



Ordinary Differential Equations

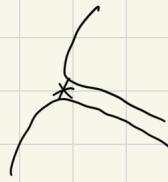
Examples

1. A, B

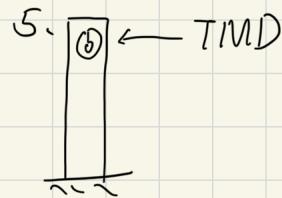
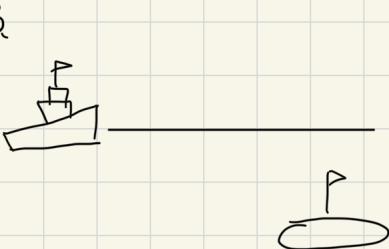
Newton's Cooling Law

4. 1940. 8. XX

Tacoma Narrows Bridge



2. Baseball



First Order Equations

$$t: y, \quad y' = \frac{dy}{dt}$$

$$F(t, y, y') = 0$$

↳ $y' = f(t, y)$ at some neighbourhood of (t_0, y_0)

(This is from the implicit function theorem)

We say the equation is linear if f is linear for y

$$y' + a(t)y = b(t)$$

(i) $b = 0$, then $\frac{y'}{y} = -a(t)$, $\ln|y| = -\int a(t)dt + C$, $|y| = C e^{-\int a(t)dt}$, $C \geq 0$

(ii) $b \neq 0$, then consider the variation of parameters

Let $y = c(t)e^{-\int a(t)dt}$

$$y' + a(t)y = c'(t)e^{-\int a(t)dt} - a(t)c(t)e^{-\int a(t)dt} + a(t)c(t)e^{-\int a(t)dt} = b(t)$$

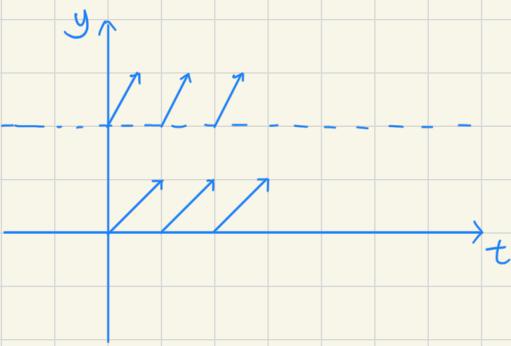
$$\text{so } c'(t)e^{-\int a(t)dt} = b(t), \quad y = e^{-\int a(t)dt} \cdot \int b(t)e^{\int a(t)dt} dt$$

Example. $y' = y^2 + 1$.

$$\frac{dy}{dt} = y^2 + 1, \quad \frac{dy}{y^2+1} = dt, \quad \arctan y = t + C$$

$$y = \tan(t + C)$$

if $y(0) = 0$, then $y = \tan t$, $-\frac{\pi}{2} < t < \frac{\pi}{2}$



mathus.

$P(t)$: population of a species at time t

$$\frac{dp}{dt} = r \cdot p \quad r: \text{birth rate - death rate}$$

Venhuist (1837)

Logistic equation.

$$\frac{dp}{dt} = rp\left(1 - \frac{p}{k}\right)$$

k : carry capacity

$$\frac{dp}{dt} = p(a - bp), \quad a, b > 0$$

$$\frac{dp}{p(a-bp)} = dt$$

$$\frac{b}{a} \int \frac{1}{bp-a} + \frac{1}{bp} dp = t$$

$$t = \frac{1}{a} \ln \frac{p}{|a-bp|} + C, \quad P(0) = P_0$$

$$|a-bp| = C \cdot e^{at}, \quad \text{so} \quad C = \frac{P_0}{|a-bP_0|}$$

$$\text{if } a-bP_0 = 0, \text{ then } P(t) \equiv \frac{a}{b}$$

$$\text{if } a-bP_0 > 0, \quad \frac{P}{a-bp} = C \cdot e^{at}$$

$$P = \frac{aCe^{at}}{1+bCe^{at}} = \frac{a}{b+\frac{1}{C}e^{at}} = \frac{aP_0}{bP_0 + (a-bP_0)e^{-at}}$$

Example. A commercial fisherman is estimated to have carrying capacity 10000kg of a certain kind of fish. Suppose the annual growth of the total fish P is governed by the equation $\frac{dp}{dt} = p(1 - \frac{P}{10000})$, and initially there are 2000kg of fish.

(a) What is the fish population after one year?

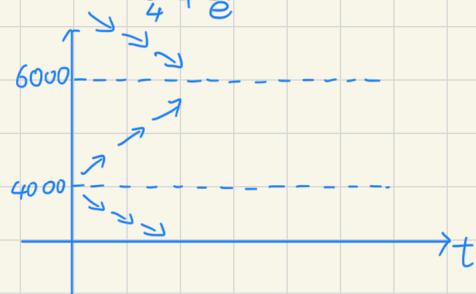
Suppose after waiting for a certain period of time, the owner of the fishery decides to harvest 2400kg of fish annually at a constant rate. Then

(b) What is the optimal waiting period?

$$a) P(t) = \frac{C}{\frac{C}{10000} + e^{-at}}$$

$$C = \frac{P_0}{|a - bp_0|} = \frac{2000}{1 - \frac{2000}{10000}} = 2500$$

$$P(t) = \frac{2500}{\frac{1}{4} + e^{-t}} \approx 4046 \text{ kg}$$

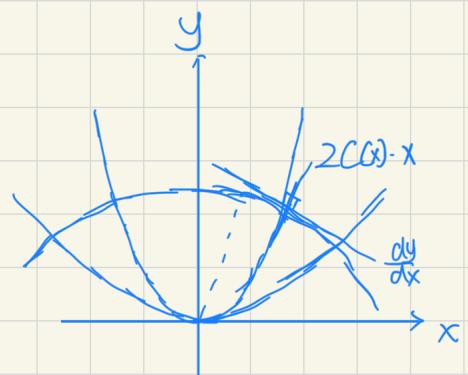


(2)

$$\frac{dP}{dt} = P(1 - \frac{P}{10000}) - 2400$$

$$\frac{dP}{dt} = -\frac{1}{10000}(P-6000)(P-4000)$$

Here we only require to harvest 2400kg of fish within a year, rather than harvest them in a single moment.



Example: Given a family of curves $y = Cx^2$, $C \in \mathbb{R}$

Find the orthogonal trajectories

$$\frac{dy}{dx} \cdot 2Cx \cdot x = -1, \quad y = C(x) \cdot x^2$$

$$2 \frac{dy}{dx} \frac{y}{x^2} \cdot x = -1, \quad 2y \, dy = -x \, dx, \quad y^2 = -\frac{1}{2}x^2 + C, \quad C > 0$$

$$\Leftrightarrow \frac{x^2}{2} + y^2 = C, \quad \text{WLOG } C=1, \quad \frac{x^2}{2} + y^2 = 1$$

we see that it's an ellipse

Newton's Cooling Law

$T(t)$: the temperature of a body immersed in a medium.

A : the temperature of the medium

The temperature changing rate is proportional to $A - T$

$$\frac{dT}{dt} = k(A - T), k > 0$$

$$\text{then } \frac{dT}{T-A} = -kdt, \ln|T-A| = -kt + C \quad |T-A| = Ce^{-kt}, C > 0$$

$$\text{if } T_0 > A, \text{ then } T = (T_0 - A)e^{-kt} + A$$

$$\text{if } T_0 < A, \text{ then } T = A - (A - T_0)e^{-kt} = (T_0 - A)e^{-kt} + A$$

$$\text{if } T_0 = A, \text{ then } T \equiv A$$

Need rigorous proof

You can actually find many examples in PHY1001

Thm. (Uniqueness) Let $\begin{cases} y' = f(y, t) \\ y(t_0) = y_0 \end{cases}$ be an initial value problem,

Suppose f & $\frac{\partial f}{\partial y}$ are continuous on a compact neighborhood of (t_0, y_0)
Then the (IVP) (*) has at most one solution

pf: Suppose we have two solution y_1, y_2

$$\text{i.e. } \frac{dy}{dt} = f(y, t), y(t_0) = y_0$$

$$y_1(t) = y_0 + \int_{t_0}^t f(y_1(s), s) ds \quad y_2(t) = y_0 + \int_{t_0}^t f(y_2(s), s) ds$$

$$|y_1(t) - y_2(t)| = \left| \int_{t_0}^t f(y_1(s), s) - f(y_2(s), s) ds \right| \leq \int_{t_0}^t |f(y_1(s), s) - f(y_2(s), s)| ds \\ = \int_{t_0}^t \left| \frac{\partial f}{\partial y}(s, \tilde{y}(s)) \right| |y_1(s) - y_2(s)| ds \leq M \int_{t_0}^t |y_1(s) - y_2(s)| ds$$

$$\text{let } w(t) = |y_1(t) - y_2(t)| \Rightarrow w(t) \leq M \underbrace{\int_{t_0}^t w(s) ds}_{z(t)}$$

$$\text{so } z'(t) \leq Mz(t) \quad \frac{d}{dt}(z(t)e^{-Mt}) = e^{-Mt}(z'(t) - Mz(t)) \leq 0$$

$$z(t_0) = 0, z(t)e^{-Mt} \geq 0, \text{ so } z(t) \equiv 0$$

therefore we have $y_1(t) \equiv y_2(t)$

$$(IVP) \begin{cases} y' = f(t, y) \\ y(t_0) = y_0 \end{cases}$$

On $R = [t_0 - a, t_0 + a] \times [y_0 - b, y_0 + b]$, $M = \max_{(t,y) \in R} |f(t, y)|$

