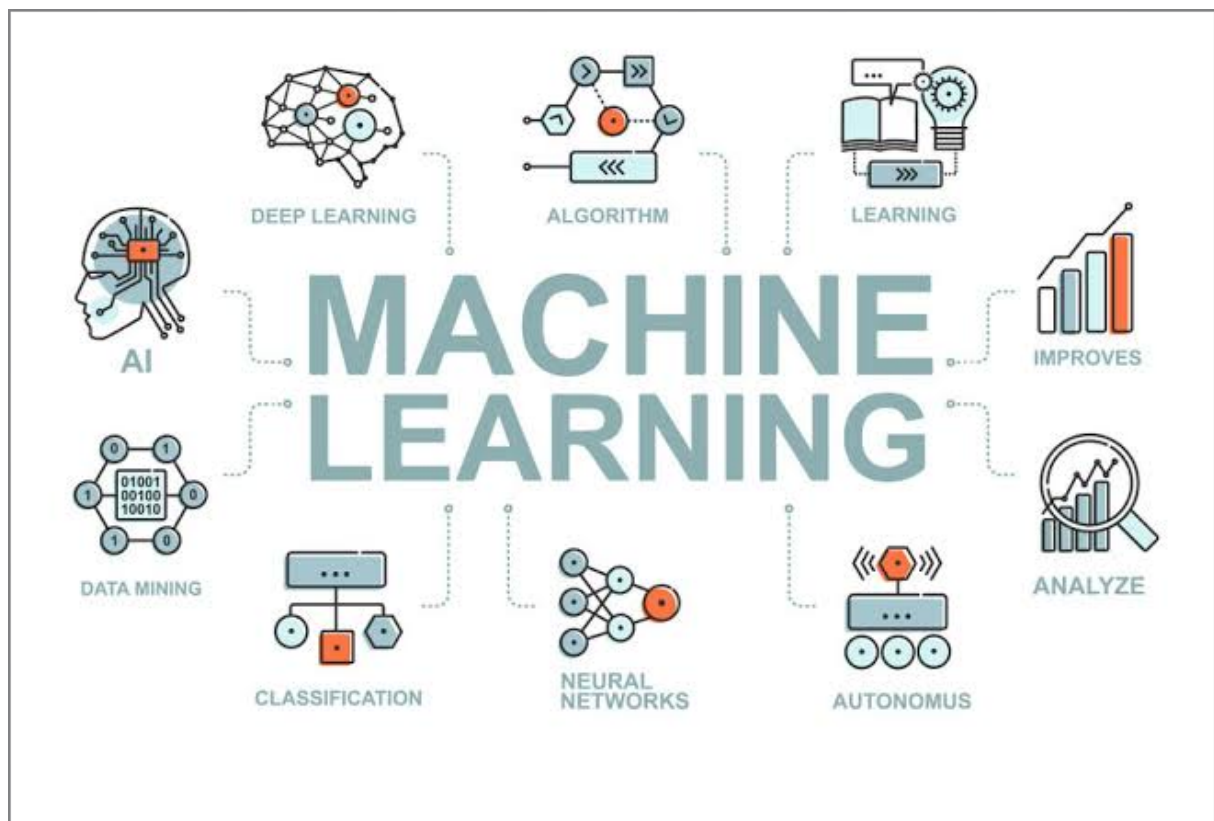


# ML REPORT FOR WOC

*31/03/2023*



By Manav Jain

22JE0538

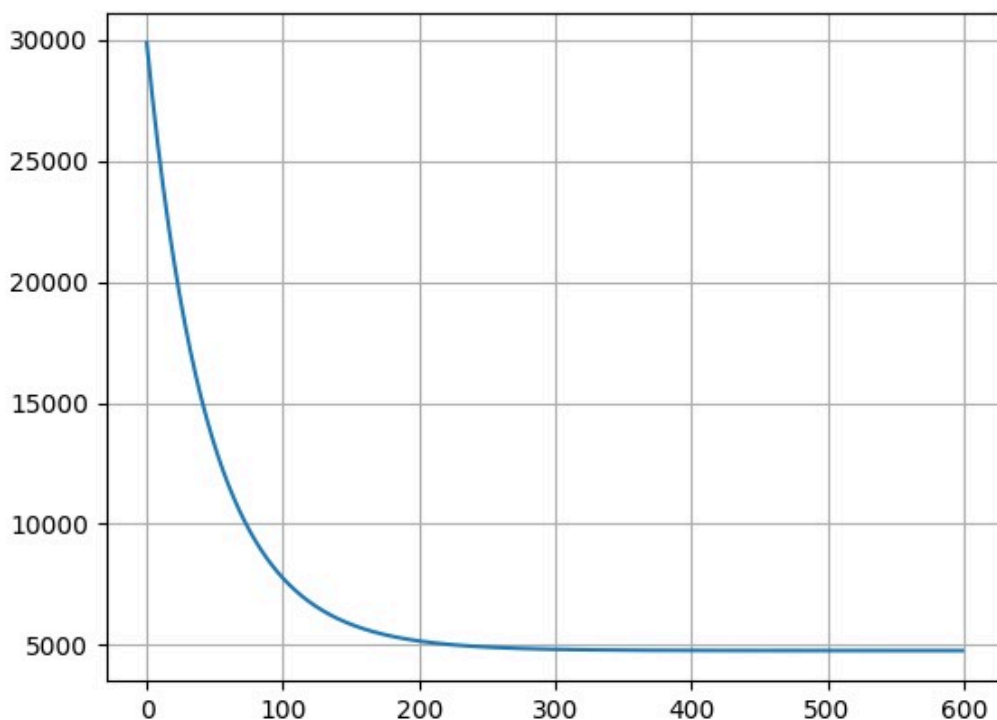
# Progress Through The Weeks

Before the start of coding phase and provision of training and test datasets, I installed jupyter notebook and started getting familiar with its working. I started implementing basic algorithms like single variable linear regression. As soon as the datasets were made available, I started working with them creating primitive algorithms for linear and polynomial regression. Similarly, I created primitive algorithms for logistic regression and started watching machine learning specialisation course. Then, I attended the mid-evaluation, where under the guidance of my mentors, I realised there were many unvectorized spots in my code. I also learnt about  $r^2$  scores and accuracy measurement for classification. Over the next week, I vectorised my code and calculated  $r^2$  scores and accuracies for various algorithms. Vectorising my code reduced the runtime drastically. I also started working on KNN and continued watching machine learning specialisation course. While analysing my code, I realised that a lot of major factors of my code were common through all the algorithms. Then, I started implementing all 5

algorithms in the same notebook. Then, I created python file in which I defined a base class labelled as “base\_class” in which I defined all common parts of all 5 algorithms and created different classes for each algorithm. Using concept of inheritance, I defined the base class as the parent class for all algorithm classes so that I can access all the common parts through base class. After this, I created a notebook in which I imported classes from python file to run my code. After this, I tried implementing neural network and figured out single layer neural network. I was also done with n-layer neural network and spent the rest of the time improving the runtime and accuracy of my code.

