

Enemy-Victim Models

“Populační ekologie živočichů“

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Predator-prey system

Acarus



Cheyletus



Predator-prey model

- ▶ continuous model of Lotka & Volterra (1925-1928)
- continuous predation
- capture several prey items, functional response Type II

H .. density of prey

r .. intrinsic rate of prey population

a .. predation rate

P .. density of predators

m .. predator mortality rate

b .. reproduction rate of predators

- ▶ in the absence of predator, prey grows exponentially → $\frac{dH}{dt} = rH$
- ▶ in the absence of prey, predator dies exponentially → $\frac{dP}{dt} = -mP$
- ▶ predation rate is linear function of the number of prey .. aHP
- ▶ each prey contributes identically to the growth of predator .. bHP

$$\frac{dH}{dt} = rH - aHP$$

$$\frac{dP}{dt} = bHP - mP$$

Analysis of the model

Zero isoclines:

- ▶ for prey population:

$$\frac{dH}{dt} = 0 \quad 0 = rH - aHP$$

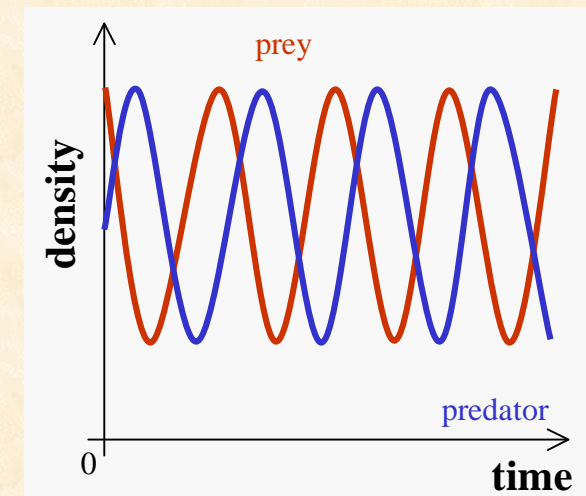
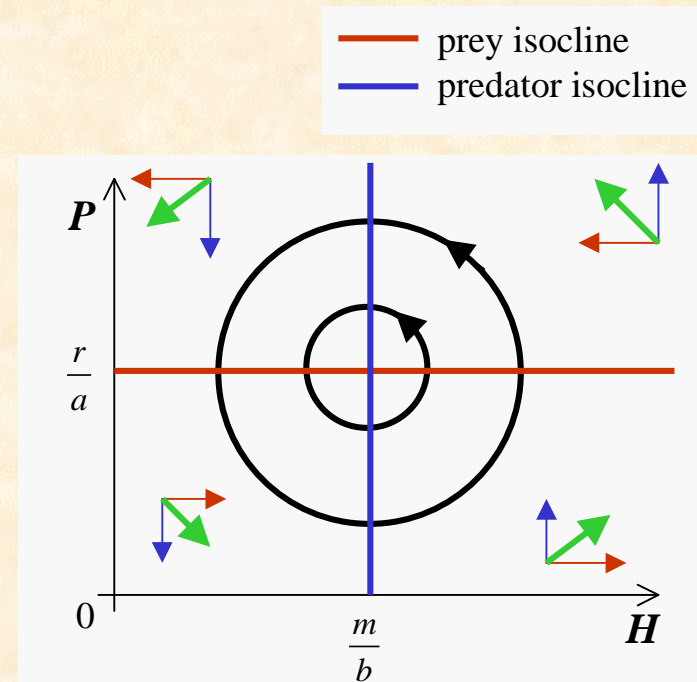
$$P = \frac{r}{a}$$

- ▶ for predator population:

$$\frac{dP}{dt} = 0 \quad 0 = bHP - mP$$

$$H = \frac{m}{b}$$

- ▶ prey population would grow to infinity
→ **neutral stability**
- ▶ do not converge, has no asymptotic stability (trajectories are closed lines)
- ▶ unstable system, amplitude of the cycles is determined by initial numbers
- ▶ POOR model



Incorporation of density-dependence

- ▶ in the absence of the predator prey population reaches carrying capacity K

$$\frac{dH}{dt} = rH \left(1 - \frac{H}{K} \right) - aHP$$

$$\frac{dP}{dt} = bHP - mP$$

- ▶ for given parameter values: $r = 3$, $m = 2$, $a = 0.1$, $b = 0.3$, $K = 10$

$$\frac{dH}{dt} = 3H \left(1 - \frac{H}{10} \right) - 0.1HP$$

$$\frac{dP}{dt} = 0.3HP - 2P$$

Zero isoclines:

