

2021. 10. 12 Density-dependence & Logistic population growth

$$\frac{dN}{dt} = [b(N) - d(N)]N \quad \left\{ \begin{array}{l} \text{model assumptions \& derivation} \\ \text{equilibrium \& graphical analysis} \\ \text{Local stability analysis} \end{array} \right.$$

② Model derivation

$$\frac{dN}{dt} = B - D + I - E$$

③ Assumptions

- { 1. closed population $I = E = 0$, ($\frac{dN}{dt} = 0$)
- 2. Identical individuals w/o internal structure
- 3. Continuous population growth w/o time lag
- 4. time independent BUT density dependent per capita rates;

$$= [b(N) - d(N)]N \xrightarrow{\text{...}} r(N) \cdot N$$

- { 5. Limited resource, so negative effect of density on growth rate ($\frac{\partial r(N)}{\partial N} < 0$)
- 6. Linear density dependence: $b(N) = b_0 - b_N \cdot N$

$$d(N) = d_0 + d_N \cdot N$$

→ time independence

$$= [b_0 - b_N \cdot N - d_0 - d_N \cdot N]N$$

$$= \left[\underbrace{(b_0 - d_0)}_{r_0} - \underbrace{(b_N + d_N)N}_{\alpha} \right] N$$

$$\Rightarrow \frac{dN}{dt} = [r_0 - \alpha \cdot N]N \quad \xrightarrow{\text{logistic population growth model}}$$

↓ intrinsic growth rate → strength of density dependence ($\alpha > 0$)

($r_0 > 0$ when $b_0 > d_0$)

[Pierre Francois Verhulst 1838]

[Raymond Pearl & Lowell Read 1920]

$$\Rightarrow \frac{dN}{dt} = r_0 N \left(1 - \frac{N}{K}\right) \dots \textcircled{2}$$

carrying capacity, maximum pop. size that can be supported

$$K_r = \frac{r_0}{\alpha}$$

To understand the dynamics

② Integrate the logistic equation

$$\frac{dN}{dt} = r_0 N \left(1 - \frac{N}{K}\right)$$

$$\Rightarrow \frac{1}{N} \times \frac{K}{K-N} dN = r_0 dt$$

$$\Rightarrow \int \left[\frac{1}{N} - \left(\frac{-1}{K-N} \right) \right] dN = \int r_0 dt$$

$$\Rightarrow \int \frac{1}{N} dN - \int \frac{1}{N-K} dN = \int r_0 dt$$

$$\Rightarrow \ln N - \ln(N-K) = \ln \left(\frac{N}{N-K} \right) = r_0 t + C_1$$

$$\Rightarrow \frac{N}{N-K} = C_2 e^{r_0 t}$$

$$\Rightarrow N(t) = \frac{K \cdot C_2 e^{r_0 t}}{C_2 e^{r_0 t} - 1} = \frac{K}{1 - \left(\frac{1}{C_2} e^{-r_0 t}\right)}$$

$$\text{at } t=0, N(t=0) = N_0 = \frac{K C_2}{C_2 - 1} \Rightarrow C_2 = \frac{N_0}{N_0 - K}$$

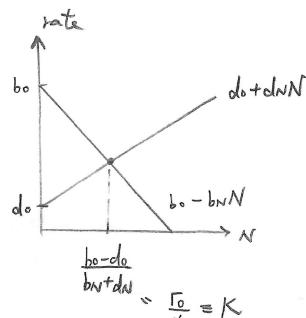
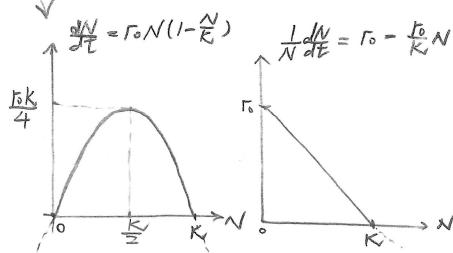
$$\Rightarrow N(t) = \frac{K}{1 - \left(\frac{N_0 - K}{N_0}\right) e^{-r_0 t}} \dots \textcircled{4}$$

$$\lim_{t \rightarrow \infty} N(t) = K \Rightarrow \text{Final fate} \dots \textcircled{5}$$

