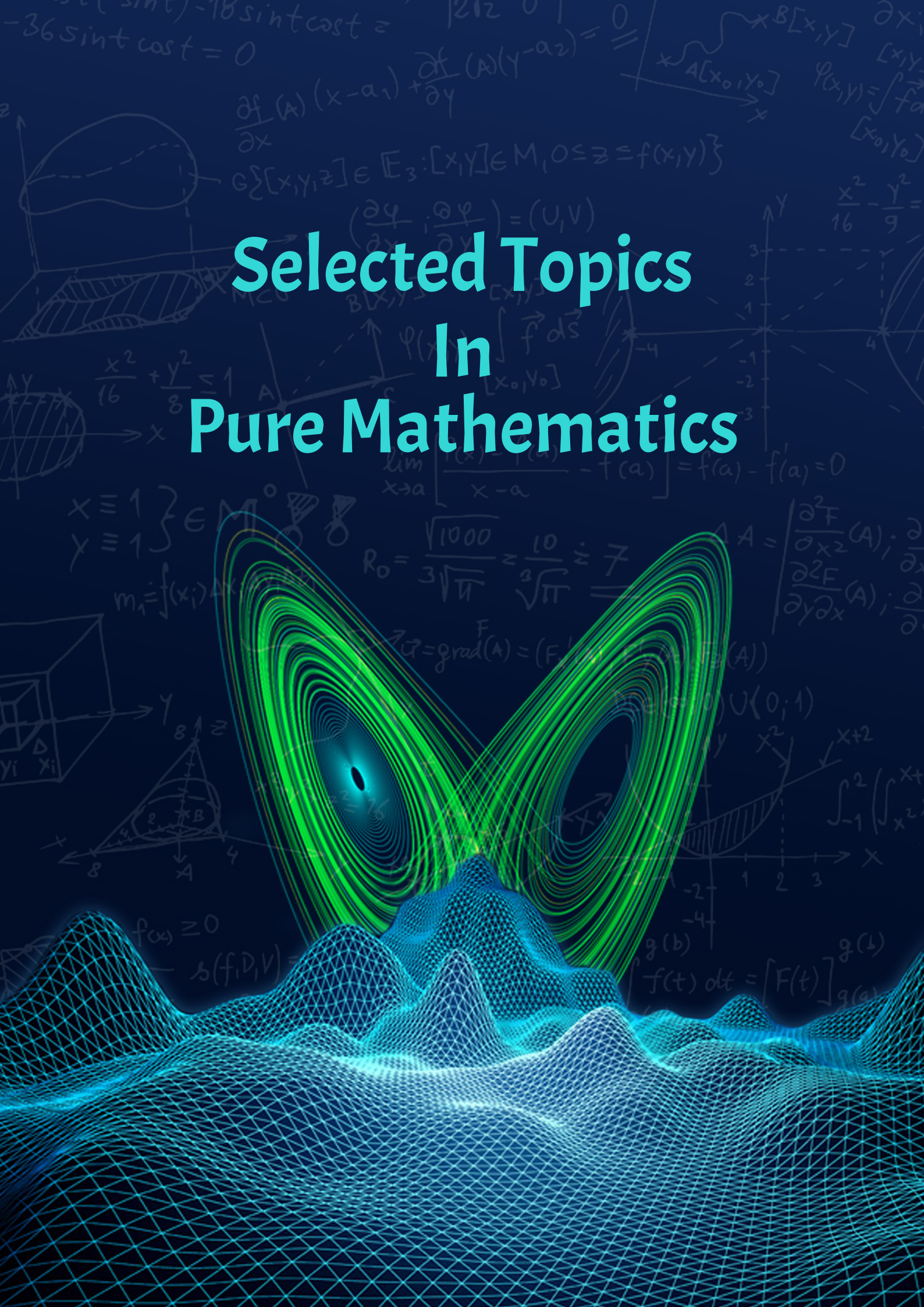


Selected Topics In Pure Mathematics





Selected Topics In Pure Mathematics

(040101401)

(Fall 2023)

By

Prof. Dr. Mahmoud M. El-Borai

Department of Mathematics and computer Sciences

Faculty of Sciences

Alexandria University

Contents

1	Fractional Calculus	3
1.1	Fractional Integral	3
1.2	Fractional Derivative	4
1.3	Laplace Transform For Fractional Integral	5
1.4	The Integral of Derivative	5
2	Stability	7
2.1	Lipschitz Condition	7
2.2	Lyapunov Function	13
3	Dynamical System And Climate Change Models	15
3.1	Kaper and Engler Climate Model	15
3.2	Adomian Decomposition Method(A.D.M)	16
4	Linear ODE System	24
4.1	Matrix and Determinant Differentiation	25

1 Fractional Calculus

1.1 Fractional Integral

Let f be a continuous function on $[a, b]$ and let I denote the integral operator

$$If(t) = \int_0^t f(s)ds = g(t) \quad , \quad t \in [a, b]$$

And if we apply it again

$$\begin{aligned} I^2 f(t) &= \int_0^t g(s)ds \\ &= \int_0^t \int_0^s f(\theta)d\theta \\ &= \int_0^t \left(\int_s^t d\theta \right) f(s)ds \\ &= \int_0^t (t-s)f(s)ds \end{aligned}$$

We can get the general formula for integrating n time by

$$\begin{aligned} I^n f(t) &= \frac{1}{(n-1)!} \int_0^t (t-s)^{n-1} f(s)ds \\ &= \frac{1}{\Gamma(n)} \int_0^t (t-s)^{n-1} f(s)ds \end{aligned}$$

Then we can say that the fractional integral of order α is defined as

$$\begin{aligned} I^\alpha f(t) &= \frac{1}{\Gamma(\alpha)} \int_0^t (t-s)^{\alpha-1} f(s)ds \\ 0 &< \alpha \leq 1 \end{aligned} \tag{1}$$

■ **Example 1.1.1** Evaluate $I^{\frac{1}{2}}(t^n)$

$$I^{\frac{1}{2}}(t^n) = \frac{1}{\Gamma(\frac{1}{2})} \int_0^t (t-s)^{-\frac{1}{2}} s^n ds$$

Let $s = t\theta \rightarrow ds = t d\theta$

$$\begin{aligned} I^{\frac{1}{2}}(t^n) &= \frac{1}{\Gamma(\frac{1}{2})} \int_0^1 (t-t\theta)^{-\frac{1}{2}} (t\theta)^n t d\theta \\ &= \frac{1}{\Gamma(\frac{1}{2})} t^{n+\frac{1}{2}} \int_0^1 (1-\theta)^{-\frac{1}{2}} (\theta)^n d\theta \\ &= \frac{1}{\Gamma(\frac{1}{2})} t^{n+\frac{1}{2}} \beta\left(\frac{1}{2}, n+1\right) \\ &= \frac{1}{\Gamma(\frac{1}{2})} t^{n+\frac{1}{2}} \frac{\Gamma(\frac{1}{2})\Gamma(n+1)}{\Gamma(n+\frac{3}{2})} \\ &= t^{n+\frac{1}{2}} \frac{\Gamma(n+1)}{\Gamma(n+\frac{3}{2})} \end{aligned}$$

■

1.2 Fractional Derivative

The fractional derivative is defined by

$$D^\alpha f(t) = \frac{1}{\Gamma(1-\alpha)} \int_0^t (t-s)^{-\alpha} \frac{df(s)}{ds} ds \quad (2)$$

Or

$$D_*^\alpha f(t) = \frac{1}{\Gamma(1-\alpha)} \frac{d}{dt} \int_0^t (t-s)^{-\alpha} f(s) ds \quad (3)$$

$$0 \leq \alpha < 1$$

Where α is the order of differentiation

The need of having 2 formulas that each has a problem that the other solves like that formula (2) need the 1st derivative to exist to get the fractional derivative and formula (3) the derivative of the constant not equal zero

$$\begin{aligned} D_*^\alpha 1 &= \frac{1}{\Gamma(1-\alpha)} \frac{d}{dt} \int_0^t (t-s)^{-\alpha} ds \\ &= \frac{-1}{\Gamma(1-\alpha)} \frac{d}{dt} \left[\frac{(t-s)^{-\alpha+1}}{-\alpha+1} \right]_0^t \\ &= \frac{-1}{\Gamma(1-\alpha)} \frac{t^{1-\alpha}}{1-\alpha} \neq 0 \end{aligned}$$

■ **Example 1.2.1** Evaluate $D^{\frac{1}{2}}(t^n)$ using the definition D_*^α

$$D^{\frac{1}{2}}(t^n) = \frac{1}{\Gamma(\frac{1}{2})} \frac{d}{dt} \int_0^t (t-s)^{-\frac{1}{2}} s^n ds$$

Let $s = t\xi \rightarrow ds = t d\xi$

$$\begin{aligned} D^{\frac{1}{2}}(t^n) &= \frac{1}{\Gamma(\frac{1}{2})} \frac{d}{dt} \int_0^1 (t-t\xi)^{-\frac{1}{2}} (t\xi)^n t d\xi \\ &= \frac{1}{\Gamma(\frac{1}{2})} \frac{d}{dt} t^{n+\frac{1}{2}} \int_0^1 (1-\xi)^{-\frac{1}{2}} (\xi)^n d\xi \\ &= \frac{1}{\Gamma(\frac{1}{2})} \frac{d}{dt} t^{n+\frac{1}{2}} \beta\left(\frac{1}{2}, n+1\right) \\ &= \left(n + \frac{1}{2}\right) \frac{\Gamma(n+1)}{\Gamma(n+\frac{3}{2})} t^{n-\frac{1}{2}} \\ &= \frac{\Gamma(n+1)}{\Gamma(n+\frac{1}{2})} t^{n-\frac{1}{2}} \end{aligned}$$