▶ for prey population: $\frac{dH}{dt} = 0$ $0 = 3H\left(1 - \frac{H}{10}\right) - 0.1HP$

if $H = 0$ (trivial solution) or if $0 = 3\left(1 - \frac{H}{10}\right) - 0.1P$

$$P = 30 - 3H$$

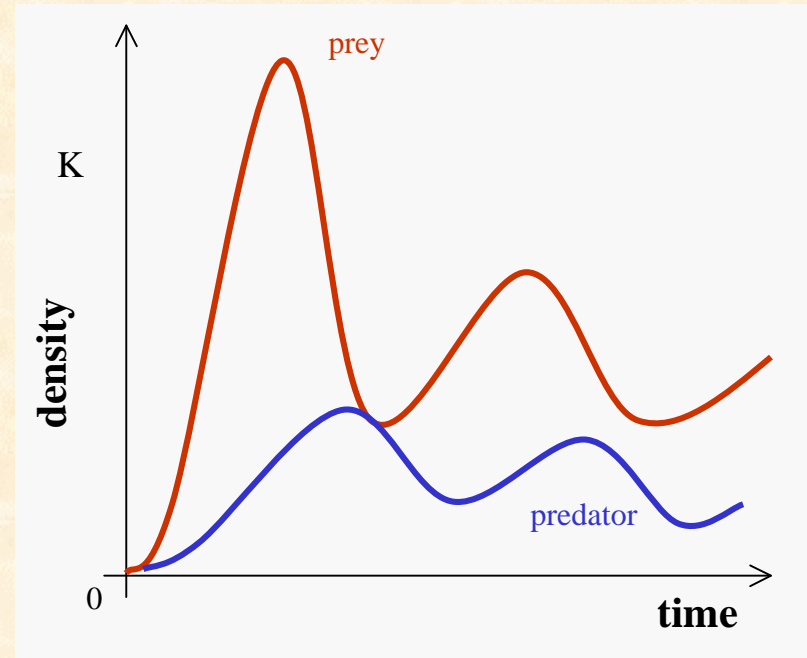
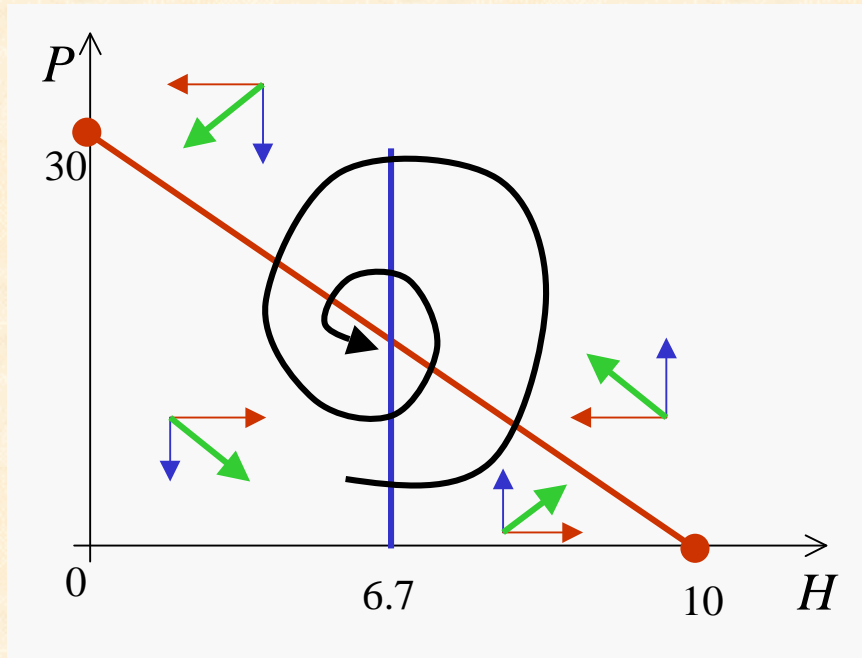
▶ for predator population: $\frac{dP}{dt} = 0$ $0.3HP - 2P = 0$

if $P = 0$ (trivial solution)
or if $0.3H - 2 = 0$

$$H = 6.667$$

▶ gradient of prey isocline is negative





.. approaches stable equilibrium

Incorporation of functional response

▶ functional response Type II: $H_a = \frac{aHT}{1 + aHT_h}$

▶ rate of consumption by all predators: $\frac{H_a P}{T} = \frac{aHP}{1 + aHT_h}$

$$\frac{dH}{dt} = r_H H \left(1 - \frac{H}{K} \right) - \frac{aHP}{1 + aHT_h} \quad \frac{dP}{dt} = bHP - mP$$