* Recall: Separation of variables

$$\frac{dN}{dt} = f(N) \cdot g(t)$$

$$\int \frac{1}{f(N)} dN = \int g(t) dt$$

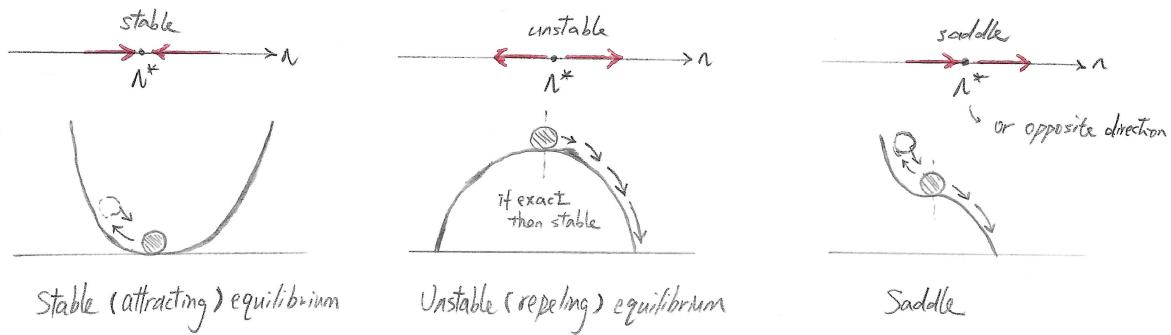


$$\frac{b_0 - b_1 N}{b_1 N + dN} = \frac{r_0}{K} = K$$

- Ways to understand final fate / longterm dynamics
- 1. Integrate $\frac{dN}{dt}$
 - 2. Graphical approach
 - 3. Stability analysis

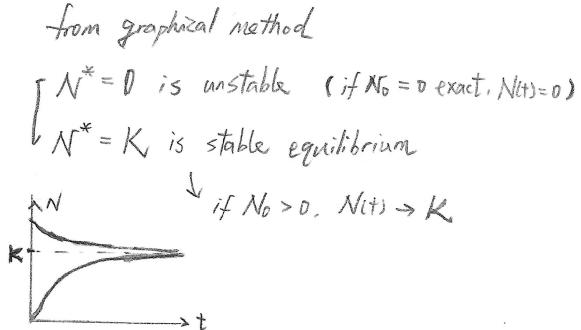
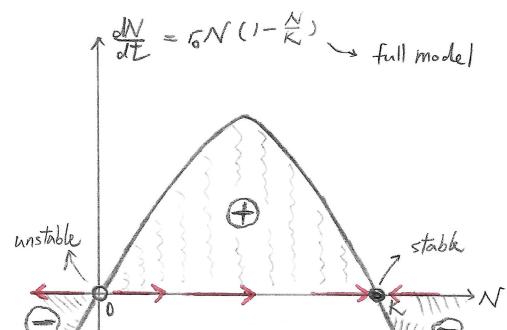
② Equilibrium (fixed point)