■

1.3 Laplace Transform For Fractional Integral

We know that

$$\mathcal{L}\{f(t)\} = \int_0^\infty e^{-st} f(t) dt = F(s)$$

Then we can do the following

$$\begin{aligned}\mathcal{L}\{I^\alpha f(t)\} &= \int_0^\infty e^{-st} \frac{1}{\Gamma(\alpha)} \int_0^t (t-\theta)^{\alpha-1} f(\theta) d\theta dt \\ &= \frac{1}{\Gamma(\alpha)} \mathcal{L}\{t^{\alpha-1} \times f(t)\}\end{aligned}$$

From the convolution property

$$\mathcal{L}\{I^\alpha f\} = \frac{1}{\Gamma(\alpha)} \mathcal{L}\{t^{\alpha-1}\} \times \mathcal{L}\{f(t)\} \quad (4)$$

Let's handle the first transformation

$$\begin{aligned}\mathcal{L}\{t^{\alpha-1}\} &= \int_0^\infty e^{-st} t^{\alpha-1} dt \\ \text{put } st = \eta &\implies dt = \frac{1}{s} d\eta \\ \mathcal{L}\{t^{\alpha-1}\} &= \int_0^\infty e^{-\eta} \eta^{\alpha-1} s^{1-\alpha} \frac{1}{s} d\eta \\ &= \int_0^\infty e^{-\eta} \eta^{\alpha-1} s^{-\alpha} \frac{1}{s} d\eta = s^{-\alpha} \Gamma(\alpha)\end{aligned}$$

Now substitute in equation (4)

$$\mathcal{L}\{I^\alpha f\} = s^{-\alpha} \mathcal{L}\{f(t)\} = s^{-\alpha} F(s)$$

1.4 The Integral of Derivative

Now that we defined the integral and the differential operator logically they suppose to cancel each other we need to proof that

$$I^\alpha D^\alpha f(t) = f(t)$$

Using the formula (2)

$$\begin{aligned}I^\alpha D^\alpha f(t) &= I^\alpha \left[\frac{1}{\Gamma(1-\alpha)} \int_0^t (t-s)^{-\alpha} \frac{df(s)}{ds} ds \right] \quad 0 < \alpha < 1 \\ &= \frac{1}{\Gamma(\alpha)\Gamma(1-\alpha)} \int_0^t (t-s)^{\alpha-1} \int_0^s (s-\theta)^{-\alpha} \frac{df(\theta)}{d\theta} d\theta ds \\ &= \frac{1}{\Gamma(\alpha)\Gamma(1-\alpha)} \int_0^t \underbrace{\int_\theta^t (t-s)^{\alpha-1} (s-\theta)^{-\alpha} ds}_J \frac{df(\theta)}{d\theta} d\theta\end{aligned} \quad (5)$$

Let's handle the inner integral first

$$\begin{aligned}J &= \int_\theta^t (t-s)^{\alpha-1} (s-\theta)^{-\alpha} ds \\ \text{put } s-\theta &= \eta \implies ds = d\eta \\ &= \int_0^{t-\theta} (t-\theta-\eta)^{\alpha-1} (\eta)^{-\alpha} d\eta \\ &= (t-\theta)^{\alpha-1} \int_0^{t-\theta} \left(1 - \frac{\eta}{t-\theta}\right)^{\alpha-1} (\eta)^{-\alpha} d\eta\end{aligned}$$

$$\begin{aligned}
& \text{put } \eta = (t - \theta)\xi \implies d\eta = (t - \theta)d\xi \\
& = (t - \theta)^{\alpha-1} \int_0^1 (1 - \xi)^{\alpha-1} (t - \theta)^{1-\alpha} \xi^{-\alpha} d\xi \\
& = \int_0^1 (1 - \xi)^{\alpha-1} \xi^{-\alpha} d\xi = \beta(\alpha, 1 - \alpha)
\end{aligned}$$

Substitute in (5) we get that

$$\begin{aligned}
I^\alpha D^\alpha f &= \frac{\beta(\alpha, 1 - \alpha)}{\Gamma(\alpha)\Gamma(1 - \alpha)} \int_0^t \frac{df(\theta)}{d\theta} d\theta \\
&= \frac{\Gamma(\alpha)\Gamma(1 - \alpha)}{\Gamma(\alpha + 1 - \alpha)\Gamma(\alpha)\Gamma(1 - \alpha)} [f(t) - f(0)] \\
&= f(t) - f(0)
\end{aligned}$$

2 Stability

Consider

$$\begin{cases} \frac{dx(t)}{dt} = f(x(t), t) & t > 0 \\ x(0) = a \end{cases} \quad (1)$$

We say that the solutions of equation (1) are **stable** if and only if

$$\forall \epsilon > 0, \exists \delta > 0 \text{ s.t } x(0) = a, x^*(0) = b$$

$$|a - b| < \delta \implies |x(t) - x^*(t)| \leq \epsilon$$

Where $x(t)$ and $x^*(t)$ are solutions of equation (1) moreover we say that the solutions of equation (1) are **asymptotically stable** if and only if they satisfies the previous Conditions and

$$\lim_{t \rightarrow \infty} (x(t) - x^*(t)) = 0$$

2.1 Lipschitz Condition

We say that $f(x)$ satisfies lipschitz condition with lipschitz constant N if and only if

$$|f(x) - f(x^*)| \leq N |x - x^*|$$

$$f(x) \text{ defined on } [a, b], x, x^* \in [a, b]$$

If $f(x)$ is differentiable and $f'(x)$ is bounded i.e. $|f'(x)| \leq M$

$$f(x) - f(x^*) = (x - x^*)f'(x^*)$$

$$x \leq x^* \leq x^*$$

$$|f(x) - f(x^*)| \leq M |x - x^*|$$

Theorem 2.1 Let $f(x, t)$ be a continues function on $G := \{(x, t) \mid a \leq x \leq b, 0 \leq t \leq T\}$ and Satisfies lipschitz condition with respect to x and with lipschitz constant N , i.e.

$$|f(x, t) - f(x^*, t)| \leq N |x - x^*|$$

Suppose that $\begin{cases} \frac{ds_1(t)}{dt} = f(s_1(t), t), s_1(0) = \beta_1 \\ \frac{ds_2(t)}{dt} = f(s_2(t), t), s_2(0) = \beta_2 \end{cases}$
if $|\beta_1 - \beta_2| \leq \delta$ then $|s_1(t) - s_2(t)| \leq \delta e^{Nt}$

proof.

$$s_1(t) = \beta_1 + \int_0^t f(s_1(\theta), \theta) d\theta \quad (1)$$

$$s_2(t) = \beta_2 + \int_0^t f(s_2(\theta), \theta) d\theta \quad (2)$$

Subtract equation (2) from (1)

$$s_1(t) - s_2(t) = \beta_1 - \beta_2 + \int_0^t [f(s_1(\theta), \theta) - f(s_2(\theta), \theta)] d\theta$$

Taking the absolute value to both sides

And because $|\beta_1 - \beta_2| \leq \delta$ and f satisfies lipschitz condition $|f(x, t) - f(x^*, t)| \leq N |x - x^*|$

Then

$$|s_1(t) - s_2(t)| \leq \delta + N \int_0^t |s_1(\theta) - s_2(\theta)| d\theta$$

Put $|s_1(t) - s_2(t)| = r(t)$

$$r(t) \leq \delta + N \int_0^t r(\theta) d\theta \quad (3)$$

Set $R(t) = \int_0^t r(\theta) d\theta$ i.e. $r(t) = \frac{dR(t)}{dt}$

And Substitute in (3)

$$\frac{dR(t)}{dt} - NR(t) \leq \delta \quad (4)$$

Multiply both sides by e^{-Nt}

$$e^{-Nt} \left[\frac{dR(t)}{dt} - NR(t) \right] \leq \delta e^{-Nt}$$

$$\frac{d}{dt} [e^{-Nt} R(t)] \leq \delta e^{-Nt}$$

Integrating both sides from $0 \rightarrow t$ we get that

$$e^{-Nt} R(t) - R(0) \leq \frac{\delta}{N} [1 - e^{-Nt}]$$

And we know that

$$R(t) = \int_0^t r(\theta) d\theta$$

Then

$$R(0) = \int_0^0 r(\theta) d\theta = 0$$

Therefore we get

$$R(t) \leq \frac{\delta}{N} [e^{Nt} - 1]$$

Substitute in (4) to get the following

$$r(t) \leq \delta + \delta [e^{Nt} - 1]$$

$$|s_1(t) - s_2(t)| \leq \delta e^{Nt}$$

■

Theorem 2.2 Let $f(x, t)$ be a continues function on $G := \{(x, t) \mid a \leq x \leq b, 0 \leq t \leq T\}$ and Satisfies lipschitz condition with respect to x and with lipschitz constant N , i.e.

$$|f(x, t) - f(x^*, t)| \leq N |x - x^*|$$

Suppose that

$$\frac{dx(t)}{dt} = -\gamma x(t) + f(x(t), t) \quad (1)$$

And let $s_1(t)$ and $s_2(t)$ be solutions for equation (1) corresponding to $\begin{cases} s_1(0) = \beta_1 \\ s_2(0) = \beta_2 \end{cases}$
If $\gamma > N$ and $|\beta_1 - \beta_2| \leq \delta$ then $\lim_{t \rightarrow \infty} |s_1(t) - s_2(t)| = 0$ and $|s_1(t) - s_2(t)| \leq \delta e^{-(\gamma-N)t}$

proof. Let $y(t) = e^{\gamma t} x(t)$

$$\frac{dy(t)}{dt} = e^{\gamma t} \frac{dx(t)}{dt} + \gamma e^{\gamma t} x(t)$$

Substitute $\frac{dx(t)}{dt}$ from equation (1)

$$\begin{aligned} \frac{dy(t)}{dt} &= e^{\gamma t} [-\gamma x(t) + f(x(t), t)] + \gamma e^{\gamma t} x(t) \\ &= e^{\gamma t} f(x(t), t) \\ \because y(t) &= e^{\gamma t} x(t) \\ \therefore x(t) &= e^{-\gamma t} y(t) \end{aligned}$$

