# CHE302 LECTURE IV MATHEMATICAL MODELING OF CHEMICAL PROCESS

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# THE RATIONALE FOR MATHEMATICAL MODELING

### Where to use

- To improve understanding of the process
- To train plant operating personnel
- To design the control strategy for a new process
- To select the controller setting
- To design the control law
- To optimize process operating conditions

### A Classification of Models

- Theoretical models (based on physicochemical law)
- Empirical models (based on process data analysis)
- Semi-empirical models (combined approach)

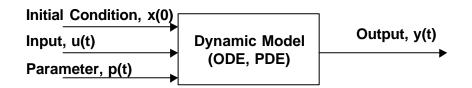
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### DYNAMIC VERSUS STEADY-STATE MODEL

### Dynamicmodel

- Describes time behavior of a process
  - Changes in input, disturbance, parameters, initial condition, etc.
- Described by a set of differential equations
   and in any (ODE) partial (PDE) differential algebra

: ordinary (ODE), partial (PDE), differential-algebraic(DAE)



### Steady-state model

- Steady state: No further changes in all variables
- No dependency in time: No transient behavior
- Can be obtained by setting the time derivative term zero

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### **MODELINGPRINCIPLES**

### Conservation law

Within a defined system boundary (control volume)

$$\begin{bmatrix} \text{rate of} \\ \text{accumulation} \end{bmatrix} = \begin{bmatrix} \text{rate of} \\ \text{input} \end{bmatrix} - \begin{bmatrix} \text{rate of} \\ \text{output} \end{bmatrix}$$
$$+ \begin{bmatrix} \text{rate of} \\ \text{generation} \end{bmatrix} - \begin{bmatrix} \text{rate of} \\ \text{disappreance} \end{bmatrix}$$

- Mass balance (overall, components)
- Energy balance
- Momentum or force balance
- Algebraic equations: relationships between variables and parameters

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### MODELING APPROACHES

#### Theoretical Model

- Follow conservation laws
- Based on physicochemical laws
- Variables and parameters have physical meaning
- Difficult to develop
- Can become quite complex
- Extrapolation is valid unless the physicochemical laws are invalid
- Used for optimization and rigorous prediction of the process behavior

#### Empirical model

- Based on the operation data
- Parameters may not have physical meaning
- Easy to develop
- Usually quite simple
- Requires well designed experimental data
- The behavior is correct only around the experimental condition
- Extrapolation is usually invalid
- Used for control design and simplified prediction model

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# **DEGREE OF FREEDOM (DOF) ANALYSIS**

### DOF

- Number of variables that can be specified independently
- $N_{\rm F} = N_{\rm V} N_{\rm E}$ 
  - $N_F$ : Degree of freedom (no. of independent variables)
  - $N_V$ : Number of variables
  - $N_E$ : Number of equations (no. of dependent variables)
  - Assume no equation can be obtained by a combination of other equations

# Solution depending on DOF

- If  $N_F = 0$ , the system is *exactly determined*. Unique solution exists.
- If  $N_F > 0$ , the system is *underdetermined*. Infinitely many solutions exist.
- If  $N_F < 0$ , the system is *overdetermined*. No solutions exist.

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### LINEAR VERSUS NONLINEAR MODELS

• Superposition principle

 $\forall a, b \in \Re$ , and for a linear operator, L

Then 
$$L(ax_1(t) + bx_2(t)) = aL(x_1(t)) + bL(x_2(t))$$

• Linear dynamic model: superposition principle holds

$$\forall \boldsymbol{a}, \boldsymbol{b} \in \Re, \ u_1(t) \to y_1(t) \text{ and } u_2(t) \to y_2(t)$$

$$\boldsymbol{a}u_1(t) + \boldsymbol{b}u_2(t) \to \boldsymbol{a}y_1(t) + \boldsymbol{b}y_2(t)$$
 $\forall \boldsymbol{a}, \boldsymbol{b} \in \Re, \ x_1(0) \to y_1(t) \text{ and } x_2(0) \to y_2(t)$ 

$$\boldsymbol{a}x_1(0) + \boldsymbol{b}x_2(0) \to \boldsymbol{a}y_1(t) + \boldsymbol{b}y_2(t)$$

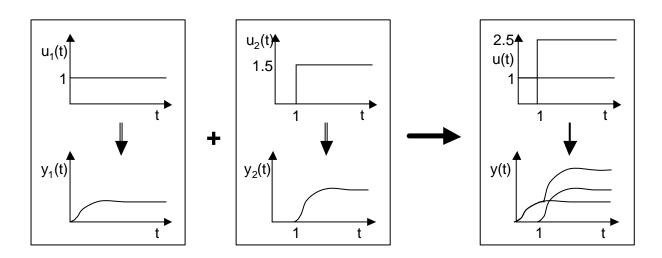
- Easy to solve and analytical solution exists.
- Usually, locally valid around the operating condition
- Nonlinear: "Not linear"
  - Usually, hard to solve and analytical solution does not exist.