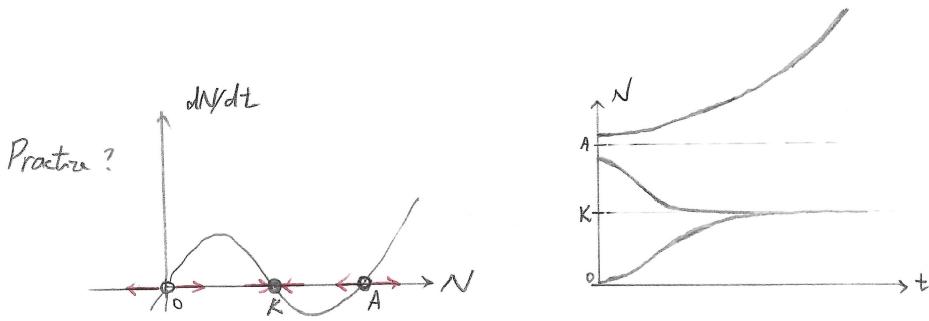
- A point where the variable does not further change
these values are good candidates where the variable will end up
- consider $\frac{dN}{dt} = f(N)$, a point N^* is an equilibrium of $\frac{dN}{dt} = f(N)$
if $f(N^*) = 0$, that is $\frac{dN}{dt}|_{N=N^*} = f(N^*) = 0$
 - full model, population growth, not per capita growth
 - do logistic example here
- Equilibrium (equilibria) can be classified based on their stability
 1. Attracting (stable) equilibrium, attracts nearby trajectories
 2. Repelling (unstable) equilibrium, repels nearby trajectories
 3. Saddle equilibrium, attract trajectories on one side but repels on the other



③ Graphical approach plot $\frac{dN}{dt}$ versus N

$$\frac{dN}{dt} = r_0 N \left(1 - \frac{N}{K}\right) \xrightarrow{\text{find equilibrium}} r_0 N^* \left(1 - \frac{N^*}{K}\right) = 0 \Rightarrow \text{equilibrium } N^* = 0 \vee K \quad \text{④}$$





② Local stability analysis

near the equilibrium N^* , do Taylor expansion of $\frac{dN}{dt} = f(N)$ about N^*

think about "displacements" from N^* , $\varepsilon = N - N^*$

think how ε changes through time (study behavior of nearby points)

$$\begin{aligned}\frac{d\varepsilon}{dt} &= \frac{d(N-N^*)}{dt} = \frac{dN}{dt} - \underbrace{\frac{dN^*}{dt}}_0 = f(N) \xrightarrow{\text{not function of } \varepsilon} \\ &= f(N^* + \varepsilon) \quad \xrightarrow{\text{Recall: Taylor expansion}} \quad f(a+\Delta x) = f(a) + \Delta x f'(a) + \frac{\Delta x^2}{2!} f''(a) \dots\end{aligned}$$

under small ε , near neighbors

$$\begin{aligned}&= f(N^*) + \varepsilon \cdot \frac{df}{dN} \Big|_{N=N^*} + O(\varepsilon^2) \\ \Rightarrow \frac{d\varepsilon}{dt} &= \left[\frac{df}{dN} \Big|_{N=N^*} \right] \varepsilon \quad \dots \textcircled{D} \\ &\quad \downarrow \text{define as } \lambda \text{ (constant)}\end{aligned}$$

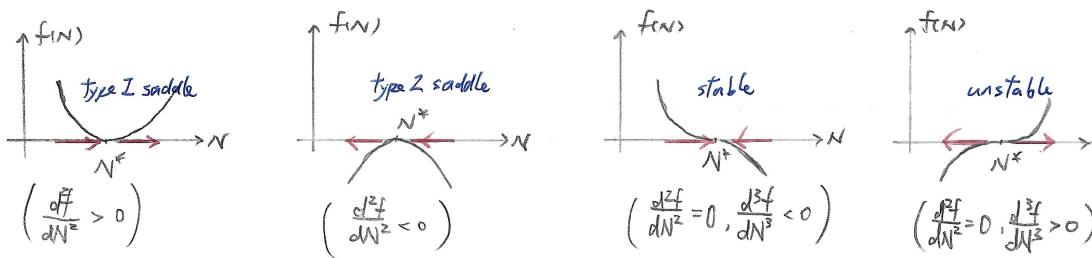
$$\Rightarrow \varepsilon(t) = \varepsilon_0 \cdot e^{\lambda t} \quad \text{with } \lambda = \frac{df}{dN} \Big|_{N=N^*} \quad \dots \textcircled{D}$$