Therefore

$$\frac{dy(t)}{dt} = e^{\gamma t} f(e^{-\gamma t} y(t), t) \quad (2)$$

Let $\mathbf{S}_1(t)$ and $\mathbf{S}_2(t)$ be solution of equation (2)

$$\begin{cases} \mathbf{S}_1(t) = e^{\gamma t} s_1(t), & s_1(0) = \beta_1 \\ \mathbf{S}_2(t) = e^{\gamma t} s_2(t), & s_2(0) = \beta_2 \end{cases} \quad (3)$$

$y(0) = x(0)$ i.e. $\mathbf{S}_1(t) = \beta_1$ and $\mathbf{S}_2(t) = \beta_2$

$$\mathbf{S}_1(t) = \beta_1 + \int_0^t e^{\gamma \theta} f(e^{-\gamma \theta} \mathbf{S}_1(\theta), \theta) d\theta \quad (4)$$

$$\mathbf{S}_2(t) = \beta_2 + \int_0^t e^{\gamma \theta} f(e^{-\gamma \theta} \mathbf{S}_2(\theta), \theta) d\theta \quad (5)$$

Subtract equation (5) from (4)

$$\mathbf{S}_1(t) - \mathbf{S}_2(t) = \beta_1 - \beta_2 + \int_0^t e^{\gamma \theta} [f(e^{-\gamma \theta} \mathbf{S}_1(\theta), \theta) - f(e^{-\gamma \theta} \mathbf{S}_2(\theta), \theta)] d\theta$$

Taking the absolute value to both sides

And because $|\beta_1 - \beta_2| \leq \delta$ and f satisfies lipschitz condition $|f(x, t) - f(x^*, t)| \leq N |x - x^*|$

Then

$$\begin{aligned} |\mathbf{S}_1(t) - \mathbf{S}_2(t)| &\leq \delta + N \int_0^t e^{\gamma \theta} |e^{-\gamma \theta} \mathbf{S}_1(\theta) - e^{-\gamma \theta} \mathbf{S}_2(\theta)| d\theta \\ &\leq \delta + N \int_0^t |\mathbf{S}_1(\theta) - \mathbf{S}_2(\theta)| d\theta \end{aligned}$$

Put $|\mathbf{S}_1(t) - \mathbf{S}_2(t)| d\theta = r(t)$

$$r(t) \leq \delta + N \int_0^t r(\theta) d\theta \quad (6)$$

Set $R(t) = \int_0^t r(\theta) d\theta$ i.e. $r(t) = \frac{dR(t)}{dt}$

And Substitute in (6)

$$\frac{dR(t)}{dt} - NR(t) \leq \delta \quad (7)$$

Multiply both sides by e^{-Nt}

$$e^{-Nt} \left[\frac{dR(t)}{dt} - NR(t) \right] \leq \delta e^{-Nt}$$

$$\frac{d}{dt} [e^{-Nt} R(t)] \leq \delta e^{-Nt}$$

Integrating both sides from $0 \rightarrow t$ we get that

$$R(t) \leq \frac{\delta}{N} [e^{Nt} - 1]$$

Substitute in (7) to get the following

$$r(t) \leq \delta + \delta [e^{Nt} - 1]$$

$$|\mathbf{S}_1(t) - \mathbf{S}_2(t)| \leq \delta e^{Nt}$$

Multiply both sides by $e^{-\gamma t}$

$$e^{-\gamma t} |\mathbf{S}_1(t) - \mathbf{S}_2(t)| \leq \delta e^{-(\gamma-N)t}$$

From equations (3) we get that

$$e^{-\gamma t} e^{\gamma t} |s_1(t) - s_2(t)| \leq \delta e^{-(\gamma-N)t}$$

$$|s_1(t) - s_2(t)| \leq \delta e^{-(\gamma-N)t}$$

Because $\gamma > N$ is given in the theorem then the power of R.H.S is negative therefore when $t \rightarrow \infty$ then $e^{-(\gamma-N)t} \rightarrow 0$ then

$$\lim_{t \rightarrow \infty} |s_1(t) - s_2(t)| \leq 0$$

$$\therefore \lim_{t \rightarrow \infty} |s_1(t) - s_2(t)| = 0$$

■

Rudolf Lipschitz

Rudolf Otto Sigismund Lipschitz (14 May 1832 – 7 October 1903) was a German mathematician who made contributions to mathematical analysis (where he gave his name to the Lipschitz continuity condition) and differential geometry, as well as number theory, algebras with involution and classical mechanics.



Theorem 2.3 Let A be a constant matrix suppose that all the characteristic roots of A with negative real part

Now consider the equation

$$\frac{dx(t)}{dt} = Ax(t) + f(t, x(t))$$

If $\|f(t, x(t))\| = o(\|x(t)\|)$ and $f(t, 0) = 0$, then the rest point is asymptotically stable

We can define A as
$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$
 And x as
$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

And the Norm as $\|x(t)\| = \sum_{i=1}^n |x_i(t)|$ Or $\|x(t)\| = \left(\sum_{i=1}^n |x_i(t)|^2 \right)^{\frac{1}{2}}$

And $\|A\| = \sum_{i,j=1}^n |a_{ij}|$ Or $\|A\| = \left(\sum_{i,j=1}^n |a_{ij}|^2 \right)^{\frac{1}{2}}$

And rest point is the zero solution of the equation (1)

The rest point is stable if

$$\forall \epsilon > 0, \exists \delta > 0 \text{ s.t } \|x(0)\| \leq \delta \implies \|x(t)\| \leq \epsilon$$

And it is asymptotically stable if it satisfies the last condition and

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

proof. We have

$$\frac{dx(t)}{dt} = Ax(t) + f(t, x(t)) \quad (1)$$

We can write it as

$$x(t) = e^{At}x(0) + \int_0^t e^{A(t-\theta)} f(x(\theta), \theta) d\theta \quad (2)$$

This is a representation for equation (1) to see that they are the same take the derivative of it with respect to t

$$\frac{dx(t)}{dt} = Ae^{At}x(0) + \frac{d}{dt} \int_0^t e^{A(t-\theta)} f(x(\theta), \theta) d\theta$$

Using Leibniz rule

$$\begin{aligned} \frac{dx(t)}{dt} &= Ae^{At}x(0) + A \int_0^t e^{A(t-\theta)} f(x(\theta), \theta) d\theta + f(x(t), t) \\ &= A \left(e^{At}x(0) + \int_0^t e^{A(t-\theta)} f(x(\theta), \theta) d\theta \right) + f(x(t), t) \\ &= Ax(t) + f(x(t), t) \end{aligned}$$

Leibniz rule

$$\frac{d}{dt} \int_{\alpha(t)}^{\beta(t)} f(t, \theta) d\theta = \frac{d\beta(t)}{dt} f(t, \beta(t)) - \frac{d\alpha(t)}{dt} f(t, \alpha(t)) + \int_{\alpha(t)}^{\beta(t)} \frac{\partial f(t, \theta)}{\partial t} d\theta$$

We can find $K > 0, \sigma > 0$ such that $\|e^{At}\| < Ke^{-\sigma t}$
 Take the Norm for equation (2)

$$\|x(t)\| \leq Ke^{-\sigma t}\|x(0)\| + K \int_0^t e^{-\sigma(t-\theta)} \|f(x(\theta), \theta)\| d\theta$$

And we know that $\|f(t, x(t))\| = o(\|x(t)\|)$ or in other word $\lim_{\|x(t)\| \rightarrow 0} \frac{\|f(x(t), t)\|}{\|x(t)\|} = 0$ i.e

$$\forall \epsilon > 0, \exists \delta > 0 \text{ such that } \|x(t)\| \leq \delta \implies \|f(x(t), t)\| \leq \epsilon \|x(t)\|$$

Thus

$$\|x(t)\| \leq Ke^{-\sigma t}\|x(0)\| + K\epsilon \int_0^t e^{-\sigma(t-\theta)} \|x(\theta)\| d\theta$$

Put $\epsilon = \frac{\epsilon}{K}$ and multiply by $e^{\sigma t}$

$$e^{\sigma t}\|x(t)\| \leq K\|x(0)\| + \epsilon \int_0^t e^{\sigma\theta} \|x(\theta)\| d\theta$$

As long as $\|x(t)\| \leq \delta$

$$\text{Set } R(t) = \int_0^t e^{\sigma\theta} \|x(\theta)\| d\theta \quad \text{i.e.} \quad \frac{dR(t)}{dt} = e^{\sigma t}\|x(t)\|$$

$$\begin{aligned} \frac{dR(t)}{dt} &\leq K\|x(0)\| + \epsilon R(t) \\ \frac{dR(t)}{dt} - \epsilon R(t) &\leq K\|x(0)\| \end{aligned} \tag{3}$$

Multiply by $e^{-\epsilon t}$

$$\begin{aligned} e^{-\epsilon t} \left[\frac{dR(t)}{dt} - \epsilon R(t) \right] &\leq K\|x(0)\| e^{-\epsilon t} \\ \frac{d}{dt} [e^{-\epsilon t} R(t)] &\leq K\|x(0)\| e^{-\epsilon t} \end{aligned}$$

Integrate with respect to t

$$e^{-\epsilon t} R(t) \leq \frac{K\sigma}{\epsilon} (1 - e^{-\epsilon t})$$

Multiply by $e^{\epsilon t}$

$$R(t) \leq \frac{K\sigma}{\epsilon} (e^{\epsilon t} - 1)$$

Substitute $\frac{dR(t)}{dt} = e^{\sigma t}\|x(t)\|$ and $R(t) \leq \frac{K\sigma}{\epsilon} (e^{\epsilon t} - 1)$ in equation (3)

$$\begin{aligned} e^{\sigma t}\|x(t)\| &\leq K\|x(0)\| + \epsilon R(t) \\ &\leq K\delta + K\delta e^{\epsilon t} - K\delta \\ &\leq K\delta e^{\epsilon t} \end{aligned}$$

$$\|x(t)\| \leq K\delta e^{(\epsilon-\sigma)t}$$

Put $\epsilon < \sigma$ and take the limit as $t \rightarrow \infty$

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

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

■

2.2 Lyapunov Function

Consider the Dynamical System or the Autonomous ODE

$$\frac{dx(t)}{dt} = f(x(t))$$

$$x(t) = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad f(x(t)) = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{bmatrix}$$

