

# Stability of Differential Equations

through the lens of power series

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# Outline

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- $\frac{dy}{dx} = \alpha y$
- $\frac{d}{dt}y = \alpha(t)y$
- $\frac{d}{dt}y = y^3$
- $\frac{d}{dt}y = \sin(y)$

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## Series - a quick introduction

Power series show up first in Calculus, and then again in ODE courses. Usually, they are used to solve linear ODEs like this one from Stewart:

$$y'' - 2ty' + y = 0 \quad y(0) = 0, y'(0) = 1$$

with solution

$$y(t) = t + \sum_{n=1}^{\infty} \frac{1 \cdot 5 \cdot 9 \cdot \dots \cdot (4n-3)}{(2n+1)!} t^{2n+1}$$

Ex 0:  $y' = \alpha y$

Consider

$$\frac{d}{dt}y(t) = \alpha y(t)$$

Let

$$y(t) = \sum_{n=0}^{\infty} y_n t^n,$$

plug it into the ODE and get

$$\sum_{n=0}^{\infty} (n+1)y_{n+1}t^n = \alpha \sum_{n=0}^{\infty} y_n t^n$$

and compare coefficients for each power of  $t$ , we see

$$y_{n+1} = \frac{\alpha}{n+1}y_n$$

Ex 0:  $y' = \alpha y$

Since

$$y_{n+1} = \frac{\alpha}{n+1} y_n$$

we have

$$y_1 = \frac{\alpha}{1} y_0 = \alpha y_0$$

$$y_2 = \frac{\alpha}{2} y_1 = \frac{\alpha}{2} (\alpha y_0) = \frac{\alpha^2 y_0}{2!}$$

and ...

$$y_n = \frac{\alpha^n}{n!} y_0$$

Noting that  $y_0$  is the IV, we have  $y(t) = y_0 \exp(\alpha t)$ .

## Ex 1: $y' = \alpha(t)y$

Now consider

$$\frac{d}{dt}y = \alpha(t)y \quad y(0) = y_0$$

a non-autonomous IODE, with solution

$$y(t) = y_0 \exp\left(\int_0^t \alpha(\tau) d\tau\right).$$

Assume

$$\alpha(t) = \sum_{n=0}^{\infty} a_n t^n \quad \text{and} \quad y(t) = \sum_{n=0}^{\infty} y_n t^n$$

and substitute:

$$\sum_{n=0}^{\infty} (n+1)y_{n+1}t^n = \left(\sum_{n=0}^{\infty} a_n t^n\right) \cdot \left(\sum_{n=0}^{\infty} y_n t^n\right)$$

## aside: Products

Power series are easy to add, subtract, differentiate and integrate - do it term by term.

If  $A = \sum_{n=0} a_n t^n$  and  $B = \sum_{n=0} b_n t^n$ , what is  $A \cdot B$ ?

$$\begin{aligned} & (a_0 + a_1 t + a_2 t^2 + a_3 t^3 + \dots) \cdot (b_0 + b_1 t + b_2 t^2 + b_3 t^3 + \dots) \\ &= (a_0 + b_0) + (a_0 b_1 + a_1 b_0) t + (a_0 b_2 + a_1 b_1 + a_2 b_0) t^2 \dots \\ &\quad + (a_0 b_3 + a_1 b_2 + a_2 b_1 + a_3 b_0) t^3 + \dots \\ &= \sum_{n=0} \left[ \sum_{i=0}^n a_i b_{n-i} \right] t^n \end{aligned}$$

We'll call this the **Cauchy Product**

## Ex 1: $y' = \alpha(t)y$

With MAPLE

```
>> restart:  
>> Order := 4:  
>> alpha := t -> sum(a[k]*t^k,k=0..Order):  
>> GROWTH := diff(y(t),t) = alpha(t)*y(t):  
>> Yseries := dsolve({GROWTH,y(0)=y[0]},y(t),type='series');
```

$$y(t) = y_0 + a_0 y_0 t + \left(\frac{1}{2} a_0^2 y_0 + \frac{1}{2} a_1 y_0\right) t^2 + \\ \left(\frac{1}{6} a_0^3 y_0 + \frac{1}{2} a_1 a_0 y_0 + \frac{1}{3} a_2 y_0\right) t^3 + O(t^4)$$

which we can check

```
>> SOLN1 := y[0] * exp(int(alpha(tau),tau=0..t));  
>> taylor(SOLN1,t=0);
```



## But why?

From

$$y(t) = y_0 + a_0 y_0 t + (1/2 a_0^2 y_0 + 1/2 a_1 y_0) t^2 + \\ (1/6 a_0^3 y_0 + 1/2 a_1 a_0 y_0 + 1/3 a_2 y_0) t^3 + O(t^4)$$

we can find

$$\partial_{y_0} y(t) = 1 + a_0 t + (1/2 a_0^2 + 1/2 a_1) t^2 + \\ (1/6 a_0^3 + 1/2 a_1 a_0 + 1/3 a_2) t^3 + O(t^4)$$

Sensitivity to initial conditions! Which we can verify...

```
>> Yp := taylor(diff(Yseries,y_0),t=0);
```

$$= 1 + a_0 t + (1/2 a_0^2 + 1/2 a_1) t^2 + \\ (1/6 a_0^3 + 1/2 a_1 a_0 + 1/3 a_2) t^3 + O(t^4)$$

# The Lyapunov exponent

The *Lyapunov* exponent,  $\lambda(t)$ , is the exponential growth rate measuring sensitivity to initial conditions. It is classically computed as:

$$\|\delta\mathbf{x}(t)\| \approx \exp(\lambda(t) \cdot t) \|\delta\mathbf{x}_0\|,$$

where  $\delta\mathbf{x}$  is the (first) variation of trajectory  $\mathbf{x}$ .