1. if $\lambda > 0$, displacements amplify, N^* is unstable

2. if $\lambda < 0$, displacements shrink, N^* is stable

(3. if $\lambda = 0$, linear term is insufficient & higher order terms are required)

nonhyperbolic equilibrium



② Link between graphical approach & stability analysis

for logistic growth: $\frac{dN}{dT} = r_0 N \left(1 - \frac{N}{K}\right) = f(N)$

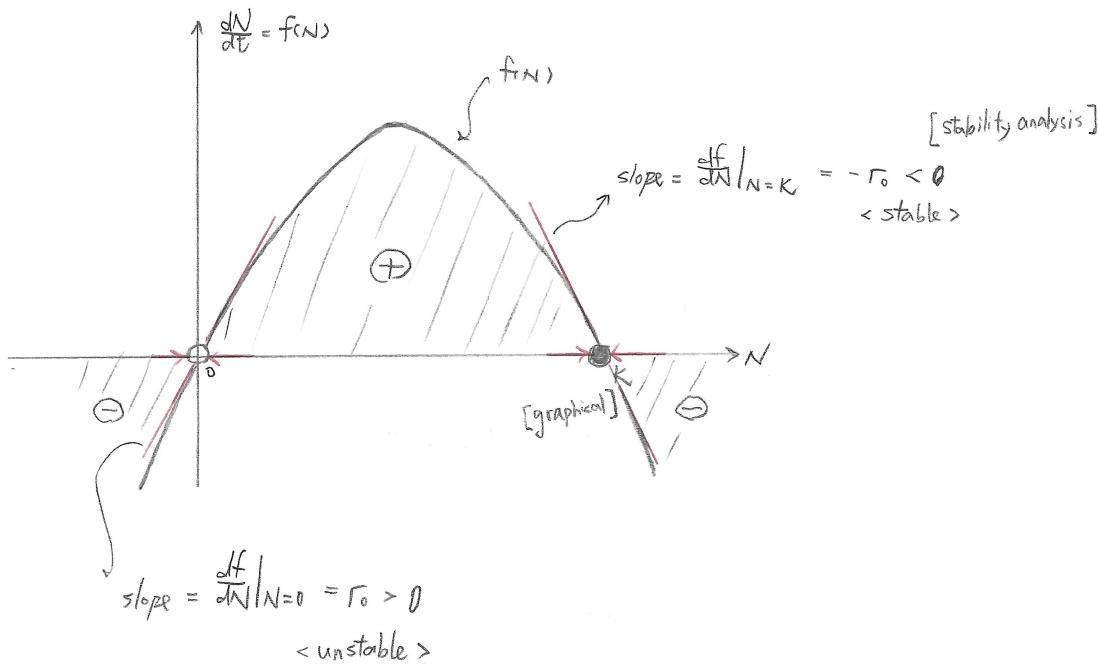
$$\frac{df}{dN} = r_0 - \frac{2r_0}{K} \cdot N = f'(N) \quad \text{...①}$$

Trick: don't expand everything

$$\frac{df}{dN} = r_0 \left(1 - \frac{N}{K}\right) + r_0 N \left(-\frac{1}{K}\right)$$