Theorem 2.4 — Lyapunov's theorem. Suppose that there exist a function $V(x)$, such that

$$||V(x)|| \geq 0, \forall x \text{ and } ||V(x)|| = 0 \text{ only at } x = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\frac{dV}{dt} = \sum_{i=1}^n \frac{\partial V}{\partial x_i} \frac{dx_i}{dt} = \sum_{i=1}^n \frac{\partial V}{\partial x_i} f_i(x, t) \leq 0$$

In some neighborhood of $\begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ It's supposed that $f(0) = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ then the rest point is stable

If also $\frac{dV}{dt} \leq -\beta, \beta > 0$ outside $||x(t)|| \leq \delta$

Then the rest point is asymptotically stable

■ **Example 2.2.1** Check the stability of the system

$$\begin{cases} \frac{dx}{dt} = -y - x^3 \\ \frac{dy}{dt} = x - y^3 \\ V = x^2 + y^2 \end{cases}$$

For sure $V \geq 0, \forall x, y$ and $V = 0$ only at $x = y = 0$

$$\begin{aligned} \frac{dV}{dt} &= 2x \frac{dx}{dt} + 2y \frac{dy}{dt} \\ &= 2x(-y - x^3) + 2y(x - y^3) \\ &= -2xy - 2x^4 + 2xy - 2y^4 \\ &= -2(x^4 + y^4) \end{aligned}$$

$\frac{dV}{dt}$ is negative and $< -\beta$ then the system is asymptotically stable ■

■ **Example 2.2.2** Check the stability of the system

$$\begin{cases} \frac{dx}{dt} = -xy^4 \\ \frac{dy}{dt} = yx^4 \\ V = x^4 + y^4 \end{cases}$$

For sure $V \geq 0, \forall x, y$ and $V = 0$ only at $x = y = 0$

$$\begin{aligned} \frac{dV}{dt} &= 4x^3 \frac{dx}{dt} + 4y^3 \frac{dy}{dt} \\ &= -4x^4 y^4 + 4x^4 y^4 \\ &= 0 \end{aligned}$$

Then the system is stable but not asymptotically stable ■

■ **Example 2.2.3** Check the stability of the system

$$\begin{cases} \frac{dx(t)}{dt} = -y(t) - x^3(t) + z(t) \\ \frac{dy(t)}{dt} = x(t) - y^3(t) - z(t) \\ \frac{dz(t)}{dt} = y(t) - x(t) - z^3(t) \end{cases}$$

Set Lyapunov function as following

$$V = x^2 + y^2 + z^2$$

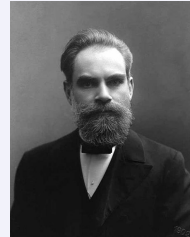
For sure $V \geq 0, \forall x, y, z$ and $V = 0$ only at $x = y = z = 0$

$$\begin{aligned} \frac{\partial V}{\partial t} &= 2x\dot{x} + 2y\dot{y} + 2z\dot{z} \\ &= 2x[-y - x^3 + z] + 2y[x - y^3 - z] + 2z[y - x - z^3] \\ &= -2xy - 2x^4 + 2xz + 2xy - 2y^4 - 2yz + 2zy - 2zx - 2z^4 \\ &= -2x^4 - 2y^4 - 2z^4 \\ \therefore \dot{V} &= -2(x^4 + y^4 + z^4) \leq 0 \end{aligned}$$

i.e. \dot{V} outside $(0, 0, 0)$ is < 0 then the system is asymptotically stable ■

Aleksandr Lyapunov

Aleksandr Mikhailovich Lyapunov was a Russian mathematician, mechanician and physicist. Lyapunov contributed to several fields, including differential equations, potential theory, dynamical systems and probability theory. His main preoccupations were the stability of equilibria and the motion of mechanical systems, especially rotating fluid masses, and the study of particles under the influence of gravity. Lyapunov's impact was significant, and the following mathematical concepts are named after him: Lyapunov equation, Lyapunov exponent, Lyapunov function, Lyapunov fractal, Lyapunov stability, Lyapunov's central limit theorem, Lyapunov vector



3 Dynamical System And Climate Change Models

Consider the following Autonomous

$$R \frac{dT(t)}{dt} = a - bT(t)$$

$$a = (1 - \alpha)Q - A \quad , \quad b = B$$

This equation represents a simple energy balance model used in climate science. This type of model is often used to study the Earth's energy budget, taking into account various factors that influence the planet's temperature over time. In this equation:

- $\frac{dT(t)}{dt}$ represents the rate of change of temperature with respect to time.
- Q represents the incoming solar radiation.
- α is the albedo, which represents the fraction of incoming solar radiation that is reflected back to space.
- R is the averaged heat capacity of the Earth/atmosphere system (heat capacity is the amount of heat required to raise the temperature of an object or substance 1 kelvin(= 1 C))
- A and B are empirically determined parameters.

Now let's try to solve it

$$\frac{dT(t)}{a - bT(t)} = \frac{1}{R} dt$$

Multiply both sides with $-b$ and integrating with respect to t

$$\int_0^t \frac{-bdT(t)}{a - bT(t)} = \frac{-bt}{R}$$

$$\ln(a - bT(t)) - \ln(a - bT(0)) = \frac{-bt}{R}$$

$$\ln\left(\frac{a - bT(t)}{a - bT(0)}\right) = \frac{-bt}{R}$$

$$a - bT(t) = (a - bT(0))e^{\frac{-bt}{R}}$$

$$T(t) = \frac{a}{b} + \frac{1}{b}(bT(0) - a)e^{\frac{-bt}{R}}$$

When taking the limit of $T(t)$ as t goes to ∞

$$\lim_{t \rightarrow \infty} T(t) = \frac{a}{b}$$

This is called the equilibrium point (or the zero solution that makes $T(t)$ constant)

3.1 Kaper and Engler Climate Model

Consider the next model

$$R \frac{dT(t)}{dt} = (1 - \alpha)Q - \sigma T^4(t) \quad 0 < \alpha < 1$$

The Kaper and Engler climate model is a simplified mathematical representation of the Earth's climate system. The model describes the rate of change of the Earth's temperature $T(t)$ over time t where :

- σ is the Stefan-Boltzmann constant, which relates the temperature of a black body (in this case, the Earth) to the amount of radiation it emits.

This equation captures two main factors influencing the Earth's temperature change:

1. Solar Radiation (First Term): The term $(1 - \alpha)Q$ represents the solar radiation absorbed by the Earth. $(1 - \alpha)$ is the fraction of incoming solar radiation that is absorbed (since α is the albedo, the fraction that is reflected), and Q represents the total incoming solar radiation.
2. Radioactive Cooling (Second Term): The term $-\sigma T^4(t)$ represents the Earth's radioactive cooling. This term describes how the Earth emits thermal radiation into space as a function of its temperature $T(t)$. According to the Stefan-Boltzmann law, the rate at which a black body radiates energy is proportional to the fourth power of its temperature.

The equilibrium point of this model is

$$(1 - \alpha)Q - \sigma T^4(t) = 0$$

$$T^4(t) = \frac{(1 - \alpha)Q}{\sigma}$$

3.2 Adomian Decomposition Method(A.D.M)

Consider the problem

$$\begin{cases} \frac{\partial u(x, t)}{\partial t} + L(u(x, t)) + N(u(x, y)) = g(x, t) \\ u(x, 0) = \phi(x) \end{cases} \quad (1)$$

This is general form of any equation where $L(u(x, t))$ is the linear part and $N(u(x, t))$ is the non-linear part

This equation can be solved by successive approximation or the method that we will discuss which is A.D.M

Now, Integrate (1) from $0 \rightarrow t$

$$u(x, t) = \phi(x) - \int_0^t L(u(x, s))ds - \int_0^t N(u(x, s))ds + \int_0^t g(x, s)ds$$

Set

$$u(x, t) = \sum_{n=0}^{\infty} u_n(x, y) \quad (2)$$

$$N(u(x, t)) = \sum_{n=0}^{\infty} A_n(x, y) \quad (3)$$

Where A_0, A_1, A_2, \dots are **Adomian polynomials** defined as:

$$A_n(x, y) = \frac{1}{n!} \frac{d^n}{d\lambda^n} \left(N \left(\sum_{j=0}^n \lambda^j u_j \right) \right) \quad (4)$$

Substitute equations (2),(3) into (4) we get

$$u(x, y) = \sum_{n=0}^{\infty} u_n(x, t) = \phi(x) + \int_0^t g(x, s)ds - \int_0^t L \sum_{n=0}^{\infty} u_n(z, t) - \int_0^t \sum_{n=0}^{\infty} A_n(x, s)ds$$

Now,

$$u_0 = \phi(x) + \int_0^t g(x, s)ds$$

$$u_1 = - \int_0^t L u_0 ds - \int_0^t A_0 ds$$

$$u_2 = - \int_0^t L u_1 ds - \int_0^t A_1 ds$$

$$\vdots$$

$$u_n = - \int_0^t L u_{n-1} ds - \int_0^t A_{n-1} ds$$

$$\therefore u(x, t) = \sum_{n=0}^{\infty} u_n(x, y) = u_0 + u_1 + u_2 + \dots$$

This will get the solution for the equation (1)

Note that The Adomian Decomposition Method (ADM) is a numerical technique used to approximate solutions of differential equations. Whether ADM converges or diverges depends on the specific problem and how it is applied. The convergence and divergence of ADM can be influenced by several factors, including the complexity of the problem, the choice of the decomposition functions, and the behavior of the nonlinear terms in the differential equation.

