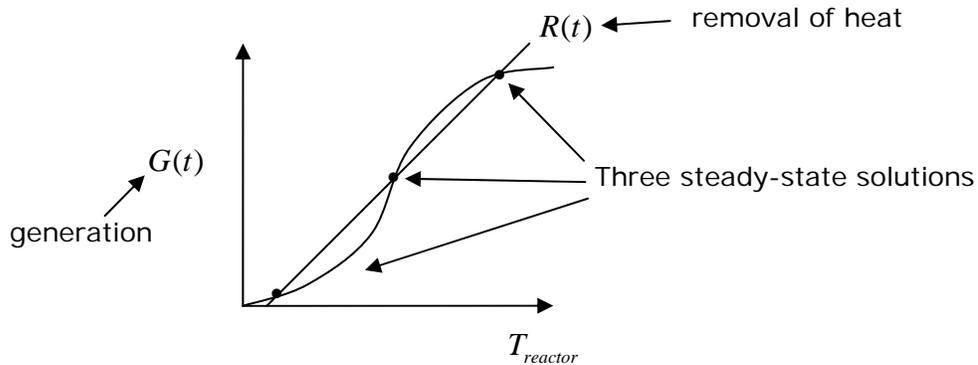


**Lecture 12: Data collection and analysis**

This lecture covers: Experimental methods for the determination of kinetic parameters of chemical and enzymatic reactions; determination of cell growth parameters; statistical analysis and model discrimination

Continuing the stability and multiple steady-state discussion from Lecture 11:



**Figure 1.** Three steady-state conditions shown on a G(T) versus T graph.

$$\begin{pmatrix} \xi_1 \\ \xi_2 \\ T \end{pmatrix}_{ss} = \underline{z}_{ss}$$

$$\begin{aligned} \frac{d\xi_1}{dt} &= 0 \\ \frac{d\xi_2}{dt} &= 0 \\ \frac{dT}{dt} &= 0 \end{aligned} \quad \rightarrow \text{steady-state}$$

$$\text{original eqns. } \left\{ \begin{aligned} \frac{d\xi_1}{dt} &= f_1(\xi_1, \xi_2, T) \\ \frac{d\xi_2}{dt} &= f_2(\xi_1, \xi_2, T) \\ \frac{dT}{dt} &= f_3(\xi_1, \xi_2, T) \end{aligned} \right\} \rightarrow \frac{d\underline{z}}{dt} = \underline{F}(\underline{z}) \quad \rightarrow \text{vector notation}$$

stability: we want any perturbation  $\delta \underline{z}$  from  $\underline{z}_{ss}$  to be self correcting

$$\text{i.e. } \frac{d}{dt}(\delta \underline{z}) = (-ve)\delta \underline{z}$$



what does perturbation cause?  
- back to steady-state or off elsewhere?

**Figure 2.** A small perturbation moves the system away from steady state. Does the system move back or does it move to elsewhere?

$$\frac{d\underline{z}}{dt} = \underline{F}(\underline{z})$$

$$\underline{z} = \underline{z}_{ss} + \delta \underline{z} \quad \delta \underline{z} = \underline{z} - \underline{z}_{ss} \quad \#$$

$$\frac{d}{dt}(\delta \underline{z}) = \frac{d\underline{z}}{dt} = \underline{F}(\delta \underline{z} + \underline{z}_{ss})$$

$$\approx \boxed{\frac{d\underline{F}}{d\underline{z}}} \delta \underline{z} + \cancel{\underline{F}(\underline{z}_{ss})} + \cancel{O(\delta \underline{z}^2)}$$

Jacobian matrix

$$\frac{d}{dt}(\delta \underline{z}) = \sum \left( \frac{dF_n}{dz_m} \right)_{\underline{z}_{ss}} \delta z_m$$

$$= J \delta \underline{z}$$

$$J = \begin{pmatrix} \frac{df_1}{d\xi_1} & \frac{df_1}{d\xi_2} & \frac{df_1}{dT} \\ \frac{df_2}{d\xi_1} & \frac{df_2}{d\xi_2} & \frac{df_2}{dT} \\ \frac{df_3}{d\xi_1} & \frac{df_3}{d\xi_2} & \frac{df_3}{dT} \end{pmatrix}$$

Jacobian

Matrix

$$\frac{d}{dt}(\delta \underline{z}) = M (\delta \underline{z})$$

if eigenvalues of  $M < 0$  then stable



$$\int r \underbrace{dx dy dz}_{dV} \rightarrow rV \text{ if homogeneous}$$

"well-stirred" reactor  
(slow reactions)

"no" conversion  
(really ~ .1% conversion)  
 $\underline{C} = \underline{C}_0 \pm .1\% \rightarrow$  can measure (output-input)  
(r barely changes)

\*need very sensitive product detection  
"differential reactor"

From data:

guess mechanism  
vary ( $\underline{k}$ ,  $\underline{k}_{eq}$ )  $\rightarrow$  make a fit

- 1) Is mechanism consistent (error bars?) w/ data?
- 2) How to regress  $\underline{k}$ ? (least squares method)