$L = \max_{R} \left| \frac{\partial}{\partial y} f(t, y) \right|$. (H) $f, \frac{\partial f}{\partial y}$ are continuous on R

Thm. (IVP) has a solution provided that (H) holds.
(i.e. $\exists d > 0$, s.t. (IVP) has a solution for $(t_0 - \alpha, t_0 + \alpha)$)

proof: $\boxed{y(t) = y_0 + \int_{t_0}^t f(s, y(s)) ds} \quad (*)$ suffice to solve $(*)$

$$y_0(t) \equiv y_0, \quad y_1(t) \equiv y_0 + \int_{t_0}^t f(s, y_0(s)) ds$$

$$\text{Set } y_2(t) = y_0 + \int_{t_0}^t f(s, y_1(s)) ds$$

...

$$y_{k+1}(t) = y_0 + \int_{t_0}^t f(s, y_k(s)) ds$$

...

(Picard iteration) Claim. $\{y_k(t)\}_{k=0}^{\infty}$ converges on $(t_0 - \alpha, t_0 + \alpha)$ for some $\alpha > 0$

$$y_{k+1}(t) = y_0(t) + (y_1(t) - y_0(t)) + \dots + (y_k(t) - y_{k-1}(t)) + (y_{k+1}(t) - y_k(t)) \quad (\text{Telescope series})$$

WTS. $\sum_{n=1}^{\infty} (y_n(t) - y_{n-1}(t))$ converges uniformly on $(t_0 - \alpha, t_0 + \alpha)$

$$\begin{aligned} |y_{n+1}(t) - y_n(t)| &= \left| \int_{t_0}^t f(s, y_n(s)) - f(s, y_{n-1}(s)) ds \right| \leq \int_{t_0}^t |f(s, y_n(s)) - f(s, y_{n-1}(s))| ds \\ &= \int_{t_0}^t \left| \frac{\partial f}{\partial y}(s, \tilde{y}) \right| |y_n(s) - y_{n-1}(s)| ds \end{aligned}$$

$$|y_1(t) - y_0| \leq \int_{t_0}^t |f(s, y_0)| ds \leq M |t - t_0| \leq M \alpha \leq b$$

Here we take $\alpha = \min\{a, \frac{b}{M}\}$, so in this case $(t, y_1(t)) \in R$

Suppose $y_n(t)$ is defined on $B_\alpha(t_0)$ for $n \leq k-1$, then when $n=k$

$$|y_k(t) - y_0| \leq \int_{t_0}^t |f(s, y_{k-1}(s))| ds \leq M |t - t_0| \leq b. \text{ so } \forall n, y_n(t) \text{ is defined on } B_\alpha(t_0)$$

$$|y_2(t) - y_1(t)| \leq \int_{t_0}^t |f(s, y_1(s)) - f(s, y_0)| ds \leq L M \int_{t_0}^t |s - t_0| ds = \frac{LM}{2!} |t - t_0|^2$$

$$\text{so } |y_{n+1}(t) - y_n(t)| \leq \frac{M}{L} \cdot \frac{|t - t_0|^n \cdot L^n}{n!}$$

Claim. $\{y_n(t)\}$ is Cauchy. w.l.o.g. suppose $n > m$
 $|f_n(t) - f_m(t)| \leq |f_n(t) - f_{n-1}(t)| + \dots + |f_{m+1}(t) - f_m(t)| \leq \frac{M}{L} \left(\frac{(\alpha L)^n}{n!} + \dots + \frac{(\alpha L)^{m+1}}{m!} \right)$

$$\leq \frac{M}{L} \left(e^{\alpha L} - \left(1 + \frac{\alpha L}{1!} + \dots + \frac{(\alpha L)^m}{m!} \right) \right) < \varepsilon \quad B_K: \frac{x}{e^x - 1} = \sum_{k=0}^{\infty} \frac{B_k}{k!} x^k$$

so $\{f_n(t)\}$ converges

n -dimensional case:

$$F(t, y, y', \dots, y^{(n)}) = 0, \quad y(x_0) = y_0, \quad y'(x_0) = y_0', \dots, \quad y^{(n-1)}(x_0) = y_0^{(n-1)}$$

$$y^{(n)} = f(t, y, y', \dots, y^{(n-1)})$$

$$\text{Then let } \vec{y} = (y, y', \dots, y^{(n-1)})^T \quad \vec{F}(t, y) = (y', \dots, y^{(n-1)}, f(t, y, \dots, y^{(n-1)}))^T$$

$$\text{Hence we have } \vec{y}' = \vec{F}(t, \vec{y})$$

So, if \vec{F} satisfies the Lipschitz condition, i.e. $\|\vec{F}(x, \vec{y}_1) - \vec{F}(x, \vec{y}_2)\| \leq L\|\vec{y}_1 - \vec{y}_2\|$ we can repeat the Picard iteration.

Hint: The norm in \mathbb{R}^n is given by $(x_1^2 + \dots + x_n^2)^{\frac{1}{2}}$

Here we have $R = [t-a_0, t+a_0] \times \dots \times [y_0^{(n-1)} - a_{n-1}, y_0^{(n-1)} + a_{n-1}]$, $M = \max_{[t, \vec{y}] \in R} \|\vec{F}(t, \vec{y})\|$

$$\vec{y}_0 = \vec{y}(t_0)$$

$$\vec{y}_1 = \vec{y}_0 + \int_{t_0}^t \vec{F}(s, \vec{y}_0(s)) ds$$

...

$$\vec{y}_n = \vec{y}_0 + \int_{t_0}^t \vec{F}(s, \vec{y}_{n-1}(s)) ds$$

$$\vec{y}_n - \vec{y}_{n-1} = \int_{t_0}^t \vec{F}(s, \vec{y}_n(s)) - \vec{F}(s, \vec{y}_{n-1}(s)) ds$$

$$\|\vec{y}_n - \vec{y}_{n-1}\| \leq \int_{t_0}^t \|F(s, y_n) - F(s, y_{n-1})\| ds \leq L \int_{t_0}^t \|y_n - y_{n-1}\| ds$$

$$\leq \dots \leq L^{n-1} \int_{[t_0, t]}^n \|\vec{y}_1 - \vec{y}_0\| ds^{n-1} \leq M L^{n-1} \int_{[t_0, t]}^n |s - t_0| ds^{n-1} = M L^{n-1} \cdot \frac{|t - t_0|^n}{n!}$$

$$\text{Choose } \alpha = \min \left\{ a_0, \frac{M}{a_1}, \dots, \frac{M}{a_{n-1}} \right\}, \text{ so } \|\vec{y}_n - \vec{y}_{n-1}\| \leq \frac{M}{L} \frac{(\alpha L)^n}{n!}$$

It is easy to check that $\{\vec{y}_n\}$ is Cauchy.

Exact equation.

intuition: Consider function $F(t, y(t)) = C$

then $\frac{dF}{dt} = \frac{\partial F}{\partial t} + \frac{\partial F}{\partial y} \cdot \frac{dy}{dt} = 0$

let $\frac{\partial F}{\partial t} = M(t, y)$, $\frac{\partial F}{\partial y} = N(t, y)$, i.e. $M(t, y) + N(t, y) \frac{dy}{dt} = 0$

Thm. Given $M(t, y)$ & $N(t, y)$. Then

$$\exists \phi(t, y) \text{ s.t. } \begin{cases} \frac{\partial \phi}{\partial t} = M \\ \frac{\partial \phi}{\partial y} = N \end{cases} \Leftrightarrow \frac{\partial M}{\partial y} = \frac{\partial N}{\partial t}$$

proof: $\Rightarrow \checkmark$

$$\Leftarrow \text{Set } \phi(t, y) = \int M(t, y) dt + \int [N(t, y) - \frac{\partial}{\partial y} M(t, y)] dy$$