1. Convergence :

- ADM is more likely to converge for problems with simple nonlinear terms. If the nonlinear terms are well-behaved, ADM tends to produce more accurate results.
- Convergence is more likely when the decomposition functions are carefully chosen and match the problem's characteristics.
- For linear problems, ADM will always converge.

2. Divergence :

- ADM may diverge if the nonlinear terms in the differential equation are very complex or ill-behaved. Chaotic behavior in the nonlinear terms can lead to divergence.
- Poor choices of decomposition functions can also result in divergence.
- In some cases, ADM may exhibit slow convergence, meaning that it requires a large number of iterations to reach an acceptable solution. If the nonlinear terms in the differential equation are highly oscillatory, slow convergence may occur.

To enhance the convergence of ADM, you can try different strategies, such as refining the choice of decomposition functions, using regularization techniques, or employing other numerical methods in combination with ADM.

It's important to note that ADM is not guaranteed to converge for all problems, and its convergence behavior can be problem-dependent. Therefore, it is often recommended to perform a convergence analysis for a specific differential equation and problem setup to determine whether ADM is a suitable method for solving it. Additionally, other numerical methods, such as finite difference, finite element, or spectral methods, may be more appropriate for certain types of problems.

George Adomian

George Adomian (March 21, 1922 – June 17, 1996) was an American mathematician of Armenian descent who developed the Adomian decomposition method (ADM) for solving nonlinear differential equations, both ordinary and partial. The method is explained among other places in his book *"Solving Frontier Problems in Physics: The Decomposition Method"*. He was a faculty member at the University of Georgia (UGA) from 1966 through 1989. While at UGA, he started the Center for Applied Mathematics. Adomian was also an aerospace engineer.



■ **Example 3.2.1** Consider the nonlinear differential equation

$$\begin{cases} \frac{\partial u(x, t)}{\partial t} = x^2 - \frac{1}{4} \left(\frac{\partial u(x, t)}{\partial x} \right)^2 \\ u(x, 0) = 0 \end{cases} \quad (1)$$

Set

$$u(x, t) = \sum_{n=0}^{\infty} u_n(x, t)$$

$$N(u) = \sum_{n=0}^{\infty} A_n(x, t)$$

Where $N(u)$ represents the nonlinear form of u in our case in equation (1) $N(u) = \left(\frac{\partial u}{\partial x} \right)^2$

$$A_n(x, t) = \left[\frac{1}{n!} \frac{d^n}{d\lambda^n} N \left(\sum_{i=0}^n \lambda^i u_i(x, t) \right) \right]_{\lambda=0}$$

$$A_n(x, t) = \left[\frac{1}{n!} \frac{d^n}{d\lambda^n} \left(\sum_{i=0}^n \lambda^i \frac{\partial u_i(x, t)}{\partial x} \right)^2 \right]_{\lambda=0}$$

Integrating equation (1) from $0 \rightarrow t$

$$u(x, t) = \sum_{n=0}^{\infty} u_n(x, t) = x^2 t - \frac{1}{4} \int_0^t \sum_{n=0}^{\infty} A_n(x, \theta) d\theta$$

Now we get A_0, A_1, A_2, \dots

$$A_0(x, \theta) = \left[\sum_{i=0}^0 \lambda^i \frac{\partial u_i(x, \theta)}{\partial x} \right]_{\lambda=0} = \left(\frac{\partial u_0(x, \theta)}{\partial x} \right)^2$$

$$A_1(x, \theta) = \left[\frac{d}{d\lambda} \left(\sum_{i=0}^1 \lambda^i \frac{\partial u_i(x, \theta)}{\partial x} \right)^2 \right]_{\lambda=0}$$

$$= \left[\frac{d}{d\lambda} \left(\frac{\partial u_0(x, \theta)}{\partial x} + \lambda \frac{\partial u_1(x, \theta)}{\partial x} \right)^2 \right]_{\lambda=0}$$

$$= 2 \left[\left(\frac{\partial u_0(x, \theta)}{\partial x} + \lambda \frac{\partial u_1(x, \theta)}{\partial x} \right) \frac{\partial u_1(x, \theta)}{\partial x} \right]_{\lambda=0} = 2 \frac{\partial u_0(x, \theta)}{\partial x} \frac{\partial u_1(x, \theta)}{\partial x}$$

$$A_2(x, \theta) = \left[\frac{1}{2!} \frac{d^2}{d\lambda^2} \left(\sum_{i=0}^2 \lambda^i \frac{\partial u_i(x, \theta)}{\partial x} \right)^2 \right]_{\lambda=0}$$

$$= \left[\frac{1}{2} \frac{d^2}{d\lambda^2} \left(\frac{\partial u_0(x, \theta)}{\partial x} + \lambda \frac{\partial u_1(x, \theta)}{\partial x} + \lambda^2 \frac{\partial u_2(x, \theta)}{\partial x} \right)^2 \right]_{\lambda=0}$$

$$= \left(\frac{\partial u_1(x, \theta)}{\partial x} \right)^2 + 2 \left(\frac{\partial u_0(x, \theta)}{\partial x} \frac{\partial u_2(x, \theta)}{\partial x} \right)$$

$$A_3(x, \theta) = 2 \frac{\partial u_1(x, \theta)}{\partial x} \frac{\partial u_2(x, \theta)}{\partial x} + 2 \frac{\partial u_0(x, \theta)}{\partial x} \frac{\partial u_2(x, \theta)}{\partial x}$$

Now because

$$u_0 + u_1 + u_2 + \dots = u(x, t) = x^2 t - \frac{1}{4} [A_0 + A_1 + A_2 + \dots]$$

Then

$$u_0 = x^2 t$$

$$u_1 = -\frac{1}{4} \int_0^t A_0 d\theta = -\frac{1}{4} \int_0^t \left(\frac{\partial u_0(x, \theta)}{\partial x} \right)^2 = -\int_0^t x^2 \theta^2 d\theta = -\frac{1}{3} x^2 t^3$$

$$u_2 = \frac{2}{15} x^2 t^5, \quad u_3 = \frac{-17}{315} x^2 t^7, \quad \dots$$

$$u(x, t) = x^2 \left[t - \frac{1}{3} t^3 + \frac{2}{15} t^5 - \frac{17}{315} t^7 \dots \right] = x^2 \tanh(t)$$

■

■ **Example 3.2.2** Consider the following hyperbolic nonlinear problem

$$\begin{cases} \frac{\partial u(x, t)}{\partial t} = u(x, t) \frac{\partial u(x, t)}{\partial x} \\ u(x, 0) = \frac{x}{10} \end{cases} \quad (1)$$

The solution of (1) is given by $u(x, t) = \frac{x}{10 - t}$

Now let's use A.D.M to solve it !

First we integrate with respect to t

$$u(x, t) = u(x, 0) + \int_0^t u(x, s) \frac{\partial u(x, s)}{\partial x} ds$$

$$u(x, t) = \frac{x}{10} + \int_0^t u(x, s) \frac{\partial u(x, s)}{\partial x} ds$$

Let $u(x, t) = \sum_{n=0}^{\infty} u_n$ & $N(u) = \sum_{n=0}^{\infty} A_n$, where

$$A_n = \left[\frac{1}{n!} \frac{d^n}{d\lambda^n} \left(\sum_{j=0}^{\infty} \lambda^j u_j \right) \left(\sum_{j=0}^{\infty} \lambda^j \frac{\partial u_j}{\partial x} \right) \right]_{\lambda=0}$$

And as before we get $A_0, A_1, A_2 \dots$

$$A_0 = u_0 \frac{\partial u_0}{\partial x}$$

$$A_1 = \left[\frac{d}{d\lambda} \left(\sum_{i=0}^1 \lambda^i u_i(x, t) \right) \left(\sum_{i=0}^1 \lambda^i \frac{\partial u_i(x, t)}{\partial x} \right) \right]_{\lambda=0}$$

$$= \left[\frac{d}{d\lambda} (u_0 + \lambda u_1) \left(\frac{\partial u_0}{\partial x} + \lambda \frac{\partial u_1}{\partial x} \right) \right]_{\lambda=0}$$

$$= u_1 \frac{\partial u_0}{\partial x} + u_0 \frac{\partial u_1}{\partial x}$$

$$A_2 = u_0 \frac{\partial u_2}{\partial x} + u_1 \frac{\partial u_1}{\partial x} + u_2 \frac{\partial u_0}{\partial x}$$

$$\vdots$$

Now

$$u_0 + u_1 + u_2 + \dots = \frac{x}{10} + \int_0^t \sum_{i=0}^{\infty} A_i ds$$

Put

$$u_0 = \frac{x}{10}$$

$$u_1 = \int_0^t u_0 \frac{\partial u_0}{\partial x} ds = \frac{x}{10} \left(\frac{t}{10} \right)$$

$$u_2 = \int_0^t u_0 \frac{\partial u_1}{\partial x} + u_1 \frac{\partial u_0}{\partial x} ds = \frac{x}{10} \left(\frac{t}{10}\right)^2$$

\vdots

$$u_n = \frac{x}{10} \left(\frac{t}{10}\right)^n$$

$$\sum_{n=0}^{\infty} u_n(x, t) = \frac{x}{10} \left[1 + \left(\frac{t}{10}\right) + \left(\frac{t}{10}\right)^2 + \dots \right]$$

$$\therefore u(x, t) = \frac{x}{10} \sum_{n=0}^{\infty} \left(\frac{t}{10}\right)^n$$

Remember that the value of the geometric series is $\sum_{n=s}^{\infty} (r)^n = \frac{r^s}{1-r}$

Then in our case $\sum_{n=0}^{\infty} \left(\frac{t}{10}\right)^n = \frac{1}{1-\frac{t}{10}} = \frac{10}{10-t}$

$$\therefore u(x, t) = \frac{x}{10} \frac{10}{10-t} = \frac{x}{10-t}$$

■

■ **Example 3.2.3** Consider the nonlinear system of equations

$$\begin{cases} \frac{\partial u}{\partial t} = u \frac{\partial u}{\partial x} + v \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial t} = u \frac{\partial v}{\partial x} + v \frac{\partial v}{\partial y} \\ u(x, y, 0) = v(x, y, 0) = x + y \implies I.C \end{cases} \quad (1)$$