▶ for parameters: $r_H = 3, a = 0.1, T_h = 2, K = 10$

$$\frac{dH}{dt} = 0 \quad 0 = 3H \left(1 - \frac{H}{10} \right) - \frac{0.1HP}{1 + 0.1H \cdot 2}$$

prey isocline:

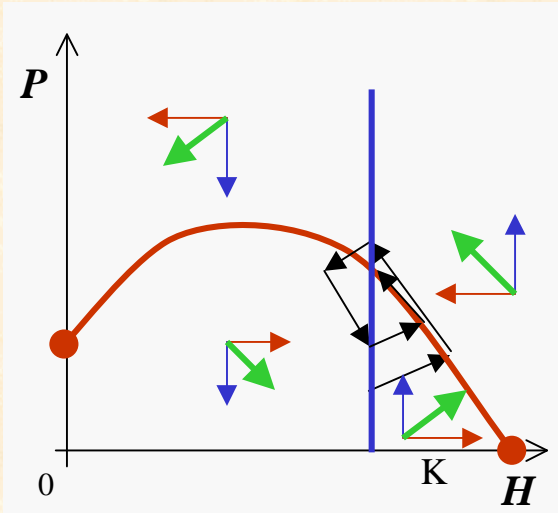
$$P = 30 + 6H - 0.6H^2$$

predator isocline:

