# Solutions to workshop 05: Collection and analysis of rate data

Lecture notes for chemical reaction engineering

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Try following problems from Fogler 5e (Fogler 2016). P 7-4, P 7-5, P 7-6, P 7-7, P 7-10

We will go through some of these problems in the workshop.

#### P 7-4

When arterial blood enters a tissue capillary, it exchanges oxygen and carbon dioxide with its environment, as shown in this diagram.



The kinetics of this deoxygenation of hemoglobin in blood was studied with the aid of a **tubular reactor** by Nakamura and Staub (*J. Physiol.*, 173, 161).

$$HbO_2 \xrightarrow[k_{-1}]{k_1} Hb + O_2$$

Although this is a reversible reaction, measurements were made in the initial phases of the decomposition so that the reverse reaction could be neglected. Consider a system similar to the one used by Nakamura and Staub: the solution enters a tubular reactor (0.158 cm in diameter) that has oxygen electrodes placed at 5-cm intervals down the tube. The solution flow rate into the reactor is 19.6 cm<sup>3</sup>/s with  $C_{A0} = 2.33 \times 10^{-6}$  mol/cm<sup>3</sup>.

| Electrode position                        | 1    | 2    | 3    | 4    | 5    | 6    | 7     |
|---|------|------|------|------|------|------|-------|
| Percent decomposition of HbO <sub>2</sub> | 0.00 | 1.93 | 3.82 | 5.68 | 7.48 | 9.25 | 11.00 |

- (a) Using the method of differential analysis of rate data, determine the reaction order and the forward specific reaction-rate constant  $k_1$  for the deoxygenation of hemoglobin.
- (b) Repeat using regression.

#### 💡 Solution

• Reaction is reversible, but the measurements were taken in the initial phases where the reverse reaction can be neglected

$$HbO_2 \xrightarrow{k} Hb + O_2$$

• % decomposition (conversion) along the length of PFR is given

Rate law:  $-r_A = kC^{\alpha}_A = kC^{\alpha}_{A0}(1-X)^{\alpha}$ For PFR: dX

$$F_{A0}\frac{dX}{dV} = kC^{\alpha}_{A0}(1-X)^{\alpha}$$

 $V = A_c Z; \Rightarrow dV = A_c dZ$ 

$$\frac{dX}{dZ} = \underbrace{\frac{kC^{\alpha}_{A0}A_c}{F_{A0}}}_{a} (1-X)^{\alpha}$$
$$\frac{dX}{dZ} = a(1-X)^{\alpha}$$

Taking log

$$\ln \frac{dX}{dZ} = \ln a + \alpha \ln(1-X)$$

A plot of  $\ln \frac{dX}{dZ}$  vs  $\ln(1-X)$  has a slope of  $\alpha$ 

```
import numpy as np
from scipy.stats import linregress
import matplotlib.pyplot as plt
# Data
delta_Z = 5 \# cm
C A0 = 2.33e-6 # mol/cm<sup>3</sup>
v 0 = 19.6 \ \text{# cm}^3/\text{s}
diameter = 0.158 # cm
Ac = np.pi * diameter**2/4
F_AO = C_AO * v_O
electrode_positions = np.array([1, 2, 3, 4, 5, 6, 7])
percent_decomposition = np.array([0, 1.93, 3.82, 5.68, 7.48, 9.25, 11.00])
Z = (electrode_positions - 1) *delta_Z
X = percent_decomposition/100
# fit polynomial to the data
coefficients = np.polyfit(Z, X, deg=2)
polynomial = np.poly1d(coefficients)
z_fit = np.linspace(Z.min(), Z.max(), 100)
x_fit = polynomial(z_fit)
plt.plot(Z, X, 'o', label='Experimental data')
plt.plot(z_fit, x_fit , '-', label='Polynomial fit')
plt.xlabel('Length (cm)')
plt.ylabel('Decomposition')
plt.title('Decomposition of $Hb0_2$ vs. Length')
plt.legend()
plt.grid(True)
plt.show()
```



```
# calculate the derivative of the polynomial
dxdz = np.polyder(polynomial)
# calculate ln(1-X) and ln(dxdz) --> ideally should do this with fitted data
# ln_1_minus_x = np.log(1-x_fit)
# ln_dxdz = np.log(dxdz(z_fit))
ln_1_minus_x = np.log(1-X)
ln_dxdz = np.log(dxdz(Z))
# fit line
res = linregress(ln_1_minus_x, ln_dxdz)
line = res.slope * ln_1_minus_x + res.intercept
plt.plot(ln_1_minus_x, ln_dxdz, 'o', label='Differential analysis')
plt.plot(ln_1_minus_x, line, '-', label='fitted line')
plt.annotate(
 f'Slope = {res.slope:.2e}\nIntercept = {res.intercept:.2f}\n$R^2$ = {res.rvalue**2:.
  xy=(0.5, 0.15),
  xycoords='axes fraction',
  fontsize=12
)
plt.xlabel('$\ln (1-X)$')
plt.ylabel('$\ln(\\frac{dX}{dZ})$')
plt.legend()
plt.grid(True)
plt.show()
# pick the closest round order
order = round(res.slope)
a = np.exp(res.intercept)
k = a * F_A0 / (C_A0**order * Ac)
<>:25: SyntaxWarning: invalid escape sequence '\1'
<>:26: SyntaxWarning: invalid escape sequence '\l'
<>:25: SyntaxWarning: invalid escape sequence '\l'
<>:26: SyntaxWarning: invalid escape sequence '\1'
C:\Users\Ranjeet\AppData\Local\Temp\ipykernel_28976\1237372626.py:25: SyntaxWarning: i
  plt.xlabel('$\ln (1-X)$')
C:\Users\Ranjeet\AppData\Local\Temp\ipykernel_28976\1237372626.py:26: SyntaxWarning: i
  plt.ylabel('$\ln(\\frac{dX}{dZ})$')
```



