

ELL784: Assignment 1

Maximum marks: 6 (+2 Extra credit)

Submission deadline: **17 September, 23:59**

1 Part 1 (3 marks)

The objective here is to implement the concepts of regression learnt in class via polynomial curve fitting. To recap, polynomial curve fitting is an example of regression. In regression, the objective is to learn a function that maps an input variable x to a continuous target variable t .

For the first part of this assignment, we provide a personalised input file that contains data of the form:

$$\begin{aligned}x_1, t_1 \\x_2, t_2 \\ \cdot \\ \cdot \\ \cdot \\x_{100}, t_{100}\end{aligned}$$

The relation between x and t is of the form

$$t = w_0 + w_1x + \dots + w_Mx^M + \epsilon$$

where the noise ϵ is drawn from a normal distribution with mean 0 and unknown (but fixed, for a given file) variance. M is also unknown. You should download your file from http://web.iitd.ac.in/~sumeet/A1/<EntryID>/Gaussian_noise.csv (for example, http://web.iitd.ac.in/~sumeet/A1/2010MT50608/Gaussian_noise.csv). The end goal is to identify the underlying polynomial (both the degree and the coefficients), as well as to obtain an estimate of the noise variance.

The tasks to be accomplished are:

- To begin with, use only the first 20 data points in your file.
- You may use standard functions/libraries in any programming language of your choice: MATLAB, Java, Python etc.
- Solve the curve fitting regression problem using error function minimisation. You can define your own error function other than sum-of-squares error (note that the error function need not be convex). Try different error formulations and report the results. Also try and use a validation approach to characterise the goodness of fit for polynomials of different order. Can you distinguish overfitting, underfitting, and the best fit? In addition to this, obtain an estimate for the noise variance.
- Introduce regularisation and observe the changes. For quadratic regularisation, can you obtain an estimate of the optimal value for the regularisation parameter λ ? What is your corresponding best guess for the underlying polynomial? And the noise variance?

- Now repeat all of the above using the full data set of 100 points. How are your results affected by adding more data? Comment on the differences.
- At the end: what is your final estimate of the true underlying polynomial? Why?

1.1 Part 1EC (extra credit, 2 marks)

You are provided with a second personalised data set, available at http://web.iitd.ac.in/~sumeet/A1/<EntryID>/NonGaussian_noise.csv. It is generated from the same type of polynomial, except that the noise is now non-Gaussian. Can you repeat the analysis for this data set, focusing in particular on characterising the noise – can you figure out what kind of noise it is? Please justify your answer using appropriate analysis.

1.2 Evaluation criteria

- You are required to give a demonstration of regression, the coefficients you've obtained and how you've done so. [You might be asked to reproduce some of the results.]
- Visualise the data and results in meaningful ways. [Gnuplot is a good tool for this.]
- How close did you get to the actual answer? (Which we know, and will tell you during your viva!) If there is a big discrepancy, why might this have happened?
- How well you understood what you are demonstrating and the concepts involved.
- How extensively you have played around with various parameters, and your analysis of the variations in the consequent results.
- The insights that you have gained from your experiments.

2 Part 2 (3 marks)

For this part, you will be trying to model a real-world time-series data set. This contains measurements of a certain quantity at different points in time. (Details of what that quantity is will be revealed later.) The provided data set should be downloaded from http://web.iitd.ac.in/~sumeet/A1/real_ts.csv. This contains 120 time points; however, for 10 of them, the measured value has been removed. In each row, the first value is the date of the measurement (in US format, so month comes before day), and the second value is the actual measurement. Your task is to train a linear regression model (using appropriate basis functions) which can predict the missing values as accurately as possible.

You are allowed to use only linear regression models for this task. Cross-validation, hyper-parameter tuning and regularisation are encouraged to produce better results.

2.1 Evaluation criteria

Evaluation on this task will be based on conceptual clarity, interpretation of results and reproducibility. Your understanding of the algorithm, and the extent of validation and tuning will also be evaluated.

Additionally, your predictions for the 10 missing points are to be submitted to a real-time leaderboard, where you will be ranked on your performance on that test set (in terms of minimising mean-squared error). Details of the leaderboard will be announced in a few days.

3 Evaluation

- You should prepare a report, compiling all your results and your interpretation of them, along with your overall conclusions. In particular, you should attempt to answer all of the questions posed above. Any graphs or other visualisations should also be included therein. If you wish, you may also include code or other materials which are relevant, though this is not required. The submission link is <http://web.iitd.ac.in/~sumeet/submit.html>; put everything into a single zip file or tarball, and name it as per the instructions given there. The submission deadline is **September 17th, 23:59**. Any late submissions will be penalised.
- The schedule for demos/vivas will be announced by your respective TAs, in advance. If for any reason you cannot attend in your scheduled slot, you must arrange for an alternative slot with your TAs well in advance. Last-minute requests for rescheduling will normally not be accepted.