$$H = \text{constant}$$

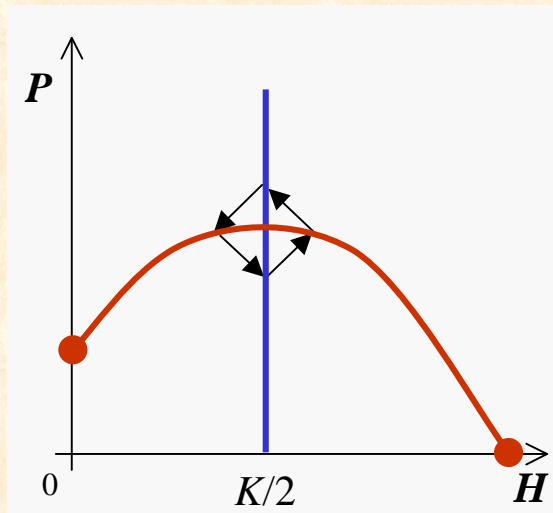
▶ predator exploits prey close to K

- isocline: $H = 9$



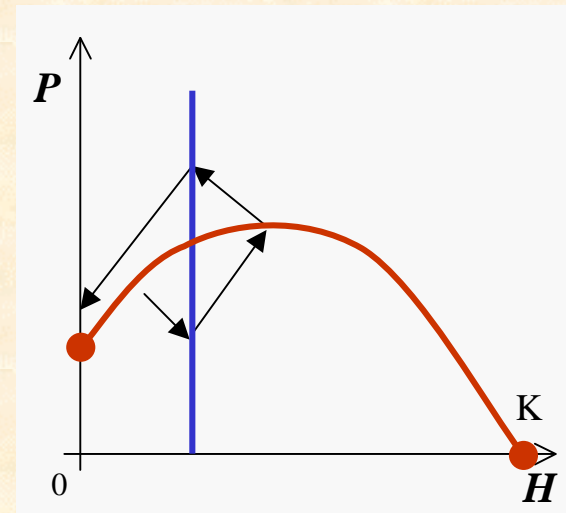
▶ predator exploits prey close to $K/2$

- isocline: $H = 5$

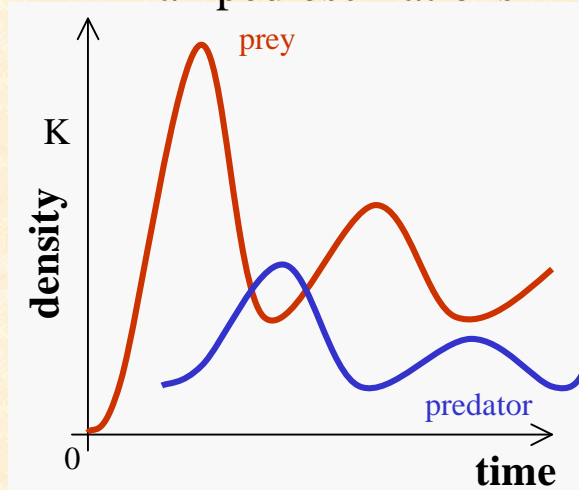


▶ predator exploits prey at low density

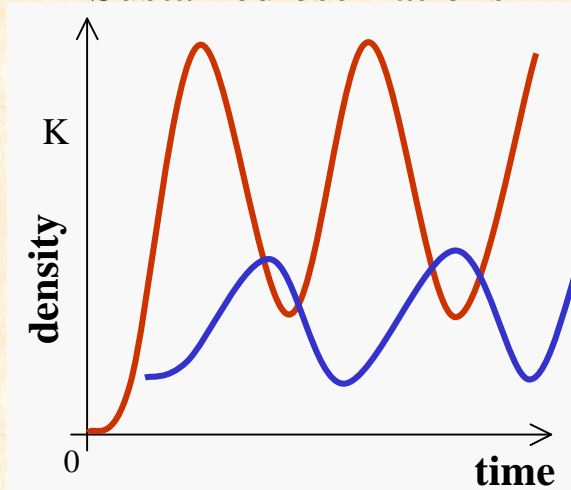
- isocline: $H = 2$



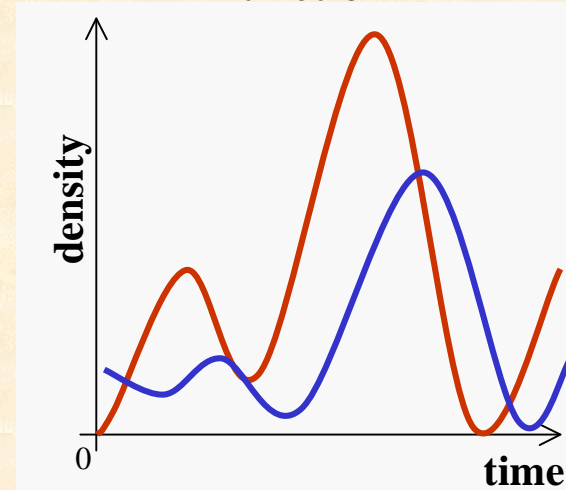
Damped oscillations



Sustained oscillations



Extinction



Incorporation of predator's carrying capacity

- ▶ logistic model with carrying capacity proportional to H
- ▶ k .. carrying capacity of the predator
- ▶ $r_p = bH - m$

$$\frac{dP}{dt} = bHP - mP$$

$$\frac{dP}{dt} = r_p P \left(1 - \frac{P}{kH} \right) \quad \frac{dH}{dt} = r_H H \left(1 - \frac{H}{K} \right) - \frac{aHP}{1 + aHT_h}$$

- ▶ for parameters: $r_p = 2, k = 0.2$

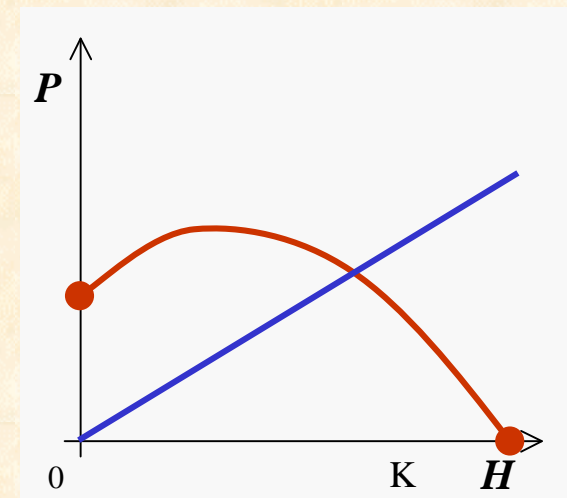
$$\frac{dP}{dt} = 0 \quad 0 = 2P \left(1 - \frac{P}{0.2H} \right)$$