Integrate each equation in (1) with respect to t

$$\begin{aligned} u(x, y, t) &= x + y + \int_0^t \left(u(x, y, \theta) \frac{\partial u}{\partial x} + v(x, y, \theta) \frac{\partial u}{\partial y} \right) d\theta \\ v(x, y, t) &= x + y + \int_0^t \left(u(x, y, \theta) \frac{\partial v}{\partial x} + v(x, y, \theta) \frac{\partial v}{\partial y} \right) d\theta \end{aligned}$$

Let

$$\begin{aligned} u(x, y, t) &= \sum_{n=0}^{\infty} u_n & \& \quad v(x, y, t) = \sum_{n=0}^{\infty} v_n \\ u \frac{\partial u}{\partial x} + v \frac{\partial u}{\partial y} &= \sum_{n=0}^{\infty} A_n & \& \quad u \frac{\partial v}{\partial x} + v \frac{\partial v}{\partial y} = \sum_{n=0}^{\infty} B_n \end{aligned}$$

Where

$$\begin{aligned} A_n &= \frac{1}{n!} \frac{d^n}{d\lambda^n} \left[\left(\sum_{j=0}^n \lambda^j u_j \right) \left(\sum_{j=0}^n \lambda^j \frac{\partial u_j}{\partial x} \right) + \left(\sum_{j=0}^n \lambda^j v_j \right) \left(\sum_{j=0}^n \lambda^j \frac{\partial u_j}{\partial y} \right) \right]_{\lambda=0} \\ B_n &= \frac{1}{n!} \frac{d^n}{d\lambda^n} \left[\left(\sum_{j=0}^n \lambda^j u_j \right) \left(\sum_{j=0}^n \lambda^j \frac{\partial v_j}{\partial x} \right) + \left(\sum_{j=0}^n \lambda^j v_j \right) \left(\sum_{j=0}^n \lambda^j \frac{\partial v_j}{\partial y} \right) \right]_{\lambda=0} \end{aligned}$$

And as before we get $A_0, A_1, A_2 \dots$ and $B_0, B_1, B_2 \dots$

$$\begin{aligned} A_0 &= u_0 \frac{\partial u_0}{\partial x} + v_0 \frac{\partial u_0}{\partial y} \\ A_1 &= u_0 \frac{\partial u_1}{\partial x} + v_0 \frac{\partial u_1}{\partial y} + u_1 \frac{\partial u_0}{\partial x} + v_1 \frac{\partial u_0}{\partial y} \\ A_2 &= u_0 \frac{\partial u_2}{\partial x} + v_0 \frac{\partial u_2}{\partial y} + u_1 \frac{\partial u_1}{\partial x} + v_1 \frac{\partial u_1}{\partial y} + u_2 \frac{\partial u_0}{\partial x} + v_2 \frac{\partial u_0}{\partial y} \\ &\vdots \end{aligned}$$

Similarly

$$\begin{aligned} B_0 &= u_0 \frac{\partial v_0}{\partial x} + v_0 \frac{\partial v_0}{\partial y} \\ B_1 &= u_0 \frac{\partial v_1}{\partial x} + v_0 \frac{\partial v_1}{\partial y} + u_1 \frac{\partial v_0}{\partial x} + v_1 \frac{\partial v_0}{\partial y} \\ B_2 &= u_0 \frac{\partial v_2}{\partial x} + v_0 \frac{\partial v_2}{\partial y} + u_1 \frac{\partial v_1}{\partial x} + v_1 \frac{\partial v_1}{\partial y} + u_2 \frac{\partial v_0}{\partial x} + v_2 \frac{\partial v_0}{\partial y} \\ &\vdots \end{aligned}$$

Now, We get that

$$\begin{aligned} u_0 &= x + y & v_0 &= x + y \\ u_1 &= (x + y)(2t) & v_1 &= (x + y)(2t) \\ u_2 &= (x + y)(2t)^2 & v_2 &= (x + y)(2t)^2 \\ &\vdots & &\vdots \\ u_n &= (x + y)(2t)^n & v_n &= (x + y)(2t)^n \end{aligned}$$

Thus

$$\begin{aligned} \sum_{n=0}^{\infty} u_n &= (x + y) [1 + (2t) + (2t)^2 + \dots] \\ u(x, y, t) &= (x + y) \frac{1}{1 - 2t} = \frac{x + y}{1 - 2t} \\ \sum_{n=0}^{\infty} v_n &= (x + y) [1 + (2t) + (2t)^2 + \dots] \\ v(x, y, t) &= (x + y) \frac{1}{1 - 2t} = \frac{x + y}{1 - 2t} \end{aligned}$$

■

■ **Example 3.2.4** Consider the following system (Archana Varsoliwala & Twinkle R. Singh 2022)



$$\begin{cases} \frac{\partial m(x, t)}{\partial t} + m \frac{\partial m(x, t)}{\partial x} - f p(x, t) + g \frac{\partial \eta(x, t)}{\partial x} = 0 \\ \frac{\partial p(x, t)}{\partial t} + m \frac{\partial p(x, t)}{\partial x} + f m(x, t) = 0 \\ \frac{\partial \eta(x, t)}{\partial t} + m \frac{\partial \eta(x, t)}{\partial x} + \eta \frac{\partial \eta(x, t)}{\partial x} = 0 \end{cases} \quad (\star)$$

The problems that we has in this system are the nonlinear terms which are the terms colored in red we will work on it as we did before but on bigger scale (system of 3 equations)

Set

$$m(x, t) = \sum_{n=0}^{\infty} m_n(x, t) \quad , \quad p(x, t) = \sum_{n=0}^{\infty} p_n(x, t) \quad , \quad \eta(x, t) = \sum_{n=0}^{\infty} \eta_n(x, t)$$

And now we set the Summation that is equivalent to each nonlinear term

$$m \frac{\partial m(x, t)}{\partial x} = \sum_{n=0}^{\infty} A_n(x, t) \quad , \quad m \frac{\partial p(x, t)}{\partial x} = \sum_{n=0}^{\infty} B_n(x, t) \quad , \quad m \frac{\partial \eta(x, t)}{\partial x} + \eta \frac{\partial \eta(x, t)}{\partial x} = \sum_{n=0}^{\infty} C_n(x, t)$$

Then we get the formula of A_n, B_n, C_n

$$\begin{aligned} A_n(x, t) &= \frac{1}{n!} \frac{d^n}{d\lambda^n} \left[\left(\sum_{i=0}^n \lambda^i m_i(x, t) \right) \left(\sum_{i=0}^n \lambda^i \frac{\partial m_i(x, t)}{\partial x} \right) \right]_{\lambda=0} \\ B_n(x, t) &= \frac{1}{n!} \frac{d^n}{d\lambda^n} \left[\left(\sum_{i=0}^n \lambda^i m_i(x, t) \right) \left(\sum_{i=0}^n \lambda^i \frac{\partial p_i(x, t)}{\partial x} \right) \right]_{\lambda=0} \\ C_n(x, t) &= \frac{1}{n!} \frac{d^n}{d\lambda^n} \left[\left(\sum_{i=0}^n \lambda^i m_i(x, t) \right) \left(\sum_{i=0}^n \lambda^i \frac{\partial \eta_i(x, t)}{\partial x} \right) + \left(\sum_{i=0}^n \lambda^i \eta_i(x, t) \right) \left(\sum_{i=0}^n \lambda^i \frac{\partial \eta_i(x, t)}{\partial x} \right) \right]_{\lambda=0} \end{aligned}$$

Now, Integrate each equation in (\star) from $0 \rightarrow t$

$$\begin{aligned} m(x, t) &= m(x, 0) - \int_0^t m(x, s) \frac{\partial m(x, s)}{\partial x} ds + f \int_0^t p(x, s) ds - g \int_0^t \frac{\partial \eta(x, s)}{\partial x} ds \\ p(x, t) &= p(x, 0) - \int_0^t m(x, s) \frac{\partial p(x, s)}{\partial x} ds - f \int_0^t m(x, s) ds \\ \eta(x, t) &= \eta(x, 0) - \int_0^t m(x, s) \frac{\partial \eta(x, s)}{\partial x} ds - \int_0^t \eta(x, s) \frac{\partial \eta(x, s)}{\partial x} ds \end{aligned}$$

Now replace each nonlinear term with it's equivalent Summation

$$\begin{aligned} \sum_{n=0}^{\infty} m_n(x, t) &= m(x, 0) - \int_0^t \sum_{n=0}^{\infty} A_n(x, s) ds + f \int_0^t \sum_{n=0}^{\infty} p_n(x, s) ds - g \int_0^t \sum_{n=0}^{\infty} \frac{\partial \eta_n(x, s)}{\partial x} ds \\ \sum_{n=0}^{\infty} p_n(x, t) &= p(x, 0) - \int_0^t \sum_{n=0}^{\infty} B_n(x, s) ds - f \int_0^t \sum_{n=0}^{\infty} m_n(x, s) ds \\ \sum_{n=0}^{\infty} \eta_n(x, t) &= \eta(x, 0) - \int_0^t \sum_{n=0}^{\infty} C_n(x, s) ds \end{aligned}$$

Set initial conditions $m(x, 0) = p(x, 0) = \eta(x, 0) = x$ and $f = g = 1$
Now,

$$m_0 = x \quad p_0 = x \quad \eta_0 = x$$

$$\begin{aligned} A_0 &= m_0 \frac{\partial m_0}{\partial x} \\ A_1 &= m_1 \frac{\partial m_0}{\partial x} + m_0 \frac{\partial m_1}{\partial x} \\ &\vdots \\ B_0 &= m_0 \frac{\partial p_0}{\partial x} \\ B_1 &= m_1 \frac{\partial p_0}{\partial x} + m_0 \frac{\partial p_1}{\partial x} \\ &\vdots \\ C_0 &= m_0 \frac{\partial \eta_0}{\partial x} + \eta_0 \frac{\partial \eta_0}{\partial x} \\ C_1 &= m_1 \frac{\partial \eta_0}{\partial x} + \eta_1 \frac{\partial \eta_0}{\partial x} + m_0 \frac{\partial \eta_1}{\partial x} + \eta_0 \frac{\partial \eta_1}{\partial x} \\ &\vdots \end{aligned}$$

