

## Wan ~~Ch.1~~ Introduction (Ch.1)

### The modeling cycle

- Identify physical problem, parameters, key variables
- Formulate initial math model (simple to start)
- Math, computational analysis
- Validation - data comparison
- Improved formulation

repeat



Hope to describe

- evolution of system
- stability
- qualitative properties (limit cycles / bifurcations)
- range of validity

add on :

- feedback
- control
- diffusion (pde)
- ~~limit~~ stochastic terms

# population modeling (Wan Ch. 2)

Let  $y(t)$  denote population time  $t$

(could be in billions or kg mass depending on problem)

- assume contin. var.

- start with a simple model (later adjust after comparisons with data)

$$\frac{dy}{dt} = f(t, y) \quad \text{1st order}$$

one hopes that population should depend on  $y$  mainly (so to start, assume autonomous)

$$\frac{dy}{dt} = f(y) \quad \text{predict world <sup>human</sup> population at a later time}$$

properties if  $y = 0$  no growth  $\Rightarrow f(0) = 0$

$$\text{Taylor's series } f(y) \approx f'(0)y + \frac{1}{2}f''(0)y^2 + \dots + \frac{1}{n!}f^{(n)}(0)y^n + \frac{1}{(n+1)!}f^{(n+1)}(c)y^{n+1} \leftarrow \text{error term}$$

where  $0 < c < y$

for small enough populations

$$f(y) = f'(0)y + \underbrace{\frac{1}{2}f''(0)y^2}_{\text{error}}$$

$$\approx \underbrace{f'(0)}_{a_0} y$$

if  $a_0 > 0 \Rightarrow$  growing population

if  $a_0 < 0 \Rightarrow$  dying population

$$\underbrace{\frac{dy}{dt} = a_0 y}_{\text{linear growth rate model}} \Rightarrow y = y_0 e^{a_0 t} = Y_0 e^{a_0 t}$$

$a_0 =$  growth rate constant

math analysis of simple population model:

$$\frac{dy}{dt} = a_0 y \quad y(0) = Y_0 \Rightarrow y = Y_0 e^{a_0 t}$$



①  $f(y) = a_0 y$ :

linear growth rate

$a_0$ : growth rate const.

$$a_0 = \frac{dy/dt}{y} = \text{"relative" growth rate}$$

ex At 10% interest rate, 1000\$ becomes 1100 after 1 year

$$\frac{\Delta y / \Delta t}{y} = \frac{100 / 1}{1000} = 0.1$$

- If  $a_0 > 0$  population grows exponentially
- ① - might not be realistic due to limited space
  - ② if  $y$  is very small, for some types of populations e.g. wolves, it does not match reality
    - $a_0 < 0$  with  $y$  small enough
- $\therefore$  probably valid for limited range of  $y$

Model validation: See how  $Y_0 e^{a_0 t}$  fits data

②. Suppose 2 data points,  $y$ : world population at time  $t$   
 $t=0$  at 2011

date	$t$	$y$
2011	0	7 (billin)
1974	-37	4

③  $t=0 \quad Y_0 e^0 = 7 \quad Y_0 = 7$

$t=-37 \quad Y_0 e^{-37a_0} = 4$

$e^{-37a_0} = 4/7$

$$a_0 = \frac{-1}{37} \ln\left(\frac{4}{7}\right)$$

Table 2.1

2.4

$$(y = 7e^{a_0 t} \quad a_0 = \frac{-1}{37} \ln(\frac{4}{7}))$$

	1804	1927	1959	1979	<del>1987</del>	1999	2011
$t_n$	-207	-84	-52	-37	-24	-12	0
$y(t_n)$	.306	1.97	3.19	<del>4</del>	4.87	5.84	7
$\bar{y}_n = \text{data}$	1	2	3	4	5	6	7

(see book)

- 7% error back to 1927 (1804 was off)

∴ would guess it is accurate for 50 years forward

Model can predict:  $y(9) =$  2020 world pop

$y(39) =$  2050 " "

when will  $p = 10$  billion?

### Improved parameter estimation

Suppose  $y(t) = Y(t, A)$  is the model.  $A = \{a_1, a_2, \dots, a_m\}$  = parameter set

(in above example,  $A = \{a_1, a_2\} = \{Y_0, a_0\}$ )

Suppose we have data  $\bar{y}_k$  at time  $t_k$   $k=1, 2, \dots, N$

and model value  $y_k = Y(t_k, A)$

residual (error) =  $E_k = Y(t_k, A) - \bar{y}_k$

Main case of interest:  $N > m$

(if  $N = m \Rightarrow$  should be a unique sol. )  
 $N < m \Rightarrow$  ~~can't~~ many sol's

$N > m$  : usually no exact sol.

Therefore minimize errors

Least squares minima. 
$$S_N^2 = \sum_{k=1}^N (y_k - \bar{y}_k)^2$$

i.e., minimize  $\| \{E_k\} \|_{l^2}^2$   
 $\| \{x_1, \dots, x_n\} \|_{l^2} = (x_1^2 + x_2^2 + \dots + x_n^2)^{1/2}$

Here,  $E_k^2 = (y_k - \bar{y}_k)^2 = (Y_0 \frac{a_0 t_k}{A_k} - \bar{y}_k)^2$

$$S_N^2 = \sum_{k=1}^N E_k^2 \quad S_N^2 \text{ minimized when } \nabla_A S_N^2 = 0$$

$$\frac{\partial S_N^2}{\partial Y_0} = \sum_{k=1}^N \frac{\partial E_k^2}{\partial Y_0} = \sum_{k=1}^N 2(Y_0 \frac{a_0 t_k}{A_k} - \bar{y}_k) \frac{a_0 t_k}{A_k} = 0$$

$$\Rightarrow Y_0 \sum_{k=1}^N A_k^2 = \sum_{k=1}^N \bar{y}_k A_k \Rightarrow Y_0 = \frac{\sum_{k=1}^N \bar{y}_k A_k}{\sum_{k=1}^N A_k^2} \quad (1)$$

$$\frac{\partial S_N^2}{\partial a_0} = \sum_{k=1}^N \frac{\partial E_k^2}{\partial a_0} = \sum_{k=1}^N 2(Y_0 \frac{a_0 t_k}{A_k} - \bar{y}_k) Y_0 t_k \frac{1}{A_k}$$

$$\Rightarrow \sum_{k=1}^N (Y_0 A_k - \bar{y}_k) t_k A_k = 0 \quad (2)$$

2 eq's, 2 unknowns. Plug (1) into (2)

(2) of the form  $f(a_0) = 0$   
 - e.g., use bisection method.