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# ILLUSTRATION OF SUPERPOSITION PRINCIPLE



- Valid only for linear process
  - For example, if  $y(t)=u(t)^2$ ,  $(u_1(t)+1.5u_2(t))^2$  is not same as  $u_1(t)^2+1.5u_2(t)^2$ .

### TYPICAL LINEAR DYNAMIC MODEL

### **Linear ODE**

$$t \frac{dy(t)}{dt} = -y(t) + Ku(t)$$
 (t and K are contant, 1st order)  

$$\frac{d^{n}y(t)}{dt^{n}} + a_{n-1} \frac{d^{n-1}y(t)}{dt^{n-1}} + \dots + a_{0}y(t)$$

$$= b_{m} \frac{d^{m}u(t)}{dt^{m}} + b_{m-1} \frac{d^{m-1}u(t)}{dt^{m-1}} + \dots + b_{0}u(t)$$
 (nth order)

### **Nonlinear ODE**

$$\mathbf{t} \frac{dy(t)}{dt} = -y(t)^{2} + Ku(t) \qquad \mathbf{t} \frac{dy(t)}{dt} y(t) = -y(t)\sin(y) + Ku(t)$$

$$\mathbf{t} \frac{dy(t)}{dt} = -y(t) + K\sqrt{u(t)} \qquad \mathbf{t} \frac{dy(t)}{dt} = -e^{-y(t)} + Ku(t)$$

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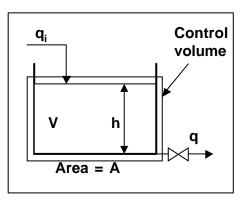
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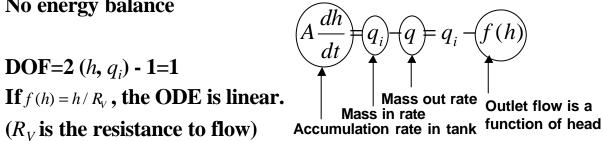
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# **MODELS OF REPRESENTATIVE PROCESSES**

# Liquid storage systems

- System boundary: storage tank
- Mass in: q<sub>i</sub> (vol. flow, indep. var)
- Mass out: q (vol, flow, dep. var)
- No generation or disappearance (no reaction or leakage)
- No energy balance
- **DOF=2**  $(h, q_i)$  1=1
- If  $f(h) = h/R_V$ , the ODE is linear.
- If  $f(h) = C_V \sqrt{rgh/g_c}$ , the ODE is nonlinear. ( $C_V$  is the valve constant)



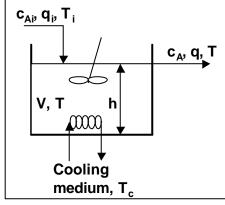


### • Continuous Stirred Tank Reactor (CSTR)

- Liquid level is constant (No acc. in tank)
- Constant density, perfect mixing
- **Reaction:**  $\mathbf{A} \rightarrow \mathbf{B}$   $(r = k_0 \exp(-E/RT)c_A)$
- System boundary: CSTR tank
- Component mass balance

$$V\frac{dc_A}{dt} = q(c_{Ai} - c_A) - Vkc_A$$

Energy balance



$$V \, \mathbf{r} C_p \, \frac{dT}{dt} = q \, \mathbf{r} C_p (T_i - T) + (-\Delta H) V k c_A + U A (T_c - T)$$

- DOF analysis
  - No. of variables:  $6(q, c_A, c_{Ai}, T_i, T, T_c)$
  - No. of equation: 2 (two dependent vars.:  $c_A$ , T)
  - DOF=6 2 = 4
  - Independent variables:  $4(q, c_{Ai}, T_i, T_c)$
  - Parameters:kineticparameters, V, U, A, otherphysical properties
  - Disturbances: any of q,  $c_{Ai}$ ,  $T_i$ ,  $T_c$ , which are not manipulatable