Example $\underbrace{\frac{1}{2}y^2 + 2ye^t}_{M} + \underbrace{(y+e^t)}_{N} \frac{dy}{dt} = 0$

$$M_y = y+2e^t, N_t = e^t$$

Suppose $\exists \mu = \mu(t)$, s.t. $(\mu M)_y = (\mu N)_t$

$$\mu M_y = \mu_t N + \mu N_t$$

$$\mu_t = \frac{1}{N} \mu (M_y - N_t)$$

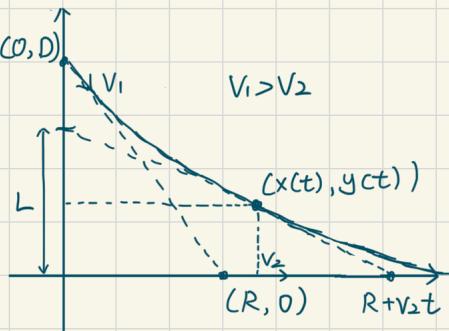
$$\frac{1}{\mu} \frac{d\mu}{dt} = \frac{M_y - N_t}{N} \frac{y+e^t}{y+e^t} = 1, \quad \frac{d\mu}{\mu} = dt$$

Therefore we take $\mu = e^t$

$$F(t, y) = \frac{1}{2}y^2 e^t + ye^{2t}, \text{ then } \frac{dF}{dt} = \frac{1}{2}y^2 e^t + 2ye^{2t} + (ye^t + e^{2t}) \frac{dy}{dt} = 0$$

$$\text{so } \frac{1}{2}y^2 e^t + ye^{2t} = C \text{ for some constant } C$$

Pursuit problem.



$$\left(\frac{dy}{dt}\right)^2 + \left(\frac{dx}{dt}\right)^2 = v_1^2$$

$$y(t) \cdot \frac{dy}{dx} = R + v_2 t$$

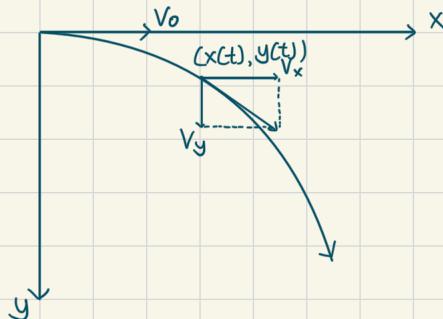
$$\frac{dy}{dx} = \frac{R + v_2 t}{y}$$

$$\frac{dy}{dt} = \frac{R + v_2 t}{y} \cdot \frac{dx}{dt} = \frac{R + v_2 t}{y} \cdot \sqrt{v_1^2 - \left(\frac{dy}{dt}\right)^2}$$

$$\left(\frac{dy}{dt}\right)^2 = \left(\frac{R + v_2 t}{y}\right)^2 \cdot \left(v_1^2 - \left(\frac{dy}{dt}\right)^2\right)$$

$$\left(1 + \left(\frac{R + v_2 t}{y}\right)^2\right) \left(\frac{dy}{dt}\right)^2 = \frac{R + v_2 t}{y}$$

Projectile Motion (with air friction)



Lec. 6

2nd order Linear Equations

$$\begin{cases} y'' = f(t, y, y') \\ y(0) = y_0 \\ y'(0) = y_1 \end{cases}$$

$$\underbrace{y'' + p(t)y' + q(t)y = r(t)}_{L[y] \equiv}$$

$$L[y] = y'' + p(t)y' + q(t)y = 0$$

$$(IVP) \begin{cases} y'' + p(t)y' + q(t)y = 0 \\ y(t_0) = \alpha \\ y'(t_0) = \beta \end{cases}$$

Thm. (IVP) has a unique solution if p, q are continuous.

V = the set of all solutions to $L[y] = 0$

Prop. V is a vector space of dimension 2

pf.) ① $0_V \in V$

② If y_1, y_2 are solutions for $L[y] = 0$, then so is $\alpha y_1 + \beta y_2$

We say that $V = \text{Ker}(L)$ $\hookrightarrow V$ is a vector space

$$\begin{cases} y_1'' + p(t)y_1' + q(t)y_1 = 0 \\ y_1(t_0) = 1 \\ y_1'(t_0) = 0 \end{cases} \quad \begin{cases} y_2'' + p(t)y_2' + q(t)y_2 = 0 \\ y_2(t_0) = 0 \\ y_2'(t_0) = 1 \end{cases}$$

Claim. $\{y_1, y_2\}$ is a basis for V

i.e. $y \in V$, let $\alpha = y(t_0)$ $\beta = y'(t_0)$

Since $(\alpha y_1 + \beta y_2)'(t_0) = \alpha$, $(\alpha y_1 + \beta y_2)''(t_0) = \beta$, then $y = \alpha y_1 + \beta y_2$

Check. $\{y_1, y_2\}$ is linearly independent

$$\alpha_1 y_1 + \alpha_2 y_2 \equiv 0, \text{ so } \alpha_1 y_1(t_0) = 0, \text{ so } \alpha_1 = 0$$

$$\alpha_1 y_1' + \alpha_2 y_2' \equiv 0, \text{ so } \alpha_2 y_2'(t_0) = 0, \text{ so } \alpha_2 = 0$$

we see that $\text{span}\{y_1, y_2\} = V$ and $\{y_1, y_2\}$ is linearly independent

so it is a basis for V , $\dim(V) = 2$

Wronskian $W[y_1, y_2] = \begin{vmatrix} y_1 & y_2 \\ y'_1 & y'_2 \end{vmatrix} = y_1 y'_2 - y'_1 y_2$, a function of t

Prop If y_1, y_2 are 2 solutions to $y'' + p(t)y' + q(t) = 0$, then

$$W' + p(t)W = 0, \text{ where } W = W[y_1, y_2], \text{ i.e.}$$

$$W(t) = W(t_0)e^{-\int_{t_0}^t p(s)ds}$$

$$\text{See that } W' = y'_1 y'_2 + y_1 y''_2 - y''_1 y_2 - y'_1 y''_2 = y_1 y''_2 - y''_1 y_2$$

$$\begin{aligned} W' + p(t)W &= y_1 y''_2 - y''_1 y_2 + p(t)(y_1 y'_2 - y'_1 y_2) = y_1(y_2'' - p(t)y_2') - y_2(y_1'' - p(t)y_1') \\ &= -q(t)y_1 y_2 + q(t)y_1 y_2 = 0 \end{aligned}$$

Prop $W \equiv 0 \Leftrightarrow \{y_1, y_2\}$ is linearly dependent.

$$(\Leftarrow): y_1 = -\frac{c_2}{c_1} y_2, \text{ so } W[y_1, y_2] \equiv 0$$

(\Rightarrow): Suppose $y_2(\bar{t}) \neq 0$ at some \bar{t} , so set $y_3 = \frac{y_1(\bar{t})}{y_2(\bar{t})} y_2(t)$
we see that $y_3(\bar{t}) = y_1(\bar{t}) \neq y'_3(\bar{t}) = y'_1(\bar{t})$, so

$$(3) W[y_1, y_2] \equiv 0 \Leftrightarrow W[y_1, y_2](t_0) = 0 \quad (y_1, y_2 \text{ are 2 solutions})$$

(\Rightarrow): \checkmark

$$(\Leftarrow): (y_1 y'_2 - y'_1 y_2)(t_0) = 0$$

$$\begin{cases} y_1'' + p(t)y'_1 + q(t)y_1 = 0 \\ y_2'' + p(t)y'_2 + q(t)y_2 = 0 \end{cases} \Rightarrow \begin{cases} y_1''(t_0) + p(t_0)y'_1(t_0) + q(t_0)y_1(t_0) = 0 \\ y_2''(t_0) + p(t_0)y'_2(t_0) + q(t_0)y_2(t_0) = 0 \end{cases}$$

$$\text{Since } W[y_1, y_2] = 0, \text{ then } \begin{cases} y_2''(t_0) = c y_1''(t_0) \\ y_2'(t_0) = c y_1'(t_0) \\ y_2(t_0) = c y_1(t_0) \end{cases}$$

We notice that $y_2 = cy_1$ is a solution, since this is an IVP
then $y_2 = cy_1$ is unique, so $\{y_1, y_2\}$ is linearly dependent, i.e. $W[y_1, y_2] \equiv 0$

Another proof

$$\begin{cases} y_1'' y_2 + p(t)y'_1 y_2 + q(t)y_1 y_2 = 0 \\ y_1 y_2'' + p(t)y_1 y'_2 + q(t)y_1 y_2 = 0 \end{cases}$$

so

$$(y_1'' y_2 - y_1 y_2'') + p(t)(y_1 y'_2 - y'_1 y_2) = 0$$

$$\text{i.e. } \frac{dW}{dt} + p(t)W = 0$$

so see $W(t_0) = 0$, and $W(t) \equiv 0$ is a solution
by the uniqueness theorem, $W(t) \equiv 0$

Lec 7.

