

# Prog Asg 1: Regression Model for Structural Strength Prediction


Start Assignment


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**Due** 3 Sep by 23:59      **Points** 200      **Submitting** a file upload  
**Available** 18 Aug at 0:00 - 4 Sep at 23:59

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The aim of this assignment is to learn how to apply various regression techniques to real-life problems. You can submit your solution as a Jupyter Notebook with comments and discussions on the results obtained in each step.

Train Data: [https://github.com/gagan-iitb/CS550/blob/main/Labs\\_M23/MaterialStrength\\_Train.csv](https://github.com/gagan-iitb/CS550/blob/main/Labs_M23/MaterialStrength_Train.csv)  
 ([https://github.com/gagan-iitb/CS550/blob/main/Labs\\_M23/MaterialStrength\\_Train.csv](https://github.com/gagan-iitb/CS550/blob/main/Labs_M23/MaterialStrength_Train.csv))

Test Data: [https://github.com/gagan-iitb/CS550/blob/main/Labs\\_M23/MaterialStrength\\_test.csv](https://github.com/gagan-iitb/CS550/blob/main/Labs_M23/MaterialStrength_test.csv)   
([https://github.com/gagan-iitb/CS550/blob/main/Labs\\_M23/MaterialStrength\\_test.csv](https://github.com/gagan-iitb/CS550/blob/main/Labs_M23/MaterialStrength_test.csv))

## [40 marks] EDA and Feature Engineering

Apply various EDA techniques to visualize, pre-process, and clean the data. Study the correlations amongst attributes, perform feature transformations etc. and prepare your dataset for modeling (machine learning). Prepare your validation set OR cross-validation approach that would be used in the remaining part of the assignment for hyper-parameter tuning and/or model selection.

## [30 marks] Exact Solution

Apply the method(s) discussed in class: Normal Equations and Pseudo-inverse to compute the optimal parameters. You may test out various features/combinations in this stage. Please provide the equation for the response variable and its interpretation.

## [20 marks] Statistical Analysis

Using OLS library, study the statistical properties of your model(s). Which attributes/features are significant? What are the confidence intervals for each of them? Write your interpretations and comparison of at least 2 good models.

## [30 marks] Gradient Descent

Write your own code to perform gradient descent and experiment with the learning rate hyper-parameter. Plot the loss and validation curves. Code your strategy for the convergence criterion.

### **[20 marks] KNN**

Use KNN (non-parametric approach). Experiment with various values of  $K$ . What do you observe? How does its validation accuracy compare to the parametric approaches? What are the pros and cons?

### **[30 marks] Generalized Linear Models**

Build at least 2 GLM models with different link functions and distributions. Provide justifications for your choice and interpret your results.

### **[30 marks] Test Output**

Use your validation approach to select the best model and provide the predictions for the test set as a .csv file with only a single column (the output). There should be no header to the file. We would set up a leaderboard and your marks here would be proportional to the accuracy of your predictions.

### **Submission instructions**

- a. No zip files are allowed. No Colab files are allowed. Multiple files can be submitted.
- b. No copying allowed. A plagiarism check will be performed. This will lead to severe penalties.
- c. Naming convention for every file you submit: Roll number\_First Name\_Asg1\_...
- d. Provide justifications and interpretations of your results.
- e. No late submissions are allowed. Start early and submit often (incrementally, unlimited attempts are allowed).
- f. If you are seriously sick (for more than a week) or have an emergency submit a medical certificate and get approval from the instructor., skip the assignment (we will re-scale marks for your other assignments).