The liquid-phase irreversible reaction

$$A \longrightarrow B + C$$

is carried out in a CSTR. To learn the rate law, the volumetric flow rate,  $v_0$ , (hence  $\tau = V/v_0$ ) is varied and the effluent concentrations of species A are recorded as a function of the space time t. Pure A enters the reactor at a concentration of 2 mol/dm<sup>3</sup>. Steady-state conditions exist when the measurements are recorded.

| Run              | 1   | 2    | 3   | 4    | 5    |
|------------------|-----|------|-----|------|------|
| au (min)         | 15  | 38   | 100 | 300  | 1200 |
| $C_A (mol/dm^3)$ | 1.5 | 1.25 | 1.0 | 0.75 | 0.5  |

- (a) Determine the reaction order and specific reaction-rate constant.
- (b) If you were to repeat this experiment to determine the kinetics, what would you do differently? Would you run at a higher, lower, or the same temperature? If you were to take more data, where would you place the measurements (e.g.,  $\tau$ )?
- (c) It is believed that the technician may have made a dilution factor-of-10 error in one of the concentration measurements. What do you think? How do your answers compare using regression (Polymath or other software) with those obtained by graphical methods?

Note: All measurements were taken at steady-state conditions.

Solution

• The liquid-phase irreversible reaction

$$A \longrightarrow B + C$$

• carried out in a CSTR.

+  $\tau\, {\rm vs.}\, C_A$  data is given

Rate law:  $-r_A = k C^{\alpha}_A$  For CSTR,

$$\begin{split} V = \frac{\upsilon_0(C_{A0}-C_A)}{-r_A} \Rightarrow \tau = \frac{(C_{A0}-C_A)}{kC_A^{\alpha}} \\ \therefore kC_A^{\alpha} = \frac{(C_{A0}-C_A)}{\tau} \end{split}$$

Taking log

$$\ln k + \alpha \ln C_A = \ln \left( \frac{C_{A0} - C_A}{\tau} \right)$$

```
import numpy as np
from scipy.stats import linregress
import matplotlib.pyplot as plt
# Data
C_A0 = 2.0 \# mol/dm^3
tau = np.array([15, 38, 100, 300, 1200]) # space time in minutes
C_A = np.array([1.5, 1.25, 1.0, 0.75, 0.5]) # concentration in mol/dm<sup>3</sup>
# plot the data
plt.plot(tau, C_A, 'o-', label='concentration')
plt.xlabel('$\\tau$')
plt.ylabel('$C_A$')
plt.legend()
plt.grid(True)
plt.xlim(0,1200)
plt.show()
\ln ca = np.log(C A)
ln_ca_by_tau = np.log((C_AO - C_A) / tau)
# fit line
res = linregress(ln_ca, ln_ca_by_tau)
line = res.slope * ln_ca + res.intercept
plt.plot(ln_ca, ln_ca_by_tau, 'o', label='Differential analysis')
plt.plot(ln_ca, line, '-', label='fitted line')
plt.annotate(
  f'Slope = {res.slope:.2e}\nIntercept = {res.intercept:.2f}\n$R^2$ = {res.rvalue**2:.
  xy=(0.5, 0.15),
  xycoords='axes fraction',
  fontsize=12
)
plt.xlabel('$\ln (C_A)$')
plt.ylabel('$\ln(\\frac{C_{A0}-C_A}{\\tau})$')
plt.legend()
plt.grid(True)
plt.show()
# pick the closest round order
order = round(res.slope)
k = np.exp(res.intercept)
```



