

Analyzing the Relationship Between the Concentration of KIO_3 and the Speed of the NaHSO_3 -based Iodine Clock Reaction

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1 Introduction

1.1 Purpose

1.2 Hypothesis

Reaction speed will decrease (increasing the time taken to react) as KIO_3 concentration decreases, as there will be fewer KIO_3 molecules available to react at any one time.

1.3 Variables

Independent Variable

- KIO_3 concentration [M]

Dependent Variable

- Reaction speed [s]

Controlled Variables

- Molarity of NaHSO_3 solution [0.020 M]
- Volume of KIO_3 solution [10.000 mL]
- Volume of NaHSO_3 solution [10.000 mL]
- Temperature of solutions [25.200 °C]
- Distilled Water

2 Materials

- 75 mL 0.020 M KIO_3
- 100 mL 0.020 M NaHSO_3
- 9 · test tubes

- 2 · 100 mL beaker
- 2 · 10 mL graduated cylinder
- 1 · 2 mL pipette
- 1 · timer
- 1 · thermometer

3 Procedure

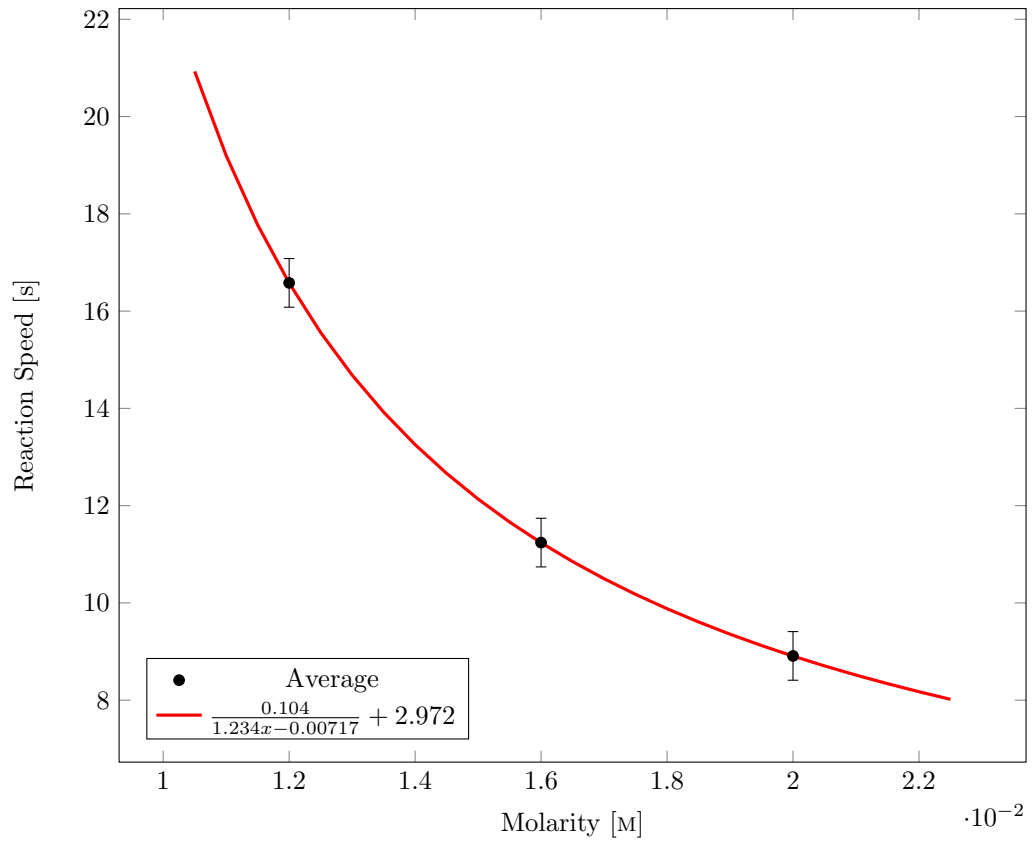
1. Pour KIO_3 and NaHSO_3 solutions into beakers
2. Measure 10 mL KIO_3 solution in graduated cylinder
3. Measure 10 mL NaHSO_3 solution in graduated cylinder
4. If KIO_3 solution is not 10 mL, fill up to 10 mL with distilled water
5. Pour both solutions into test tube
6. Start timer
7. Wait until solution turns blue/black
8. Stop timer
9. Record time taken to react
10. Repeat steps 2–9 2 times with 8 mL and 6 mL KIO_3 solution, respectively
11. Repeat steps 1 – 10 as necessary for data collection

4 Data

Table 1: Data

Molarity	Trial 1 s	Trial 2 s	Trial 3 s	avg
0.020±0.50	8.030±0.50	9.180±0.50	9.510±0.50	8.91±0.50
0.016±0.50	11.000±0.50	11.120±0.50	11.590±0.50	11.24±0.50
0.012±0.50	18.460±0.50	15.820±0.50	15.450±0.50	16.58±0.50

Reaction Speed Relative to the Molarity of KIO₃ Solution



5 Calculations

5.1 Average

$$\langle s(M) \rangle \quad (1)$$

$$\frac{\sum_i s(M)}{i} \quad (2)$$

$$\frac{s(M_1) + s(M_2) + \dots + s(M_{i-1}) + s(M_i)}{i} \quad (3)$$

5.2 Error and Minimization

Error was calculated using sum squared error, which is defined as follows:

$$E = \sum_{i=1}^n (y_i - f(x_i))^2 \quad (4)$$

A program carried out the minimization of the above function, which entails tracing the derivative of the function with respect to a given parameter—defined here as k —to the local minima of the function

$$E' = \frac{\partial}{\partial k} \left(\sum_{i=1}^n (y_i - f(x_i))^2 \right) \quad (5)$$

5.3 Line of Best Fit

When choosing my line of best fit, I first tried to piece together all of the information that I could gather without data. First I knew that my function $?(x)$ had to satisfy the following conditions for the following scenarios:

$$?(x) = \begin{cases} \text{Zero Order:} & [A] = -kt + [A]_0 \\ \text{First Order:} & \ln [A] = -kt + \ln [A]_0 \\ \text{Second Order:} & \frac{1}{[A]} = kt + \frac{1}{[A]_0} \\ \text{nth Order:} & \frac{1}{[A]^{n-1}} = (n-1)kt + \frac{1}{[A]_0^{n-1}} \end{cases} \quad (6)$$

which I simplified to

$$?(x) = \begin{cases} f(x) = a(bx + c) + d \\ g(x) = a \ln(bx + c) + d \\ h(x) = \frac{a}{bx + c} + d \end{cases} \quad (7)$$

I ended up having to assume that the rate of reaction was less than or equal to 2 primarily due to the fact that ternary (third-order or more) reactions are extremely rare, and testing for one would subsequently be impractical.

Judging by the appearance of the graphed data—steep slope for smaller x , small slope for larger x —I had an intuition that the best model would be the second order reaction:

$$h(x) = \frac{a}{bx + c} + d \quad (8)$$

Testing my prediction proved me correct, as the error was much lower for the function $h(x)$ —representing a second order reaction—than for $g(x)$ or $f(x)$ —representing first and zeroeth order reactions, respectively. These results are shown in the table below.

Error		
$f(x)$	$g(x)$	$h(x)$
1.510 02	13.954 68	$6.513\,84 \cdot 10^{-8}$