Suppose $ay'' + by' + cy = 0$, let $D = \frac{d}{dt}$

$$\text{so } (aD^2 + bD + c)y = 0$$

$$(D - r_1)(D - r_2)y = 0, \quad r = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

First, suppose $\Delta = b^2 - 4ac > 0$

then $\overset{\uparrow}{e^{r_1 t}}, \overset{\uparrow}{e^{r_2 t}}$ are two solutions, and $W[y_1, y_2] = (r_2 - r_1)e^{r_1 t}e^{r_2 t} \neq 0$

so $\{y_1, y_2\}$ forms a basis

if $\Delta < 0$, we consider complex roots, so $r_1 = \alpha + \beta i, r_2 = \alpha - \beta i$

$$\text{we have } y_1 = e^{\alpha t}(\cos(\beta t) + i \sin(\beta t)), \quad y_2 = e^{\alpha t}(\cos(\beta t) - i \sin(\beta t))$$

Since we are more used to real solutions, we change the basis into $\{e^{\alpha t} \cos(\beta t), e^{\alpha t} \sin(\beta t)\}$

if $\Delta = 0$, $(D - r)^2 y = 0$

Consider the Jordan Canonical

if $(D - r)y = 0$, then we find a solution $y_1 = e^{rt}$

if $(D - r)y = e^{rt}$, then we find another solution $y_2 = te^{rt}$

$$W[y_1, y_2] = \begin{vmatrix} e^{rt} & te^{rt} \\ re^{rt} & e^{rt} + te^{rt} \end{vmatrix} = e^{2rt} \neq 0$$

Check. $(D - r_1)(D - r_2)y = (D - r_2)(D - r_1)y$

$$(D - r_1)\left(\frac{dy}{dt} - r_2 y\right) = \frac{d^2y}{dt^2} - (r_1 + r_2)\frac{dy}{dt} + r_1 r_2 y$$

$$(D - r_2)\left(\frac{dy}{dt} - r_1 y\right) = \frac{d^2y}{dt^2} - (r_1 + r_2)\frac{dy}{dt} + r_1 r_2 y$$

Def. D is said to be a differential operator on a field K if for any $a, b \in K$

$$D: K \rightarrow K \cdot D(a + b) = D(a) + D(b)$$

$$\cdot D(ab) = aD(b) + bD(a)$$

Then for the n 'th order case

$$\mathcal{L}_n[y] := a_0 y + a_1 y^{(1)} + \dots + a_n y^{(n)} = 0$$

\mathbb{C} is an algebraic closed field, any thus we can write the equation as

$$(D - \lambda_1)^{m_1}(D - \lambda_2)^{m_2} \dots (D - \lambda_k)^{m_k} y = 0 \quad (m_1 + \dots + m_k = n)$$

Claim. $\{e^{\lambda_1 t}, \dots, t^{m_1-1}e^{\lambda_1 t}, e^{\lambda_2 t}, \dots, t^{m_2-1}e^{\lambda_2 t}, \dots, \dots, e^{\lambda_k t}, \dots, t^{m_k-1}e^{\lambda_k t}\}$ forms a basis of solutions to $\mathcal{L}_n[y] = 0$

Check. $\{e^{\lambda_1 t}, \dots, t^{m_1-1}e^{\lambda_1 t}\}$ is linearly independent sets of solution

Sec. 8

Variation of parameters

Example. $(1-t^2)y''+2ty'-2y=0 \quad -1 < t < 1$

Notice that $y=t$ is a solution

Consider $\frac{y}{t}$. $\frac{d}{dt}\left(\frac{y}{t}\right) = \frac{1}{t}y' - \frac{1}{t^2}y \quad \frac{d^2}{dt^2}\left(\frac{y}{t}\right) = \frac{1}{t}y'' - \frac{2}{t^2}y' + \frac{2}{t^3}y = \frac{1}{t^3}(t^2y'' - 2ty' + 2y) = \frac{y''}{t^3}$

$$\frac{d}{dt}\left(\frac{y}{t}\right) = \frac{1}{t^2}(ty' - y) = \frac{t^2 - 1}{2t^2}y'' \quad \text{so} \quad \frac{d}{dt}\left(\frac{y}{t}\right) = \frac{1}{2}t(t^2 - 1)\frac{d^2}{dt^2}\left(\frac{y}{t}\right)$$

$y=t^2+1$ is also a solution

$$W[t, t^2+1] = \begin{vmatrix} t & t^2+1 \\ 1 & 2t \end{vmatrix} = -t^2 - 1 < 0 \quad \checkmark$$

Suppose we have 2 linearly independent solutions to

$y'' + P(t)y' + Q(t)y = 0$, denoted $y_1(t), y_2(t)$

Then we consider the equation $y'' + P(t)y' + Q(t)y = r(t)$

$$y(t) = C_1(t)y_1(t) + C_2(t)y_2(t)$$

$$y' = (C_1'y_1 + C_2'y_2) + (C_1y_1' + C_2y_2')$$

To simplify the equation, let $C_1'y_1 + C_2'y_2 = 0$

$$\text{so } y' = C_1y_1' + C_2y_2', \quad y'' = C_1y_1'' + C_1y_1' + C_2y_2'' + C_2y_2'$$

$$y'' + P(t)y' + Q(t)y = C_1(y_1'' + P(t)y_1' + Q(t)y_1) + C_2(y_2'' + P(t)y_2' + Q(t)y_2) + C_1'y_1' + C_2'y_2' = r(t)$$

$$\begin{cases} C_1'y_1 + C_2'y_2 = 0 \\ C_1'y_1' + C_2'y_2' = r \end{cases}$$

$$C_1 = \frac{\begin{vmatrix} 0 & y_2 \\ r & y_2' \end{vmatrix}}{W[y_1, y_2]} = \frac{-ry_2}{W[y_1, y_2]}$$

$$C_2 = \frac{\begin{vmatrix} y_1 & 0 \\ y_1' & r \end{vmatrix}}{W[y_1, y_2]} = \frac{ry_1}{W[y_1, y_2]}$$

$$(D^n + P_1(t)D^{n-1} + \dots + P_{n-1}(t)D + P_n) \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = 0$$

$$W[y_1, \dots, y_n] = \det \begin{pmatrix} y_1 & \dots & y_n \\ y_1^{(1)} & \dots & y_n^{(1)} \\ \vdots & & \vdots \\ y_1^{(n-1)} & \dots & y_n^{(n-1)} \end{pmatrix} \neq 0 \iff \{y_1, \dots, y_n\} \text{ linearly indep.}$$

Lec 11.

Series solutions

$$(*) \quad P(t)y'' + Q(t)y' + R(t)y = 0$$

Thm. If $\frac{Q(t)}{P(t)}$ & $\frac{R(t)}{P(t)}$ have Taylor series expansions converges on $(t_0 - \rho, t_0 + \rho)$, then every solution of (*) must have a Taylor series expansion which converges on $(t_0 - \rho, t_0 + \rho)$

Euler's Equation. $t^2y'' + aty' + by = 0$

Bessel's equation of order $\frac{1}{2}$

$$t^2y'' + ty' + (t^2 - \frac{1}{4})y = 0$$

$$y = \sum_{n=0}^{\infty} a_n t^{n+r}, \quad (a_0 \neq 0)$$

$$y' = \sum_{n=0}^{\infty} (n+r)a_n t^{n+r-1}$$

$$y'' = \sum_{n=0}^{\infty} (n+r)(n+r-1)a_n t^{n+r-2}$$

$$t^2y'' + ty' + (t^2 - \frac{1}{4})y = \sum_{n=0}^{\infty} [(n+r)(n+r-1)a_n + (n+r)a_n - \frac{1}{4}a_n] t^{n+r} + \sum_{n=2}^{\infty} a_{n-2} t^{n+r}$$

$$= t^r \left[\sum_{n=0}^{\infty} t^n [(n+r)^2 - \frac{1}{4}] a_n + \sum_{n=2}^{\infty} a_{n-2} t^n \right] = \left\{ a_0 \left(r^2 - \frac{1}{4} \right) + a_1 \left[(r+1)^2 - \frac{1}{4} \right] t + \sum_{n=2}^{\infty} t^n \left\{ [(n+r)^2 - \frac{1}{4}] a_n + a_{n-2} \right\} \right\} t^r = 0$$

$$\text{so } r = \pm \frac{1}{2}$$

$$\textcircled{1} \quad r = \frac{1}{2}, \quad \left[(1+\frac{1}{2})^2 - \frac{1}{4} \right] a_1 = 0, \quad \text{so } a_1 = 0$$

$$\text{For } n \geq 2, \quad \left[(n+\frac{1}{2})^2 - \frac{1}{4} \right] a_n + a_{n-2} = 0, \quad a_n = \frac{(-1)a_{n-2}}{n(n+1)}$$

$$\text{w.t.o.g. } a_0 = 1, \quad \text{then } y = \frac{1}{\sqrt{t}} \left(t - \frac{t^3}{3!} + \frac{t^5}{5!} - \dots \right) = \frac{\sin t}{\sqrt{t}}$$

o

Def. Suppose $f(t, y)$ is continuous on G , if $\forall (t, y_1), (t, y_2) \in G$, $|f(x, y_1) - f(x, y_2)| \leq F(|y_1 - y_2|)$

where $F(r) > 0$ is a continuous function ($r > 0$), with $\int_0^{\infty} \frac{1}{F(r)} dr = +\infty$ for $\forall \varepsilon > 0$