Substitute in system, we get

$$\begin{aligned} m_1 &= - \int_0^t A_0 ds + \int_0^t P_0 ds - \int_0^t \frac{\partial \eta_0}{\partial x} ds \\ &= - \int_0^t x.1 ds + \int_0^t x ds - \int_0^t 1 ds = -T \\ m_2 &= \dots \\ &\vdots \\ p_1 &= - \int_0^t B_0 ds - \int_0^t m_0 ds \\ &= - \int_0^t x.1 ds - \int_0^t x ds = -2xt \\ p_2 &= \dots \\ &\vdots \\ \eta_1 &= - \int_0^t c_0 ds \\ &= - \int_0^t x.1 + x.1 ds = -2xt \\ \eta_2 &= \dots \\ &\vdots \end{aligned}$$

Getting $m_1, m_2, m_3 \dots$, $p_1, p_2, p_3 \dots$, $\eta_1, \eta_2, \eta_3 \dots$ will get us the solution of the system (★) ■

4 Linear ODE System

Consider the system

$$\begin{cases} \frac{dX(t)}{dt} = A(t)X(t) \\ X(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix}, \quad A(t) = \begin{bmatrix} a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\ a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\ a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t) \end{bmatrix} \end{cases} \quad (1)$$

Where $a_{ij}(t)$ are continuous functions on $[\alpha, \beta]$

Suppose that $\mathbf{S}(t)$ is a square matrix of order n such that

$$\frac{d\mathbf{S}(t)}{dt} = A(t)\mathbf{S}(t) \quad , \quad \mathbf{S}(t) = \begin{bmatrix} s_{11}(t) & s_{12}(t) & \dots & s_{1n}(t) \\ s_{21}(t) & s_{22}(t) & \dots & s_{2n}(t) \\ \vdots & \vdots & & \vdots \\ s_{n1}(t) & s_{n2}(t) & \dots & s_{nn}(t) \end{bmatrix}$$

If the columns of $\begin{bmatrix} s_{11}(t) \\ s_{21}(t) \\ \vdots \\ s_{n1}(t) \end{bmatrix}, \begin{bmatrix} s_{12}(t) \\ s_{22}(t) \\ \vdots \\ s_{n2}(t) \end{bmatrix}, \dots, \begin{bmatrix} s_{1n}(t) \\ s_{2n}(t) \\ \vdots \\ s_{nn}(t) \end{bmatrix}$ are Linearly independent then $\mathbf{S}(t)$ is called a fundamental solution of system (1)

Theorem 4.1 The solution $\mathbf{S}(t)$ is fundamental iff $\det(\mathbf{S}(t)) \neq 0$ on the interval $[\alpha, \beta]$
i.e Linearly independent $\iff \det(\mathbf{S}(t)) \neq 0$

Theorem 4.2 If $\mathbf{S}(t)$ is a fundamental solution of system (1) and if \mathbf{C} is constant matrix such that $\det(\mathbf{C}) \neq 0$ then $\mathbf{S}(t)\mathbf{C}$ is a fundamental solution of system (1)

Proof. Set

$$f(t) = \mathbf{S}(t)\mathbf{C}$$

Where $\det(\mathbf{C}) \neq 0$

$$\begin{aligned} \det f(t) &= \det[\mathbf{S}(t)\mathbf{C}] \\ &= \det \mathbf{S}(t) \det \mathbf{C} \\ &\neq 0 \end{aligned}$$

Now we proof that it's a solution

$$\frac{df(t)}{dt} = \frac{d\mathbf{S}(t)}{dt}\mathbf{C} = A(t)\mathbf{S}(t)\mathbf{C} = A(t)f(t)$$

Then $\mathbf{S}(t)\mathbf{C}$ is a fundamental solution ■

Theorem 4.3 If $\mathbf{S}_1(t)$ and $\mathbf{S}_2(t)$ are fundamental solution of system (1) then $\mathbf{S}_2(t) = \mathbf{S}_1(t)\mathbf{C}$ where \mathbf{C} is constant matrix such that $\det(\mathbf{C}) \neq 0$

Proof. Let $\mathbf{S}_1(t)$ and $\mathbf{S}_2(t)$ be fundamental solutions
Set

$$\mathbf{S}_2(t) = \mathbf{S}_1(t)\mathbf{G}(t)$$

Then

$$\begin{aligned}\frac{d\mathbf{S}_2(t)}{dt} &= \mathbf{S}_1(t)\frac{d\mathbf{G}(t)}{dt} + \frac{d\mathbf{S}_1(t)}{dt}\mathbf{G}(t) \\ A(t)\mathbf{S}_2(t) &= \mathbf{S}_1(t)\frac{d\mathbf{G}(t)}{dt} + A(t)\mathbf{S}_1(t)\mathbf{G}(t) \\ A(t)\mathbf{S}_1(t)\mathbf{G}(t) &= \mathbf{S}_1(t)\frac{d\mathbf{G}(t)}{dt} + A(t)\mathbf{S}_1(t)\mathbf{G}(t) \\ \therefore \mathbf{S}_1(t)\frac{d\mathbf{G}(t)}{dt} &= 0\end{aligned}$$

Because $\det(\mathbf{S}_1(t)) \neq 0$ then it has inverse
Then multiply by this inverse from the left

$$\begin{aligned}\mathbf{S}_1^{-1}(t)\mathbf{S}_1(t)\frac{d\mathbf{G}(t)}{dt} &= 0 \\ \frac{d\mathbf{G}(t)}{dt} &= 0 \\ \therefore \mathbf{G}(t) &= \mathbf{C}\end{aligned}$$

Then $\mathbf{G}(t)$ is a constant matrix ■

4.1 Matrix and Determinant Differentiation

If a differentiation applied on matrix then the result will be as following

$$\frac{d}{dt} \begin{bmatrix} a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\ a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\ a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t) \end{bmatrix} = \begin{bmatrix} \dot{a}_{11}(t) & \dot{a}_{12}(t) & \dots & \dot{a}_{1n}(t) \\ \dot{a}_{21}(t) & \dot{a}_{22}(t) & \dots & \dot{a}_{2n}(t) \\ \vdots & \vdots & & \vdots \\ \dot{a}_{n1}(t) & \dot{a}_{n2}(t) & \dots & \dot{a}_{nn}(t) \end{bmatrix}$$

On the other hand if it's applied on Determinant

$$\begin{aligned}\frac{d}{dt} \begin{vmatrix} a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\ a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\ a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t) \end{vmatrix} &= \begin{vmatrix} \dot{a}_{11}(t) & \dot{a}_{12}(t) & \dots & \dot{a}_{1n}(t) \\ a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\ a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t) \end{vmatrix} + \begin{vmatrix} a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\ \dot{a}_{21}(t) & \dot{a}_{22}(t) & \dots & \dot{a}_{2n}(t) \\ \vdots & \vdots & & \vdots \\ a_{n1}(t) & a_{n2}(t) & \dots & a_{nn}(t) \end{vmatrix} + \dots \\ &+ \begin{vmatrix} a_{11}(t) & a_{12}(t) & \dots & a_{1n}(t) \\ a_{21}(t) & a_{22}(t) & \dots & a_{2n}(t) \\ \vdots & \vdots & & \vdots \\ \dot{a}_{n1}(t) & \dot{a}_{n2}(t) & \dots & \dot{a}_{nn}(t) \end{vmatrix}\end{aligned}$$

Lemma 4.4 Let $\mathbf{S}(t) = e^{\mathbf{A}t}$ where \mathbf{A} is a constant matrix then

$$\frac{d\mathbf{S}(t)}{dt} = \mathbf{A}e^{\mathbf{A}t} = \mathbf{A}\mathbf{S}(t)$$

Where $\det(e^{\mathbf{A}t}) \neq 0$ then $e^{\mathbf{A}t}$ is a fundamental solution of $\frac{dX(t)}{dt} = \mathbf{A}X(t)$

Theorem 4.5 If $\mathbf{S}(t)$ is fundamental solution of system (1) i.e

$$\frac{d\mathbf{S}(t)}{dt} = \mathbf{A}(t)\mathbf{S}(t)$$

Then $\frac{d}{dt} \det(\mathbf{S}(t)) = \text{tr}[\mathbf{A}(t)] \det(\mathbf{S}(t))$

And $\det(\mathbf{S}(t)) = \det(\mathbf{S}(t_0)) e^{\int_{t_0}^t \text{tr}[\mathbf{A}(\theta)] d\theta}$

Where $\text{tr}[\mathbf{A}(t)] = a_{11}(t) + a_{22}(t) + \dots + a_{nn}(t)$

Proof. Because

$$\frac{d\mathbf{S}(t)}{dt} = \mathbf{A}(t)\mathbf{S}(t)$$

Let

$$\mathbf{A}(t)\mathbf{S}(t) = b_{ij}(t) = \sum_{k=1}^n a_{ik}(t)s_{kj}(t)$$

Then

$$\begin{aligned} \frac{d}{dt} \det(\mathbf{S}(t)) &= \begin{vmatrix} \sum_{k=1}^n a_{1k}(t)s_{k1}(t) & \sum_{k=1}^n a_{1k}(t)s_{k2}(t) & \dots & \sum_{k=1}^n a_{1k}(t)s_{kn}(t) \\ s_{21}(t) & s_{22}(t) & \dots & s_{2n}(t) \\ \vdots & \vdots & \dots & \vdots \\ s_{n1}(t) & s_{n2}(t) & \dots & s_{nn}(t) \end{vmatrix} + \dots \\ &+ \begin{vmatrix} s_{11}(t) & s_{12}(t) & \dots & s_{1n}(t) \\ s_{21}(t) & s_{22}(t) & \dots & s_{2n}(t) \\ \vdots & \vdots & \dots & \vdots \\ \sum_{k=1}^n a_{nk}(t)s_{k1}(t) & \sum_{k=1}^n a_{nk}(t)s_{k2}(t) & \dots & \sum_{k=1}^n a_{nk}(t)s_{kn}(t) \end{vmatrix} \end{aligned}$$

