

# Biomeng 261: Lab 1 (2022)

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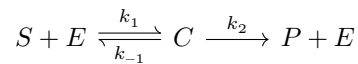
Due: **Monday 8 August 11.59pm** (*online only!*)

Create an **answer document** (e.g. in Word) and write your answers in that as you go. Please put your **name** and **ID** at the top of this document.

This lab is assessed individually, with **2 Parts worth 10 marks each** for a total of **20 marks**. These marks will be converted to **2.5% of your grade for the course**.

## Part I: Simulating ODEs [10 marks total]

In lectures we discussed the enzyme-catalysed reaction



where  $E$  is the enzyme and  $C$  is the enzyme-substrate complex formed when the substrate  $S$  binds to the enzyme's active site, forming product  $P$ . Using conservation of mass and the law of mass action we derived

$$\begin{aligned}\frac{d[S]}{dt} &= -J_1 + J_{-1} \\ \frac{d[E]}{dt} &= -J_1 + J_{-1} + J_2 \\ \frac{d[C]}{dt} &= J_1 - J_{-1} - J_2 \\ \frac{d[P]}{dt} &= J_2\end{aligned}$$

where

$$\begin{aligned}J_1 &= k_1[S][E] \\ J_{-1} &= k_{-1}[C] \\ J_2 &= k_2[C]\end{aligned}$$

In laboratory enzyme kinetics experiments, the rate of reaction is measured for different initial amounts of substrate and enzyme. Initial conditions appropriate for modeling these experiments are

$$[S](0) = S_0, \quad [E](0) = E_0, \quad [C](0) = 0, \quad [P](0) = 0.$$

Using the quasi-steady state *approximation*, we derived the Michaelis-Menten rate equation

$$v(t) = \frac{d[P]}{dt} = \frac{V_{\max}[S]}{K_M + [S]}$$

where

$$V_{\max} = k_2 E_0,$$

and  $K_M$ , the Michaelis-Menten constant, is given by

$$K_M = \frac{k_{-1} + k_2}{k_1}.$$

We are going to simulate this system with the following parameter set and initial conditions:

- $k_1 = 10 \mu M^{-1} s^{-1}$ ,  $k_{-1} = 19 s^{-1}$ ,  $k_2 = 11 s^{-1}$ .
- Initial conditions:  $[S](0) = 16 \mu M$ ,  $[E](0) = 1 \mu M$ , all other species set to  $0 \mu M$ .

The MATLAB script **Biomeng261\_L1\_Part1.m** (download from Canvas) integrates the ODEs to provide solutions  $[S]$ ,  $[E]$ ,  $[C]$  and  $[P]$  as functions of time. It also produces plots of  $[S]$  and  $[E]$ . There is also a Python version if you want to use that.

In this lab you will use and modify the MATLAB/Python script to explore the concepts of enzyme kinetic models covered in the lectures.

### Ia. Getting familiar with the script

- Check that script runs and provides a graph of  $[S]$  and  $[E]$ . Check that workspace has variables for each species. What are the end values of  $[S]$ ,  $[C]$ ,  $[E]$ ,  $[P]$ ?

[1 mark]

- Modify the script to also plot  $[C]$  and  $[P]$ . Make sure to include the correct labels. **Paste these plots into your answer document.**

[1 marks]

- Have the results reached steady state? If not modify the script to change the integration time and provide steady state values. State your integration time, the steady state values and **paste a figure into your answer document that justifies your answer.**

[1 marks]

- Does solution show that  $[E] + [C] = E_0$ ? **Create a plot to support your answer.**

[1 mark]

### Ib. Exploring assumptions

- Do results indicate that  $[C]$  reaches quasi-steady state during the simulations (with your new time range from the previous question). If so, how quickly? **Justify with a plot.** (Hints: plot the RHS of two of the ODEs on the same graph and compare. You will need to use MATLAB's `.*` operator to multiply vectors element-wise. Remember to label your plots!)

