

# Application Of Computational Methods in Chemical Engineering

## Introduction

This report explores the pervasive impact of computational methods in chemical engineering. Examining advanced algorithms, simulations, and modeling, we uncover how these techniques revolutionize traditional processes, offering innovative solutions for design, analysis, and optimization. In a landscape driven by efficiency and sustainability, the synergy between computational methods and chemical engineering plays a pivotal role. This report provides a concise overview of key advancements and future prospects in this dynamic intersection of science and technology. Hence there are a lot off diverse and wide spread approach of computational methods in Chemical Engineering industry.

## Materials and Methods

In our pursuit to elevate chemical engineering practices, our team embarked on a transformative project leveraging the power of computational methods. This report sheds light on our journey, where Microsoft Excel and MATLAB served as integral tools in our arsenal. Microsoft Excel, with its versatile data handling capabilities, played a crucial role in swift data analysis and visualization, facilitating informed decision-making. Meanwhile, MATLAB, with its computational prowess, enabled us to conduct intricate simulations and modeling, pushing the boundaries of our project's innovation. This report unfolds the story of how our team harnessed the combined strengths of these platforms, showcasing their collaborative impact on the design, analysis, and optimization of chemical processes.

## Results

Our project, employing Microsoft Excel and MATLAB, delivered impactful outcomes. Linear regression effectively addressed bubble point and dew point problems, ensuring accurate thermodynamic solutions. The MATLAB-based Newton-Raphson method enhanced computational efficiency and root-finding precision. Graphical analyses in Excel revealed key trends, notably in Reynolds number versus coefficient of friction and petroleum fraction versus temperature. The high coefficients of determination ( $R^2$ ) ranged from 0.996 to 0.998, affirming the robustness of our models.

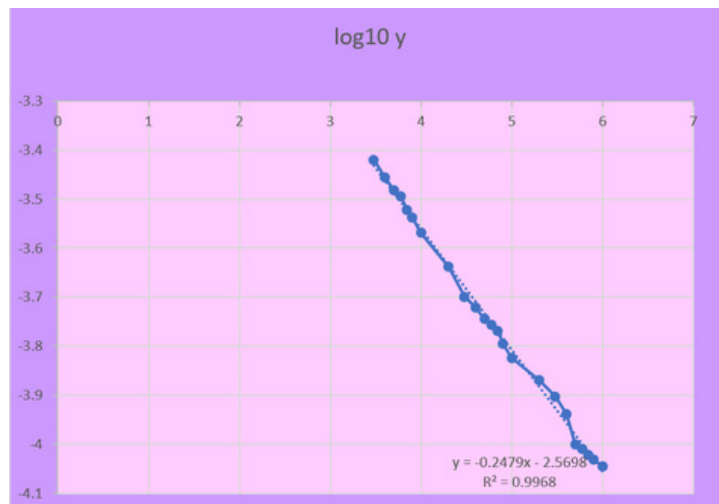
## Conclusion

In summary, our work showcased the efficacy of computational tools in tackling complex chemical engineering challenges. Accurate data extraction, graphical representation, and statistical analysis were pivotal. Discussions on the potential integration of artificial intelligence and neural networks in the industry provided a forward-looking perspective. This project underscores the transformative role of computational methods in shaping the future of chemical engineering.

## Objective:

In our project, the application of computational methods extended across a spectrum of challenges in chemical engineering. Notably, we employed linear regression techniques to tackle bubble point and dew point problems, offering precise solutions to complex thermodynamic scenarios. Utilizing the Newton-Raphson numerical technique, our team developed a MATLAB computational code to efficiently find roots, enhancing the accuracy and efficiency of our calculations.

Data extraction from diverse datasets facilitated comprehensive analyses, and graphical representation in Microsoft Excel enabled us to visualize critical trends, such as the relationship between Reynolds number and coefficient of friction, and petroleum fraction versus temperature. The calculated coefficient of determination ( $R^2$ ), ranging between 0.996 to 0.998, underscored the reliability of our models. We not only extracted valuable equations from these plots but also engaged in insightful discussions on the integration of artificial intelligence and neural networks in the chemical engineering industry during our concluding sessions, exploring avenues for further innovation in the field.



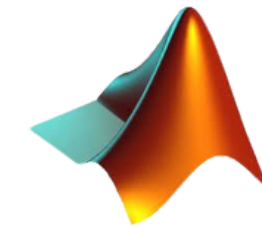
### Matlab Code for parameter estimation

```
% Define your data
mu = [0.25; 0.31; 0.36; 0.43; 0.45; 0.47;
      0.50; 0.52];
x = [0.1; 0.15; 0.25; 0.50; 0.75; 1.00; 1.50;
     3.00];

% Perform linear regression
Y = 1 ./ mu;
a = polyfit(1 ./ mu, 1 ./ x, 1);

% Extract parameters
mu_max = 1 / a(2);
k_max = a(1) * mu_max;

% Display results
fprintf('mu(max) = %f hr^-1\n', mu_max);
fprintf('k(max) = %f g/liter\n', k_max);
```



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