The reaction

#### $A \longrightarrow B + C$

was carried out in a constant-volume batch reactor where the following concentration measurements were recorded as a function of time.

| t (min)          | 0 | 5   | 9    | 15  | 22   | 30   | 40   | 60   |
|------------------|---|-----|------|-----|------|------|------|------|
| $C_A (mol/dm^3)$ | 2 | 1.6 | 1.35 | 1.1 | 0.87 | 0.70 | 0.53 | 0.35 |

- (a) Use nonlinear least squares (i.e., regression) and one other method to determine the reaction order,  $\alpha$ , and the specific reaction rate, k.
- (b) Nicolas Bellini wants to know, if you were to take more data, where would you place the points? Why?
- (c) Prof. Dr. Sven Köttlov from Jofostan University always asks his students, if you were to repeat this experiment to determine the kinetics, what would you do differently? Would you run at a higher, lower, or the same temperature? Take different data points? Explain.
- (d) It is believed that the technician made a dilution error in the concentration measured at 60 min.

What do you think? How do your answers compare using regression (Polymath or other software) with those obtained by graphical methods?

Solution

Reaction:

 $A \longrightarrow B + C$ 

• carried out in a constant-volume batch reactor

$$-\frac{dC_A}{dt} = kC_A^{\alpha}$$

• concentration measurements were recorded as a function of time.

```
import numpy as np
from scipy.integrate import solve_ivp
from scipy.optimize import minimize
from scipy.optimize import curve_fit
import matplotlib.pyplot as plt
# Given data
t = np.array([0, 5, 9, 15, 22, 30, 40, 60]) # min
C_A = np.array([2, 1.6, 1.35, 1.1, 0.87, 0.7, 0.53, 0.35]) # mol/dm<sup>3</sup>
C AO = C A[0]
# Define the system of ODEs
def batch_reactor(t, y, *args):
    C_A = y[0]
    k, n = args
    rate = -k*C_A**n
    dca dt = rate
    return [dca_dt]
t_span = (t.min(), t.max())
# initial conditions
y0 = [C_A0]
# Initial guesses of k and n
k = 1
n = 1
# Objective function to minimize: the difference between C_A (experimental) and C_A (m
def objective(params):
    k,n = params
    # Solve the ODE using solve_ivp
    solution = solve_ivp(
     batch_reactor,
                           # The ODE function to solve
                          # Time interval
     t span,
                       # Initial condition in a list
     y0,
      args=(k, n),
                          # Additional arguments passed to the ODE function
      dense output=True  # Generate a continuous solution
    )
    sol specific = solution.sol(t)
    C A model = solution.sol(t)[0]
    ssr = np.sum((C_A - C_A_model)**2) # Sum of squared residuals
    return ssr
```



The following data were reported [from C. N. Hinshelwood and P. J. Ackey, *Proc. R. Soc. (Lond)., A115,* 215] for a gas-phase constant-volume decomposition of dimethyl ether at 504°C in a batch reactor. Initially, only  $(CH_3)_2O$  was present.

| Time (s)              | 390 | 777 | 1195 | 3155 | $\infty$ |
|-----------------------|-----|-----|------|------|----------|
| Total Pressure (mmHg) | 408 | 488 | 562  | 799  | 931      |

- (a) Why do you think the total pressure measurement at t = 0 is missing? Can you estimate it?
- (b) Assuming that the reaction

$$(CH_3)_2O \longrightarrow CH_4 + H_2 + CO$$

is irreversible and goes virtually to completion, determine the reaction order and specific reaction rate k.