[2 marks]

- Do results agree with Michaelis-Menten kinetics? For all times? **Justify with a plot.** (Hint: you should compare an expression involving  $[C]$  with an expression involving  $[S]$ . Plot both on the same graph and remember to use `./` whenever multiplying or dividing vectors by vectors!)

[2 marks]

- Repeat the above two questions with  $E_0 = 10 \mu M$  (instead of  $E_0 = 1 \mu M$ ). **Include the two modified plots and comment on the difference.** What important dimensionless parameter has changed?

[2 marks]

## Part II [10 marks total]. Hill curves

Recall from lectures that the Hill curve relates how many apparent binding sites  $n$  an enzyme has to the reaction velocity  $v$

$$v = \frac{V_{\max}[S]^n}{K_d + [S]^n}$$

where here we denote  $K_d = K_M^n$ , i.e. for this term we have lumped the power into the definition of  $K_d$ .

The following Table shows some  $v$  and  $S$  data for an enzyme known to have **more than one cooperative binding site**. The  $V_{\max}$  is known to be  $30 \mu M/s$  and the  $K_d$  is known to be  $60 \mu M^n$ .

$[S] (\mu M)$	$v (\mu M/s)$
0.5	0.07
1.0	0.47
2.0	2.76
3.0	6.74
4.0	11.4
6.0	19.1
10	26.2
20	29.1
30	30.0

What we don't know is how many binding sites  $n$  it has.

Create a new script to answer the following questions.

### IIa. Plotting

- Plot the data points above (put reaction velocity  $v$  on the y-axis and  $[S]$  on the x-axis, and use scatter) with appropriate axis ranges etc. Note the shape of the implied curve. **Paste the resulting figure into your answer document.**

[2 marks]

Hint: modifying the following (MATLAB) commands might be helpful...

```
plot(x,y,'ko');  
xlabel('x');  
ylabel('y');  
title('my cool plot')
```

Python has similar plotting methods e.g.

```
import matplotlib.pyplot as plt  
plt.plot(x,y,'ko');  
plt.xlabel('x');  
plt.ylabel('y');  
plt.title('my cool plot')
```

### IIb. Curve fitting

We are going to try fitting a curve to these data, where the curve is defined by the Hill function above.

First, let's define a function that takes a vector of  $[S]$  values, a single  $n$  and returns a predicted velocity  $v_{\text{predicted}}$ . (We will use the  $V_{\text{max}}$  and  $K_d$  given above.). We use MATLAB's `'` operator to do vector operations. If you want to use Python then you should use numpy.

To do this, copy the following function definition into your new script (note: for MATLAB you probably need to put it at the very end of your script, after any other commands you add):

```
function v_hill = hill(S,n)
    Kd=60;
    Vmax=30;
    v_hill = Vmax*S.^n./(Kd+S.^n);
end
```

- Evaluate this function for  $n = 2$  and the given  $[S]$  data above. Paste the resulting set of values into your answer document.

[2 marks]

- Next write a simple for loop to repeat the above for  $n = 1, 2, \dots, 5$  and plot all corresponding **predicted**  $v$  values on the same graph. Then add the **measured**  $v$  values to the graph too. Use lines for predicted values and markers only for the measured values. **Paste this figure into your answer document** and state which integer  $n$  value you think appears to fit best. Do you think a non-integer  $n$  would fit better? Guess a reasonable value.

[4 marks]

We can try to **quantify** our judgement of distance or discrepancy between predicted and observed data as follows. A simple distance can be defined using the (Euclidean) norm of the difference of two vectors, e.g.

```
X = [-2 3 -1];
Y = [1 4 5];
distance = norm(X-Y)
```

Here the default norm MATLAB uses is the so-called Euclidean norm (which equals the square-root of the sum of squared differences between the elements of the vectors).

- For each integer  $n$  in the above loop, include a calculation of the **distance** (or error) between the predicted and the measured  $v$  vectors. **Paste the differences between predicted and measured**  $v$  vectors for  $n = 1, \dots, 5$  into your answer document. Do these agree with your intuition and visual inspection?

[2 marks]