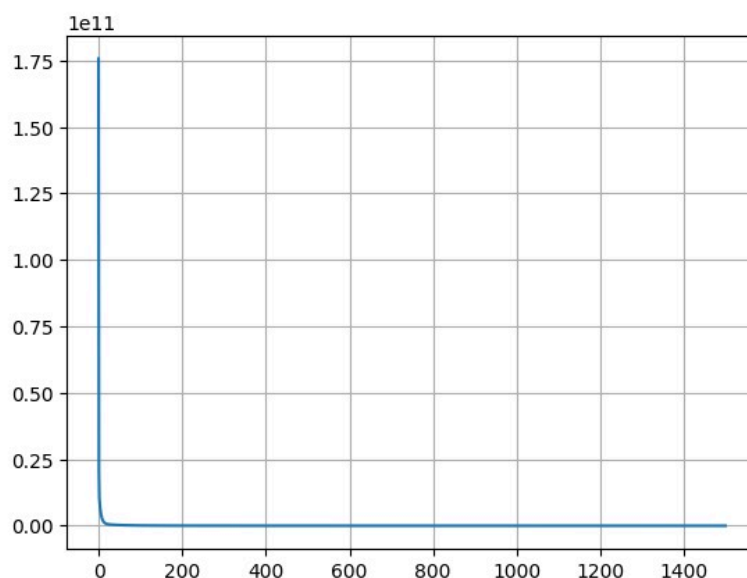
# Linear Regression

For linear regression, I defined the algorithm class as “**linear\_regression**” connected to “**base\_class**” as its parent class. The code is running for 600 iterations with learning rate of 0.01. I splitted my data in the ratio of 80:20(40000:10000) for training and testing. I achieved r2 score of **0.8418251736444053**. The reduced cost function after 600 iterations was **4774.883533808007**. The cost vs iterations graph that I obtained with 600 iterations and learning rate 0.01 is attached below.



# Polynomial Regression

For polynomial regression, I defined the algorithm class as “**polynomial\_regression**” connected to “**base\_class**” as its parent class. The code is running for 1500 iterations with learning rate of 0.0009. I set the degree of the polynomial to be considered as an user-defined input. I splitted my data in the ratio of 80:20(40000:10000) for training and testing. I achieved r2 score of **0.999635677846322** for polynomial of degree 5. The reduced cost function after 1500 iterations at degree 5 was **949981.3327363413**. The cost vs iterations graph that I obtained with 1500 iterations and learning rate 0.01 is attached below.



# Logistic Regression

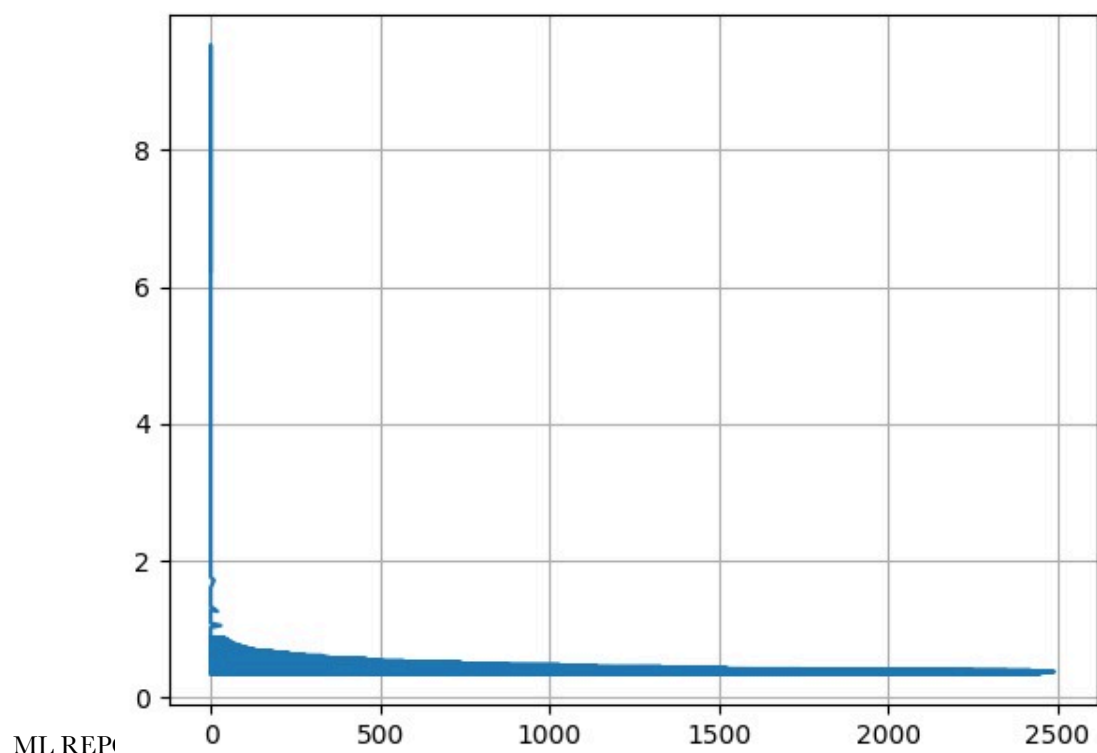
For logistic regression, I defined the algorithm class as “**logistic\_regression**” connected to “**base\_class**” as its parent class. The code is running for 250 iterations each label with learning rate of 0.099. I splitted my data in the ratio of 80:20(24000:6000) for training and testing. I achieved **79.08333333333333%** accuracy using **one v/s all data classification**.