Then say $f(t, y)$ is of Osgood condition about y

Def. Given $I \subseteq \mathbb{R}$, $|A| = \infty$, the $\{f_\alpha\}_{\alpha \in A}$ is uniformly bounded if $\exists M > 0$, s.t. $|f_\alpha(x)| \leq M$ for $\forall \alpha \in A$, $x \in I$

Def. $\{f_\alpha\}_{\alpha \in A}$ is equi-continuous if $\forall \varepsilon > 0$, $\exists \delta > 0$, s.t. $|f_\alpha(x) - f_\alpha(y)| < \varepsilon$ whenever $|x - y| < \delta$ for $\forall \alpha \in A$

Sturm Comparison Thm Let y, z be solutions to $y''(t) + q(t)y = 0$ & $z''(t) + r(t)z = 0$, where $q(t) > r(t) > 0$

Then y vanishes at least once between any 2 consecutive zeros of z
proof: $w(t) = yz' - y'z$, $w'(t) = yz'' - y''z$

Suppose y has no zeros in (α, β)

$w'(t) = zqy - yrz = yz(q - r) \neq 0$, so w.l.o.g. $w'(t) > 0$

$w(t) \nearrow$ $w(\alpha) = y(\alpha)z'(\alpha)$, $w(\beta) = y(\beta)z'(\beta)$

Since y has no zeros, then w.l.o.g. $y(t) > 0$ on (α, β)

Also $w'(\beta) > 0$, therefore $w(\beta) > 0$ strictly, $z'(\beta) > 0$

This is impossible since $z(\alpha) = z(\beta) = 0$

Examples

$$1. \quad y'' + 3y' + 2y = e^t$$

$$\begin{cases} x_1 = y \\ x_2 = y' \end{cases} \quad \begin{pmatrix} x_1' \\ x_2' \end{pmatrix} = \begin{pmatrix} x_2 \\ -2x_1 - 3x_2 + e^t \end{pmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}' = \begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 0 \\ e^t \end{pmatrix}, \quad X' = AX + b$$

$$\begin{vmatrix} -2 & 1 \\ 2 & -3-2 \end{vmatrix} = x^2 + 3x + 2 = (x+1)(x+2)$$

$$\begin{pmatrix} 1 & 1 \\ -2 & -2 \end{pmatrix} \begin{pmatrix} C_1 \\ C_2 \end{pmatrix} = \begin{pmatrix} C_1 + C_2 \\ -2C_1 - 2C_2 \end{pmatrix} = 0 \quad . \text{ so } A = PDP^{-1}. \quad P = \begin{pmatrix} 1 & 1 \\ -1 & -2 \end{pmatrix}$$

$$\begin{pmatrix} 2 & 1 \\ -2 & 1 \end{pmatrix} \begin{pmatrix} C_1 \\ C_2 \end{pmatrix} = \begin{pmatrix} 2C_1 + C_2 \\ -2C_1 + C_2 \end{pmatrix} = 0$$

$$\text{Let } \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = P^{-1} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \text{ then } \begin{pmatrix} z_1' \\ z_2' \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} + \begin{pmatrix} 2 & 1 \\ -1 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ e^t \end{pmatrix}$$

$$\begin{pmatrix} z_1' \\ z_2' \end{pmatrix} = \begin{pmatrix} -z_1 + e^t \\ -2z_2 - e^t \end{pmatrix} \quad \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \begin{pmatrix} \frac{1}{2}e^t \\ -\frac{1}{3}e^{2t} \end{pmatrix} + \begin{pmatrix} C_1 e^{-t} \\ C_2 e^{-2t} \end{pmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = P \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \begin{pmatrix} \frac{1}{2}e^t - \frac{1}{3}e^{2t} \\ -\frac{1}{2}e^t + \frac{2}{3}e^{2t} \end{pmatrix} + \begin{pmatrix} C_1 e^{-t} + C_2 e^{-2t} \\ -C_1 e^{-t} - 2C_2 e^{-2t} \end{pmatrix}$$

$$\text{Let } \mathbb{X} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} . \quad A(t) = \begin{pmatrix} a_{11}(t) & \cdots & a_{1n}(t) \\ \vdots & \ddots & \vdots \\ a_{n1}(t) & \cdots & a_{nn}(t) \end{pmatrix} \quad b(t) = \begin{pmatrix} y_1(t) \\ \vdots \\ y_n(t) \end{pmatrix}$$

$$(*) \quad \begin{cases} \mathbb{X}' = A\mathbb{X} + b \\ \mathbb{X}(t_0) = \mathbb{X}_0 \end{cases}$$

Thm. (*) has exactly one solution in the entire interval.

(Use Picard iteration directly)

$$(+) \quad \mathbb{X}' = A\mathbb{X} \quad , \text{ where } \mathbb{X} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

Thm. Let V be the set of all solutions to (+).

Then V is a n -dimensional vector space.

pf:) $\mathbb{X}_1' = A\mathbb{X}_1, \mathbb{X}_2' = A\mathbb{X}_2$, then $(\alpha\mathbb{X}_1 + \beta\mathbb{X}_2)' = \alpha\mathbb{X}_1' + \beta\mathbb{X}_2' = \alpha A\mathbb{X}_1 + \beta A\mathbb{X}_2 = A(\alpha\mathbb{X}_1 + \beta\mathbb{X}_2)$

Claim. $\{\mathbb{X}_1, \dots, \mathbb{X}_n\}$ forms a basis for V , \mathbb{X}_i is given by $\mathbb{X}_i(t_0) = e_i$, unique

Check. $V = \text{span}\{\mathbb{X}_1, \dots, \mathbb{X}_n\}$, $\forall \mathbb{X} \in V, \mathbb{X}(t_0) = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix} = c_1 e_1 + \dots + c_n e_n = c_1 \mathbb{X}_1(t_0) + \dots + c_n \mathbb{X}_n(t_0)$

$\mathbb{X} = c_1 \mathbb{X}_1 + \dots + c_n \mathbb{X}_n$ is a unique solution

Check. Linearity independence, let $(c_1 \mathbb{X}_1 + \dots + c_n \mathbb{X}_n)(t_0) = 0$, then $c_1 = \dots = c_n = 0$

Def. The wronskian of $\{x_1, \dots, x_n\}$ is given by

$$W[x_1, \dots, x_n] = \det(x_1, \dots, x_n)$$

Prop. $W(t) = W(0)e^{\text{tr}(A)t}$

$$\frac{d}{dt} W(t) = \frac{d}{dt} \begin{vmatrix} x_{11} & \cdots & x_{1n} \\ \vdots & \ddots & \vdots \\ x_{n1} & \cdots & x_{nn} \end{vmatrix} = \sum_{i=1}^n \begin{vmatrix} x_{11} & \cdots & x_{1n} \\ \vdots & \ddots & \vdots \\ x_{i1} & \cdots & x_{in} \\ \vdots & \ddots & \vdots \\ x_{n1} & \cdots & x_{nn} \end{vmatrix}$$

proof: $(v_1, 1 \cdots, v_n) = \det(v_1, \dots, v_n)(e_1, 1 \cdots, e_n)$

$$\frac{d}{dt} (v_1, 1 \cdots, v_n) = \sum_{i=1}^n (v_1, 1 \cdots, 1 \frac{dv_i}{dt} 1 \cdots, v_n) = \frac{d}{dt} \det(v_1, \dots, v_n)(e_1, 1 \cdots, e_n)$$

Also, $v_1, 1 \cdots, 1 \frac{dv_i}{dt} 1 \cdots, v_n = \det(v_1, \dots, \frac{dv_i}{dt}, \dots, v_n)(e_1, 1 \cdots, e_n)$

so $\frac{d}{dt} \det(v_1, \dots, v_n)(e_1, 1 \cdots, e_n) = \sum_{i=1}^n \det(v_1, \dots, \frac{dv_i}{dt}, \dots, v_n)(e_1, 1 \cdots, e_n)$

Since $\dim(A^n V) = 1$, then $\frac{d}{dt} \det(v_1, \dots, v_n) = \sum_{i=1}^n \det(v_1, \dots, \frac{dv_i}{dt}, \dots, v_n)$

$$\frac{dW}{dt} = \sum_{i=1}^n \det(v_1, \dots, \frac{dv_i}{dt}, \dots, v_n) = \sum_{i=1}^n \det(v_1, \dots, Av_i, \dots, v_n) = \sum_{i=1}^n a_{ii} W = \text{tr}(A) \cdot W$$

so $W(t) = W(0)e^{\text{tr}(A)t}$

Suppose A has eigenvalue λ with corresponding eigenvector v

i.e. $Av = \lambda v$

$$X = e^{\lambda t} \vec{v} \Rightarrow X' = \lambda e^{\lambda t} \vec{v} = \lambda X, \text{ which is a solution.}$$

$$\{e^{\lambda t} v_1, \dots, e^{\lambda t} v_n\}$$

Consider matrix with 1 real eigenvalue and 2 complex conjugate eigenvectors. $\lambda, \alpha \pm \beta i$

$$\dot{x} = Ax, \quad \tilde{x}_1 = e^{(\alpha+\beta i)t} (w_1 + iw_2) = e^{\alpha t} (\cos \beta t + i \sin \beta t) (w_1 + iw_2)$$

$$= e^{\alpha t} (\cos \beta t \cdot w_1 - \sin \beta t \cdot w_2 + i(\cos \beta t \cdot w_2 + \sin \beta t \cdot w_1))$$

$$\tilde{x}_2 = e^{\alpha t} (\cos \beta t w_1 - \sin \beta t w_2 - i(\cos \beta t w_2 + \sin \beta t w_1))$$

Then $\frac{\tilde{x}_1 + \tilde{x}_2}{2}$ & $\frac{\tilde{x}_1 - \tilde{x}_2}{2i}$ are solutions

i.e. $\operatorname{Re}(\tilde{x}_1) \rightarrow x_1$ & $\operatorname{Im}(\tilde{x}_1) \rightarrow x_2$

Here we see that $x_3(t) = e^{\alpha t} w_3$

Since $W(t) = W(0) e^{\operatorname{tr}(A)t}$, we only need to check that $\{x_1(0), x_2(0), x_3(0)\}$ is linearly independent.