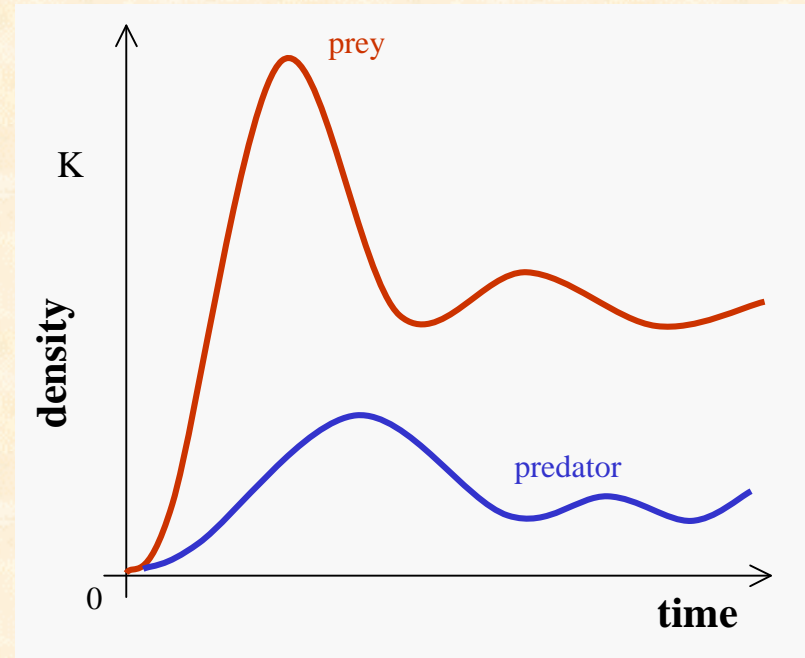
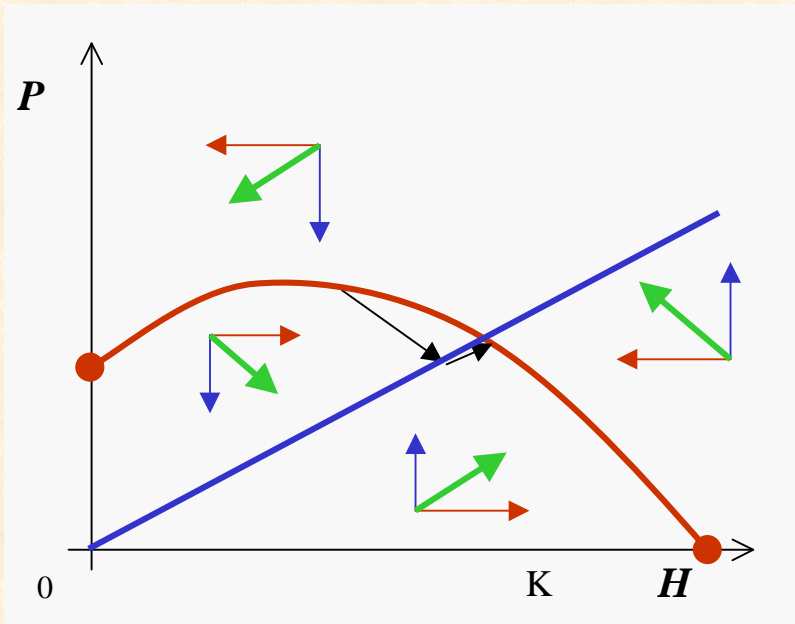
predator isocline:

$$H = 5P$$

prey isocline:

$$P = 30 + 6H - 0.6H^2$$





.. quick approach to stable equilibrium

Host-parasitoid system

Zatypota



Theridion



Host-parasitoid model

- ▶ discrete model of Nicholson & Bailey (1935)
- discrete generations
- 1, .., several, or less than 1 host
- random host search and functional response Type III
- lay eggs in aggregation

H_t = number of hosts in time t

H_a = number of attacked hosts

λ = finite rate of increase of the host

P_t = number of parasitoids

c = conversion rate, no. of parasitoids for 1 host

$$H_{t+1} = \lambda(H_t - H_a)$$

$$P_{t+1} = cH_a = H_a$$

Incorporation of random search

- ▶ parasitoid searches randomly
- ▶ encounters (x) are random (Poisson distribution)

$$p_x = \frac{\mu^x e^{-\mu}}{x!} \quad x = 0, 1, 2, \dots \quad p_0 = e^{-\mu}$$

p_0 = proportion of not encountered, μ .. mean number of encounters

E_t = total number of encounters

a = searching efficiency (proportion of hosts encountered)

$$E_t = a H_t P_t \longrightarrow \mu = \frac{E_t}{H_t} = a P_t \longrightarrow p_0 = e^{-a P_t}$$