- (c) What experimental conditions would you suggest if you were to obtain more data?
- (d) How would the data and your answers change if the reaction were run at a higher temperature? A lower temperature?

## Solution

- constant volume batch reactor
- Data on pressure with time is given

$$(CH_3)_2O \longrightarrow CH_4 + H_2 + CO$$

$$\begin{split} y_{A0} &= 1\\ \delta &= 3 - 1 = 2\\ \epsilon &= y_{A0}\delta = 2\\ V &= V_0 = V_0 \frac{P_0}{P}(1 + \epsilon X)\\ P &= P_0(1 + \epsilon X)\\ X &= \frac{P - P_0}{\epsilon P_0}\\ \text{At }t &= \infty, X = 1\\ P_\infty &= P_0(1 + 2)3P_0\\ P_0 &= P_\infty/3\\ \text{Rate law:} &- r_A = kC_A^\alpha\\ \text{As pressure data is given, we need an equation for dP/dt}\\ \frac{dP/dt = d/dt(P_0(1 + \epsilon X))}{dP/dt = P_0\epsilon dX/dt} \end{split}$$

• For constant volume batch reactor

$$\begin{split} \frac{dC_A}{dt} &= -kC_A^\alpha \\ C_A &= C_{A0}^\alpha (1-X)^\alpha \\ \frac{dX}{dt} &= k(1-X)^\alpha \end{split}$$

```
import numpy as np
from scipy.integrate import solve_ivp
from scipy.optimize import minimize
from scipy.optimize import curve_fit
import matplotlib.pyplot as plt
# Given data
delta = 2
y A0 = 1
epsilon = y_A0 * delta
t = np.array([390, 777, 1195, 3155, 10000]) # s, Replace infinity with a large number
P = np.array([408, 488, 562, 799, 931]) # mmHg
PO = P[-1]/(1 + epsilon)
t_eval = t[:-1]
P eval = P[:-1]
# Define the system of ODEs
def batch_reactor(t, y, *args):
   p = y[0]
   k, n = args
   X = (p-P0)/(epsilon * P0)
    dxdt = k * (1-X) **n
    dpdt = P0 * epsilon* dxdt
    return [dpdt]
t_span = (min(t_eval), max(t_eval))
# initial conditions
y0 = [P0]
# Initial guesses of k and n
k = 6e-4
n = 1
# Objective function to minimize: the difference between C_A (experimental) and C_A (m
def objective(params):
   k,n = params
    # Solve the ODE using solve_ivp
    solution = solve_ivp(
      batch_reactor,  # The ODE function to solve
      t_span,
                          # Time interval
      y0,
                       # Initial condition in a list
                           # Additional arguments passed to the ODE function
      args=(k, n),
                           # Generate a continuous solution
      dense output=True
```



Tests were run on a small experimental reactor used for decomposing nitrogen oxides in an automobile exhaust stream. In one series of tests, a nitrogen stream containing various concentrations of  $NO_2$  was fed to a reactor, and the kinetic data obtained are shown in Figure P7-10. Each point represents one complete run. The reactor operates essentially as an isothermal backmix reactor (CSTR). What can you deduce about the apparent order of the reaction over the temperature range studied?

The plot gives the fractional decomposition of NO<sub>2</sub> fed versus the ratio of reactor volume V (in cm<sup>3</sup>) to the NO<sub>2</sub> feed rate,  $F_{NO_{2,0}}$  (g mol/h), at different feed concentrations of NO<sub>2</sub> (in parts per million by weight). Determine as many rate law parameters as you can.

Solution
$$X = \frac{\% \text{ decomposition}}{100}$$
Rate law:  $-r_A = kC_A^n$ 
•  $V/F_{A0}$  vs. X data is given
• Reaction carried out in CSTR



Figure 1: Figure P7-10 - Auto exhaust data.

$$V = \frac{F_{A0}X}{-r_A}$$
$$\frac{V}{F_{A0}} = \frac{X}{kC_A^n}$$

The linear nature of experiment at a given temperature in Figure 1 indicates that  $X \propto \frac{V}{F_{A0}}$ Therefore, apparent order must be zero ( $C_A^n = 1$ ).

$$k\frac{V}{F_{A0}} = X$$

## References

Fogler, H. Scott. 2016. Elements of Chemical Reaction Engineering. Fifth edition. Boston: Prentice Hall.