This is clear since $x_1(0) = w_1, x_2(0) = w_2, x_3(0) = w_3$

Consider Jordan Normal Form

Example.

$$\dot{x} = \begin{pmatrix} 3 & -1 & 0 \\ 1 & 1 & 0 \\ -1 & 0 & 2 \end{pmatrix} x \quad \det(A - \lambda I) = -(2-\lambda)^3$$

$$(A - 2I)v = \begin{pmatrix} 1 & -1 & 0 \\ 1 & -1 & 0 \\ -1 & 0 & 0 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} = 0 \Rightarrow v = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

Fundamental Matrix Solution

For $\dot{X} = AX$, $A \in \mathbb{M}_{n \times n}(\mathbb{R})$

$\Xi(t)$ is a FMS if $\Xi(t)$ is a $n \times n$ matrix where each column is a solution while n columns are linearly independent.

Thm. Let $\Xi(t)$ be a FMS to $\dot{X} = AX$. Then $\Xi(t)\Xi(t_0)^{-1} = e^{At-t_0}$

(This can be used to deal with nonhomogeneous linear systems)

pf.) Consider $Y(t) = X(t)X^{-1}(t_0)$, this is also a solution, $Y(t_0) = I$

$$\text{So } X(t)X^{-1}(t_0) = e^{At-t_0}$$

Let $x_1(t), \dots, x_n(t)$ be n linearly independent solutions to $\dot{X} = AX$

$\Rightarrow C_1x_1(t) + \dots + C_nx_n(t)$ is a solution to $\dot{X} = AX$

Suppose $X(t) = u_1(t)x_1(t) + \dots + u_n(t)x_n(t)$ is a solution to $\dot{X} = AX + \vec{f}$

$$\Xi(t) = \begin{pmatrix} x_1(t) & \dots & x_n(t) \end{pmatrix}, \quad x(t) = \Xi(t) \begin{pmatrix} u_1(t) \\ \vdots \\ u_n(t) \end{pmatrix} = \Xi(t)u(t)$$

$$x(t_0) = \Xi(t_0)u(t_0)$$

$$x'(t) = \Xi'(t)u(t) + \Xi(t)u'(t) = A\Xi(t)u(t) + \Xi(t)u'(t) = AX + \Xi(t)u'(t) = AX + \vec{f}$$

$$\text{so } \vec{f}(t) = \Xi(t)u'(t), \quad u'(t) = \Xi^{-1}(t)\vec{f}(t)$$

$$\begin{aligned} x(t) &= \Xi(t)u(t) = \Xi(t) \int_{t_0}^t \Xi^{-1}(s)f(s)ds + \Xi(t)u(t_0) = \Xi(t) \int_{t_0}^t \Xi^{-1}(s)f(s)ds + \Xi(t)\Xi^{-1}(t_0)x(t_0) \\ &= e^{A(t-t_0)}x(t_0) + \Xi(t) \int_{t_0}^t \Xi^{-1}(s)f(s)ds = e^{A(t-t_0)}x(t_0) + \int_{t_0}^t e^{A(t-s)}f(s)ds \end{aligned}$$

$$x(t) = e^{A(t-t_0)}x(t_0) + \int_{t_0}^t e^{A(t-s)}f(s)ds$$

If $\Xi(t)$ is given & we hope to find A

we see that $e^{At} = \Xi(t)\Xi^{-1}(0)$

$$\frac{d}{dt}e^{At}\Big|_{t=0} = \Xi'(0)\Xi^{-1}(0) = A\Xi(0)\Xi^{-1}(0)$$

$$\text{i.e. } A = \Xi'(0)\Xi^{-1}(0)$$

$$(*) \quad \vec{x}' = \vec{f}(\vec{x}), \quad \vec{x} \in \mathbb{R}^m$$

Def. If $\vec{f}(x_0) = 0$, then $x(t) \equiv x_0$ is a solution to (*) & is called an equilibrium (Steady State)

Def. A solution $\vec{x}(t) = \vec{\phi}(t)$ of (*) is said to be stable if $\forall \varepsilon > 0$, $\exists S > 0$, s.t. for any solution $\vec{\psi}(t)$ of (*) with $\|\vec{\psi}(0) - \vec{\phi}(0)\| < S$ has the property that $\|\vec{\psi}(t) - \vec{\phi}(t)\| < \varepsilon$ for $\forall t > 0$

Continuous dependence on initial value

$$\forall \varepsilon > 0, T > 0, \exists S = S(\varepsilon, T), \text{ s.t. } |y_0 - y_1| < S \Rightarrow |y_2(t) - y_1(t)| < \varepsilon \text{ for } \forall 0 < t < T$$

Stability for linear systems $(t) \quad \dot{x} = Ax$

Prop. (Reduction) All solutions to (t) are stable $\Leftrightarrow x(t) \equiv \vec{0}$ is stable
proof: (\Rightarrow) \checkmark

(\Leftarrow) Let $\vec{\varphi}(t)$ be a solution of (t)

Since $x(t) \equiv 0$ is stable, $\forall \varepsilon > 0, \exists S > 0$, s.t. $\|\vec{\varphi}(0)\| < S \Rightarrow \|\vec{\varphi}(t)\| < \varepsilon, \forall t > 0$

Then let $\vec{\rho}(t)$ be an arbitrary solution s.t. $\|\vec{\rho}(0) - \vec{\varphi}(0)\| < S$

Set $\vec{\psi} = \vec{\rho} - \vec{\varphi}$. Then $\vec{\psi}$ is a solution (By linearity of (t))

Therefore $\|\vec{\rho}(t) - \vec{\varphi}(t)\| < \varepsilon$ for $\forall t > 0$, i.e. $\vec{\varphi}$ is stable

$X = AX$, $A: n \times n$ constant matrix

- (1) At least 1 eigenvalue of A has positive real parts
- (2) All eigenvalues of A have negative real parts
- (3) All eigenvalues of A have non-positive real parts, with some of them being 0 or purely imaginary

Prop. In case (1), zero solution is unstable

pf: Suppose $\lambda > 0$ is an eigenvalue of A

Let \vec{v} be a corresponding eigenvector of λ , then $\vec{\psi}(t) = ce^{\lambda t} \vec{v}$ is a solution to (t)

$\forall S > 0$, pick $c > 0$ s.t. $\|\vec{\psi}(0)\| = c\|\vec{v}\| = \frac{1}{2}S < S$, $\|\vec{\psi}(t)\| = \frac{1}{2}Se^{\lambda t} \rightarrow \infty$

Now consider $\lambda = \alpha + \beta i$, $\alpha > 0$

$\vec{\psi}(t) = ce^{\alpha t}(\cos \beta t \vec{v} - \sin \beta t \vec{w})$, Take $t_k = \frac{2k\pi}{\beta}$, then $\|\vec{\psi}(t)\| \rightarrow \infty$ as $k \rightarrow \infty$

In case (2), all solutions tends to 0 as $t \rightarrow \infty$

pf.) Let $\lambda_1, \dots, \lambda_k$ be negative eigenvalues of A

$\lambda_{k+1}, \dots, \lambda_m$ be eigenvalues with negative real parts

$\lambda_j = \alpha_j + i\beta_j$ for $j = k+1, \dots, m$, $\alpha_j < 0$. $v_j = w_j + iu_j$

Then the solution must be a linear combination of the following:

(1) $e^{\lambda_l t} (\vec{v}_l + t \vec{v}'_l + \dots + t^{p-1} \vec{v}'_l)$. $l = 1, \dots, k$

(2) $e^{\alpha_j t} (\cos \beta_j t \vec{w}_j - \sin \beta_j t \vec{u}_j) (1 + t + \dots + t^{q-1})$, $j = k+1, \dots, m$

Then all solution tends to 0 as $t \rightarrow \infty$

(3) Assume A has

$\lambda_1 = i\beta_1, \dots, \lambda_k = i\beta_k$, $\beta_j \in \mathbb{R}$,

$\lambda_{k+1}, \dots, \lambda_m$ having negative real parts

Prop. Suppose λ_j , $j = 1, \dots, k$ has multiplicity s_j

Then all solution of $\dot{x} = Ax$ are stable \Leftrightarrow A has s_j linearly independent eigenvectors for $\lambda_j = i\beta_j$, $j = 1, \dots, k$

proof: (\Rightarrow) Suppose $\lambda_1 = i\beta_1$ doesn't have s_1 linearly independent eigenvectors, then $te^{i\beta_1 t}$ is a solution

This implies that 0 is not stable, \times .

