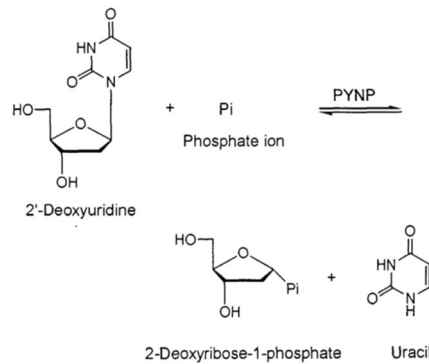


# DEOXYURIDINE METABOLISM: URACIL PRODUCTION SPEED

## INTRODUCTION

In an experiment, we constructed samples with differing concentrations of deoxyuridine in addition to a constant amount of enzyme ( $E_0 = 0.1 \mu\text{M}$ ) and the appropriate buffers. After a set period, we stopped the reaction (with sodium hydroxide in this case) and measured the amount of uracil produced via spectrophotometry.

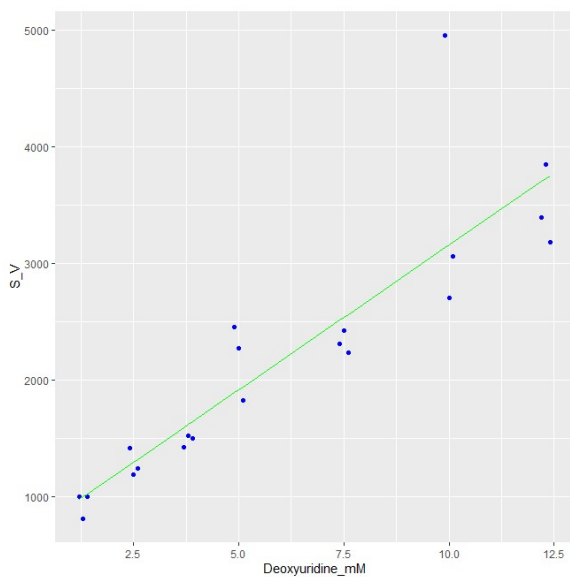


## RESULTS AND CONCLUSIONS

Applying the Michaelis-Menten model, and, using linear models in R, inference of the Michaelis-Menten constant  $K_m$  for the system, as well as the maximum uracil production  $V_{\text{max}}$ .

The starting point of the procedure is the Michaelis-Menten equation:

$$\frac{S}{V(\text{uracil})} = \frac{K_m}{V_{\text{max}}} + \frac{S}{V_{\text{max}}}$$



```
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    675.59      206.67   3.269  0.00404 **
Deoxyuridine_mM 248.00       29.02  8.545 6.21e-08 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 498 on 19 degrees of freedom
Multiple R-squared:  0.7935,    Adjusted R-squared:  0.7827 
F-statistic: 73.02 on 1 and 19 DF,  p-value: 6.212e-08

> ## Get confidence intervals for model coefficients
> confint(modelo)
              2.5 %    97.5 %
(Intercept)  243.0280 1108.1577
Deoxyuridine_mM 187.2524 308.7413
```

```
7 library(ggplot2)
8
9 ## load (mock) example data
10 table <- read.table("uracil_production.txt")
11 table$S_V <- (table$Deoxyuridine_mM) / (table$Velocity_mM_per_s)
12
13 ## Run a least squares linear regression
14 modelo <- lm(S_V ~ Deoxyuridine_mM, data = table)
15 ## Get model summary statistics
16 summary(modelo)
17 ## Get confidence intervals for model coefficients
18 confint(modelo)
19
20 ## Add extra column to the table containing fitted values
21 table$fitted_y_modelo <- fitted.values
22
23 ## make a figure using ggplot2
24 pl <- ggplot(table, aes(x = Deoxyuridine_mM, y = S_V, color = "blue")) +
25   geom_line(aes(x = Deoxyuridine_mM, y = fitted_y), color = "green")
26
27 ## Save the figure in a pdf file.
28 pdf("example_line_fit.pdf")
29 print(pl)
```

$$\text{Slope} = 248.0 \text{ s}^{-1} \rightarrow V_{\text{max}} = \text{Slope}^{-1} = 0.00403225 \text{ s}$$

$$\text{indep. term} = 675.6 \text{ s}^{-1} \rightarrow K_m = \text{i. term} * V_{\text{max}} = 0.00403225 * 675.6 = 2.72 \text{ mM}$$

