

6. Conclusion

Model Performance Summary:

Model	MSE	MAE	R-squared
Multiple Linear Regression model	5.005	0.78	0.912
Lasso Regression model	4.75	1.28	0.917
Ridge Regression model	4.85	0.77	0.915
Elastic Net Regression model	4.144	1.568	0.927

Based on the analysis of the Algerian forest fire data using multiple linear regression and the provided results for various regression models (Multiple Linear Regression, Lasso Regression, Ridge Regression, and Elastic Net Regression), the following findings can be summarized:

1. Model Performance:

- All regression models achieved relatively high R-squared values, indicating good explanatory power of the models. The R-squared values range from 0.912 to 0.927, suggesting that between 91.2% to 92.7% of the variance in forest fire severity can be explained by the independent variables included in the models.
- The mean squared error (MSE) and mean absolute error (MAE) vary among the models but generally indicate reasonable predictive accuracy, with lower values indicating better performance.

2. Comparison of Models:

- The Elastic Net Regression model achieved the lowest MSE (4.144) among all models, indicating the lowest average squared difference between predicted and actual values.
- The Lasso Regression model had the lowest MAE (1.28), suggesting slightly lower average errors in predictions compared to the other models.
- The Ridge Regression model had the lowest MAE (0.77), indicating even lower average errors in predictions compared to Lasso Regression.

3. Limitations:

- Overfitting: The model may have overfit to the training data, leading to poor performance on unseen data. This could be due to the inclusion of irrelevant variables, insufficient regularization, or too complex a model relative to the amount of training data.

4. Strengths and Recommendations:

- Despite the variations in performance between training and unseen data, the regression models still demonstrate valuable predictive capabilities and offer insights into forest fire severity prediction.
- To improve generalization to unseen data, model refinement strategies such as feature selection, regularization, and cross-validation should be considered.