## Knn

For KNN, I defined the algorithm class as “**KNN**” connected to “**base\_class**” as its parent class. I have taken the value of k to be a user-defined input. I splitted my data in the ratio of 80:20(24000:6000) for training and testing. I achieved **84.0%** accuracy for value of k set as 5.

# Neural Network

For single layer neural network, I defined the algorithm class as “**single\_layer\_neural\_network**” connected to “**logistic\_regression**” as its parent class. The purpose of doing so was that now I could access functions defined in both “**logistic\_regression**” and “**base\_class**”. This was necessary as there were certain functions defined in “**logistic\_regression**” that were needed to be accessed in neural network. I splitted my data in the ratio of 80:20(24000:6000) for training and testing. I achieved **83.28333333333333%**. The cost v/s iterations graph is attached below.



This accuracy was achieved after running 2500 iterations at a learning rate of 0.0003. The single layer was of **28 neurons of sigmoid activation**.

For n-layer neural network, I defined the algorithm class as “**n\_neural\_network**” connected to

“**single\_layer\_neural\_network**” as its parent class.

With this, I was able to access functions from

“**single\_layer\_neural\_network**”,

“**logistic\_regression**” and “**base\_class**”. I splitted

my data in the ratio of 80:20(24000:6000) for training

and testing. I achieved **38.2%** accuracy. This accuracy

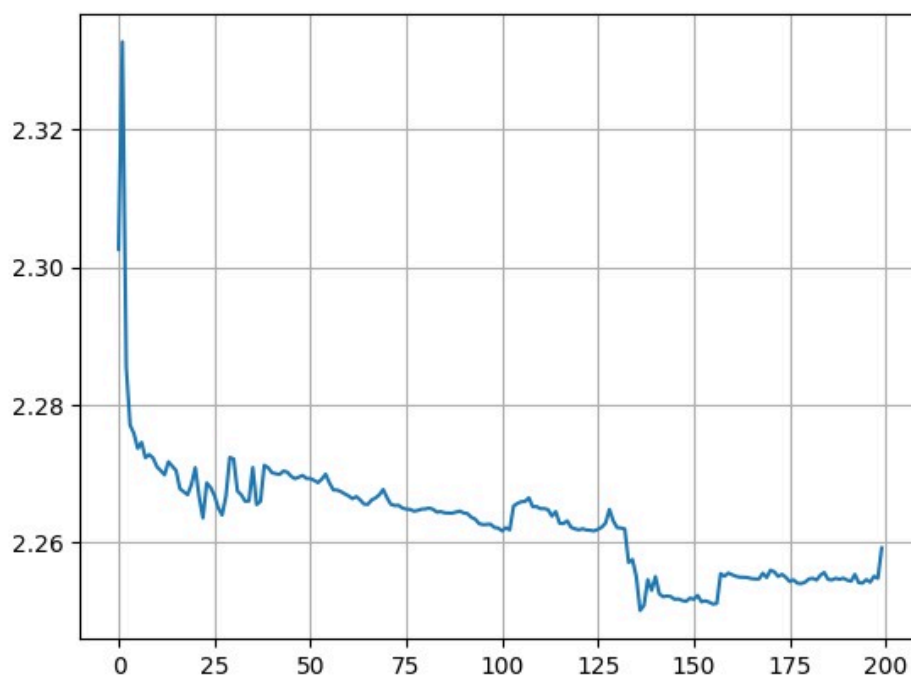
was achieved after running 200 iterations at a learning

rate of 0.09. There were two hidden layers in which the

first hidden layer was of **112 neurons of tanh**

**activation** and the second hidden layer was of **28**

**neurons of tanh activation**.





# About Me

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