- ▶ proportion of encounters (1 or more times): $p = (1 - p_0)$

$$p = (1 - e^{-a P_t})$$

$$H_a = H_t (1 - e^{-a P_t})$$

$$H_{t+1} = \lambda(H_t - H_a)$$

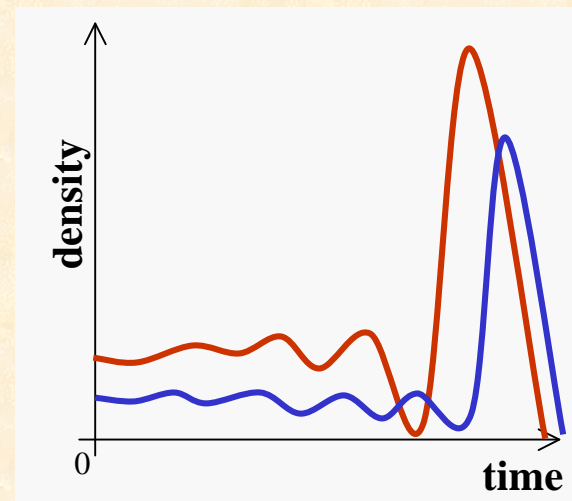
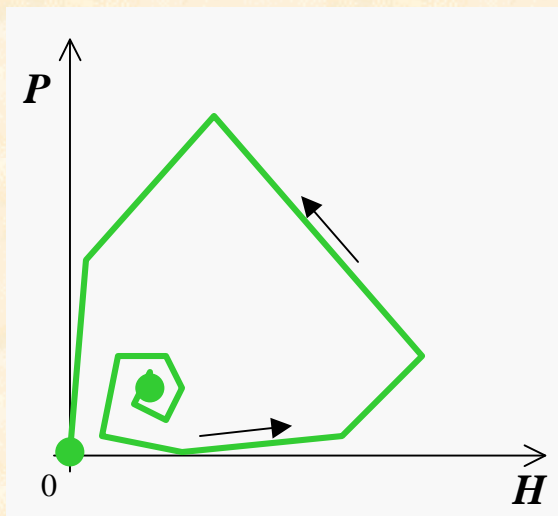
$$P_{t+1} = H_a$$



$$H_{t+1} = \lambda H_t e^{-aP_t}$$

$$P_{t+1} = H_t (1 - e^{-aP_t})$$

- ▶ highly unstable model for all parameter values:
 - equilibrium is possible but the slightest disturbance leads to divergent oscillations (extinction of parasitoid)



Incorporation of density-dependence

- ▶ exponential growth of hosts is replaced by logistic equation

$$H_{t+1} = \lambda H_t e^{\left(1 - \frac{H_t}{K}\right) - aP_t}$$

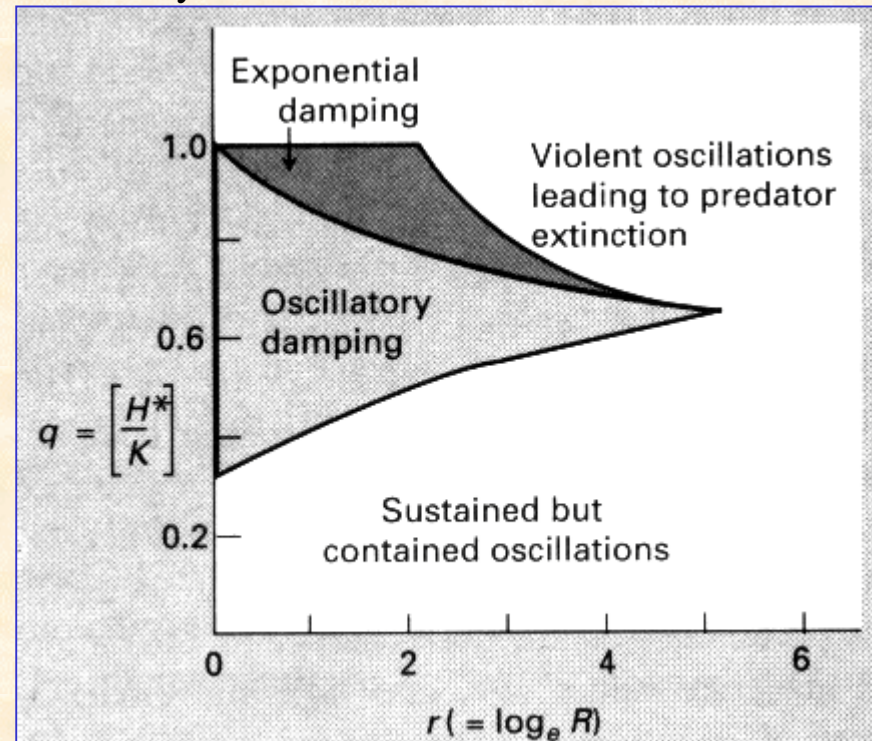
$$P_{t+1} = H_t \left(1 - e^{-aP_t}\right)$$

$$q = \frac{H^*}{K}$$

H^* .. new host carrying capacity

- ▶ depends on parasitoids' efficiency
 - when a is low then $q \rightarrow 1$
 - when a is high then $q \rightarrow 0$
- ▶ density-dependence have stabilising effect for moderate r and q