(\Leftarrow) trivial

Example. $\dot{x} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} x$

$$\det(A - \lambda I) = \lambda^2 + 1 = 0, \quad \lambda = \pm i$$

$$(A - iI)v = \begin{pmatrix} i & -1 \\ 1 & -i \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = 0, \quad \vec{v} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} + i \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\vec{x}(t) = e^{it} \vec{v} = \begin{pmatrix} -\sin t \\ \cos t \end{pmatrix} + i \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}$$

$$\text{so } x(t) = c_1 \begin{pmatrix} -\sin t \\ \cos t \end{pmatrix} + c_2 \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}, \quad \|x(0)\| = \sqrt{c_1^2 + c_2^2}$$

$$\|x(t)\| = \sqrt{c_1^2 + c_2^2} = \|x(0)\|$$

Take $S = \epsilon$, then \forall solution that $\|x(0)\| < S$, we have $\|x(t)\| < \epsilon$

Def. A solution $\vec{\phi}(t)$ is said to be asymptotically stable if

$\exists S > 0$, s.t. as $t \rightarrow \infty$, $\|\vec{\phi}(t) - \vec{\psi}(t)\| \rightarrow 0$, \forall solution $\vec{\psi}(t)$

s.t. $\|\vec{\psi}(0) - \vec{\phi}(0)\| < S$

Stability for Nonlinear systems $\dot{x} = Ax + \vec{g}(x)$

where $A = n \times n$ constant matrix, $\vec{g}(x) = o(\|x\|)$ near $\vec{x} = 0$

Thm.

- (1) All eigenvalues of A have negative real parts
- (2) At least 1 eigenvalue of A has positive real parts
- (3) Otherwise

For nonlinear system $\dot{x} = Ax + \vec{g}(x)$. where $\vec{g}(x) = o(\|x\|)$ at $x = \vec{0}$

Then consider the stability of $X \equiv 0$, one has

- (1) Asymptotically stable
- (2) unstable

proof: (1) $x(t) = e^{At}x(0) + \int_0^t e^{A(t-s)}\vec{g}(x(s)) ds$

Claim. $\exists \alpha > 0, K > 0$, s.t. $\|e^{At}\vec{c}\| \leq K e^{-\alpha t} \|\vec{c}\|, \forall \vec{c}, t \geq 0$

$$e^{-\alpha t} (C_1 \vec{v}_1 + C_2 t \vec{v}_2 + \dots + C_p t^{p-1} \vec{v}_p) = O(e^{-\alpha t} t^{p-1}) = O(e^{-\alpha t})$$

Thus take $\alpha = \max\{\lambda_1, \dots, \lambda_k\}$

So one has $\|e^{At}\vec{c}\| \leq K e^{-\alpha t} \|\vec{c}\|$ for some $K > 0$

$\forall \varepsilon > 0, \exists S > 0$, s.t. $\|\vec{g}(x)\| \leq \varepsilon \|x\|$ when $\|x\| < S$

Goal. $\exists \eta > 0$, s.t. if $\|x(0)\| < \eta$, then $x(t) \rightarrow \vec{0}$ as $t \rightarrow \infty$

Strategy. Pick $\varepsilon > 0, S = \delta \varepsilon$, & $\eta > 0$, s.t. $\|x(0)\| < \eta \Rightarrow \|x(t)\| < S$

$$\|x(t)\| \leq \|e^{At}x(0)\| + \int_0^t \|e^{A(t-s)}\vec{g}(x(s))\| ds \leq K e^{-\alpha t} \|x(0)\| + K \int_0^t e^{-\alpha(t-s)} \|x(s)\| ds$$

$$e^{\alpha t} \|x(t)\| \leq K \|x(0)\| + K \varepsilon \int_0^t e^{-\alpha s} \|x(s)\| ds \quad (*)$$

let $u(t) = \int_0^t e^{-\alpha s} \|x(s)\| ds$, then $u'(t) = e^{\alpha t} \|x(t)\| \leq K \|x(0)\| + K \varepsilon u(t)$

$$u'(t) - \varepsilon K u(t) \leq K \|x(0)\|, \text{ i.e. } (u(t)e^{-\varepsilon K t})' \leq K \|x(0)\| e^{-\varepsilon K t}$$

$$\text{Hence } u(t) \leq \frac{1}{\varepsilon} \|x(0)\| (e^{\varepsilon K t} - 1)$$

plug in (*), then $e^{\alpha t} \|x(t)\| \leq K \|x(0)\| + K \|x(0)\| (e^{\varepsilon K t} - 1)$

$$\|x(t)\| \leq K e^{(K\varepsilon - \alpha)t} \|x(0)\|$$

Here, set $\varepsilon = \frac{\alpha}{2K}$, then $K\varepsilon - \alpha = -\frac{1}{2}\alpha < 0$

Choose $\eta = \frac{S}{K}$, then $\|x(t)\| < S e^{-\frac{1}{2}\alpha t}$ whenever $\|x(0)\| < \eta$

Hence $\lim_{t \rightarrow \infty} \|x(t)\| = 0$

General Nonlinear System.

(1) $\dot{\vec{X}} = \vec{f}(\vec{X})$, $\vec{f}(X_0) = 0$. i.e. $X(t) \equiv X_0$ is a solution.

Set $y = X - X_0$, then $y' = X'$

$$\vec{f}(X) = \vec{f}(X_0) + D\vec{f}(X_0)(X - X_0) + o(||X - X_0||^2)$$

$$y' = D\vec{f}(X_0)y + o(||y||)$$

where $D\vec{f}(X_0)$ is the Jacobean matrix

$$D\vec{f}(X_0) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}, \dots, \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1}, \dots, \frac{\partial f_n}{\partial x_n} \end{pmatrix}_{X=\vec{X}_0}$$

Phase Plane (2-dimensional)

(x, y) : phase plane

$$\frac{dy}{dx} = \frac{g(x, y)}{f(x, y)} \Rightarrow \frac{dy}{dx} = h(x, y)$$

Def. An orbit (trajectory) is the solution curve $\{(x(t), y(t)) | t \in \mathbb{R}\}$ on the phase plane.

Example. (1) $\begin{cases} x' = y \\ y' = -x \end{cases}$

$$\text{so } \frac{dy}{dx} = -\frac{x}{y} \quad y dy = -x dx, \quad x^2 + y^2 = C^2$$

$$\vec{x} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \vec{x}, \text{ then } \vec{x}_1 = \begin{pmatrix} \cos t \\ -\sin t \end{pmatrix} \quad \vec{x}_2 = \begin{pmatrix} \sin t \\ \cos t \end{pmatrix}$$

$$(2) \begin{cases} x' = y(1+x^2+y^2) \\ y' = -x(1+x^2+y^2) \end{cases} \Rightarrow x^2 + y^2 = C^2, \quad C > 0$$

$$\text{Then } \vec{x}' = \begin{pmatrix} 0 & (1+C^2) \\ -(1+C^2) & 0 \end{pmatrix} \vec{x}, \text{ all solutions has period } \frac{2\pi}{1+C^2}$$

Qualitative Properties of Orbits

$$\vec{x}' = \vec{f}(\vec{x}) = \begin{pmatrix} f_1(x_1, \dots, x_n) \\ \vdots \\ f_n(x_1, \dots, x_n) \end{pmatrix} \quad (*)$$

Prop. 1. Suppose f_1, \dots, f_n are continuously differentiable. Then for every $x_0 \in \mathbb{R}^n$, $(*)$ has a unique orbit going through x_0 .

Cor. No two orbits can intersect with each other

Prop. 2 Let $\vec{\varphi}(t) = \vec{x}(t)$ be a solution of $(*)$. If for some t_0 , $\exists T > 0$ s.t. $\vec{\varphi}(t_0 + T) = \vec{\varphi}(t_0)$, then $\forall t > 0$, $\vec{\varphi}(t) = \vec{\varphi}(t + T)$

Cor. Every closed orbit is periodic

$$(*) \begin{cases} x' = f(x, y) \\ y' = g(x, y) \end{cases}$$

Poincaré-Bendixson Thm.