Value of the catalysis constant  $k_{cat}$ :

$$V_{max} = k_{cat} * E_0 \rightarrow k_{cat} = \frac{0,00403225 \text{ s}}{0,0001 \text{ M}} = 40,32 \text{ s/mM}$$

For the case in which each molecule of complex has exactly the same probability of disintegrating originating a molecule of product or doing so giving back one molecule of substrate, values of  $k_r$  and  $k_f$ .

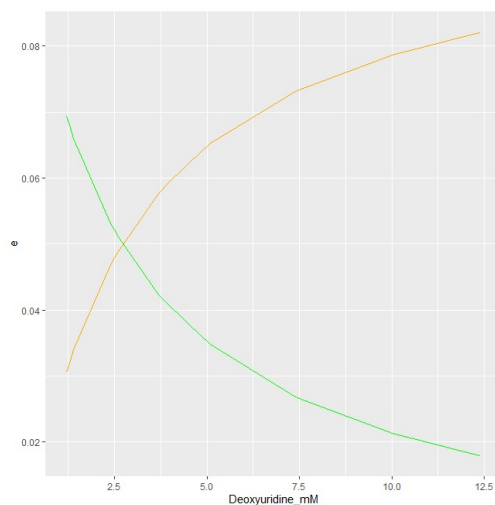
We assume  $k_r = k_f$ , then if:

$$K_M = \frac{(k_r + k_{cat})}{k_f} \rightarrow K_M = \frac{(k_f + k_{cat})}{k_f} \rightarrow k_f = \frac{k_{cat}}{K_M + 1} = \frac{40,32 \frac{\text{s}}{\text{M}}}{2,72 \text{ M} + 1} = 10,84 \text{ mM}$$

$$= k_r$$

Graphical representation of the predicted concentrations of complex and enzyme as a function of the remaining substrate:  $C(t)$  and  $E(t)$  in the range of deoxyuridine concentrations present in the input data.

$$E(t) = \frac{K_M E_0}{(K_M + S)}; C(t) = (E_0 * S)/(K_M + S)$$



```
7 library(ggplot2)
8
9 ## Load (mock) example data
10 table=read.table("uracil_production.txt")
11 table$S_V=(table$Deoxyuridine_mM)/(table$Velocity_mM_per_s)
12
13 ## Run a least squares linear regression
14 modelo=lm(S_V~Deoxyuridine_mM,data=table)
15 ## Get model summary statistics
16 summary(modelo)
17 ## Get confidence intervals for model coefficients
18 confint(modelo)
19
20 ## Add extra column to the table containing fitted values
21 table$fitted_y=modelo$fitted.values
22
23 ## make a figure using ggplot2
24 p1=ggplot(table)+geom_point(aes(x=Deoxyuridine_mM,y=S_V),color="blue")+
25 geom_line(aes(x=Deoxyuridine_mM,y=fitted_y),color="green")
26
27 ## Save the figure in a pdf file.
28 #pdf("example_linear_fit.pdf")
29 print(p1)
30 #dev.off()
31
32 table$c=(table$Deoxyuridine_mM*0.1)/(table$Deoxyuridine_mM+2.720)
33 table$e=(2.720*0.1)/(table$Deoxyuridine_mM+2.720)
34 #data=rbind(0,e,c)
35 print(table)
36
37
38 p12=ggplot(table)+
39 geom_line(aes(x=Deoxyuridine_mM,y=e),color="green")+
40 geom_line(aes(x=Deoxyuridine_mM,y=c),color="orange")
41
42
43 print(p12)
```