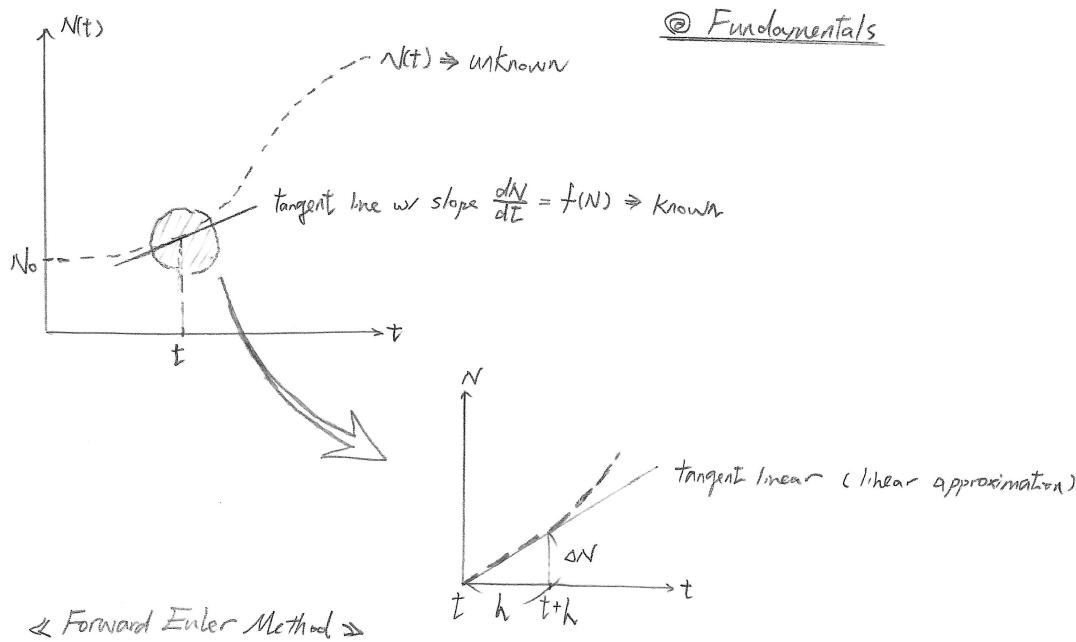
① for $N^* = 0$, $\frac{df}{dN}|_{N=0} = r_0 > 0 \rightsquigarrow \text{unstable}$

② for $N^* = K$, $\frac{df}{dN}|_{N=K} = -r_0 < 0 \rightsquigarrow \text{stable}$



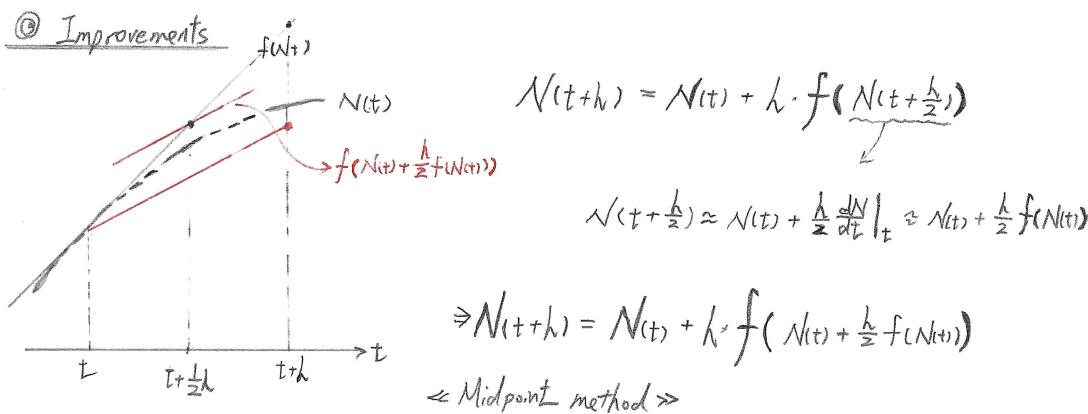
③ Final Note: Parameter constraint is sometimes important

2021.10.12. Basic Intro to Numerical Methods



$$\begin{aligned}
 N(t+h) &= N(t) + h \cdot \text{slope of tangent line at time point } t \\
 &= N(t) + h \cdot \underline{f(N(t))} \\
 &\quad \downarrow \quad \text{your ODE, provide this to solve} \\
 &\text{Recall: Taylor expansion} \\
 N(t+h) &= N(t) + h \cdot \frac{dN}{dt} \Big|_{N=N_t} + O(h^2)
 \end{aligned}$$

only works when h is small enough (but this is costly)



Commonly used algorithm: 4th order Runge-Kutta methods ... → weighted average of slopes
 Multistep methods, implicit methods, adaptive timesteps.