Suppose Ω is a compact domain which contains no equilibrium point of (*). If (*) has one solution that completely stays in Ω for all time t large (i.e. $\exists t_0 > 0$, s.t. $(x(t), y(t)) \in \Omega$ for $t \geq t_0$), then either the solution $(x(t), y(t))$ is periodic or it spirals into a periodic solution.

Example.. $z'' + (z^2 + 2z'^2 - 1)z' + z = 0$

Set $\begin{cases} x = z \\ y = z' \end{cases}$, then $\begin{cases} x' = y \\ y' = -y(z^2 + 2z'^2 - 1) - x \end{cases}$

$$\frac{1}{2} \frac{d(x^2 + y^2)}{dt} = xy - y^2(z^2 + 2z'^2 - 1) - xy = y^2(1 - x^2 - 2y^2)$$

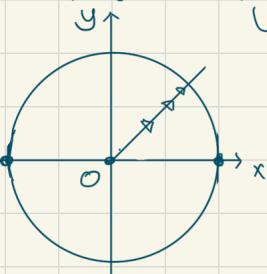
Limit Cycle.

Def. A closed orbit \mathcal{C} is called a limit circle of (*) if orbits of (*) spiral into it or away from it

Example. Find all limit cycles of $\begin{cases} x' = x(1 - x^2 - y^2) \\ y' = y(1 - x^2 - y^2) \end{cases}$

$$x' = 0 \Rightarrow x = 0 \text{ or } x^2 + y^2 = 1$$

$$\hookrightarrow y(1 - y^2) = 0 \Rightarrow y = 0 \text{ or } y = \pm 1$$



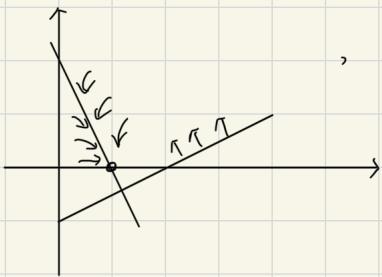
$$\begin{cases} x' = xy + x \cos(x^2 + y^2) \\ y' = -x^2y + y \cos(x^2 + y^2) \end{cases}$$

$$\begin{cases} xy + x \cos(x^2 + y^2) = 0 \\ x^2y = y \cos(x^2 + y^2) \end{cases}, \text{ so } x = 0 \text{ or } \cos(x^2 + y^2) + y = 0$$

If $x = 0$, we have $y \cos(y^2) = 0$, so $y = 0$ or $y^2 = (n + \frac{1}{2})\pi$

$$\begin{cases} x' = x(a - by - ex) \\ y' = y(c - c + dx - fy) \end{cases}, \quad a, b, c, d, e, f > 0$$

x : prey (猎物) y : predator (捕食者)



, $\frac{a}{e} < \frac{c}{d}$ implies that all solutions with $(x_0) > 0, y_0 > 0$ will tend to $(0, 0)$.

$$\begin{cases} x' = x(a - by) \\ y' = y(c - dx) \end{cases} \quad (*)$$

$$\frac{dy}{dx} = \frac{yc - c + dx}{x(a - by)}, \quad \int \frac{a - by}{y} dy = \int \frac{-c + dx}{x} dx$$

$$ye^{-by} = x^c e^{dx} \cdot C$$

$$\text{Set } \varphi(x, y) = \frac{ye^{-by}}{x^c} = f(x)g(y) = k$$

Consider the equilibrium points of $(*)$, we have $(0, 0)$ & $(\frac{c}{d}, \frac{a}{b})$

$$\nabla \varphi = (f'(x)g(y), f(x)g'(y))^T$$

If $\|\nabla \varphi\| = 0$, we have $f'(x) = 0$ & $g'(y) = 0$, and therefore $x = \frac{c}{d}$, $y = \frac{a}{b}$

So for all $0 < k < \varphi(\frac{c}{d}, \frac{a}{b})$, we are getting a closed & bounded region without equilibrium point.

Therefore $(x(t), y(t))$ forms a periodic solution

Let $(x(t), y(t))$ be such periodic solution with period T

$$\text{Lemma. } \frac{1}{T} \int_0^T x(t) dt = \frac{c}{d}, \quad \frac{1}{T} \int_0^T y(t) dt = \frac{a}{b}$$

$$\text{proof: } \frac{x'}{x} = a - by, \quad \frac{y'}{y} = -c + dx$$

$$0 = \ln x \Big|_0^T = \int_0^T \frac{x'}{x} dt = \int_0^T a - by dt = aT - b \int_0^T y(t) dt$$

$$\text{so } \frac{1}{T} \int_0^T y(t) dt = \frac{a}{b}, \quad \text{similarly } \frac{1}{T} \int_0^T x(t) dt = \frac{c}{d}$$

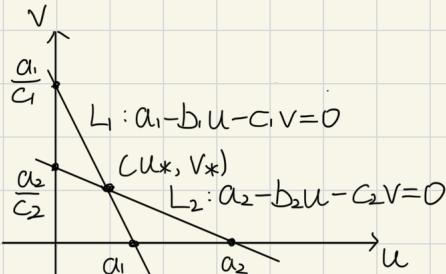
Lotka - Volterra Competition System

$$\begin{cases} u' = u(a_1 - b_1 u - c_1 v) \\ v' = v(a_2 - b_2 u - c_2 v) \end{cases} \quad (\dagger)$$

a_i, b_i, c_i 's are positive constants

Weak Competition: $\frac{b_1}{b_2} > \frac{a_1}{a_2} > \frac{c_1}{c_2}$

Strong Competition: $\frac{b_1}{b_2} < \frac{a_1}{a_2} < \frac{c_1}{c_2}$



$$\text{Here } u_* = \frac{a_1 c_2 - a_2 c_1}{b_1 c_2 - b_2 c_1}, \quad v_* = \frac{a_2 b_1 - a_1 b_2}{b_1 c_2 - b_2 c_1}$$

Linearize (\dagger) at (u_*, v_*)

$$J = \begin{pmatrix} \frac{\partial f}{\partial u} & \frac{\partial f}{\partial v} \\ \frac{\partial g}{\partial u} & \frac{\partial g}{\partial v} \end{pmatrix}_{(u_*, v_*)} = \begin{pmatrix} a_1 - 2b_1 u_* - c_1 v_* & -c_1 u_* \\ -b_2 v_* & a_2 - b_2 u_* - 2c_2 v_* \end{pmatrix} = \begin{pmatrix} -b_1 u_* & -c_1 u_* \\ -b_2 v_* & -c_2 v_* \end{pmatrix}$$

$$D = |\lambda I - J| = \begin{vmatrix} \lambda + b_1 u_* & c_1 u_* \\ b_2 u_* & \lambda + c_2 v_* \end{vmatrix} = \lambda^2 + (b_1 u_* + c_2 v_*)\lambda + (b_1 c_2 - b_2 c_1)u_* v_*$$

Thm. (u_*, v_*) is globally asymptotically stable in the weak competition case with $u(0), v(0) > 0$

proof: Let $E(u(t), v(t)) = b_2(u(t) - u_*) - u_* \ln \frac{u(t)}{u_*} + c_1(v(t) - v_*) - v_* \ln \frac{v(t)}{v_*}$

At (u_*, v_*) , $E(u_*, v_*) = 0$

$$\frac{\partial E}{\partial u} = b_2(1 - \frac{u_*}{u}) = 0 \quad \frac{\partial E}{\partial v} = c_1(1 - \frac{v_*}{v}) = 0, \text{ then } (u, v) = (u_*, v_*)$$

so (u_*, v_*) is the global minimal

$$\frac{dE}{dt} = b_2(u' - u_* \frac{u'}{u}) + c_1(v' - v_* \frac{v'}{v}) = \frac{b_2}{u} u'(u - u_*) + \frac{c_1}{v} v'(v - v_*)$$

$$= b_2 \underbrace{(a_1 - b_1 u - c_1 v)}_{b_1 u_* + c_1 v_* - b_1 u - c_1 v} (u - u_*) + c_1 \underbrace{(a_2 - b_2 u - c_2 v)}_{b_2 u_* + c_2 v_* - b_2 u - c_2 v} (v - v_*)$$

$$= b_1(u_* - u) + c_1(v_* - v) = b_2(u_* - u) + c_2(v_* - v)$$

$$= b_1 b_2 (u_* - u)(u - u_*) + 2b_2 c_1 (u_* - u)(v - v_*) + c_1 c_2 (v - v_*)(v_* - v)$$

$$= -[b_1 b_2 (u - u_*)^2 + 2b_2 c_1 (u - u_*)(v - v_*) + c_1 c_2 (v - v_*)^2] \leq 0$$

equality holds iff $(u, v) \equiv (u_*, v_*)$

Since $E(t)$ is strictly decreasing, then $\exists c \geq 0$ s.t. $\lim_{t \rightarrow \infty} E(t) = c$

Hence $\lim_{t \rightarrow \infty} E'(t) = 0$, so $\lim_{t \rightarrow \infty} (u(t), v(t)) = (u_*, v_*)$