We can instead calculate  $\lambda(t)$  directly via:

$$\partial_{y_0} y(t) \approx \exp(\lambda(t) \cdot t)$$

# The Lyapunov exponent

We have

$$\partial_{y_0} y(t) = \sum_{n=0}^{\infty} f_n(y_0) t^n$$

and so  $\lambda(t)$  is easy to compute:

```
>> simplify(taylor(ln(Yp))/t);
```

$$\lambda(t) = a_0 + 1/2 a_1 t + 1/3 a_2 t^2 + O(t^3)$$

For our problem, a direct calculation verifies this:

$$\frac{1}{t} \int_0^t \alpha(\tau) d\tau.$$

This time average is the mean coefficient.

$$\text{Ex 2: } \frac{d}{dt}y = y^3$$

Consider

$$\frac{d}{dt}y(t) = y^3(t).$$

Now let  $v(t) = y^2$  and notice that

$$y' = y^3 = v \cdot y$$

$$v' = 2y \cdot y' = 2v^2$$

The variational problem for this would be

$$\delta \mathbf{w}' = \begin{bmatrix} v & y \\ 0 & 4v \end{bmatrix} \delta \mathbf{w}$$

from which we find the exponents to be  $\lambda_1 = \frac{1}{t} \int_0^t v(\tau)$ ,  $\lambda_2 = \frac{1}{t} \int_0^t 4v(\tau)$ .

# Stability

If  $y(0) = y_0$ , then  $v(0) = y_0^2$ , and our series solution is (from MAPLE)

$$y(t) = y_0 + y_0^3 t + 3/2 y_0^5 t^2 + 5/2 y_0^7 t^3 + O(t^4)$$

$$v(t) = y_0^2 + 2 y_0^4 t + 4 y_0^6 t^2 + 8 y_0^8 t^3 + O(t^4),$$

and we have complete stability information.

$$\partial_{y_0} y(t) = 1 + 3y_0^2 t + (15/2)y_0^4 t^2 + (35/2)y_0^6 t^3 + O(t^4)$$

with the corresponding exponent ( $\lambda(t) = \frac{1}{t} \ln(\partial_{y_0} y(t))$ ),

$$\lambda(t) = y_0^2 + y_0^4 t + (4/3)y_0^6 t^2 + O(t^3)$$

at **any point** in the flow!

### Ex 3: $y' = \sin(y)$

Consider

$$\frac{d}{dt}y(t) = \sin(y)$$

and the auxiliary variables  $u = \sin(y)$  and  $v = \cos(y)$ . Then

$$y' = 1 \cdot u$$

$$u' = v \cdot u$$

$$v' = -u \cdot u$$

We can build a series solution of  $y(t)$  and compute stability as before.

$$i.e. \quad y(t) = \sum_{n=0}^{\infty} f_n(y_0)t^n$$

# The PROCESS

Generate a polynomial (quadratic) system.(Reverse Polish)

Construct series solution  $Y(t; y_0)$  using  $(y_0, t)$ .

Construct sensitivity to ICS  $M(t; y_0) := \partial_{y_0} Y(t; y_0)$

$t =$  large value

20 WHILE  $T < T_{\max}$

Evaluate  $Y(T)$  to generate IC

Evaluate  $M(T; IC)$

Compute  $\lambda(T)$  (or mean) for local time interval

(compute local radius of convergence,  $r$ ?)

Advance:  $T = T + r$

GOTO 20

# THEORY

Carothers et. al. 2005 [2]

## Theorem

*A function  $v$  is the solution to an arbitrary component of a polynomial system of differential equations if and only if for some  $n$  there is a polynomial  $Q$  in  $n + 1$  variables so that  $Q(v, v', \dots, v^{(n)}) = 0$ .*

This implies that the motion of our moon may be described  
**without**  
reference to the earth, sun, or any other planets!



# THEORY

## Error Bound

Warne et. al. 2006 [3]

If we have (at  $a = 0$ ) a system  $\mathbf{y}'(t) = \mathbf{f}(\mathbf{y}(t))$ ,  $\mathbf{y}(0) = \mathbf{b}$ , then

$$\left\| \mathbf{y}(t) - \sum_{k=0}^n \mathbf{y}_k t^k \right\|_{\infty} \leq \frac{\|\mathbf{b}\|_{\infty} |Kt|^{n+1}}{1 - |Mt|} \quad \text{for } m \geq 2 \quad (1)$$

Where the parameters  $K$  and  $M$  depend on immediately observable quantities of the original system;

$M$  is the largest row sum of coefficients, and  $K = (m - 1)c^{m-1}$ , where  $c = \max\{1, \|\mathbf{b}\|_{\infty}\}$  and  $m = \deg(\mathbf{f})$ .

# Conclusions

- Easy to find approximate solution operator as a function of  $IV$ .
- Easy to compute stability and exponents (and spectrum?).
- No need to evolve tangent space (a la Wolf).
- Non-autonomous? No fear!
- Non-linear? No problem!

Questions??

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# References



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## For more information



James Sochacki

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