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### STANDARD FORM OF MODELS

### From the previous example

$$\frac{dc_A}{dt} = \frac{q}{V}(c_{Ai} - c_A) - kc_A = f_1(c_A, T, q, c_{Ai})$$

$$\frac{dT}{dt} = \frac{q}{V}(T_i - T) + \frac{q}{rC_p}(-\Delta H)kc_A + \frac{UA}{rC_p}(T_c - T) = f_2(c_A, T, q, T_c, T_i)$$

### State-space model

$$\dot{\mathbf{x}} = d\mathbf{x} / dt = \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d})$$

where 
$$\mathbf{x} = [x_1, \dots, x_n]^T$$
,  $\mathbf{u} = [u_1, \dots, u_m]^T$ ,  $\mathbf{d} = [d_1, \dots, d_l]^T$ 

- x: states,  $[c_{\Delta} T]^T$
- **u: inputs,**  $[q T_c]^T$
- **d: disturbances,**  $[c_{Ai} T_i]^T$
- y: outputs can be a function of above, y=g(x,d,u),  $[c_A T]^T$
- If higher order derivatives exist, convert them to 1st order.

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### CONVERT TO 1ST-ORDER ODE

Higher order ODE

$$\frac{d^{n}x(t)}{dt^{n}} + a_{n-1}\frac{d^{n-1}x(t)}{dt^{n-1}} + \dots + a_{0}x(t) = b_{0}u(t)$$

Define new states

$$x_1 = x$$
,  $x_2 = \dot{x}$ ,  $x_3 = \ddot{x}$ , ...,  $x_n = x^{(n-1)}$ 

A set of 1st-order ODE's

$$\dot{x}_1 = x_2 
\dot{x}_2 = x_3 
\vdots 
\dot{x}_n = -a_{n-1}x_n - a_{n-2}x_{n-1} - \dots - a_0x_1 + b_0u$$

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### SOLUTION OF MODELS

- ODE (state-space model)
  - Linear case: find the analytical solution via Laplace transform, or other methods.
  - Nonlinear case: analytical solution usually does not exist.
    - Use a numerical integration, such as <u>RK method</u>, by defining initial condition, time behavior of input/disturbance
    - Linearize around the operating condition and find the analytical solution
- PDE
  - Convert to ODE by discretization of spatial variables using <u>finite difference approximation</u> and etc.

$$\frac{\partial T_{L}}{\partial t} = -v \frac{\partial T_{L}}{\partial z} + \frac{1}{\boldsymbol{t}_{HL}} (T_{w} - T_{L}) \qquad \frac{dT_{L}(j)}{dt} = -\frac{v}{\Delta z} T_{L}(j-1) - \left(\frac{v}{\Delta z} + \frac{1}{\boldsymbol{t}_{HL}}\right) T_{L}(j) + \frac{1}{\boldsymbol{t}_{HL}} T_{w} \\ \frac{\partial T_{L}}{\partial z} \approx \frac{T_{L}(j) - T_{L}(j-1)}{\Delta z} \qquad (j=1, \dots N)$$

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### **LINEARIZATION**

# **Equilibrium (Steady state)**

- Set the derivatives as zero:  $0 = f(\overline{x}, \overline{u}, \overline{d})$
- Overbar denotes the steady-state value and  $(\bar{x}, \bar{u}, \bar{d})$  is the equilibrium point. (could be multiple)
- Solve them analytically or numerically using *Newton method*

# Linearization around equilibrium point

Taylor series expansion to 1<sup>st</sup> order

$$f(\mathbf{x}, \mathbf{u}) = f(\overline{\mathbf{x}}, \overline{\mathbf{u}}) + \frac{\partial \mathbf{f}}{\partial \mathbf{x}}\Big|_{(\overline{\mathbf{x}}, \overline{\mathbf{u}})} (\mathbf{x} - \overline{\mathbf{x}}) + \frac{\partial \mathbf{f}}{\partial \mathbf{u}}\Big|_{(\overline{\mathbf{x}}, \overline{\mathbf{u}})} (\mathbf{u} - \overline{\mathbf{u}}) + \cdots$$

$$- \text{ Ignore higher order terms}$$

$$- \text{ Define deviation variables: } \mathbf{x}' = \mathbf{x} - \overline{\mathbf{x}}, \ \mathbf{u}' = \mathbf{u} - \overline{\mathbf{u}}$$

$$\frac{\partial \mathbf{f}}{\partial \mathbf{x}} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

$$\dot{x}' = \frac{\partial f}{\partial x}\bigg|_{(\overline{x},\overline{u})} x' + \frac{\partial f}{\partial u}\bigg|_{(\overline{x},\overline{u})} u' = Ax' + Bu'$$

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