Stability boundaries



Incorporation of the refuge

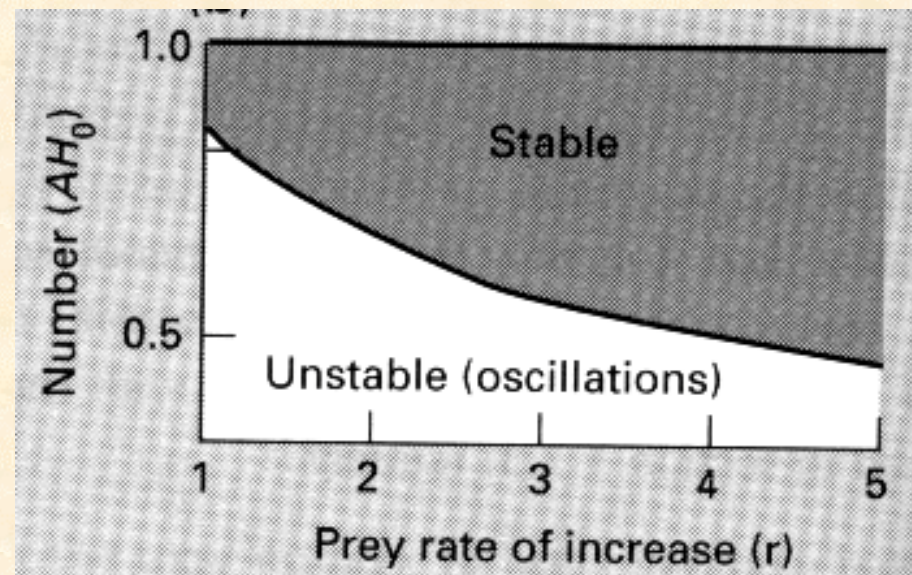
- ▶ if hosts are distributed non-randomly in the space

Fixed number in refuge: H_0 hosts are always protected

$$H_{t+1} = \lambda H_0 + \lambda(H_t - H_0)e^{-aP_t}$$

$$P_{t+1} = (H_t - H_0)(1 - e^{-aP_t})$$

- ▶ have strong stabilising effect even for large r



Incorporation of aggregated distribution

► distribution of encounters is not random but aggregated (negative binomial distribution)

- proportion of hosts not encountered (p_0):
$$p_0 = \left(1 + \frac{aP_t}{k}\right)^{-k}$$

where k = degree of aggregation

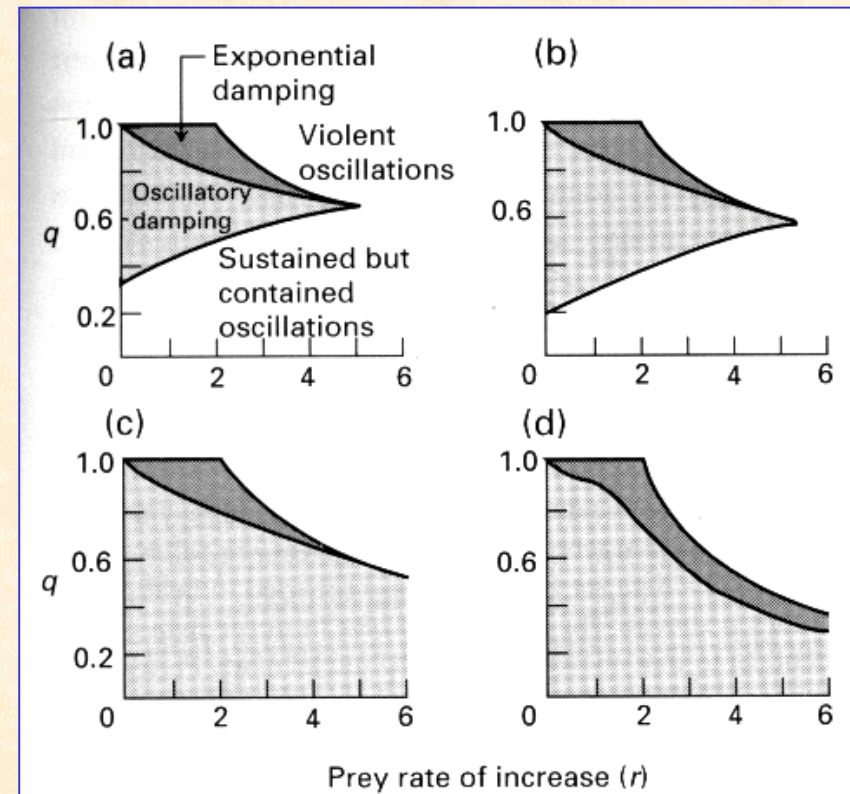
$$H_{t+1} = \lambda H_t e^{\left(1 - \frac{H_t}{K}\right) \left(1 + \frac{aP_t}{k}\right)^{-k}}$$