Algebraic representation of each of them:

$$N1 = \begin{pmatrix} -1 \\ -1 \\ 1 \\ 2 \end{pmatrix}$$

$$N2 = \begin{pmatrix} 1 & -1 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 & -1 \end{pmatrix}$$

$$N3 = (1 \quad -1 \quad -1)$$

$$N4 = \begin{pmatrix} 1 & -1 & 0 & -1 \\ 0 & 2 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$N5 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & -1 & 1 \\ 0 & 1 & -1 \end{pmatrix}$$

System 1:

$$\frac{dS1}{dt} = -V1; \frac{dS2}{dt} = -V1; \frac{dS3}{dt} = V1; \frac{dS4}{dt} = 2V1$$

System 2:

$$\frac{dS1}{dt} = V1 - V2; \frac{dS2}{dt} = V3 - V2; \frac{dS3}{dt} = V3 - V4; \frac{dS4}{dt} = V4 - V5$$

System 3:

$$\frac{dS1}{dt} = V1 - V2 - V3$$

System 4:

$$\frac{dS1}{dt} = V1 - V2 - V4; \frac{dS2}{dt} = 2V3 - V2; \frac{dS3}{dt} = V4$$

System 5:

$$\frac{dS1}{dt} = V1; \frac{dS2}{dt} = V3 - V2; \frac{dS3}{dt} = V2 - V3$$



```

55 N = matrix(rep(0,45), nrow=9, ncol=5)
56
57 N[1,]=c(1,0,0,0,-1,0,0,0,0,0)
58 N[2,]=c(0,1,0,0,-1,-1,0,0,0,0)
59 N[3,]=c(0,0,1,0,0,-2,0,0,0,0)
60 N[4,]=c(0,0,0,1,0,-1,0,0,0,0)
61 N[5,]=c(0,0,0,0,1,0,0,-1,0,0)
62 N[6,]=c(0,0,0,0,1,0,-4,0,0,0)
63 N[7,]=c(0,0,0,0,1,-1,0,0,0,0)
64 N[8,]=c(0,0,0,0,1,0,0,-1,0,0)
65 N[9,]=c(0,0,0,0,0,1,0,0,-1,0)
66 N[10,]=c(0,1,1,0,0,0,0,0,0,0)
67 N[11,]=c(1,0,0,0,0,0,0,0,0,0)
68 N[11,]=c(0,1,0,0,0,0,0,0,0,0)
69 N[13,]=c(0,0,1,0,0,0,0,0,0,0)
70 N[14,]=c(0,0,0,1,0,0,0,0,0,0)
71 N[15,]=c(0,0,0,0,1,0,0,0,0,0)
72 N[16,]=c(0,0,0,0,1,0,0,0,0,0)
73 N[17,]=c(0,0,0,0,0,1,0,0,0,0)
74 N[18,]=c(0,0,0,0,0,0,1,0,0,0)
75 N[19,]=c(0,0,0,0,0,0,0,1,0,0)
76 N[20,]=c(0,0,0,0,0,0,0,0,1,0)
77
78 constraints_matrix=N
79 constraints_direction=c("=", "=", "=", "=", "=", "=", "=", "=", "<=", ">=", ">=", ">=", ">=", ">=", ">=")
80
81 constraints_values=c(0,0,0,0,0,0,0,0,0,0,56,0,0,0,0,0,0,0,0,0,0)
82
83 objective=c(0,0,0,0,0,0,0,0,0,0)
84
85
86
87 optimum <- lp(direction="max",
88               objective.in = objective,
89               const.mat = constraints_matrix,
90               const.dir = constraints_direction,
91               const.rhs = constraints_values)
92
93 optimum$objval
94 optimum$solution
95
96 table<-data.frame(constraints_matrix)
97 table$signo=constraints_direction
98 table$res=constraints_values

```

```

> optimum$objval
[1] 8
> optimum$solution
[1] 32 40 16 8 32 8 8 32 8 8

```

The output indicates the respective optimal velocities for maximum V10 which is 8 mM/s.