Using the properties of determinants we get

$$\frac{d}{dt} \det(\mathbf{S}(t)) = a_{11} \det(\mathbf{S}(t)) + a_{22} \det(\mathbf{S}(t)) + \dots + a_{nn} \det(\mathbf{S}(t)) = \text{tr}(\mathbf{A}(t)) \det(\mathbf{S}(t))$$

Then we can write that

$$\begin{aligned} \frac{d}{dt} \det(\mathbf{S}(t)) &= \text{tr}(\mathbf{A}(t)) \det(\mathbf{S}(t)) \\ \frac{d \det(\mathbf{S}(t))}{\det(\mathbf{S}(t))} &= \text{tr}(\mathbf{A}(t)) dt \\ \ln \left(\frac{\det(\mathbf{S}(t))}{\det(\mathbf{S}(t_0))} \right) &= \int_{t_0}^t \text{tr}[\mathbf{A}(\theta)] d\theta \\ \det(\mathbf{S}(t)) &= \det(\mathbf{S}(t_0)) e^{\int_{t_0}^t \text{tr}[\mathbf{A}(\theta)] d\theta} \end{aligned}$$

■

special case

If \mathbf{A} is a constant matrix and $t_0 = 0$ then

$$\det(\mathbf{S}(t)) = \det(\mathbf{S}(0)) e^{t \times \text{tr}[\mathbf{A}]}$$

Consider the linear homogeneous system

$$\left\{ \frac{dX(t)}{dt} = \mathbf{A}X(t) \right. \quad (2)$$

We try to find a solution $X(t)$ of the form

$$X(t) = e^{\lambda t} \mathbf{v} \quad , \quad \text{Where } \mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

We notice that

$$\frac{dX(t)}{dt} = \lambda e^{\lambda t} \mathbf{v} = e^{\lambda t} \mathbf{A} \mathbf{v}$$

Hence $X(t) = e^{\lambda t} \mathbf{v}$ is a solution of system $\iff \mathbf{A} \mathbf{v} = \lambda \mathbf{v}$

A non zero vector \mathbf{v} satisfying the last equation is called **eigenvector** of matrix \mathbf{A} corresponding to **eigenvalue** λ

Now, to find $e^{\mathbf{A}}$, \mathbf{A} is of order n

$$\frac{dX(t)}{dt} = \mathbf{A}X(t)$$

we find the n^{th} solution $\begin{bmatrix} x_{11}(t) \\ x_{21}(t) \\ \vdots \\ x_{n1}(t) \end{bmatrix}, \begin{bmatrix} x_{12}(t) \\ x_{22}(t) \\ \vdots \\ x_{n2}(t) \end{bmatrix}, \dots, \begin{bmatrix} x_{1n}(t) \\ x_{2n}(t) \\ \vdots \\ x_{nn}(t) \end{bmatrix}$ and by compining them in one matrix

we get the solution

$$X(t) = e^{\mathbf{A}} = \begin{bmatrix} x_{11}(t) & x_{12}(t) & \dots & x_{1n}(t) \\ x_{21}(t) & x_{22}(t) & \dots & x_{2n}(t) \\ \vdots & \vdots & & \vdots \\ x_{n1}(t) & x_{n2}(t) & \dots & x_{nn}(t) \end{bmatrix}$$

■ **Example 4.1.1** Find $\det e^{\mathbf{A}}$ where $\mathbf{A} = \begin{bmatrix} 8 & 5 & 2 \\ 9 & 1 & 3 \\ 4 & -5 & 6 \end{bmatrix}$

$$\det e^{\mathbf{A}} = \text{tr}(\mathbf{A}) = 8 + 1 + 6 = 15$$

■

■ **Example 4.1.2** Solve the IVP

$$\begin{cases} \frac{dx(t)}{dt} = \mathbf{A}X(t) \\ A = \begin{bmatrix} 1 & 12 \\ 3 & 1 \end{bmatrix} \\ x(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{cases} \quad (3)$$

Characteristic polynomial of \mathbf{A} is given by $(\mathbf{A} - \lambda I) = 0$ which is

$$\begin{aligned} \begin{vmatrix} 1 - \lambda & 12 \\ 3 & 1 - \lambda \end{vmatrix} &= 0 \\ (1 - \lambda)^2 &= 36 \\ 1 - \lambda &= \pm 6 \\ \lambda &= -5, \quad \lambda = 7 \end{aligned}$$

At $\lambda = 7$, we seek a non zero vector \mathbf{v} such that

$$(\mathbf{A} - 7I)\mathbf{v} = \begin{bmatrix} -6 & 12 \\ 3 & -6 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

By solving this system we get

$$v_1 = 2v_2$$

Consequently, every vector $\mathbf{v} = \mathbf{c} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ is an eigen vector of \mathbf{A} corresponding to eigenvalue $\lambda = 7$
Thus,

$$X(t) = \mathbf{c}_1 e^{7t} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

Is a solution of our system

Similarly, At $\lambda = -5$

$$(\mathbf{A} + 5I)\mathbf{v} = \begin{bmatrix} 6 & 12 \\ 3 & 6 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

By solving this system we get

$$v_1 = -2v_2$$

Consequently, every vector $\mathbf{v} = \mathbf{c} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$ is an eigen vector of \mathbf{A} corresponding to eigenvalue $\lambda = -5$ Thus,

$$X(t) = \mathbf{c}_2 e^{-5t} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

Then the general solution is given by

$$X(t) = \mathbf{c}_1 e^{7t} \begin{bmatrix} 2 \\ 1 \end{bmatrix} + \mathbf{c}_2 e^{-5t} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

The constants \mathbf{c}_1 and \mathbf{c}_2 are determined from initial condition

$$X(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \mathbf{c}_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + \mathbf{c}_2 \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

We get

$$\begin{aligned} 2c_1 - 2c_2 &= 0 \\ c_1 + c_2 &= 1 \end{aligned}$$

We get that $c_1 = c_2 = \frac{1}{2}$

Thus, The general solution is given by

$$X(t) = \frac{1}{2}e^{7t} \begin{bmatrix} 2 \\ 1 \end{bmatrix} + \frac{1}{2}e^{-5t} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

■

Moreover, Evaluate $e^{\mathbf{A}}$

Remember we considered that the solution of the form $X(t) = e^{\lambda t}\mathbf{v}$ and we know that $\mathbf{A}\mathbf{v} = \lambda\mathbf{v}$
 $X(t) = e^{\mathbf{A}t}\mathbf{v}$

We can get \mathbf{v} easily from initial condition $\mathbf{v} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$$e^{\mathbf{A}t} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} e^{7t} - e^{-5t} \\ \frac{1}{2}e^{7t} + \frac{1}{2}e^{-5t} \end{bmatrix}$$

Assume $e^{\mathbf{A}t} = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$ We can deduce that

$$\begin{aligned} b_{12} &= e^{7t} - e^{-5t} \\ b_{22} &= \frac{1}{2}e^{7t} + \frac{1}{2}e^{-5t} \end{aligned}$$

If we consider a new initial condition. It is easy to obtain that

$$e^{\mathbf{A}t} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}e^{7t} + \frac{1}{2}e^{7t} \\ \frac{1}{4}e^{7t} - \frac{1}{4}e^{-5t} \end{bmatrix}$$

Thus, we get

$$\begin{aligned} b_{11} &= \frac{1}{2}e^{7t} + \frac{1}{2}e^{7t} \\ b_{21} &= \frac{1}{4}e^{7t} - \frac{1}{4}e^{-5t} \end{aligned}$$

Therefore we get that

$$e^{\mathbf{A}t} = \begin{bmatrix} \frac{1}{2}e^{7t} + \frac{1}{2}e^{7t} & e^{7t} - e^{-5t} \\ \frac{1}{4}e^{7t} - \frac{1}{4}e^{-5t} & \frac{1}{2}e^{7t} + \frac{1}{2}e^{-5t} \end{bmatrix}$$

At $t=1$

$$e^{\mathbf{A}} = \begin{bmatrix} \frac{1}{2}e^7 + \frac{1}{2}e^7 & e^7 - e^{-5} \\ \frac{1}{4}e^7 - \frac{1}{4}e^{-5} & \frac{1}{2}e^7 + \frac{1}{2}e^{-5} \end{bmatrix}$$

Selected Topics In Pure Mathematics

In this book embark on a captivating journey through the intricate web of pure mathematics, climate change models, and dynamical systems. Delve into the depths of stability theory, where the delicate balance between order and chaos shapes the world around us.

Explore the mystique of climate change models, unraveling the secrets hidden within the Earth's complex ecosystems.

At the heart of this compelling narrative are the groundbreaking methods of solving nonlinear systems, where successive approximation and the revolutionary Adomian decomposition method converge.

Witness the fusion of theory and application as mathematicians and scientists collaborate to decipher the enigmas of our changing climate.

Through the lens of rigorous mathematical analysis, this book illuminates the profound connections between stability and climate change, offering fresh perspectives on the challenges that lie ahead.

This book is a testament to the power of mathematics, revealing its crucial role in understanding the world and shaping our future. Prepare to be enthralled, enlightened, and inspired as you explore the elegant symphony of mathematics, climate science, and dynamical systems

Author :

Prof. Dr. Mahmoud M. El-Berai

Prepared by :

Ahmed Mohamed Habib & Hazem Hassam

Cover Designed by :

Hazem Hassam