$$P_{t+1} = H_t \left(1 - \left(1 + \frac{aP_t}{k}\right)^{-k}\right)$$

► very stable model system if $k \leq 1$

Stability boundaries:

a) $k=\infty$, b) $k=2$, c) $k=1$, d) $k=0$



Example 21

You want to control population of mites. Before introduction of predatory mites you want to simulate the predator-prey dynamic using the following model:

$$\frac{dH}{dt} = r_H H \left(1 - \frac{H}{K} \right) - \frac{aHP}{1 + aHT_h} \quad \frac{dP}{dt} = \frac{acHP}{1 + aHT_h} - dP$$

Parameter estimates are obtained experimentally:

1. Rear prey population without predators. You find $r_H = 0.4$ and $K = 500$.
2. Rear predators at constant prey densities. You find predators' mortality $d = 0.08$ and conversion efficiency $c = 0.8$.
3. Perform functional response experiment. You find that $a = 0.001$ and $T_h = 0.5$.

How long it takes for the predatory mite to control mite pests. Initial densities are 200 individuals of pests and 10 individuals of predators?

```
predprey<-function(t,y,pa){  
H<-y[1]  
P<-y[2]  
with(as.list(pa),{  
dH.dt<-rH*H*(1-H/K)-a*H*P/(1+a*H*Th)  
dP.dt<-a*c*H*P/(1+a*H*Th)-d*P  
return(list(c(dH.dt,dP.dt)))})}
```

```
H<-200;P<-10  
time<-seq(0,200,0.1)  
pa<-c(rH=0.4,K=500,a=0.001,Th=0.5,c=0.8,d=0.08)  
library(deSolve)  
out<-data.frame(ode(c(H,P),time,predprey,pa))  
matplot(time,out[,-1],type="l",lty=1:2,col=1)  
legend("right",c("H","P"),lty=1:2)
```

Example 22

Caterpillars increased their population density in flour to 50 individuals/100 kg. You observed that their $\lambda = 3$ and $K = 800$. You need to control these pests using a parasitoid. You can choose from three parasitoid species (A, B, C). The three species differ in the number of eggs/host (c) and in their search efficiency (a):

	A	B	C
c	1	3	2
a	0.003	0.1	0.005

1. Use the discrete Nicholson-Bailey host-parasitoid model with density-dependence. Introduce a single parasitoid per 100 kg. Find which of the three species will achieve the quickest control.

```
time<-20
HP<-data.frame(H<-numeric(time),P<-numeric(time))
a=0.003;L=3;K=800;c=1
HP[1,]<-c(50,1)
for (t in 1:20) HP[t+1,]<-{
H<-L*HP[t,1]*exp((K-HP[t,1])/K-a*HP[t,2])
P<-c*HP[t,1]*(1-exp(-a*HP[t,2]))
c(H,P)}
matplot(HP,type="l",lty=1:2)
legend(10,2000,c("H","P"),lty=1:2)
```

```
HP<-data.frame(H<-numeric(time),P<-numeric(time))
a=0.1;L=3;K=800;c=3
HP[1,]<-c(50,1)
for (t in 1:20) HP[t+1,]<-{
H<-L*HP[t,1]*exp((K-HP[t,1])/K-a*HP[t,2])
P<-c*HP[t,1]*(1-exp(-a*HP[t,2]))
c(H,P)}
matplot(HP,type="l",lty=1:2)
legend(10,2000,c("H","P"),lty=1:2)
```

```
HP<-data.frame(H<-numeric(time),P<-numeric(time))
a=0.005;L=3;K=500;c=2
HP[1,]<-c(50,1)
for (t in 1:20) HP[t+1,]<-{
H<-L*HP[t,1]*exp((K-HP[t,1])/K-a*HP[t,2])
P<-c*HP[t,1]*(1-exp(-a*HP[t,2]))
c(H,P)}
matplot(HP,type="l",lty=1:2)
legend(10,4000,c("H","P"),lty=1:2)
```