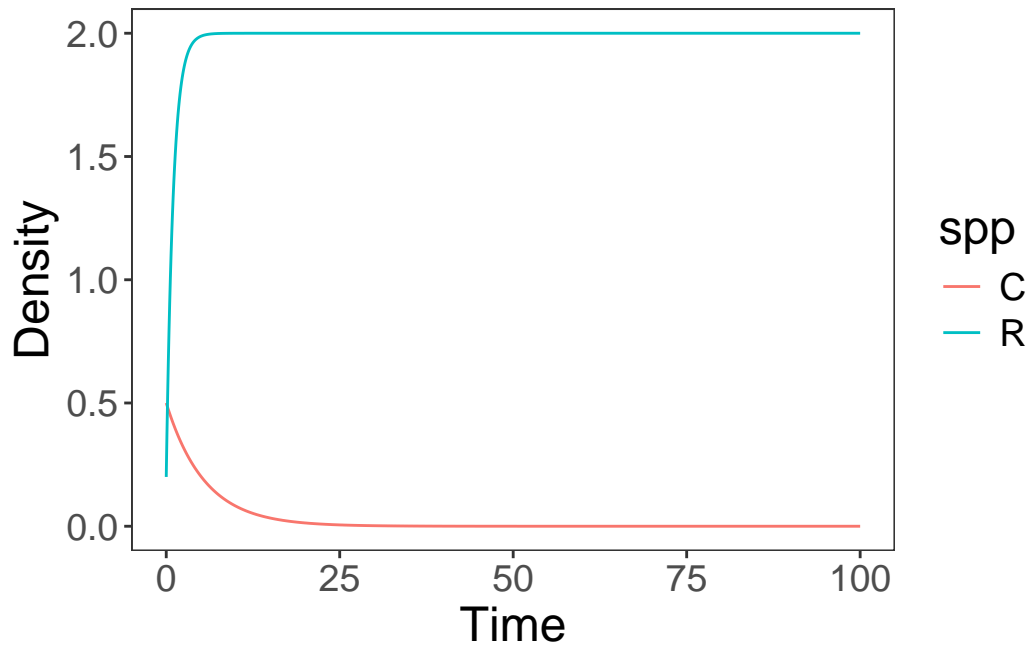
# Define initial conditions for R and C
state <- c(R = 0.2, C = 0.5)

# Set time points for the simulation
times <- seq(0, 100, by = 0.1)

# Run the simulation using deSolve's ode function
output <- data.frame(deSolve::ode(
  y = state, times = times,
  func = consumer_resource_model,
  parms = parameters
)) |>
  tidyr::pivot_longer(
    cols = c(R, C),
    names_to = "spp"
  )

# Plot the results
ggplot2::ggplot(
  data = output
) +
  geom_line(aes(x = time, y = value, colour = spp)) +
  theme_bw() +
  theme(
    text = element_text(size = 18),
    panel.grid.major = element_blank(),
    panel.grid.minor = element_blank(),
    strip.background = element_blank()
  ) +
  labs(x = "Time", y = "Density")

```



Source: [Article Notebook](#)

Now we do it drawing distributions and we can look at the equilibrium conditions:

```
consumer_equil_resource_model <- function(time, state, parameters) {
  R <- state[1]
  C <- state[2]
  T <- parameters["T"]
  D <- parameters["D"]
  S <- parameters["S"]
  delta <- parameters["delta"]
  m_a <- parameters["m_a"]
  m_b <- parameters["m_b"]
  m_c <- parameters["m_c"]
  a <- parameters["a"]
  h <- parameters["h"]
  gamma <- parameters["gamma"]
  T_I <- parameters["T_I"]
  R_half <- parameters["R_half"]

  # Functional response f(R, T)
  I_max_T <- exp(-(T - T_I)^2 * gamma)
  f_R_T <- I_max_T * R / (R + R_half)
```

```

# Respiration rate m(T)
m_T <- m_a * exp(m_b * T) + m_c

# Differential equations
dR_dt <- D * (S - R) - f_R_T * C
dC_dt <- (1 - delta) * f_R_T - m_T * C

return(list(c(dR_dt, dC_dt)))
}

# Define the parameter ranges (means and sds for S and T)
S_mean <- 1.5
S_sd <- 0.2
T_mean <- 20
T_sd <- 1

# Set other model parameters (fixed values)
parameters_base <- c(
  D = 0.1, # Outflow rate
  delta = 0.2, # Consumption efficiency
  m_a = 0.1,
  m_b = 0.05,
  m_c = 0.02,
  a = 0.5, # Attack rate
  h = 0.1, # Handling rate
  gamma = 0.05, # Breadth of response
  T_I = 20, # Optimum temperature for consumption
  R_half = 0.1 # Half-saturation density
)

# Define initial conditions for R and C
state <- c(R = 1, C = 0.5)

# Time points for the simulation
times <- seq(0, 500, by = 0.1) # Extended time for equilibrium

# Number of draws for S and T
num_draws <- 100

# Storage for equilibrium C values
equilibrium_C <- numeric(num_draws)
T_vals <- numeric(num_draws)

```

```
S_vals <- numeric(num_draws)
```

Source: [Article Notebook](#)

Vinton, Anna C., and David A. Vasseur. 2022. “Resource Limitation Determines Realized Thermal Performance of Consumers in Trophodynamic Models.” *Ecology Letters* 25 (10): 2142–55. <https://doi.org/10.1111/ele.14086>.